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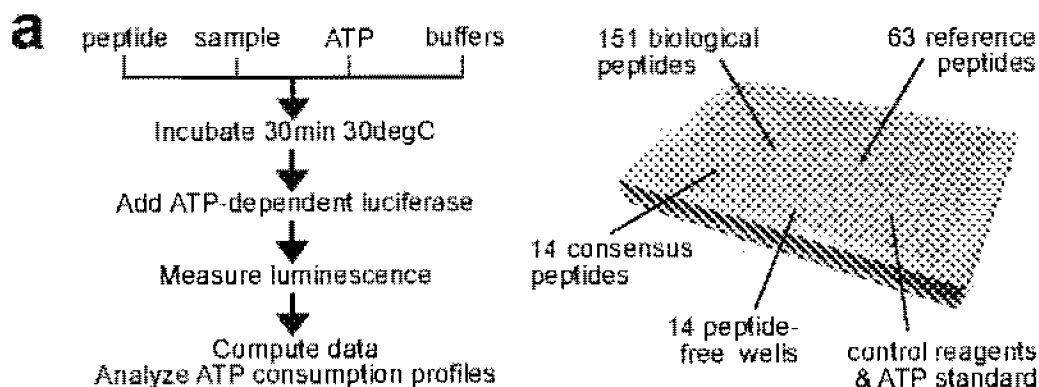


Fig. 1

(57) Abrégé/Abstract:

The disclosure provides methods and compositions for determining the kinase activity profile of a biological sample, e.g., that contains multiple kinases.

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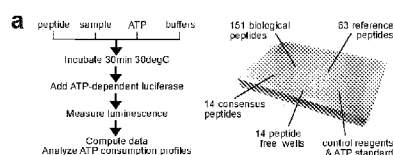


Fig. 1

(57) Abstract: The disclosure provides methods and compositions for determining the kinase activity profile of a biological sample, e.g., that contains multiple kinases.



WO 2019/200245 A1

DETECTION OF PHOSPHOKINASE SIGNATURES

5 CROSS-REFERENCE TO RELATED APPLICATIONS

[0001] This application claims priority benefit of U.S. provisional application no. 62/657,620, filed April 13, 2018, which is incorporated by reference in its entirety for all purposes.

BACKGROUND OF THE INVENTION

- 10 [0002] In a functional sense, cancer is a proteomic disease that arises from selectively diverted signaling pathways^{1,2}. Kinase phosphorylation cascades function as adaptive networks that are re-wired by oncogenic processes³⁻⁷. Half of all cancer-drugs that are FDA-approved or in clinical trials target hyperactive kinases, such as BCR-ABL, BRAF^{V600E} or HER2^{8,9}. While therapeutic decisions increasingly rely on the detection of mutated kinase
- 15 genes or aberrantly expressed/phosphorylated proteins, few experimental platforms are practical enough to directly and comprehensively monitor the activity of kinase enzymes¹⁰, and thus the key actionable dependencies of tumors remain often unclear^{11,12}. A technology capable of identifying the phospho-catalytic signatures of kinases in biological samples could benefit the bio-medical field and improve therapeutic guidance.
- 20 [0003] Proteomic detection systems use the phosphorylatable regions of proteins to infer kinase activity. Antibody-based assays measure (phospho-)protein levels, which depend on availability and specificity of antibodies^{1,13-15}. Mass spectrometry techniques¹⁶⁻²¹, sometimes combined with kinase inhibitors²²⁻²⁵, allow detection of raw amounts of (phospho-)proteins, but remain restricted due to cost, equipment and protocols. Alternatively, generic amino-acid
- 25 sequences are used as individual biochemical probes to directly detect the phospho-catalytic activity of kinases in radioactive-labeling assays, microfluidic electrophoresis systems, ATP-consumption tests, hybrid peptide/phospho-antibody platforms, or SPR and FRET techniques²⁶⁻³³. Readouts from these approaches, however, rely on broad-spectrum consensus peptides originally designed for one-probe-to-many-kinases detection methods, well suited
- 30 for pharmacological drug screens, but not intended to specifically identify or differentiate between kinases' activity in biological extracts.

BRIEF SUMMARY OF SOME ASPECTS OF THE DISCLOSURE

[0004] Provided herein is a new technological resource to distinguish and measure the phospho-catalytic activity of many kinases in parallel. This strategy relies on collections of peptide probes that are derived from the biological target sites of kinases^{34,35}, and are

5 physically used as distinct combinatorial sets of sensors to monitor the activity of kinases in samples. The technology is modular by design: users can adapt probe libraries and assay conditions to their needs. Using a proof-of-concept 228-peptide library, computational methods to analyze phospho-catalytic signatures established from high throughput ATP-consumption measurements are provided in the examples section, which illustrate the present

10 invention. Further, aspects of the invention relate to identification of kinase targets in specific types of cancer. For example, an analysis of BRAF^{V600E} tumors for kinase activity is provided as an illustration. BRAF^{V600E} colorectal cancers (CRC) and melanomas (MEL) remain clinical challenges, with poor prognosis and mostly palliative treatment options, that likely involve kinase-signaling pathways, which can hold the key to new therapeutic

15 opportunities. In one aspect, this approach is used to identify new druggable kinase nodes that drive the unresponsiveness of CRC and MEL to anti-BRAF^{V600E} therapy in cell models and patient tumors. Additional benefits of the present invention will also be apparent to one of skill in the art, *e.g.*, as discussed in the EXAMPLES section.

[0005] In some aspects, the disclosure is thus based, in part, on a high-throughput system to

20 measure the activity of kinase enzymes using their biological peptide targets as phospho-sensors. In some embodiments, the disclosure provides a target peptide set to detect the activity of multiple kinase/kinase families, including, *e.g.*, ABL, AKT, CDK, EGFR, GSK3B, MAPK and SRC, using a multiplex assay, *e.g.*, an ATP-consumption screen, to identify the activity of kinases of interest to be evaluated, *e.g.*, kinase activities in cancer

25 cells. Further, in some aspects, the disclosure provides methods and compositions for the analysis of drug sensitivity, *e.g.*, drug-sensitivity to BRAFV600E-targeted therapy in colorectal cancer; and identifies phosphocatalytic signatures of melanomas for designing and implementing therapies.

30 BRIEF DESCRIPTION OF THE FIGURES

[0006] **Fig. 1. Libraries of biological peptides function as combinatorial sensors to identify, differentiate, and measure the phosphorylation activity of kinases.**

(a) Schematic of the assay procedure.

- (b) Unsupervised hierarchical clustering of phospho-catalytic activity signatures of 25 recombinant kinases established from 228 peptides. For each experimental run, the average value of ATP consumption across the 228 peptides was used for internal normalization. The kinase activity per-peptide was then calculated as the difference in ATP consumption
- 5 between individual peptide-derived values and the overall internal mean. Next, peptide-specific activity values were averaged across independent repeats to establish the phospho-activity signature of each kinase for all peptide sensors. Finally, the 228-peptide-derived phospho-catalytic signatures of the 25 recombinant kinases were analyzed using unsupervised hierarchical clustering. Phospho-catalytic profiles are color-coded based on the relative level
- 10 of activity measured in presence of each peptide, from blue for low-or-no activity, to white for intermediate-or-mean activity, to red for high phospho-catalytic activity. The red/gold/grey peptide color-key on the right side of the phospho-catalytic heatmap indicates the origin of peptides (biological, generic positive control, or random). The number of independent experimental repeats per kinase (n) is listed underneath the heatmap. The same
- 15 concentration of recombinant kinases was used (see Supplemental Methods).
- (c) Pearson-correlation heatmap highlighting the functional relationship between kinase enzymes established from their differential spectrum of 228-peptide-specific phospho-catalytic activities. Kinases are arranged by alphabetical order within their respective families.
- 20 (d) Profile of AUC values obtained for an increasing number of randomized sampling combinations of peptide sensors to predict the identity of HCK. AUC values (y-axis) reflect the performance of HT-KAM assay for predicting HCK's identity by comparing the 7x 228-peptide phospho-signatures of HCK versus the 113x 228-peptide phospho-signatures measured for all other 24 kinases, when relying on one or multiple peptide sensors (x-axis;
- 25 random peptide sampling of a combination of up to 50 peptides out of 228). The AUC values corresponding to Diagonal Linear Discriminant Analysis (DLDA) classifiers that match HCK's specific all-biological peptide subset is shown as a red dot; HCK's all-generic positive control peptide subset as a gold dot; all-random peptides as a grey dot; exact AUC values and number of peptides for each combination are indicated.
- 30 (e) Distinct subsets of peptides can be identified as functional predictors of the differential activity signature of each kinase family. The number and origin of peptides composing the differential phospho-signature of kinases is shown in the bar graph. Peptides were classified as 'predicted' (dark shade) or not (i.e. 'other'; light shade), where 'predicted' defines a peptide sequence previously identified in the literature as a target of a given kinase. For a

biological peptide, it is an amino acid sequence that corresponds to a known phosphorylatable protein region³⁴, and for a generic positive control peptide it is an amino acid sequence that is commercially available. The AUC of each peptide set per kinase is listed underneath the graph.

- 5 (f) Phospho-catalytic profiles measured in presence of the biological peptide targets of AKT1, MAPK1/ERK2 or JAK2 kinase. The average activity per indicated biological peptide across '*n*' independent experiments is color-coded blue-to-white-to-red from low-to-medium-to-high kinase activity. Biological peptide probes are organized by decreasing phospho-catalytic activity per kinase. Each phospho-catalytic heatmap is shown side-by-side with a
10 grey-scale of significance (p-values for a pairwise t-test comparing the activity measured from '*n*' independent experimental runs for each biological peptide vs. the average activity measured with 5-random peptides). Average phospho-catalytic activities obtained from four distinct control peptide groups are shown for comparison at the bottom of each panel (random, Y/S/T-free, reference, or generic positive control).
- 15 (g) Box plots showing that the subsets of biological peptides specific to each kinase are associated with significantly higher levels of phosphorylation activities than activities measured from the pool of 5-random peptides. Box plot distribution of data displays minimum, first quartile, median, third quartile, and maximum. The difference in kinase activity experimentally measured with biological vs. reference peptides was highly
20 significant ($p < 2E-100$; left vs. right set of box plots).
- (h) Functional clustering of kinases based on the 151-biological peptides included in the assay. The analysis is based on all HT-KAM readouts, but only phospho-activity values derived from the 151 biological peptides were used for ranking analysis. Kinases belonging to distant genetic families are distinguishable. ABL and SRC kinases from non-receptor
25 tyrosine kinase families are identified as functionally related. Individual kinases of known distant biological/biochemical relationship with the rest of their genetic families can be singled out (e.g. ABL1^{T315I}, AKT2; MAPK14; SRMS).
- (i) AUC forest-plots comparing the specificity/sensitivity of biological (red) vs. generic control positive (gold) vs. random (grey) peptide subsets. Biological peptides behave as good
30 functional sensors of kinases, and the identity of a kinase can be effectively predicted from the phospho-signature monitored with its subset of biological peptides.
- (j) Enzymatic activity of ABL1 measured in presence of its 11-biological peptides, and either without treatment (left column; UNT) or treated with serial dilutions of ABL1-targeting

inhibitors imatinib and dasatinib (2 sets of 3 columns on the right; concentrations indicated at the top).

- (k) Measurable effects of inhibitors on the activity of ABL1, ABL1^{T315I}, LYN A, AKT1 kinases using their biological peptides as reporters. Shifts in activity are represented as the average (colored dot) +/- StDev of the change in ATP consumption across all biological peptides in comparison to untreated control. While the activity of ABL1 was most reduced by these inhibitors, the anticipated specific effects of inhibitors was apparent by their relative low or lack of effect on ABL1^{T315I}, LYN A and AKT1 kinases (e.g. LYN A was affected by higher doses of Dasatinib). Staurosporine was used as a non-specific control of kinase activity inhibition.

[0007] Fig. 2. Mapping the phospho-catalytic signatures of cancer cells in their native or drug-treated/resistant states, identify their specific kinase dependencies and reveal their differential vulnerabilities to single and combinatorial kinase-targeting therapies.

- (a) Identification of kinases known to participate in the response of BRAF^{V600E} CRC cells (WiDr) to BRAF^{V600E}-targeting therapy (VEM). Western blots for total and phospho- protein MEK1/2, ERK1/2 and EGFR are shown as controls on the left. Change in activity of these kinases after VEM treatment was measured using their respective subsets of biological peptides (shown as heatmaps and bar graphs). Briefly, for each experimental run, the average value of ATP consumption across the 228 peptides and 14 data-points from cell extract alone (i.e. established from 14 peptide-free control wells per 384-well plate) was used for internal normalization. Each sample's kinase activity per-peptide was then calculated as the difference in ATP consumption between individual peptide-derived read outs and the internal mean. Next, the peptide-specific activity values were averaged across independent repeats (3 independent biological replicates tested in 4 independent technical replicates). Finally, the difference in phosphorylation activity per peptide between VEM and UNT profiles was calculated across all 228 peptides. Results are represented as a series of kinase-focused heatmaps using their particular subset of biological peptides (top right panel). Each colored bar represents the differential activity in presence of a biological peptide of the indicated kinase, ranging from blue-to-white-to-red to respectively indicate lower-to-unchanged-to-higher activity after VEM treatment. The number of biological peptide sensors per kinase is indicated underneath the name of the kinase. A relative cumulative index of activity for each kinase was plotted as an indicator of the differential kinase signatures found in VEM-treated samples versus their control counterpart (bar plot on the bottom right; average of differential

- activity values across all kinase-specific biological peptides, divided by the number of peptide sensors). Significance is measured in pairwise Student t-test comparisons between VEM and UNT across all runs (*/#: $p < 0.05$; **/##: $p < 0.01$), and either including all biological peptide sensors (*/**), or 75% of peptides that followed the main activity trend
- 5 (#/###) to elude cross-reaction effects due to parallel feedback activation loops such as AKT or EGFR as illustrated from the biological peptide targets of ERK2 that paradoxically display higher activity (red bars at the top of the ERK2-heatmap) but relate to proteins directly involved in mechanism of therapeutic resistance, such as the EGFR-pathway^{4,36} or TGFBR-pathway⁴⁷.
- 10 (b) Identification of new kinases that mediate intrinsic resistance to BRAF^{V600E}-targeted therapy in WiDr CRC cells.
- (c) Validation of changes in activity of kinase enzymes using western blot. For example, increased phospho-S473 AKT1 relates to increased AKT1 activity, while increased phospho-S9 GSK3B foretells decreased GSK3B activity.
- 15 (d) Response of WiDr cells to combinatorial-targeted therapies. Survival graphs show cell growth sensitivity profiles to dual targeting of BRAF^{V600E} + AKT1 or PDPK1 or PRKCA. Table insets provide characteristics of response to drug combinations: C.I. = combination index following Loewe Additivity model and measured as the average growth inhibition for VEM $\leq 2\mu\text{M}$ and 2nd drug $\leq \text{GI50}$ concentration (arbitrary threshold: synergy $\text{CI} \leq 0.6$; additivity
- 20 $0.6 < \text{CI} \leq 1.0$); D.I. = drug interaction (two-way ANOVA p-val). Drug responses serve as a validation of kinase activity signatures found by HT-KAM screening, as well as a discovery of new vulnerabilities.
- (e) Peptide phosphorylation activity signature of cancer cell lines. For each experimental run, the average value of ATP consumption in sample-containing wells measured across 228
- 25 peptides and 14 peptide-free controls, was used for internal normalization. The activity per-peptide was then calculated as the difference in ATP consumption between individual peptide-derived read outs and the internal mean. Next, phosphorylation activity values measured for each peptide were averaged from ≥ 3 independent replicates for each cell line. Finally, the phospho-catalytic activity signatures measured across the 228-peptide sensors
- 30 were subjected to unsupervised hierarchical clustering. Phospho-catalytic activities are color-coded based on the relative level of activity measured in presence of each peptide for each cell line, from blue for low activity, to white for intermediate-or-mean activity, to red for high activity. The peptide class is indicated as a red/gold/grey color streak on the right side of the heatmap.

- (f) Kinase activity signature of cancer cell lines. For each cell line, the activity of kinases was calculated as the average of the phosphorylation activities measured in presence of their respective biological peptide subsets. The profiles of 60 individual kinases or kinase families detected with ≥ 4 biological peptides are shown. Unsupervised hierarchical clustering was applied across both kinases and cell lines, and color-coded as a blue-to-black-to-yellow scale from low-to-medium-to-high activity. Profiles highlight the heterogeneity of kinases' activity across cancer cells. Control for ATP levels and protein concentrations across cells, and validation of phospho-signatures using complementary computational analysis, immuno-blotting, or drug treatment responses are available in **Figs. 18-22**.
- (g) Comparison of kinase activity and drug sensitivity in two phenotypically distinct BRAF^{V600E} cell lines: A375 (melanoma) vs. WiDr (colorectal), which are respectively considered to be inherently sensitive and resistant to BRAF-therapy. The activity of 10 different kinases was mean-centered between cell lines (y-axis) and compared to their GI50's (x-axis; drug concentration causing 50% inhibition of cell growth in 3-day assays) for 14 inhibitors. Color-coded drug names are indicated underneath each graph. Validation using a second, alternative drug was tested for four of the kinases. All drug concentrations are in μM , except dinaciclib and trametinib in nM.
- (h) Graphical correlation of kinase activity with drug sensitivity. Kinase activity values (y-axis) correspond to activities shown in (c). The x-axis shows the sensitivity of A375 calculated as the Log_{10} of $\text{GI50}(\text{WiDr}) / \text{GI50}(\text{A375})$.

[0008] Fig. 3. Mapping the phospho-catalytic signatures of melanoma tumors reveals druggable kinases predictive of poor outcome and vemurafenib-unresponsiveness in BRAF^{V600E} patients.

- (a) Schematic of the procedure.
- (b) Peptide phosphorylation signatures of patient tissues. Data analysis followed the same steps as for cancer cell extracts. HT-KAM data were clustered using Euclidean distance and ward linkage. Phospho-catalytic activity and peptide annotation are color-coded as indicated in the bottom right legend. Retrospective knowledge of survival outcome, recurrence and treatment-resistance are indicated above each map, along with their BRAF^{V600E} mutational status. Within the deceased group, patients #4-5-9 were VEM-resistant (red squares), and patients #2-4-8 were ipilimumab-resistant (i.e. refractory to anti-CTLA4 immuno-therapy); see **Figs. 23-24** for details.

- (c) Kinase activity signatures of tumor tissues. Analysis followed the same steps used to deconvolute the profiles of cell lines. For each tumor, the activity of kinases was calculated as the average of phosphorylation activities measured in presence of their respective biological peptide subsets (only kinases detected with ≥ 4 biological peptides are shown).
- 5 Kinase activity profiles were mean-centered per tumor and then per kinase across samples. Semi-supervised hierarchical clustering was applied across kinases (tree on the left; Euclidean distance), while maintaining the order of patients established in (b). Kinases' catalytic activity is color-coded as a blue-to-black-to-yellow scale of relative low-to-high activity. Particular kinases are indicated by an arrow/letter on the right, and underline the
- 10 balance between up/down-activation of proto-oncogenic kinases (RPS6KB, PIM, AKT) versus tumor suppressor kinase (GSK3B), either within each patient (top vs. bottom; e.g. VEM-resistant BRAFV600E mutated melanomas), or between patient groups (left vs. right; e.g. good vs. poor outcome).
- (d) Table of kinases whose biological peptides are significantly represented among most
- 15 differential peptides associated with survival outcome. The middle column shows significance based on enrichment analysis EASE – Fisher one-sided test using 34 out of 228 peptide sensors ($p < 0.05$ as significance cut-off). The right column provides significance values (FDR-corrected Student t-test) using all (unselected) biological peptides per kinase subset, and comparing all experimental runs between surviving and deceased groups.
- 20 (e) Biological peptide-derived activity signatures of AKT, PIM, RPS6KB and GSK3B kinases averaged across tumors from VEM-resistant patients #4,5,9.
- (f) Response of BRAF^{V600E} melanoma cell lines to drugs that target vulnerabilities identified by HT-KAM assay. We chose to target AKT, PIM, RPS6KB and GSK3B kinases based on the kinase profiles of VEM-resistant patients shown in panels (c-e). The growth of A375 and
- 25 Sk-Mel-28 cells was tested in presence of a range of concentrations of VEM (horizontal axis; concentrations indicated at the top) combined with a 2nd inhibitor (vertical axis; kinase target, drug name, range of concentrations indicated on the left) in 3-day survival assays. We elected to use Sk-Mel-28 cells because they are inherently highly resilient to BRAF-therapy. The relative scale of growth inhibition (GI) is color-coded (legend on the top left). GI heatmaps
- 30 serve as visual indicators of whether the therapeutic index of BRAF-therapy alone can be improved when combined with a 2nd kinase-targeting drug.
- (g) Streamlined side-by-side comparison of drugs' effects on cells' response to BRAF^{V600E}-targeting. Graphs show the percentage of growth inhibition (y-axis) resulting from combining VEM at various concentrations (x-axis) with a 2nd inhibitor at its GI50 concentration alone

(curves of VEM+2nd drug are color-coded; black lines represent the control VEM alone). The 2nd and 4th graph serve as control response profiles obtained from MTOR- and MEK-inhibitor treatment in A375 and Sk-Mel-28. Rectangles highlight the highest achievable response elicited by these various drug-combinations. Percent cell death at GI50

5 concentrations are provided underneath the graphs.

(h) Characteristics of drug combination effects. D.I. = drug interaction. C.I. = combination index following the Bliss Independence model and measured as the average of all experimental CI values (threshold: synergy CI<0; additivity CI=0; antagonism CI>0)⁴⁸.

10 (i) Primary tumor cells derived from a VEM-resistant melanoma PDX that displays high level of phospho-RPS6KB1 protein (used as readout of kinase activity), are sensitive to RPS6KB-targeting drug in 3-week colony formation assay.

[0009] Fig. 4. Source of peptide probes and arrangement of kinase assay plates.

(a) List of 228 peptides used in the assay. The table provides the connectivity details between kinases and peptide substrates (columns 3-5) for the 151 biological peptides, 14 generic
15 positive control peptides, and 63 reference peptides included in the assay. Peptide ID's and categories are listed in the first two columns. The 25 kinases listed above the color-coded area (right side) are those tested in biochemical assay using purified recombinant kinases.

(b) Numerical-/color-coded connectivity used in table (a). This defines the 'functional' relationship between each peptide lane and each tested recombinant kinase shown in table
20 (a). For instance, a peptide/kinase intersection coded 1-black corresponds to a biological peptide predicted from literature to be phosphorylated by the kinase indicated at the top of table. However, this same biological peptide is coded 2-yellow in adjacent columns, which indicates that other kinases are not predicted to phosphorylate this peptide/substrate target sites. Similarly, a peptide/kinase intersection coded 3-orange corresponds to a generic
25 positive control peptide predicted from literature to be phosphorylated by the kinase indicated at the top of table. However, this same generic positive control peptide is coded 4-teal in adjacent columns, which indicates that other kinases are not predicted to phosphorylate this peptide/substrate target sites.

(c-d) Origin of peptide sensors. The functional relationship between kinases and peptides was
30 previously established from computational curation resources built to mine public databases (see: website at [http address cancer.ucsf.edu/phosphoatlas](http://cancer.ucsf.edu/phosphoatlas); US20120296880^{1,2,15}). Panel (c) summarizes the curation process and molecular connectivity between kinases, substrate proteins, and phosphorylatable peptide target sequences. Panel (d) is a projection of the

coverage of biological peptide targets (left) and generic positive control peptides (right) per kinase across all human kinases/kinase families. Generic positive control peptides were curated from commonly available/advertised single-probe kinase assays or screens. The quantitative and qualitative coverage of peptides per kinase is represented as discs, which are dimensioned based on absolute number of peptide targets per kinase, and color-coded based on uniqueness of peptide targets per kinase.

(e) Biological peptide/kinase connectivity.

(f) Overall arrangement of individual 384-well plates. Besides the 228 peptides, each experimental plate includes: (i) 14 'peptide-free' wells (i.e. sample with buffers but without any peptide), (ii) internal duplicates for a series of peptides-containing wells and peptide-free wells, (iii) assay controls: ATP alone (ATP-loading control), kinase assay buffer alone, and ATP dilution standard.

[0010] Fig. 5. Activity signatures established for all recombinant kinases across all peptide sensors and all experiments.

(a) Repeatability of the assay across 120 different 384-well-plate assays. The graph shows a run-to-run comparison of activity profiles for individual kinases. The reproducibility of the assay was high, as measured by Pearson's correlation coefficient for replicate runs of 228-peptide-derived kinase activity profiles of 25 recombinant kinases, and based on >40,000 single ATP-consumption data points.

(b-e) Example of peptide-derived phospho-catalytic signatures for the JAK2 kinase. Four independent experimental runs are compared. ATP consumption profiles derived from luminescence readouts are shown in (b) and compared in (c). Results transformed as internally normalized data by centering ATP consumption profiles against their overall 228-peptide-derived mean, are shown in (d) and compared in (e). The stack bar graph in (d) visually shows that ATP consumption profiles follow similar trends across all different runs: most bars follow the same activity trend of "all-up/higher-than-the-mean", or "all-down/lower-than-the-mean", or "close-to-the-experimental-mean". This relates to the good correlation (>0.9) found between experiments as shown in table (e), and in graph (a).

(f-k) Cumulative bar plots for the other 24 recombinant kinases we tested. Each plot represents the stacked results of activities observed in experimental repeats and in presence of each individual peptide. In each graph, most bars follow the same "all medium" (around '0', i.e. mean kinase activity measured across all 228-peptides per experimental run), or "all-high" (above the mean), or "all-low" (below the mean) activity trends, which further

demonstrates run-to-run reproducibility and data consistency. Graph-to-graph comparisons (i.e. kinase-to-kinase) allow discerning kinases' specific catalytic signature.

(l) Cumulative bar plot resulting from merging all profiles of all 25 kinases/120 experiments together. Together, the comparison of panels (d,f-l) show that some peptides are associated with high kinase activity for a particular kinase, or a kinase family, or broadly across multiple kinases/kinase families. Some other peptides are associated with minimal kinase activity (see most negative stacked bars). Peptides from the pool of 63 reference peptides were broadly associated with low kinase activity. As also shown in main Fig. 1b, panels (d,f-m) show that differences in pattern and intensity of activity between members of a same kinase sub-family can be identified (e.g. ABL2 vs. ABL1; AKT2 versus AKT1/3; BRK, FRK or SRMS versus other SFK's; ERK2 vs. p38a). The differential activity of distinct kinase isoforms (LYN A versus LYN B) or 'wild type' versus oncogenic kinase (ABL1 versus mutated ABL1^{T315I}) can be found, which functionally corroborates what has been so far inferred from the differential interactomes (PPI) of distinct protein isoforms¹⁶. This shows that kinases affected by alternative-splicing events or 'minimal' genetic / oncogenic mutations can exhibit highly specific and distinctive phospho-catalytic functions (as if encoded by unrelated genes). Such results expand the anticipated complexity of cell signaling networks and their alternative cancer-state¹. These data also illustrate the broad dynamic range (over 4 logs of magnitude) of measurable catalytic activities offered by the use of a large variety of peptides, and observable between different kinases. The differential distribution of 228-peptide-derived activities per kinase reveals that each kinase displays distinctive catalytic capabilities. These data show that the HT-KAM strategy is a robust discovery platform that detects and discerns the unique catalytic properties of different kinase enzymes.

[0011] **Fig. 6. Dilution and time course assays.** A subset of kinases and generic positive control peptide sensors commonly used in the field, were interrogated to assess the quality of the results generated with the HT-KAM assay platform.

(a) Activity of kinases measured at different concentrations and over time.
 (b) Activity of kinases measured over time in presence of generic positive control peptides versus negative control (non-phosphorylatable) peptides. Significance is shown at the top of the box plot. Kinase-dead recombinant proteins or inactive kinase spliced isoforms (e.g. FYN C) were also used as negative controls to validate results from the assay (data not shown). Profiles in (a-b) match standards in the field, and show the quality of the assay was excellent.

[0012] **Fig. 7. Comparison of dynamic range to data variation across peptides.**

(a) Heatmaps of the top-20 peptides associated with the highest measurable activity per tested recombinant kinase. The first column of each kinase sub-panel indicates three peptide categories: (i) an orange square is for a kinase's generic positive control peptide (as advertised/commonly used); (ii) a yellow square is for any other generic positive control peptide out of the 14 generic positive control peptides included in the HT-KAM assay; (iii) a fuchsia square is for a biological peptide. The second column of each kinase sub-panel shows the peptide-phosphorylation activities for the top-20 peptides (red color scale). Activity profiles were normalized following the method applied in **Fig. 1b** and **Fig. 5**, i.e. raw luminescence data were transformed into ATP-consumption data, and then transformed in kinase activity profiles by mean-centering results across 228-peptide sensors per run and per kinase, before averaging profiles for each kinase and comparing significance of peptide phosphorylation profiles across experimental repeats. Activities are shown as a white-to-red scale (saturated red at +100nM above the mean ATP consumption measured across 228-peptide per kinase). In the third column of each kinase sub-panel, FDR-corrected t-test values per peptide probe are shown to assess the significance of activity profiles. The phosphorylation activities measured with each of the top-20 peptides were compared to the pool of 63-reference peptides included in HT-KAM assays across all experimental repeats for each kinase. FDR-corrected t-test values (BH_p) are shown as a white-to-grey-to-black scale (white color cut-off at 0.05; saturated black at 1E-10).

These results indicate that, for many kinases, their best generic positive control peptides have a good, but not necessarily optimal, ability to report on their enzymatic activity. More specifically, the 'best' positive control peptide for 17 out of the 25 recombinant kinases did not correspond to one of their originally advertised generic positive control peptides (i.e. yellow surpasses orange for BLK, BRK, FRK, HCK, LCK, LYN A, SRC, YES1, ABL1, ABL1^{T315I}, EGFR, ErbB2, ErbB4, JAK2, CSK, AKT2, MAPK14). Furthermore, for 18 out of 25 kinases, biological peptides were better reporters than their best, originally advertised, generic positive control peptide (i.e. fuchsia surpasses orange for BLK, BRK, FGR, FRK, HCK, LCK, LYN A, SRMS, YES1, ABL1^{T315I}, EGFR, JAK2, CSK, AKT1, AKT2, AKT3, MAPK1/ERK2, MAPK14/p38a). In many instances, biological peptides can be considered as at least as good sensors of kinases' activity as currently available generic positive control peptides.

(b) Individual comparisons of FDR-corrected t-test across top activity reporting peptides per peptide category. This graph compares the significance (BH-p values) for the highest phosphorylation activities measured with the top peptide belonging to each of the three

categories: (i) best of the advertised/commonly used generic positive control peptide(s) for each kinase (orange dot); (ii) any other best generic positive control peptide (i.e. best of any of the 14 generic peptides other than the ones advertised for the indicated kinase; yellow triangle); (iii) best biological peptide (fuchsia lozenge). Along with data in panel (a), kinase assays show that positive control peptides work well (as expected), but performance can be improved by using/including biological peptide sequences.

(c) Z-factor profiles. Comparing the dynamic range to data variation of 'positive' versus 'negative' controls is a standard method in the field to evaluate the performance of an enzymatic assay (i.e. Z-factor or Z'). We used this method to assess the quality of individual assays included in HT-KAM experimental sets, and to further assess how using different peptide probes impacts assay readout performance. Z' was evaluated by comparing ATP consumption values measured in presence of a peptide probe versus in absence of peptide ($Z' = 1 - (3 * (StDev Pos + StDev Neg) / |Ave Pos - Ave Neg|)$). Z' values from kinase activity profiles were calculated for the three kinds of peptide probes shown in panels (a-b).

Comparing the extent of the assay's dynamic range (~0.1 to ~250nM ATP consumption) to experimental variation (<1 to 20nM ATP standard deviation) using generic positive control peptides versus no-peptide negative controls (i.e. orange dots; measured across the 14 peptide-free wells), showed overall excellent performance of the HT-KAM assay at the level of individual peptide assays (average Z'=0.53). Yet, Z' could be significantly increased up to 0.59 when using "any" best peptide out of the 228 peptides included in the HT-KAM screen, and in particular when including biological peptide sensors (fuchsia lozenges). Along with results from activity and significance profiles shown in (a-b), these data show that peptide sequences derived from biological substrate proteins are well suited to measure the phosphorylation activity of kinases.

[0013] Fig. 8. Profiles of Area Under the Curve (AUC) values obtained for an increasing number of randomized combinations of peptide sensors to predict the identity of an individual kinase or a kinase family.

(a-c) Progression profiles of AUCs for individual kinases and kinase families. In the analysis described below, the 'outcome' we wanted to predict was the identity of a kinase based on peptide activity profiles. For example, for AKT, the analysis outcome was "is this kinase in the AKT family: yes or no?" when using the activity profile from one or multiple peptides where peptide(s) were 'drawn' from any of the 228 peptides, and activities compared across all 25 recombinant kinases and all experimental repeats. To do so, Diagonal Linear

Discriminant Analysis (DLDA) class predictors were built for each kinase or kinase family using randomized combinations of an increasing number of peptides to discriminate one given kinase (or kinase family) from others. The performance of the DLDA classifier was defined as the AUC for predicting the identity of a kinase (or kinase family) from repeated iteration ($i=1,000$) of random peptide sampling. The confidence intervals of AUC's were computed using the DeLong method (from the pROC package of R). AUC values (y-axis) obtained for any number of combined peptides (x-axis) are shown as progression profiles of AUC's (box plots) for individual kinases (**a-b**) and kinase families (**c**). Plots show the effect of increasing n on AUC values, where n random peptides are drawn out of 228 peptides, and where n spans from 1 to a combination of 50 peptides (graphs in panels (**a,c**)) or from 1 to a combination of 100 peptides (graph in (**b**)). We analyzed individual kinases for which enough experimental repeats were executed (here we chose a stringent cut-off of ≥ 6 repeats; i.e. ABL1, AKT1, EGFR, MAPK1/ERK2, MAPK14/p38a, HCK, SRC; panels (a-b)), and for kinases grouped by subfamilies (with sufficient (≥ 6) repetitions per family; i.e. ABL, AKT, HER, MAPK, SFK; panel (c)). For each of the 50 boxplots per graph shown in (a,c), or for each of the 100 boxplots in the graph shown in (b), the line in the middle of each 'box' is the median of the data, the length of the box is 25% and 75% of the data, and the whisker is $(25\% - 1.5 \times (\text{InterQuartile Range}))$ (IQR) and $(75\% + 1.5 \times \text{IQR})$. The outliers (points) are anything that falls outside of that range.

[0014] Results in (**a-c**) show that, for every kinase or kinase family, the sensitivity and specificity of a kinase-assay intended to measure their phospho-catalytic activity, is systematically improved when: (i) including an increasing number of combined peptide sensors; and/or (ii) when using particular subsets of peptides that perform significantly better than other combinations within a particular iterative number of peptides (i.e. higher AUC's within a given n -number of peptides). These results also show that some unique combinations of few peptides can behave as highly specific and sensitive reporters of a given kinase (i.e. these peptide subsets would be located toward the top left corner of each plot; note that the results of this method/analysis do not –however– define which peptide sets differentiate with the highest probability and highest specificity/sensitivity between different kinases, which is a question we resolve later in **Figs. 10, 11, 14, 15** and shows that biological peptide subsets happen to be exquisitely well-suited for). It can be noted from these data that, in the perspective of expanding the coverage of the HT-KAM platform to monitor more/other kinases, the computational analyses and modeled data in (**a-c**) already demonstrate that

including additional peptide probes would only increase the capability of the HT-KAM system to accurately predict the identity of more/different kinases out of their phospho-catalytic activity signatures. Together, these results demonstrate that our strategy of including a multiplicity of peptide sensors considerably improves the sensitivity and specificity of any kinase assay.

[0015] Fig. 9. Kinase phosphorylation activity measured in presence of individual generic positive control peptides. To further show that including a multiplicity of peptide sensors improves the performance of a kinase assay, we asked whether any individual generic positive control peptide could provide the specificity needed to accurately identify individual kinases.

(a) List of kinases whose activities are ‘expected’ to be measured with the positive control peptides included in the HT-KAM assay. The 14 generic peptides included in the 228-peptide library are commonly used as reporters of the phospho-catalytic activity of >63 individual kinases and >27 kinase families (e.g. ABL, AKT, ERK, HER, p38, SFK, TK, etc).

(b) Comparison of kinases’ phosphorylation activity patterns across positive control peptides based on either what is anticipated from literature (top panel), or experimentally measured (bottom panel). In the top panel, the red-colored areas correspond to peptides expected to report on the activity of the indicated kinases, whereas the blue areas are not advertised as such. In the bottom panel, the blue-to-white-to-red heatmap corresponds to the experimental profile of recombinant kinases’ activity levels measured in presence of these peptides. The bottom panel is an excerpt from the complete 228-peptide phospho-catalytic activity signature of kinases. In principle, the red areas from the top panel should match the ‘redder’ areas from the bottom panel, and, as critical, the blue areas from the top panel should match the ‘bluer’ areas from the bottom panel.

(c) Detailed comparison of kinases’ activity level for each generic positive control peptide. The 14 individual graphs display the activity of all 25 recombinant kinases for each of the 14 peptides. In each graph, the red bars indicate catalytic activities expected to be the highest/most positive. Conversely, black bars are not anticipated to display high activity or any activity.

(d) Concordance between expected and experimental activities. The % concordance was evaluated based on whether experimental readouts belong or not to their expected activity groups (vertically: per kinase; horizontally: per peptide). The overall average concordance is ~52%. Cross-reactivity between kinases/peptides observed in **(b-d)** indicates that it would not

be possible to use any individual positive control peptide as a mean to specifically identify or differentiate a particular kinase from other kinases.

- (e) Comparison of individual peptides' AUC to evaluate how specific and sensitive each of the 14 positive control peptides is to identify their respective kinases. AUCs were calculated from repeated iteration of single peptide sampling for kinases tested ≥ 6 times (i.e. HCK, SRC, ABL1, EGFR, AKT1, MAPK1/ERK2, MAPK14/p38a). This analysis compared all 228-peptide phospho-catalytic profiles across all 25 recombinant kinases and all experimental repeats (method described in **Fig. 8**). This method answers the question: "how good is an individual peptide at predicting the identity of a kinase?", and results can be interpreted as:
- "the higher a peptide's AUC is for a particular kinase, the more this peptide (and this particular level of phosphorylation for this peptide) is good at discriminating this kinase from all other kinases". The top section shows all calculated AUCs for all 14 positive control peptides across all interrogated kinases (color-coded scale of AUCs shown on the right; kinase name on top). A black side-bar next to AUC(s) designates positive control peptide(s) expected to report on the indicated kinase. To help compare the performance of individual positive control peptides' specificity and sensitivity, the bottom section displays the AUCs for 14 individual biological peptides with the highest AUCs for each kinase.
- Results show that most individual positive control peptides do not provide the highest possible specificity and sensitivity for the kinases they are expected to report on.
- Furthermore, most individual positive control peptides are outperformed by individual biological peptides. These data provide new leads on how to improve single-peptide kinase assays using (biological) peptides found with the HT-KAM strategy. More critical to the concept underlying our study, these results underline why a multi-peptide approach is a valuable alternative to single peptide-based activity measurements when investigators want to specifically and differentially identify a given kinase from other kinases.

- [0016] **Fig. 10. Systematic identification of combinatorial peptide sets that best differentiate between kinase families.** We implemented computational methods to define which unique set of peptides most significantly distinguishes a kinase from all other kinases. (a-e) Differential phospho-catalytic signatures of kinase families. The phospho-catalytic activity heatmaps of ABL (a), AKT (b), HER (c), MAPK (d), SFK (e) are established from the peptide subsets that best differentiate each kinase family from all other kinases. Peptide-IDs are listed on the right side of each heatmap. A scale of catalytic intensity is shown underneath (blue-to-white-to-red of mean-centered activities). The two color-coded columns

on the left of each heatmap show the type/origin of peptides, and the mean difference in activity: (i) peptide types or origin are shown in red shade (biological), yellow shade (generic positive control), grey (reference); (ii) the mean difference in activity is shown on the left, as an indicator of how significantly more (pink) or less (green) active a kinase family is in presence of the listed peptide that was included in their differential signature.

[0017] The first step of the computational method was to systematically compare each of the 228-peptide activity signatures from all kinases belonging to a given family with ≥ 6 experimental repeats (i.e. ABL, AKT, HER, MAPK, SFK), to the 228-peptide signatures of all other kinases belonging to all other families, and then select peptides associated with the most differential activities based on whether any of the 228 peptide-associated activity values passed or not a significance threshold of $p < 0.05$ for both FDR-corrected t-test and Wilcoxon rank sum test. This implies that: (i) the selected, most differential peptides can be associated with either low, or high, or 'average' phospho-catalytic activities specific to a kinase family if it significantly contrasts with activities observed across all other kinases (following this principle, a peptide can be found as part of the differential signature of multiple kinase families owing activity levels and significances that are specific to the differential signature of its given kinase family versus all other kinases); (ii) the activities from the selected, most significantly differential peptides specific to a kinase family follow a trend that may vary from one individual kinase to another within that family (and between experimental read outs), and some individual kinases may also cluster away from the majority of other kinase family members (which underlines the functional precision of the combinatorial measurements provided by the HT-KAM assay toward the systematic identification of the specific functional attributes unique to a kinase family yet can distinguish sub-family members).

[0018] The second step of this analysis, was to extract all phospho-activity values that match the significantly differential peptides out of the 228-peptide-derived activity values for each individual family, and then apply unsupervised hierarchical clustering to group peptides and kinases based on their functional relationship (classification trees at the top and far left of each heatmap).

[0019] Each panel (a-e) also includes a graph on the top right that shows the Receiver Operating Characteristic (ROC) curve and AUC value calculated for the specific subset of most differential peptides for each kinase family. The method used to assess the performance of each peptide set (i.e. its sensitivity and specificity) was the same as the one used to predict

the performance of increasing numbers of random peptide sets described in **Fig. 8**. We used the 228-peptide / 25 kinases / 120 experiments dataset to build classifiers, and then predict and compare performance using the exact differential peptide set associated with each kinase.

[0020] As an example, panel **(b)** shows that a unique combination of 89 peptides out of 228 peptide sensors best differentiates AKT-family kinases from all other kinases/kinase families. This 89-peptide subset includes both significantly high, low and intermediate activity features of AKT that uniquely differentiate AKT from all other tested kinases. This unique set of 89 peptides is capable of detecting/identifying AKT with a high degree of sensitivity and specificity (i.e. AUC = 0.945). AKT's differential signature includes every type of peptides, whether biological, generic or reference, and whether phosphorylated or not by AKT. We complemented our analyses using Monte Carlo cross validation to further estimate how accurately our predictive model performed, which confirmed the validity of the differential peptide signature (data not shown).

[0021] Results in **(a-e)** show that differential peptide phosphorylation signatures can be assigned to every kinase families. This also demonstrates that using a multi-peptide activity-screening assay such as the HT-KAM strategy can find specific combinations of peptides that can serve as predictors of kinases' identity. As displayed in **(a-e)** and summarized in **Fig. 1e**, these results also indicate that the majority of peptides most capable of discerning kinases' unique phospho-catalytic activities are biological peptides.

[0022] **Fig. 11. Peptide sets that best differentiate individual kinases.** We applied the computational methods designed in **Fig. 10** to identify the subsets of peptides that most significantly distinguish an individual kinase from other kinases.

(a-e) Differential phospho-catalytic signatures of kinases. Differential signatures and AUCs were calculated for individual kinases with ≥ 6 experimental repeats: ABL1 **(a)**, AKT1 **(b)**, MAPK1/ERK1 **(c)**, MAPK14/p38a **(d)**, HCK **(e)**. We also identified the differential activity signatures of EGFR and SRC kinases, which respectively included 42 and 43 peptide features associated with AUC's of 0.840 and 0.893, and that successfully passed the significance threshold ($p < 0.05$) for either FDR-corrected t-test or Wilcoxon rank sum test, but since they did not concurrently pass both criteria, we decided not to include their profiles here.

(f) Bar plot summarizing the number and origin of peptides composing each kinase-specific differential signature. Biological peptides constitute the majority of peptide sensors making the differential catalytic signatures of individual kinases.

[0023] Together, results from **Figs. 8-11** and **Fig. 1b-e** show that using an array of peptides to measure the activity of kinases is an effective approach to functionally distinguish and identify different kinases. Such multi-peptide sensor-based kinase identification system is superior to any single probe enzymatic assay, including assays relying on a generic positive control peptide or any individual peptide in general. The level of differentiability, sensitivity and specificity offered by our multi-peptide platform approach is such that all kinases/kinase families can be predictably identified based on their activity signatures. Even though some of the kinases we chose to examine can be considered as more difficult to distinguish since they are biologically closely related (e.g. SFKs or ABLs), our system was capable of identifying and differentiating them (a functional distinction that no individual peptide was able to predictably achieve). As well, the signatures of kinases of more distant biological functions/unrelated genetic families were more distinct from other kinases and systematically distinguishable (e.g. MAPKs or HERs). In fact, our computational analyses and experiential data demonstrate that if we were to test larger numbers of recombinant kinases belonging to more distant kinase families, their activities would be more divergent from other kinases, and thus easier to measure/identify. While these results imply that the HT-KAM strategy could be used as a discovery tool to identify peptide sensors most uniquely differential and predictive of the identity of a kinase/kinase family, these results also indicate that the biological peptides of kinases are an already available subsets of specific/sensitive/differential sensors that are well-suited to decipher between kinases and their respective levels of activity.

[0024] **Fig. 12. Phospho-catalytic signatures of kinases established from their respective biological peptides.** The computational analyses developed in **Figs. 10-11** and **Fig. 1e** show that biological peptides contribute most to the differential peptide signatures of kinases, so we asked whether kinases phosphorylate their respective biological peptide sequences.

(a-g) Activity profiles of recombinant kinases measured in presence of their respective biological peptides. Individual kinases are organized by kinase sub-families: AKTs (a), MAPKs (b), SFKs (c), ABLs (d), HERs (e), and include JAK2 (f) and CSK (g). The origin of peptides is indicated on the left of each panel. The phosphorylation activity per indicated biological peptide is averaged from 'n' independent experiments. Kinase activity is color-coded blue-to-white-to-red from low-to-medium-to-high activity (i.e. a white/dim color indicates measurable enzymatic activity within the spectrum of low-to-high activity). For each kinase heatmap, activities are organized by decreasing intensity from top/high to

bottom/low. The average catalytic activities obtained from three different peptide groups (random, Y/S/T-free, reference) are shown for comparison at the bottom of each panel. The t-test p-values comparing the activity profiles of each biological peptide to random or Y/S/T-free or reference peptide groups across all experimental runs for a given kinase (indicated by 'n' at the top of each kinase-specific panel), are represented as a grey scale.

(h) Box plots of kinases' activity measured in presence of either their biological peptide subset, or 5-random, or 16-Y/S/T-free, or 63-reference peptides. Each box plot distribution of data displays: minimum, first quartile, median, third quartile, and maximum. Each box plot includes all experimental values across all repeats. Normalization and analyses of profiles follow previously described steps. The number of biological peptides per kinase is indicated underneath each box plot in the top graph. The t-test p-values comparing the overall activity profiles established from biological peptide versus either of the three other peptide groups measured across all kinases, are indicated in brackets on the right side of the graphs (respectively $p < 2E-100$, $p < 2E-110$, $p < 6E-121$).

(i) Table of significance comparing kinases' phosphorylation activity profile established from their biological peptide subsets versus either of the other 'control' peptide groups (i.e. random, or Y/S/T-free, or reference).

Results in (a-i) show that biological peptide probes established from substrate protein regions of kinases were overall associated with measurably and significantly higher levels of

phosphorylation activities than 'control' peptides. More than 94% of all kinase activities measured in presence of their biological peptides were greater than activities measured in presence of the 5-random peptide set. Differences in kinase activity between biological and reference peptide sets were highly significant ($p < 7E-11$ for AKT1, AKT3, MAPK1/ERK2, FYN A, HCK, LYN A, ABL1, and JAK2; $p < 4E-2$ for AKT2, MAPK14/p38a, BLK, BRK, LCK, LYN B, SRC, HER2, HER4). Individual kinase activity values derived from their individual biological peptides were significantly different from the reference peptide set (27.2% with $p \leq 0.01$ and 66.7% with $p \leq 0.05$). Kinase activity levels were also significantly higher when measured in presence of their 'non-modified' biological peptides than with their mutated or pre-phosphorylated counterparts included in the assay (see **Fig. 13** for details).

Finally, phosphorylation activities measured with biological peptides of nearly identical sequences were similar (e.g. MTOR T2446 and MTOR S2448; CDKN1A T145 and CDKN1A S146; RAF1 Y340 and RAF1 Y341; JAK2 Y1007 and JAK2 Y1008; GSK3B S9 and GSK3A S21), which further demonstrates the repeatability of the kinase activity measurements made with their biological peptides.

(j) Unsupervised hierarchical clustering of kinases' functional signatures using only biological peptides. The clustering of kinases (top) demonstrates that the 151 biological peptides included in the assay are excellent functional discriminators of kinases' catalytic signatures. Kinases belonging to a same sub-family most closely resemble each other, while most functionally distant kinases segregate away from each other.

Together, results in (a-j) demonstrate that biological peptides are effective combinatorial sensors to functionally differentiate kinases from each other, and to measure the enzymatic activity of their respective kinases. Kinases are significantly and specifically more capable of phosphorylating a vast majority of their predicted biological peptide targets than reference peptide sets.

[0025] Fig. 13. Comparing the activity of kinases measured in presence of their biological peptides versus modified peptide counterparts.

To further assert the preference of kinases to phosphorylate their biological peptide targets, we compared how significantly different the levels of phosphorylation activity of kinases were when measured in presence of their biological targets versus in presence of the mutated or pre-phosphorylated biological peptide counterparts.

(a) Difference in the activity of kinases comparing pairs of non-modified versus modified biological peptide sequences. In the waterfall plot, each teal bar shows the difference in activity between a biological peptide and its mutated ($Y/S/T \rightarrow G$) counterpart. Each violet bar shows the difference in activity between a biological peptide and its pre-phosphorylated ($Y/S/T \rightarrow pY/pS/pT$) counterpart. Each peptide pair substrate's origin and target site along with its kinase is indicated underneath the graph. Results are derived from all experimental measures for all kinase. The significance of the difference in activity in presence of two different peptides was assessed in pairwise comparisons between experimental runs (grey scale of t-test values underneath the graph).

(b-c) Overall comparison of activity profiles of Tyrosine kinases or Serine/Threonine kinases measured in presence of their predicted Y/S/T-containing biological peptides versus any Y/S/T-free biological or reference peptide. The phosphorylation activities of both Tyrosine kinases and Serine/Threonine kinases are significantly higher in presence of their Y/S/T-containing biological peptides than any Y/S/T-free biological or reference peptides.

[0026] Along with results in Fig. 12, data in (a-c) indicate that the variety of biological peptides and control peptides included in our assay provides a practical framework to systematically validate and compare activity profiles within individual assays, or across

technical repeats, or between different samples. These analyses show that kinases significantly prefer to phosphorylate their ‘un-modified’ biological peptide targets.

[0027] Fig. 14. Using HT-KAM as a ‘discovery platform’ to identify best targets of kinases, finds that biological peptides are systematically included among the most significantly high activity profiles of kinases.

(a) Binary heatmap showing peptides associated with high (red) or low (blue) activity for each kinase. For this analysis, each kinase was tested independently of all others, and we used two separate computational approaches to compare levels of ATP consumption per individual peptide to the pool of 63-reference peptides.

[0028] In the first approach, the average 228-activity data points from all experimental repeats was used in a Kalmagorov-Smirnov (KS) test comparing each 165-non-reference peptides (i.e. 151 biological peptides, and 14 positive control peptides) to the 63-reference peptides (p values with or without BH correction controlling for false-discovery rate). In parallel, the mean and standard deviation (SD) of the 63-reference peptides was computed to then identify which peptides among the 165-non-reference peptides display activity signals >2 fold SD from the mean (>highest 2.5% of reference).

[0029] In the second approach, all experimental replicates (instead of averaging them as in the first method) were used in either a linear additive model with BH corrected p-values from each 165-non-reference peptide versus 63-reference peptide (BH.p.lam<0.05 threshold), or an ANOVA model with BH corrected replicate error.

[0030] The overlapping results of these two approaches and their statistical cut-offs identify the most significantly and stringently selected high –and low– activities per peptide per kinase. This highly conservative method finds that 110 biological peptides act as best probes/sensors of kinases’ individual phospho-catalytic activities. Binomial high/low activity levels are represented as a heatmap in this panel (a; high in red, low in blue), along with the 25-tested recombinant kinases organized by subfamily (indicated at the top), and the identified 124 peptides organized based on unsupervised clustering (correlation tree on the left; peptides probe ID’s/origin on the right).

(b) Analysis of activity profiles established in presence of inhibitors to validate results found in panel (a). Here, the underlying postulate is that, when the activity of a kinase is measured in presence of an inhibitor, any peptide associated with a significant decrease in activity may

be considered as a suitable probe to detect the activity of this kinase. Such data would also retrospectively validate (or not) the previously identified peptides from the method in (a). In the three graphs shown in (b), the model system uses ABL1 or LYN A kinases, and serial dilutions of imatinib or dasatinib.

- 5 [0031] For instance, quadrant A shows a strong correlation between levels of inhibition (captured by the Pearson correlation coefficient between imatinib concentration and ATP consumption for each peptide; y-axis), and the activity level per peptide in an untreated context (x-axis) (R^2 (Fisher(inhibition), activity) = -0.48; $p = 2.75e-14$). Importantly, the relationship between the activity levels of (all 228) peptides incubated with ABL1 in an
- 10 untreated setting and their level of inhibition under increasing concentrations of imatinib, also showed that the peptides that report higher activity levels of ABL1 kinase (i.e. dots located toward the right end of the x-axis; indicating highest ATP consumption), exhibit greater inhibition of that activity in presence of increasing concentrations of imatinib (i.e. dots located toward the bottom end of the y-axis; indicating strongest negative correlation and
- 15 thus strongest inhibition). These peptides (bottom right area) largely overlap with the peptides found with the previous method in (a) (indicated as red dots in quadrant A). For instance, biological peptides JUN Y170, CDK5 Y15, WASL Y256, or BTK Y223 are at the top of the list of peptides most significantly associated with high ABL1 activity, and they also show strong inhibition of activity in presence of imatinib (shown as red dots in (b)), which
- 20 confirms that they are ABL1 substrates targetable by anti-ABL1 therapeutic drug. So, these data are a finding consistent with the initial validation purpose of the analysis. Importantly, this method also finds that the biological peptide targets of ABL1 are most systematically and significantly associated with the measurable response of ABL1 to ABL1-inhibitors (besides the 4 peptides previously mentioned, this method also finds JAK2 Y1007, MAP4K1 Y232,
- 25 ABL1 Y226, TP73 Y99, CDKN1B/p27 Y88, MDM2 Y394, RAD51 Y54 as good reporters of ABL1's activity; shown in main Fig. 1j).

[0032] Similar results were obtained for dasatinib-treated ABL1 (quadrant B) and dasatinib-treated LYN A (quadrant C) using the same experimental validation approach and statistical analysis.

- 30 [0033] (c-d) Comparison of the most repeatable peptide-derived activities of ABL1 to the spectrum of activities established from the differential signature of ABL1. Following the x-axis of the graph shown in (c), the 22 peptides that behave as the most internally robust

sensors of ABL1 activity are represented as 14 green squares + 8 black triangles (data derived from the analysis used to generate panel (a)). Following the y-axis of this graph, the 40 peptides that best differentiate ABL1 activity signature from all other kinases' signatures are represented as 32 purple lozenges + 8 black triangles (data established in Fig. 11a). The dashed rectangles (i) and (ii) outlined in graph (c) match the dashed areas (i) and (ii) in panel (d), where peptides' IDs/origins are listed. The red-filled or red-highlighted-margin squares or lozenges or triangles or circles in (c) correspond to biological peptide targets of ABL1 as defined by PhosphoAtlas and included in the HT-KAM peptide library. Noticeably, all ABL1's biological peptide targets are located on the top right corner of the graph (i.e. reporting on high ABL1 activity). This analysis also finds that peptides associated with significantly low/minimal ABL1 activity correspond to biological targets of Ser/Thr-kinases (dashed rectangle (ii) in the lower left area in (c) and listed in panel (d) as biological peptides established from phospho-target sites: AKT1 T308, MAP2K1 T192, SOS1 S1193, BRCA1 S988, NFkB1 S907, SMAD2 T8, CREB1 S133, EGFR T678). These observations directly support why combinations of 'disparate' biological peptides are intrinsically good discriminators and identifiers of different kinases.

[0034] (e-f) Venn diagrams intersecting results from analysis in (a) (i.e. robustness of peptide-phosphorylation activities) and analysis in Fig. 11 (i.e. most differential peptide activities). Results similar to ABL1 data (e) were found for other kinases, such as AKT1 (f), or SFK's, HER's, MAPK's (data not shown). In all cases, biological peptides act as robust sensors of the differential and individual phospho-catalytic activity signatures of kinases.

[0035] Based on the computational methods we developed and peptide library we designed to showcase our system, we find that biological peptides are systematically associated with most significantly and measurably high activity profiles of kinases.

25 [0036] **Fig. 15. Comparison of the specificity and sensitivity of kinases' signatures established with their biological peptide subsets, or their generic positive control peptide sets, or random peptide sequences.**

(a-b) ROC curves and AUC profiles of kinase families and individual kinases. The ROC curves and AUC values of each kinase measured with their specific subsets of either all-biological (red), or all-positive (gold), or all-random (grey) peptides are shown as lines and annotated within each graph. ROC curves and AUC values were established using the method explained in Fig. 8.

(c) Comparison of AUCs measured across kinases and measured from all-biological (red) vs. all-positive (gold) vs. all-random (grey) vs. differential (violet) peptide sets. Averages, standard deviations and p-val are calculated from AUC's values shown in (a,b) for biological (red), positive (gold) and random peptides (grey), and from AUC's values for the differential peptide subsets (violet) that were analyzed in **Figs. 10-11**.

[0037] Results from these plots show that the combinations of biological peptides are very good predictor of kinases' identity. Specifically, for kinase families ABL, AKT, HER, and MAPK, and for individual kinases AKT1, EGFR, MAPK1/ERK2, HCK, SRC, the AUC values obtained from their biological peptides are excellent (on average >0.94). It can be noted that the somewhat lower AUC observed for SFK's biological peptides (i.e. 0.859) was attributable to peculiar activity differences from individual SRC family members. Indeed, the AUC derived from the predicted biological peptides of SFK's was increased when the most dissimilar members of the SFK family and their related peptides were 'excluded' (AUC augments from 0.859 to >0.92, while singling out FGR and FRK kinases along with SRMS and BRK as majorly distinctive functional subclasses among SRC kinases, and in fact display little overlap among subsets of biological peptide target pools, hence the differential effect on specificity and sensitivity of SFK's catalytic signatures; analysis/data not shown). Following such principle, entire predictions on the functional distance between kinase enzymes (instead of kinase genes) could be made, and could reshape kinase classifications. This concept also relates to the good-yet-somewhat-lower AUC found for the MAPK family, since the two recombinant MAPK's (ERK2 vs. p38a) we studied are functionally/biologically quite distant from each other, and thus their enzymatic activities for their respective biological peptides are fairly dissimilar (which was already noticeable in earlier analyses/figures).

[0038] In addition, it can be noted from panel (c) that the specificity/sensitivity (AUC's) from biological peptide subsets performed almost to the same degree as the differential peptide sets computationally identified in **Figs. 10-11** (e.g. ABL AUC(biological peptide set; n=11) = 0.974 vs. ABL AUC(differential peptide set; n=34) = 0.911; AKT1 AUC(biological peptide set; n=20) = 0.986 vs. AKT1 AUC(differential peptide set; n=27) = 0.977; HER AUC(biological peptide set; n=22) = 0.967 vs. HER AUC(differential peptide set; n=76) = 0.948)). AUC's from positive control peptide subsets performed systematically and significantly less well.

[0039] These results demonstrate that biological peptide subsets are well-suited predictive sensors to specifically, sensitively, and differentially detect their respective kinases. These data underline the strict precision of the HT-KAM assay/data analysis system we developed, and its ability to accurately account for peculiar functional differences between kinases, including between individual kinases otherwise considered as genetically related.

[0040] **Fig. 16. Measuring the effects of kinase inhibitors using biological peptides.** Measuring the effects of inhibitors on kinases' activity is a good way to evaluate the performance of biological peptides as sensors of kinases. So, we tested whether biological peptides of kinases reported on the inhibited activity of their kinases in presence of drugs at multiple concentrations.

(a) Table of IC50 concentrations of imatinib, dasatinib, and staurosporine for ABL1, ABL1^{T315I}, LYN A and AKT1 kinases. IC50 values are indicative averages established from IC50 estimations measured in biochemical assays reported in the literature.

(b) Monitoring the effects of kinase inhibitors on the activity of ABL1, ABL1^{T315I}, LYN A and AKT1 kinases using their respective subsets of biological peptides as sensors of changes in activity. Experimental concentrations of imatinib and dasatinib correspond to 0.01x, 1x and 100x IC50 concentrations specific to ABL1 (i.e. 0.002uM, 0.2uM, 20uM, and 0.01nM, 1nM, 100nM, respectively). These concentrations are thus anticipated to affect the other recombinant kinases tested here (ABL1^{T315I}, LYN A and AKT1) with less or little-to-no effect on their levels of activity. This means that the relative differences in reported IC50 concentrations between ABL1 and LYN A for imatinib (0.2 versus 100uM) and dasatinib (1 versus 9nM) can be used to compare the measurable differences in effects of inhibitors. As well, AKT1 can be used as a 'negative control' for lack of response. Every experimental phospho-signature was initially measured across all-228 peptides, and either in presence or absence of inhibitor (i.e. each sample profile was normalized against the mean activity level measured across 228 peptides, and then averaged across all 3 independent replicates of each of the 8 different conditions for all 4 kinases). Only activities measured in presence of the predicted biological peptide targets of each kinase are shown (11 biological peptides for ABL1; 8 for LYN A; 20 for AKT1). The effect of increasing concentrations of inhibitors can be compared to control untreated (UNT) activity profiles (compare far left column to the two sets of three columns on the right). The sensitivity of ABL1, ABL1^{T315I}, LYN A and AKT1 kinases to imatinib and dasatinib can be assessed from changes in activity measured from their individual biological peptides. The averages of the kinase activity values measured

across their biological peptides in each experimental condition are provided at the bottom-end of the panel (blue-to-black-to-yellow scale of average ATP consumptions).

(c-e) Evaluation of the effects of inhibitors on kinase activity profiles using complementary analytical methods. In (c), each 228-peptide-activity profile of each sample was first

5 normalized against the baseline activity level measured from the 14 peptide-free well controls included in each experimental 384-well plate, and then each single peptide-derived activity level measured in presence of inhibitor was compared to the activity level found in untreated sample (i.e. all activities from all peptides in untreated sample become the baseline '0' value). Results of the average of the differential ATP consumption comparing inhibitor-

10 treated versus untreated for each biological peptide are shown as color-coded kinase activity profiles in (c). (These data were used to plot Fig. 1k.) High concentration (10uM) of staurosporine is included as control of general shut down of kinases' activity.

In (d), entire 228-peptide activity profiles of ABL1^{T315I} comparing untreated versus imatinib- or dasatinib- treated samples were investigated for potential trends in inhibitory effects

15 following the principles and statistical methods used in Fig. 14b. These data validate the overall lack of inhibition of ABL1^{T315I} observed with imatinib using its biological peptides in (b-c). A general dampening effect of dasatinib on ABL1^{T315I} activity was observed (bottom graph in (d); almost all 228 peptide probes display a negative R²), which may relate to lower activity levels measured with some of the biological peptides shown in (c).

20 In (e), the analysis used ATP consumption levels measured in presence of peptides most robustly associated with high activity of ABL1 (derived from the analysis described in Fig. 14a, and including biological peptides). Results directly validate the gradually inhibited activity profiles found in panels (b-c).

[0041] These data also enabled cross comparing the differential sensitivity of kinases to

25 these different drugs. For instance, the mean correlation coefficient (meanR²) relating ATP consumption and dasatinib concentration was -0.997 (SD=0.003), -0.766 (SD=0.21), -0.33 (SD=0.09) for ABL1, LYN A and ABL1^{T315I} respectively (calculated using peptides associated with kinases' highest catalytic activity, including their biological peptides). So, dasatinib had the strongest inhibitory effects on ABL1, followed by LYN A, and had some

30 limited but measurable effect on ABL1^{T315I}. As an additional validation of these data, a formal comparison of the t-test to the Fisher-transformed correlation coefficients established from the levels of drug inhibition measured either with the high-activity peptide group, or with the 63-reference peptide group (which is one of the principal peptide control groups

used to analyze HT-KAM readouts), showed that dasatinib inhibited significantly more the subsets of (biological) peptide targets of ABL1 and LYN A than the reference set (respective T-test (fisher): $p=1.087e-07$ and $p=3.081e-06$), but the effects of dasatinib did not differ significantly between these peptide groups for ABL1^{T315I} (T-test (fisher) $p=0.41$; which is explained by the general dampening of activity profiles in (d)). Based on experimental conditions tested here, data indicate that the IC50 concentration of dasatinib for ABL1 may be relatively higher than the anticipated 1nM (but far lower than 100nM), while IC50(dasatinib) for LYN A is between 1 and 100nM (and probably close to the anticipated 9nM). Using the same series of comparisons and peptide sets, the imatinib drug could be identified as a strong inhibitor of ABL1 (mean(R^2)=-0.847; SD(R^2)=0.111), with no effect on ABL1^{T315I} (mean(R^2)=0.66; SD(R^2)=0.025; which is the basis of the failure of imatinib therapy in leukemia), but with measurable, although limited, inhibitory effect on LYN A (mean(R^2)=-0.462; SD(R^2)=0.49). T-test to the Fisher-transformed correlation coefficients measuring levels of drug inhibition formally showed that imatinib inhibited significantly more the subsets of (biological) peptide targets of ABL1 and LYN A than the reference peptide set ($p=0.00252$ and $p=2.054e-06$, respectively). Conversely, the peptide sensor group appeared to be more activated by imatinib for ABL1^{T315I} ($p=0.027$), which may relate to some of the observed increased activities monitored with biological peptides in (c). (Such phenomenon was also observed for imatinib- and dasatinib-treated AKT1 (bottom heatmaps in (b-c)), while AKT-targeting drug MK2206 treatment effectively decreased AKT1 activity (data not shown) in control assays using the same analytical methods presented in (b-e).) Based on the set of drug concentrations used in this experiment, the IC50(imatinib) for ABL1 may be higher than the anticipated 0.2uM (but far lower than 20uM), while IC50(imatinib) for LYN A may be lower than the anticipated 100uM (possibly close to 20uM).

[0042] Such kind of analyses also allowed comparing the effects of different drugs on a given kinase. In particular, these results showed that dasatinib is a more potent drug than Imatinib for ABL1, and this is also true for LYN A or ABL1^{T315I}. For example, the biological peptides of ABL1 revealed a stronger response to dasatinib than imatinib, which is corroborated by statistical analysis (dasatinib vs. imatinib: mean(R^2)= -0.997 vs. -0.847; T-test (fisher) $p = 1.087e-07$ vs. 0.00252). Moreover, representations in (b,c,e) indicate that the kinetics of response to drug treatment are not identical between different peptides, which suggests that the avidity for, and the phosphorylation of, particular peptide sensors may be differentially affected by an inhibitor.

[0043] These results indicate that drug-responses established from arrays of peptide probes provide the confidence and reliability that any single-peptide readout may not offer. These data also demonstrate that the HT-KAM can be used as a functional screen to assess the differential response of different kinases to inhibitors, which further illustrates the performance of the HT-KAM assay and its potential utility toward pharmacological screens. All these results directly show that biological peptides are excellent sensors to measure the activity of their cognate kinases.

[0044] **Fig. 17.** Measuring the effects of SFK-inhibitors using SFK's biological peptides as sensors of drug response.

- (a) Activity of FYN A, HCK, LCK, and SRC kinases treated with PP2 (20nM) and SU6656 (250nM) inhibitors. Literature reports indicate that IC50 concentrations of PP2 and SU6656 are (5nM; 170nM) for FYN A; (5nM; not available) for HCK; (4 to 30nM; >5uM) for LCK; (35 to 100nM; 280nM) for SRC. Differences in drug sensitivity can be observed between kinases. Data overall corroborate previously reported ranges of sensitivities for these drugs.
- (b) Activity of HCK treated with serial dilutions of SU6656 (1nM to 100uM). Of note, the sensitivity of HCK to SU6656 was more pronounced when measuring it with biological peptides than when monitoring it using HCK's generic control positive peptides (n=4; data not shown). In (a-b), results were processed using the same analysis described in Fig. 16c. Together, these results show that the biological peptides of SRC kinases are adequate sensors to monitor the effects of SRC kinase-inhibitors.

[0045] **Fig. 18. Multi-peptide-based identification of kinases that mediate intrinsic resistance to BRAF^{V600E}-targeted therapy in colorectal cancer (CRC) cells.**

- (a) Schematic of the experimental procedure.
- (b) Previously identified mechanisms involved in resistance to BRAF^{V600E}-targeted therapy^{3,17}. The anticipated reduction in MEK1/2 and ERK1/2 protein phosphorylation (blue shading), and increase in EGFR protein phosphorylation (red shading), served as validated knowledge to assess whether the HT-KAM platform can effectively detect MEK / ERK / EGFR kinases' activity in biological samples after vemurafenib (VEM) treatment.
- (c) Baseline levels of protein and ATP concentrations comparing untreated (UNT) vs. VEM-treated WiDr cell extracts across all samples. VEM and UNT samples were comparable.
- (d) Comparison of the activity levels of MAPKs, HERs, AKTs derived from kinases' biological peptide subsets (x-axis), versus those derived from peptides associated with kinases' most differential signature (y-axis; peptide sets previously computationally

identified in **Figs. 10-11**). The table on the right of the graph indicates the number of peptides per kinase. In the graph, the x- and y- axes represent a scale of the differential activity profile between VEM and UNT samples across all experimental repeats (total of 12 UNT and 12 VEM HT-KAM runs). Data plotted on the x-axis display the average of the activity differences across all experimental repeats and all biological peptides per kinase. Data plotted on the y-axis use kinase activity levels calculated as the difference between (i) the mean activity measured from the peptide subset that specifically differentiates the kinase in question from all other kinases, and that is associated with greater phosphorylation by the recombinant kinase in question, 'minus' (ii) the mean activity measured from the differential peptide subset specifically associated with lower phosphorylation by this kinase.

[0046] For example, after VEM treatment, the 'increase' in activity of AKT1 kinase (red mark) is +82 based on the 20 biological peptide targets of AKT1 (x-axis), and +46 based on the difference between the average activity measured from AKT1's 21 most differentially high-activity peptides 'minus' the average activity measured from AKT1's 6 most differentially low-activity peptides (y-axis).

[0047] The plot also incorporates activity profiles established from two different normalization methods. One normalization uses the average value of ATP consumption across 228 peptides + 14 peptide-free cell extracts (circles). The other normalization uses the average across 16 Y/S/T-free peptides (squares). Both normalizations lead to overall similar outputs of VEM-UNT differential activity profiles. Correlation between peptide subsets and for each normalization is indicated below the table, and the trend line of the linear regression between all data points is shown (overall correl. = 0.757).

[0048] (e) Comparison between changes in phospho-protein levels measured by western blot (top section), versus the changes in kinase activity measured with generic positive control peptides (middle section), versus the changes in kinase activity measured with their subsets of biological peptides (bottom section). Each blue-to-white-to-red color band represents the average of the difference in phospho-activities per peptide between VEM and UNT samples and across all experiments (i.e. UNT equals '0'; not shown). A gray side bar is drawn by the top-75% biological peptides that follow the main activity trend. Western blots serve as validation of kinases activity profiles.

[0049] (f) Formal comparison between changes in phospho-protein levels by western blot (x-axis), and the differences in levels of phosphorylation activity for individual kinases (y-

axis) based on either their generic positive control peptides (triangles), or biological peptide subsets (circles), or differential peptide subsets (squares). The identity of kinases is color-coded (markers' legend are shown in between the graph and the table). We used Image J to quantify intensities of bands identified by western blots. Phospho-protein amounts (x-axis) are plotted against the mean difference in kinase activity (y-axis; VEM 'minus' UNT) measured with 'n' peptides belonging to each category (indicated in the table on the right) across all experiments. In the case of GSK3B, we used the opposite value of the western blot quantification since higher phosphorylation of GSK3B is associated with a decrease in its activity.

[0050] Results show that kinase activity values established from either biological or differential peptide subsets correlate with protein levels: circles and squares are located either in the top right quadrant (i.e. high phospho-protein amount is associated with high kinase activity after VEM; which includes: EGFR, AKT1, PDPK1, SGK1, PRKCA, PKN1/2), or in the bottom left quadrant (i.e. low phospho-protein amount is associated with low kinase activity after VEM; which includes: MEK1, ERK2, p38a, GSK3B). Unlike kinase signatures established from biological or differential peptide subsets, positive control peptides did not correlate with kinases' activity (see triangles in the top left quadrant and the bottom right quadrant; e.g. positive control peptides did not identify ERK2 and EGFR kinase as respectively less and more active after VEM treatment). This is also visible in panel (e) (phosphorylation profiles of positive control peptides are shown below western blots).

[0051] Together, results in (d-f) show that the HT-KAM system can identify and differentiate individual kinases and kinase families from each other in cell extracts, which validates the utility of the HT-KAM platform. The different biological peptide subsets of different kinases can effectively be used to measure the activity of their related kinases in biological samples.

[0052] (g) Summary of the functional state of kinases found in WiDr BRAF^{V600E} CRC cells treated with VEM using HT-KAM. Arrow linkers represent the functional connectivity between kinases, and colors indicate activity (e.g. VEM induces higher AKT activity, which leads to GSK3B phosphorylation, which reduces GSK3B activity).

[0053] **Fig. 19. Levels of phosphorylation activity measured with peptides derived from functionally related proteins or pathways.** We provide examples of phospho-

activities measured with individual biological peptides that we purposely selected and organized by protein of origin or by signaling pathway of interest.

(a) Peptide sensors related to EGFR-signaling (EGFR, PLCG1, GAB1, CBL, GRB10).

(b) Peptide sensors related to TGFBR-signaling (SMAD2 transcription factor).

5 (c) Peptide sensors related to transcription factors/regulators (CREB's, FOXO's, cFOS, STAT's, SPI, JUN's, NFkB1, cMYC).

(d) Peptide sensors related to kinase proteins: MTOR, JAK's.

(e) Peptide sensors related to phosphatase proteins: CDC25C, PTEN.

10 [0054] In all panels, the levels of phosphorylation activity measured with individual peptide sensors correspond to the differences in activity measured between VEM-treated and UNT-control samples. Significance per peptide (or peptide group) comparing all experimental read outs across samples and repeats is indicated (pairwise t-test p-val comparing UNT vs. VEM profiles).

15 [0055] Based on these observations, it is tempting to link levels of phosphorylation measured with biological peptides to the functional state of the signaling circuits these peptides originate from, and how they relate to mechanisms of VEM-resistance (e.g. EGFR, CDC25C³, TGFBR¹⁸, MTOR). Some of these biological peptides also happen to be those that are 'paradoxically' more phosphorylated within the peptide-phospho-signatures of kinases that otherwise exhibit low activity profiles after VEM treatment (e.g. ERK2's biological
20 peptide subset includes peptides related to EGFR-reactivation circuit (GAB1 T476, GAB1 T312, GRB10 S150 peptides) and TGFBR-pathway (SMAD2 S250 peptide); see **Fig. 18e** first panel on the left). It would be interesting to investigate whether such peptides are early signs of re-activation of particular kinases (e.g. re-activation of MAPK signaling contributes to insensitivity to RAF inhibition^{17,19}).

25 [0056] Inspection of the results for individual biological peptides demonstrate the robustness of the phospho-catalytic profiles established from cell extracts: (i) similar activity levels measured with biological peptides of similar sequences (e.g. peptides from MTOR, or JAK1, or CDKN1A, or CDC25C), (ii) systematically higher activity levels measured with biological peptides than with their mutated/pre-phosphorylated counterparts (e.g. EGFR,
30 PLCG1, GAB1, GRB10, STAT1, GRIN2B, ABL1, MTOR; data not shown). The phosphorylation state of many biological peptides could also be corroborated by western blots, for example for EGFR (Y1068) (see **Fig. 18e** third panel), MTOR (S2448) (d), and

others for which western blots are not shown (e.g. upregulation of phospho- EGFR (Y869), PLCG1 (Y1253), CBL (Y774), FOXO3 (S253), and downregulation of phospho- cMYC (T58), JUN (Y170), CDC25C (T48)).

[0057] Fig. 20. Response of BRAF^{V600E} CRC cells to targeted therapy combinations.

5 To validate the kinase activity signatures of VEM-resistant WiDr cells found by HT-KAM screen, and as a mean to assess the role of these kinases (AKT1, PDPK1, PRKCA, SGK1) as mediators of the response to VEM and their potential value as druggable vulnerabilities in our model, we used 3-day cell viability assays to monitor the response of cells to drug combinations.

10 (a) Summary graph of WiDr cell growth response to VEM at 2uM (=GI50 concentration) or 0.25uM, and either used alone or in combination with other inhibitors. The bar graph is derived from complete cell growth response data curves (as shown in **Fig. 2d** or in the VEM+EVE graph shown underneath the main panel in (a)). All results were confirmed with PLX4702, a compound related to VEM (data not shown). Results show that lower than
15 single-agent-GI50-doses resulted in effective combinatorial cell growth inhibition. These data validate that the significantly hyperactive kinases found by HT-KAM are biologically meaningful targetable vulnerabilities.

(b) Table summarizing the characteristics of WiDr cells' response to drug combinations. The significance of drug interaction (two-way ANOVA; p-val) and combination index (CI; following the Loewe Additivity model^{6,7}; arbitrary threshold: synergy $CI \leq 0.6$; additivity $0.6 < CI \leq 1.0$; average for VEM $\leq 2\mu M$ and 2nd drug $\leq IC_{50}$ concentrations) are listed.

(c) List of other drugs and additional BRAF^{V600E} cell lines intrinsically resistant to VEM that we tested to further validate kinase hits identified by HT-KAM profiling.

25 **[0058]** These results confirm that the kinases found by HT-KAM are involved in the response and resistance to VEM in BRAF^{V600E} CRC. Together with data in **Figs. 18-19** and **Fig. 2a-b**, these results show that a finite number of key functional dependencies can be identified as additional mechanisms participating in therapeutic resistance, which offers a selective choice of effective targets to explore. So, HT-KAM can be used as an exploratory/discovery platform to survey kinases and their activity levels, which is especially
30 valuable in the case of diseases that cannot be defined by a single driver mutation or individual genetic dependencies such as BRAF^{V600E} CRC. As such, the HT-KAM system can help select drug candidates that target orthogonal modes of resistance with high likelihood of

restoring therapeutic sensitivity, thus providing a rational approach to design combination therapies.

[0059] Fig. 21. Ranges and levels of peptide phosphorylation measured across 20 cancer cell lines.

5 (a) Peptide phosphorylation activity signature of cancer cell lines. Protein extracts were generated for 20 cancer cell lines. Their phosphorylation activity was tested on the HT-KAM platform. For each experimental run, the average value of ATP consumption across the 228 peptides and 14 data-points from cell extract alone (i.e. established from 14 peptide-free control wells per 384-well plate) was used for internal normalization, and then the activity
10 per-peptide was calculated as the difference in ATP consumption between individual peptide-derived read outs and the internal mean. Next, phosphorylation activity values measured for each peptide were averaged across experimental repeats for each cell line. Finally, the phospho-catalytic activity signatures measured across the 228-peptide sensors were subjected to unsupervised hierarchical clustering. Phospho-catalytic activities are color-coded based on
15 the relative level of activity measured in presence of each peptide for each cell line, from blue for low activity, to white for intermediate-or-mean activity, to red for high activity. This analysis showed that each cancer cell displayed a unique phospho-catalytic fingerprint. On the right side of the main heatmap, the peptide class is indicated as a red/gold/grey color streak. At the bottom of the main heatmap, the phosphorylation activity profiles of cell
20 extracts without peptide (i.e. measured in the 14 peptide-free control wells) are shown.

[0060] (b-d) Examining the patterns of phosphorylation activity across cancer cells. The phospho-catalytic activity profile of cancer cell lines was examined for three technical variables: range of activity per peptide probe, level of phosphorylation intensity per peptide probe, and peptide class. The underlying reason is that, for an assay to best distinguish the
25 phosphorylation activities of different samples, it would be most appropriate to rely on phospho-sensing probes that capture the widest possible dynamic range of phosphorylation activities between cells, and/or provide the overall highest level of phosphorylation activity across cells. While asking this question, we also considered the possibility that the results of these two first variables may be different depending on the class of peptide used in the HT-
30 KAM assay, that is: biological peptides, or positive control peptide probes, or reference peptides.

[0061] In (b), results are sorted by peptide class, and then by highest to lowest range of phosphorylation activity per peptide.

[0062] In (c), the table lists the values calculated from (b) to compare the three-peptide classes.

[0063] In (d), results are sorted by highest to lowest average phosphorylation activity per peptide. Peptide classes are then grouped and counted in the lower graphs to provide a sense of which peptide class reports on the highest to lowest levels of phosphorylation activity.

[0064] Results in (b-d) show that biological peptides provide the broadest ranges and highest levels of measurable phosphorylation activities. Reference peptides provide the lowest activity measurements overall, while still providing a broad range of phospho-activities. Generic positive control peptides provide the lowest range of activities, and the measured phospho-activity levels remain lower than the overall mean activity per cell line across all cells. We conclude that, although virtually all peptides participate in defining the unique peptide-phosphorylation activity signature of every cancer cell line, the set of biological peptides are particularly well-suited catalytic activity sensors to measure and differentiate the unique functional phospho-fingerprint of cancer cells.

[0065] (e) ATP levels in peptide-free wells and protein concentrations per cell line. In the graph on the left, each bar represents the average $\pm$ st dev of ATP concentration of cell extract alone measured across the 14 peptide-free control wells available for each experimental 384well/plate. All samples for each cell line were tested at least three times. The range of ATP-standard across all HT-KAM assays is shown on the far right of the graph (averages and standard deviations across all HT-KAM experiments are shown). These results show that ATP profiles of all samples fit within the limits of the range of ATP standard. ATP profiles were comparable between different cell lines or samples or experimental repeats. There is no evidence for any ATPase or phosphatase contamination. So, variations in phospho-catalytic activity measured in presence of peptides were both peptide-dependent and cell-specific (results shown across all figures). As such, phospho-catalytic activity profiles can be interpreted with confidence.

[0066] **Fig. 22. The kinase activity signature of cancer cells identifies druggable vulnerabilities.**

(a) Kinase activity signature of cancer cell lines. The activity of kinases was calculated as the average of the phosphorylation activities measured in presence of their respective subset of biological peptides. The kinase activity profile of each cell line was deconvoluted from the corresponding peptide phosphorylation profile shown in **Fig. 21a**, and for individual kinases

or kinase families with ≥ 3 biological peptides. Unsupervised hierarchical clustering was applied across cells and kinases. Each cancer cell line displayed a distinguishable kinase signature. Kinase activity is color-coded blue-to-black-to-yellow from low-to-medium-to-high activity.

- 5 (b) Pearson-correlation heatmap highlighting the functional relationship between cancer cells established from their kinase activity signature. Cells are arranged by tumor tissue of origin (MEL: melanoma; CRC: colorectal cancer; BC: breast cancer; LC: lung cancer; PC: prostate cancer).

- (c-d) Comparing kinase activity and drug sensitivity of cancer cell lines. A good way to
 10 evaluate the performance of HT-KAM-defined kinase signatures is to measure the effect of a kinase inhibitor on cell growth and then correlate it to kinase activity level. This would assess whether kinase signatures are predictive of drug sensitivity and indicative of cell-specific kinase dependencies. We tested the effects of 28 kinase-inhibiting drugs (list in (c)) in 3-day dose-response cell viability assays using a core set of cell lines (A375, AU565, H3122,
 15 HCC70, HCT116, HT29, MCF7, MDA231, MDA436, SK-CO-1, SkMel2, T47D, WiDr; additional cell lines were tested on a case by case basis where it was deemed useful, such as PC9 to compare EGFR activity versus gefitinib sensitivity²⁰). We intentionally chose to test more than one inhibitor per kinase for some of the kinases in order to validate results (the 28 drugs inhibit 17 different kinases/kinase families; see table in (c)). We also further
 20 corroborated GI50 results from literature²¹⁻²³. Kinase activity levels were extracted from signatures available in (a). We then correlated cancer cells' GI50's with their kinase activity for each drug. Examples are shown in panel (d). The table in (c) provides all correlations. The negative correlation between kinase activity and drug-GI50 across cells indicates that, overall, cell lines were more susceptible to inhibitors for which they displayed higher kinase
 25 activity. These results directly verify that profiles measured by HT-KAM assays accurately predict differences in kinase activity.

- (e) Correlation between kinase activity and drug sensitivity. Results from both the analysis in (c-d) and from Fig. 2g are included (total of 373 data points established from 37 inhibitors that target 22 kinases). For each kinase/drug pair, we compared the differential kinase
 30 activities and drug-GI50s across tested cancer cell lines. These results show that the kinase signature of cancer cells can reveal their actionable vulnerabilities, and thus their kinase dependencies.

[0067] Along with **Figs. 18-21**, this demonstrates that the peptide phosphorylation signatures can be reasonably converted in kinase activity profiles with translational relevance. The HT-KAM platform is a practical solution to find active, druggable kinases in cell culture models.

- 5 [0068] **(f-g)** Controls for the comparison of A375 vs. WiDr cells. Protein concentration, baseline ATP levels, and protein expression/phosphorylation levels are provided for validation of Fig. 2g-h.

[0069] **Fig. 23. Melanoma specimens.**

- (a) Table with details about treatments received by patients before and after the
10 biopsy/excised biospecimen that was used for histo-pathology/diagnostic and retrospectively tested on the HT-KAM platform. Treatments (muphoran; dacarabazin; interferon (IFN; low/high dose); ipilimumab (anti-CTLA4); vemurafenib (BRAF^{V600E} inhibitor); interleukin; irradiation) are color-/letter-coded. Numbers of treatment cycles are specified. Tumor stage, BRAF status, recurrence and survival are indicated. The flash-frozen tumor biospecimen of
15 patient #1 was a superficial spreading melanoma (SSM) excised from the trunk; patient #2: SSM metastasis from the trunk; patient #3: subcutaneous metastasis from a recurrent cutaneous tumor from the leg; patient #4: nodular melanoma metastasis from the leg; patient #5: cutaneous metastasis from the cubital fossa; patient #6: cutaneous metastasis from the neck; patient #7: inguinal lymph node metastasis (axillary lymphadenectomy of positive
20 sentinel); patient #8: cutaneous metastasis from a recurrent cutaneous tumor of the temporal area, patient #9: cutaneous metastasis from a recurrent cutaneous tumor located on the back.
(b) Baseline ATP levels measured across HT-KAM assays, and protein concentrations from tissue extracts. Levels are comparable between tumors and experiments.

[0070] **Fig. 24. Analysis of the phospho-catalytic signatures of tumors.**

- 25 (a) Peptide phosphorylation activity signature of melanoma specimens. Unsupervised clustering of 36 activity profiles measured across 228 peptides is shown (9 patient tumor tissues tested in 4 independent technical replicates).
(b-h) Principal component analysis (PCA). We investigated the potential association between 'variables of interest' and principal components (PCs) that define individual phospho-
30 signatures (linear regression, overall fit of univariate model PC(i) for variable (j)). The table in (b) provides a summary of the results visually represented in graphs (c-g).

In (c), results show that replicate runs from the same patient sample were significantly similar (9 patients are shown as 9 different colors, with 4 dots per patient for the 4 experimental replicates), with an association between patient ID and PC1 of $p=1.44E-06$. In (d), results show that days at which assays were run were not associated with the primary PCs of melanoma kinomes (3 different days of experimentation are shown as 3 different colors; $p>0.05$). So, the results of the PCA from the two technical variables in (c-d) demonstrate the excellent performance and high reproducibility of the HT-KAM system (i.e. experimental procedure, instrumentation, data analysis).

In (e-f), results show that survival status (f) and recurrence status (g) were highly associated with the PCs of the kinome signatures (respectively PC1: $p=5.14E-06$ and PC2: $p=0.028$; and PC1: $p=2.31E-07$). So, the results of the PCA from the two clinical variables in (f-g) reveal the strong predictive and prognostic value of the phospho-catalytic signature of tumors.

In (g), results show that PCs of melanoma phospho-signatures do not associate with BRAF mutational status. This demonstrates that, even if a BRAF^{V600E} mutation is a dominant oncogenic driver, HT-KAM effectively discerns different types of BRAF^{V600E} tumors depending on other critical variables that the phospho-signatures of tumors reflect. HT-KAM differentiates fatal BRAF^{V600E}-positive conditions (patients #4,5,8,9) from non-fatal BRAF^{V600E}-positive conditions (patients #3,7). And similarly, HT-KAM distinguishes fatal non-BRAF^{V600E} conditions (patient #2) from non-fatal non-BRAF^{V600E} conditions (patients #1,6).

In (h), we overlaid PC results from (e-g) and annotated patients groups for outcome and therapeutic resistance. This shows that the PCs from VEM treated-but-resistant patients (#4,5,9) cluster together, indicating that tumor phospho-signatures reflect patients' unresponsiveness to BRAF-targeted therapy. The strong association between BRAF-therapy resistant lethal melanoma tumors and the PCs of their phospho-signatures fully validates the results generated from the clustering analysis available in Fig. 3b.

[0071] The results of the PCA shown in (b-g) were established from phospho-signatures normalized to the 228-peptides + 14 peptide-free readouts, and were closely recapitulated when signatures were normalized using the 63-reference peptides, or the 14 peptide-free tissue extract alone readouts, or 16-Y/S/T-free peptides, or when using raw ATP-consumption data (data not shown). Collectively, this formal investigation of the PCs of tumor phospho-signatures demonstrates the robustness of the HT-KAM system, and the reliability of its output to map the phospho-catalytic signatures of tumor tissues.

[0072] Fig. 25. Identifying the predictive peptide signature of patients with poor survival outcome, including BRAF^{V600E}-mutated patients resistant to BRAF-therapy.

Since survival is a significant clinical variable identified from patients' phospho-signatures (clustering analysis in **Fig. 3b**, and PCA in **Fig. 24**), we asked whether we could find peptides
 5 that would qualify as best predictors of patients' survival outcome. To do so, we applied two methods.

(a-c) Method #1: analysis of peptide phosphorylation signatures using dual significance threshold selection. In **(a)**, the waterfall plot shows the differential activity profile across all 228-peptides comparing patients #1,3,6,7 (alive) versus patients #2,4,5,8,9 (dead).

10 Significance values (t-test and Wilcoxon rank sum test; FDR-adjusted or not) are provided underneath the graph. Peptide phosphorylation signatures were normalized to the 228-peptides + 14 peptide-free readouts. In **(b)**, peptide phosphorylation activities were selected when concurrently passing FDR-adjusted t-test $p < 0.05$ and Wilcoxon rank sum test $p < 0.05$. This rigorous dual selection threshold identified 34 peptides as the most significantly
 15 differentially phosphorylated peptides associated with poor survival. In **(c)**, peptide phosphorylation activities shown in **(b)** are detailed across all patient signatures. Two thirds of the peptides are biological peptides (listed and color-coded on the right side of the heatmap), most of which are associated with higher phosphorylation activity in tumors from patients with poor survival outcome. We further validated these data using alternative
 20 normalization schemes (data not shown), and by applying this same computational analysis to identify peptides qualifying as best predictors of recurrence (identifying an overlapping set of 27 peptides; data not shown).

(d-e) Method #2: analysis of peptide phosphorylation signatures using top-25%-activity threshold selection. The top 25% peptides displaying phospho-activities that were most
 25 variable between all patients and all experimental repeats were selected. Panel **(d)** shows the unsupervised hierarchical clustering of these 57 most differential signals (same normalization as in **a-c**). In **(e)**, the waterfall plot shows the differential activity per peptide across these 57 peptides and between patient groups (differential survival outcome as black bars; differential recurrence outcome in violet; differential BRAF status in blue; data sorted by highest to
 30 lowest differential activities comparing survival outcome).

[0073] Results in **(a-e)** show that the most significant and consistently high signals associated with fatal outcome were measured with biological peptides such as SMAD2 S465 and S245 and S250, KHDRBS1/SAM68 Y440 and Y435, MTOR T2446 and S2448,

- CDKN1A/p21 T145 and S146, BRCA1 S988 and T509, ABL1 Y226 (corresponding to Y245 in another ABL1 isoform), FCGR2B Y292, or CHEK1 S280. Likewise, peptide sensors associated with consistently low activity in fatal outcome were biological peptides NOTCH2 S2070, JUN Y170, TERT Y707, GAB1 Y627, and reference peptides from modified
- 5 biological peptides PA__128, PA__134, or PA__230. Interestingly, the phospho-signatures of poor survival outcome included peptide sensors related to kinases such as SFK's, MTOR, or TGFBR-signaling, some of which were found to confer acquired resistance to BRAF or MEK inhibitors in cell culture models of melanoma²⁴⁻²⁸. Phospho-catalytic activities measured with peptides of related sequences behaved similarly and were useful internal controls validating
- 10 assay repeatability (e.g. MTOR T2446 and S2448; CDKN1A T145 and S146; CDK5 Y15 and generic positive control PA__240 which is originally derived from CDK1 Y15 and is mostly conserved between these two CDK's). So, the peptides included in the HT-KAM platform are robust sensors that can be used to generate clinically valuable signatures to diagnose cancer specimens.
- 15 **[0074] Fig. 26. Identifying the kinase signature of patient tumors.** Determining the activity of kinases in tumors could reveal tractable candidates for therapeutic interventions. So, we explored the potential clinical utility of the kinase activity signatures of melanomas. (a) Waterfall plot showing the differential activity of kinases between patient groups. For each tumor, the activity of kinases with ≥ 4 biological peptides was calculated as the average
- 20 of the phosphorylation activities measured in presence of their respective biological peptide subsets across all experimental repeats. Next, we calculated the mean activity per kinase for each patient group, and then plotted the differential activity between groups of interests, i.e. survival / black bars, recurrence / violet, BRAF status / blue. Significance was calculated using all experimental repeats between patient groups (shown underneath the graph).
- 25 (b) Differential kinase activity signature of tumors. Kinase activity profiles were mean-centered per tumor tissue and then mean-centered across patient tumors. Semi-supervised hierarchical clustering was applied across the 60 individual kinases or kinase families. As a validation of the results in (a-b), the differential activities of some of the kinases we found in patient tumors are corroborated by gene over-expression screen for RAF-inhibitor resistance
- 30 in melanoma cell lines 25 (e.g. upregulated kinase activity of PRKC(E), RAF(1), MAP3K8/COT1, PAK(1)).
- (c) Comparison of the activity of ABL1, AKT1, ERK2, HCK and p38a deconvoluted from kinases' biological peptide subsets (x-axis), versus activity levels converted from kinases'

differential peptide signature (y-axis; method explained in **Figs. 11, 18d**). Results show good concordance across kinases and tumor groups (e.g. the correlation between kinases' activity related to survival (circles) was 0.847).

(d) Validation of differential kinase signatures using enrichment analysis. We applied an

5 enrichment analysis (EASE – Fisher one-sided test; $p < 0.05$) to identify kinases whose biological peptides were most represented among the 34 peptides most significantly associated with survival outcome (identified in **Fig. 25b-c**). Results are shown in the table, and formally identify AKT's, PIM's, and RPS6KB's as most significantly related to the melanoma survival outcome phospho-signatures.

10 (e) Differential biological peptide-activity profiles comparing different patient groups. We chose to show AKT, PIM, RPS6KB, and GSK3B kinase families because analyses in (a-b,d) showed their changes were most significant and most profound (increase or decrease). Significance per peptide across patients groups (grey scale) are shown. The bottom graph compares the kinase profiles of BRAF^{V600E} tumors from patients retrospectively known to
15 survive (#3,7) versus patients who died (#4,5,8,9; which includes all patients who did not respond to BRAF-therapy). This means that the intrinsic vulnerabilities of these tumors were not inhibited, and that alternative treatment options would have been available at the time of biopsy based on kinase signatures revealed by HT-KAM.

[0075] Together, these results demonstrate that differences in the activity of kinases can be
20 measured within a tumor biospecimen (i.e. for each patient), and between different tumors (i.e. across patients and their individual malignancies). As well, significant changes in levels of kinase activity can be identified between different groups of patients (e.g. survival outcome). So, the HT-KAM assay successfully maps the oncogenic kinase signatures of tumor tissues, and reveals the significantly hyperactive kinases that are most predictive of
25 poor outcome. Accordingly, HT-KAM identifies druggable kinase vulnerabilities most tractable to treat patients with highest likelihood of recurrence and poorest survival outcome, including patients who are unresponsive to BRAF-therapy. Specifically, our results indicate that AKT1, PIM1 and RPS6KB1 kinases are most conserved and overly active in poor outcome melanoma.

30 [0076] **Fig. 27. Analysis of gene expression data as an indirect mean to corroborate phospho-catalytic signatures.** Even though gene expression is a relatively distant factor influencing the oncogenic activity of kinase enzymes and phospho-signaling circuits^{29,30}, we asked whether melanoma patients with poorer outcome available in the TCGA resource¹⁴

displayed changes in mRNA levels that match kinases we found by HT-KAM. We confirmed that patients who express high levels of AKT1 or PIM1 displayed significantly worse overall survival or disease-free survival, as shown in (a-b) and (c-e). Kaplan-Meier curves and logrank t-test p-values are provided in (a,c). Not shown here, we also found from the TCGA data set: (i) co-occurrence of AKT1 and RPS6KB1 mRNA alteration in poor outcome melanoma (log odds ratio of +2.66; $p < 0.001$); (ii) up-regulation of CDKN1A mRNA expression in worse progression-free survival which is interesting since CDKN1A-derived peptide sensors are among top predictors of poor outcome (see Fig. 25c).

[0077] **Fig. 28. Response of BRAF^{V600E} melanoma cell lines to drug combinations.**

(a) Summary of possible cell responses to the combination of two drugs. The heatmap on the left represents the experimental effects of combining increasing concentrations of drug A+B as changes in cancer cell growth. The middle graph is a conceptual representation of how differences in effects of drugs alone or combined can be interpreted as either synergistic, additive, or antagonistic (derived from ¹⁰). The heatmap on the right represents the calculated combination index from all experimental cell growth data points. We decided to analyze experimental profiles using the Bliss Independence model to calculate combination indices (CI) and avoid inaccuracies from dose-effect curve estimations ($CI = \log_2 (E_{ab}/(E_a * E_b))$; synergy $CI < 0$; additivity $CI = 0$; antagonism $CI > 0$) 10-12.

(b) Maps of cell growth responses (left) and combination indices (right). We tested the dependency of BRAF^{V600E} melanoma cell lines for kinases found as differentially active by HT-KAM in patient tumors (Fig. 26a,b). We included everolimus or trametinib (bottom two rows) as experimental 'positive' controls and for side-by-side comparison of the effects of other kinase-targeting drugs we tested. Arrows located above and on the left of each cell growth heatmaps serve as indicators of GI50 concentrations for each drug alone. The GI50s of A375 maintained in 5% or 0.25% FBS media for drug alone were: VEM (0.15uM; 0.02uM); MK2206 (10uM; 2.5uM); AZD1208 (12.5uM; 12.5uM); LY2584702 (50uM; >200uM); PF-4708671 (32uM; 16uM); 1-Azakenpaullone (20uM; 2.5uM); BI-D1870 (12.5uM; >50uM); everolimus (5uM; 10uM); trametinib (5nM; 2.5nM). The GI50s of Sk-Mel-28 maintained in 5% or 0.25% FBS media for drug alone were: VEM (0.63uM; 0.31uM); MK2206 (2.5uM; 5uM); AZD1208 (25uM; 25uM); LY2584702 (100uM; 100uM); PF-4708671 (32uM; 16uM); 1-Azakenpaullone (20uM; 10uM); BI-D1870 (25uM; 12.5uM); everolimus (10uM; 10uM); trametinib (10nM; 10nM).

[0078] Results of growth responses (left) and combination indices (right) show that strong growth inhibitory effects were found when combining VEM with inhibitors of RPS6KB (LY2584702 or PF-4708671) or PIM kinases (AZD1208). Overall, synergistic effects of these drugs were superior to, or at least as good as, those of drugs targeting MTOR (everolimus) or MEK (trametinib) that are currently used in the clinic, but whose effects in patients remain variable and often transient.

[0079] Our results also show that other kinases whose catalytic activities were originally identified by HT-KAM as less highly or less significantly upregulated than RPS6KB, PIM or AKT kinases in BRAF^{V600E} melanoma tumors were indeed less effective therapeutic targets. This is exemplified by RPS6KA inhibition (additive effects only of the BI-D1870 drug; see heatmaps), or PKC inhibition (limited additivity effects of Go6983 or Sotrastaurin; data not shown).

[0080] Furthermore, we found that inhibiting the GSK3B kinase, which is a tumor suppressor kinase we identified as systematically less catalytically active in melanoma tumors of poor outcome patients, resulted in antagonizing the effects of VEM treatment, which means that melanoma cells became more resistant to BRAF^{V600E}-targeting drug when it was combined with the GSK3B inhibitor 1-Azakenpaullone, as visually noticeable from the patterns of cell growth and CI in heatmaps. This also strongly suggests that some of the 'good' effects of VEM rely in part on 'proper' kinase activity of the GSK3B tumor suppressor.

[0081] (c-e) Evaluating the effects of drug combinations on melanoma cell death. We used Fluorescence-Activated Cell Sorting (FACS) and measured cell death in melanoma cells treated at GI50 concentration of drug alone in 5% and 0.25% FBS culture conditions. Panel (c) shows examples of flow cytometry profiles obtained for Sk-Mel-28. Graphs in (d-e) show percentages of apoptosis for Sk-Mel-28 and A375 are graphically (averages of cell death measured in 5% or 0.25% FBS conditions). Numbers above grey bars in (d-e) indicate gain or loss in apoptosis when combining VEM with a 2nd kinase-targeting drug.

[0082] Increase in apoptosis was systematically found when combining VEM with inhibitors of RPS6KB or PIM, and these effects were overall more potent than when combined with MTOR or MEK targeting drugs. Decrease in apoptosis was found when combining VEM with GSK3B inhibitor (i.e. GSK3B inhibition rescues BRAF^{V600E} melanoma cells from cell death upon VEM treatment). Highest level of apoptotic cell death was induced

by RPS6KB-targeting combined with VEM in Sk-Mel-28, thus achieving a critical endpoint in the perspective of eradicating tumor cells.

[0083] Altogether, these results demonstrate that the top kinase hits identified in patient tumors as predictive of their therapeutic failure (see **Fig. 3c-e** and **Fig. 26**) correspond to
 5 central kinase dependencies in BRAF^{V600E} melanoma cells. We conclude that inhibitors of RPS6KB, PIM or AKT are strong therapeutic candidates to restore therapeutic sensitivity. The HT-KAM platform successfully identifies new exploitable vulnerabilities of melanoma tumors based on the kinome profiling of patient tumor tissues.

[0084] **Fig. 29. A375 expressing a constitutively active AKT1 oncogenic kinase remain**
 10 **sensitive to RPS6KB- or PIM-targeting.** We exogenously expressed myrAKT1 (i.e. a constitutively active oncogenic AKT1) in A375 cells. We then tested the sensitivity of A375 myrAKT1 cells to drug combinations.

(a) Control for myrAKT1 expression. myrAKT1-expressing A375 cells were significantly less sensitive to VEM than their control counterpart ($p=2.77E-06$; 65% increase in GI50).

15 (b-c) Cell response of A375 myrAKT1 cells to drug treatments (same experimental settings and analysis as in **Fig. 28**). Cell growth is shown in (b), cell death is shown in (c).

(d) Ranking treatment effects by comparing CI values between melanoma cell lines. Data from both **Figs. 28-29** are included.

[0085] Results indicated that BRAF^{V600E} melanoma cells that acquired the expression of
 20 the AKT oncogene remained sensitive to combinations of VEM with inhibitors of RPS6KB or PIM kinases. As well, mimicking the loss of tumor suppressive activity by inhibiting GSK3B in myrAKT1-expressing cells to recapitulate the coordinated inactivation of GSK3B kinase and hyper-activation of AKT1 found in tissues from BRAF^{V600E} melanoma patients with poorest outcome, led to strong resistance to VEM treatment. The growth response of
 25 myrAKT1-expressing A375 cells treated with combinations of VEM + MTOR or MEK targeting drugs was comparable to wild type A375 cells. These results support the notion that these multiple oncogenic phospho-signaling hubs can function independently of each other's. These kinases are suitable alternative targets to restore response in melanoma cells.

[0086] **Fig. 30. Restoring therapeutic sensitivity by targeting the RPS6KB kinase in**
 30 **melanoma tumor cells from patients that acquired resistance to BRAF^{V600E}-therapy.**

(a-b) Phospho-RPS6KB profiles of two patient-derived xenografts (PDXs) tumor tissues (a) and related cell lines (b). PDXs were previously^{4,5} established from patients refractory to BRAF^{V600E}-therapy. A limited number of primary cell lines were derived from these PDX's. (c) Sensitivity of primary melanoma cell lines to RPS6KB-inhibitor PF-4708671 in 3-week colony formation assays.

[0087] The PDX tumor M032R6.X1 displayed high levels of RPS6KB1 in tumor tissue, and the PDX-derived cell line M032R6.X1.CL maintained high levels of RPS6KB1, and the cell line M032R6.X1.CL was sensitive to RPS6KB1 targeting. This corresponds to data shown in Fig. 3i.

[0088] Conversely, the PDX tumor M061R.X1 did not display high levels of RPS6KB1 in tumor tissue, and the PDX-derived cell line M061R.X1.CL also did not display high levels of RPS6KB1, and the cell line M061R.X1.CL was not sensitive to RPS6KB1 targeting. This can be considered as a control for results in Fig. 3i.

[0089] The results indicate that targeting RPS6KB can be a successful therapeutic intervention in BRAF^{V600E} melanoma tumors/tumor cells where RPS6KB1 is elevated. Such vulnerability may be valuable to restore therapeutic sensitivity in patients who do not respond to, or relapse from, current therapies.

DETAILED DESCRIPTION OF THE INVENTION

[0090] The disclosure is based, in part, on the discovery of an effective, sensitive assay for determining a phospho-kinase activity profile of cells, *e.g.*, tumor cells, or other biological samples in which it is desirable to evaluate kinase activity. The method comprises using a panel of sensor peptides that comprise biological substrate regions for different kinases. Evaluation of all of the members of the panel provide an indication of the kinase pathways that are active in the sample. In one embodiment, provided herein is a 228-peptide panel developed to detect the activity of over 60 kinases/kinase families, including ABL, AKT, CDK, EGFR, GSK3B, MAPK, or SRC. Such a panel provides a method of defining new mechanisms of resistance to targeted therapy, *e.g.*, BRAF^{V600E}-targeted therapy in colorectal cancer, can identify new druggable targets, *e.g.*, RPS6KB1 and PIM1 as new druggable vulnerabilities predictive of poor outcome in BRAF^{V600E} melanomas patient; and otherwise provides for comprehensive evaluation of kinase activity in cancer and in other complex or previously uncharacterized samples of interest to identify kinase type.

[0091] Accordingly, provided herein are panels of sensor peptides that can be used in combination to assay for the activities of multiple kinases at the same time to identify the kinase activity and/or determine the kinase signature of a biological sample of interest, *e.g.*, a cancer sample obtained from a patient. The invention further comprises methods of assigning
5 a kinase activity identified in a biological sample, *e.g.*, a cancer sample, to a kinase family and/or of identifying the kinase activity as belonging to a particular kinase. Such information can be used, for example, to further characterize the kinases/kinase activities from biological samples of interest. In some embodiments, *e.g.*, evaluation of kinase activity in cancer or other disease states, analysis of a combinatorial peptide panel in accordance with the
10 invention can be used to develop therapeutic strategies or for further development of drug candidates.

[0092] A “sensor” peptide in the context of this invention refers to any peptide that is contained in a kinase peptide activity panel as described herein which is used to assess kinase activity. These include biological sensor peptides and sensor peptides that have mutations
15 relative to the naturally occurring sequences that influence kinase activity. As used herein, the terms “kinase activity” and “phosphorylation activity” are used interchangeably to refer to the ability of a phosphokinase to phosphorylate a sensor peptide.

[0093] The term “peptide” is used herein to refer to a polymer of amino acid residues. The term applies to amino acid polymers in which one or more amino acid residue is an artificial
20 chemical mimetic of a corresponding naturally occurring amino acid, as well as to naturally occurring amino acid polymers and non-naturally occurring amino acid polymers. Thus, a peptide for used in a panel of peptides to assess kinase activity as described herein can comprise naturally occurring and/or synthetic amino acids, including analogs and amino acid mimetics that function in a manner similar to the naturally occurring amino acids. “Naturally
25 occurring” amino acids are those encoded by the genetic code, as well as those amino acids that are later modified, *e.g.*, hydroxyproline, γ -carboxyglutamate, and O-phosphoserine. “Synthetic amino acids” or “amino acid analogs” refers to compounds that have the same basic chemical structure as a naturally occurring amino acid, *i.e.*, an α carbon that is bound to a hydrogen, a carboxyl group, an amino group, and an R group, *e.g.*, homoserine, norleucine,
30 methionine sulfoxide, methionine methyl sulfonium. Such analogs have modified R groups (*e.g.*, norleucine) or modified polypeptide backbones, but retain the same basic chemical structure as a naturally occurring amino acid. Amino acid mimetics refers to chemical compounds that have a structure that is different from the general chemical structure of an

amino acid, but that functions in a manner similar to a naturally occurring amino acid. Amino acid is also meant to include α -amino acids having L or D configuration at the α -carbon.

[0094] Amino acids may be referred to herein by either their commonly known three letter symbols or by the one-letter symbols recommended by the IUPAC-IUB Biochemical Nomenclature Commission.

Biological Sensor Peptides

[0095] In the context of this disclosure, a “biological sensor peptide” refers to a peptide that comprises a substrate region that is derived from a naturally occurring substrate region. Thus, a biological sensor peptide as used herein refers to a peptide a known substrate region phosphorylated by a kinase belonging to a kinase family. Such substrate regions can be identified, *e.g.*, from the PhosphoAtlas or PhosphoSitePlus databases (see, *e.g.*, Olow *et al.*, “An Atlas of the Human Kinome Reveals the Mutational Landscape Underlying Dysregulated Phosphorylation Cascade in Cancer” *Cancer Res.* 76(7):1733-45, April 1, 2016; Epub 2016 Feb 26; and Hornbeck *et al.*, “PhosphositePLus, 2014: Mutations, PTMs and recalibrations. *Nucl. Acids Res.* 43:D512-D520, 2015).

[0096] Biological sensor peptide can vary in length, *e.g.*, from 7 to 25 amino acids in length, or from 9 to 21 amino acids in length. In some embodiments, a peptide may be up to 50 amino acids in length, or even up to 100 amino acids in length and at least 6 amino acids in length. In some embodiments, a biological peptide is 9, 10, 11, or 12 amino acids in length. In some embodiments, a biological sensor peptide is 11 amino acids in length. In some embodiments, biological peptides can comprise additional amino acids at the C-terminal and/or N-terminal ends of the peptide. For example, an 11-mer may comprise 9 amino acids from the naturally occurring substrate region sequence and 2 amino acids, *e.g.*, GC at the C-terminus or CG at the N-terminus. In other embodiments, the biological peptide does not contain a GC or CG tag at the end terminus.

[0097] Although the sequence of a biological sensor peptide is typically identical to the corresponding region of the native substrate phosphorylation site, in some embodiments, the biological sensor peptide may contain one or more amino acid residues at the end of the peptide, relative to the native sequence, that does not influence kinase activity, as explained in the preceding paragraph. In further embodiments, a biological sensor peptide may contain an amino acid change relative to the sequence of a naturally occurring substrate region that

does not influence activity of the kinase that phosphorylate that region. In some embodiments, a biological peptide may have a conservative substitution that does not influence kinase activity. In some embodiments, a biological sensor peptide may include modified amino acids. For example, modifying a peptide to have a pre-phosphorylated peptide on one Y residue may make it easier for a kinase to phosphorylate a 'free' Y nearby within that peptide sequence. This is also applicable to other post-translational modifications that may be biologically relevant and thus, a peptide that mimics such a state may improve the sensitivity/specificity/differentiability of detection for one or more kinases.

[0098] The number of biological sensor peptides employed in the panel per kinase is variable, but the panel comprises a plurality of biological sensor peptides for the majority of the kinases to be represented in the panel. For example, in some embodiments, a panel of comprises, 8, 9, 10, 11, 12, 13, 14, or 15 biological sensor peptide per kinase, but may range from 3 to 50 biological sensor peptides per kinase. In some embodiments, a panel of kinases may employ at least 4 biological sensor peptides per kinase. In some embodiments, the number of biological sensor peptides per kinase that is included in a panel is 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 16, 17, 18, 19, or 20. In some embodiments, the number of biological sensor peptides per family ranges from 4 to 20 or more.

[0099] In some embodiments, a panel of sensor peptide comprises at least 15 biological sensor peptides per the majority of kinase families to be represented in the panel. In some embodiments, at least 10 biological sensor peptide, or at least 15 biological sensor peptides are included per kinase family represented in the panel. In some embodiments, not more than 50 peptides is included per kinase family.

Control sensor peptides

[0100] Various control sensor peptides are included in a peptide panel of the present invention for experimental controls and statistical analysis, as detailed in the Examples section and accompanying figures and figure descriptions. The controls, also referred to herein as reference peptides, include generic positive controls, and mutated peptides, in which a naturally occurring substrate region is mutated such that a known kinase that phosphorylates the substrate in nature exhibits modified activity towards the mutant peptide, *e.g.*, is inactive against the mutated peptide. Mutated control sensor peptides provide increased confidence in the assay and improves analysis/interpretation and visualization of data from the screening assay. For example, a mutated reference peptide may contain an

amino acid substitution at a phosphorylatable tyrosine, serine, or threonine, *e.g.*, a tyrosine, serine, or threonine, may be substituted with a glycine or other residue that is not longer able to be phosphorylated. In some embodiments, a control peptide may have a phosphorylatable site that is pre-phosphorylated to render it insensitive to kinases that may be present in a given sample. Control peptides may also include peptides that have random peptide sequences. Finally, known generic peptides that can serve as positive controls are available, including commercially available. Such peptides may also be included in the panel as further controls.

Sensor peptide panels

[0101] A ‘sensor peptide panel’ or ‘kinase activity panel’ as used herein refers to an assemblage of peptides that collectively provides phosphorylation activity data (for a sample of interest being evaluated to characterize the kinases in the sample) for multiple kinases at the same time and thus is sufficient to assign kinase activity in the sample to a kinase family, or one or more specific kinases; or can be used to characterize kinases in a biological sample of interest. In a kinase activity assay panel of the present invention, it is the combination of different kinase-specific sets of peptides that provides the read out of activity for multiple kinases at once.

[0102] Further, even though many kinases may have few identified target sites or even substrate protein targets, a kinase activity panel assessment as described herein can be used to identify peptides that best match a kinase and thus provide greater insight into the kinase function.

[0103] In some embodiments, a kinase activity panel of the present invention comprises a multiplicity of sensor peptides to determine kinase activity for at least two, at least three, at least four, or at least five, or more kinases families. In some embodiments, a kinase activity panel of the present invention comprises a multiplicity of sensor peptides to determine kinase activity for at least ten, at least fifteen, at least twenty, at least twenty five, or more kinase families. In some embodiments, a kinase activity panel of the present invention comprises a multiplicity of sensor peptides to determine kinase activity for at least two, at least three, at least four, or at least five, or more specific kinases. In some embodiments, a kinase activity panel of the present invention comprises a multiplicity of sensor peptides to determine kinase activity for at least ten, at least fifteen, at least twenty, at least twenty five, or more specific kinases. Accordingly, a panel may comprise any number of peptides. Indeed, a panel and

assay of the present invention is modular by design in that users can adapt panels and assay conditions to their needs. Indeed, in some embodiments, a set of peptide sensors may be employed to study particular kinases under specific conditions, e.g., pH, ion, etc.; however, the panel still employs multiple peptides to capture kinase activity for multiple kinases.

- 5 [0104] In some embodiments, a kinase activity panel of the present invention can comprise a multiplicity of sensor peptides to determine kinase activity for multiple kinases, e.g., at least five, at least ten, at least fifteen, at least twenty, at least thirty, at least forty, at least fifty, or more, kinases selected from the group consisting of ABL1, ABL1(H396P), ABL1(Q252H), ABL1(T315I), BCR-ABL (fusion gene, Philadelphia translocation), ABL2, BLK, BRK
- 10 (PTK6), FGR, FRK, FYN, HCK, LCK, LYNA, LYNB, SRC, SRMS, YES1, CSK, BTK, BTK(E41K), FER (TYK3), FES, ITK, LTK (TYK1), SYK, TEC, TEK (TIE2), TXK, TYRO3, ZAP70, JAK1, JAK2, JAK3, TYK2, PTK2 (FAK, FAK1), PTK2B (FAK2, PYK2), ALK, AXL, AXL(R499C), CSF1R, DDR1 (NTRK4, RTK6, PTK3), DDR2 (NTRKR3, TYRO10), EGFR, ERBB2, ERBB3 (in its dimeric state with another HER fam. member),
- 15 ERBB4, FGFR1, FGFR2, FGFR2(N549H), FGFR3, FGFR4, FLT1 (VEGFR1), FLT3, FLT3(D835Y), FLT4 (VEGFR3), KDR (VEGFR2), IGF1R, INSR (IRR), INSR, ROS1, KIT (SCFR), MET, PDGFRA, PDGFRB, RET, ROR1 (ROR1 (NTRKR1), ROR2 (NTRKR2)), ROR2 (NTRKR2), EPHA1, EPHA2, EPHA3, EPHA4, EPHA5, EPHA8, EPHB1, EPHB2, EPHB3, EPHB4, RON (MST1R), NTRK1 (TRKA), NTRK2 (TRKB),
- 20 NTRK3 (TRKC), MERTK (MER), WEE1 (WEE1A), WEE2 (WEE1B), MUSK, ACVR1 (ACVR1A, ALK2), ACVR1B (ALK4), TGFB1 (ALK5), TGFB2, BMP1B (ALK6), AKT1, AKT1(E17K), AKT2, AKT2(E17K), AKT3, AKT3(E17K), AKT3(G171R), ARAF, BRAF, RAF1 (cRAF), ATM, ATR, AURKA, AURKB, AURKC, CHEK1, CHEK2, PRKDC (DNA-PKcs, DNAPK), PLK1, PLK3, BRSK1, BRSK2, BUB1, C11orf7, CAMK1, CAMK1beta, CAMK1delta, CAMK1gamma, CAMK2, CAMK2A, CAMK2B, CAMK2D,
- 25 CAMK2G, CAMK4, CAMKK1, CDC42BPA (MRCK alpha), CDC42BPB (MRCK beta), CDK1 (CDC2), CDK2, CDK3, CDK4, CDK5, CDK6, CDK7, CDK8, CDK9, CDK11, CHUK (IKK alpha), IKBKB (IKK beta), IKBKE (IKK-E), CLK1, CLK2, CSNK1A1 (CK1A), CSNK1D (CK1D), CSNK2A1 (CK2A1), CSNK2A2 (CK2A2), CSNK2B (CK2B),
- 30 DAPK3 (ZIPK), DMPK (DM1), DYRK1A, DYRK2, IRAK1, IRAK4, EIF2AK1, EIF2AK2, EIF2AK3 (PEK), EIF2AK4, GRK3 (ADRBK2), GRK5, GSK3A, GSK3B, HIPK2, ILK (p59integrin-linked kinase RAF-like kinase), STK3 (MST2, KRS1), STK4 (MST1, KRS2), STK10 (LOK), STK11 (LKB), MTOR (FRAP1), KSR1, KSR1(A635F), KSR1(L639F),

KSR2, KSR2(R676S), MAPK1 (ERK2), MAPK3 (ERK1), MAPK7 (ERK5/6), MAPK8 (JNK1), MAPK9 (JNK2), MAPK10 (JNK3), MAPK11 (p38b), MAPK12 (p38g), MAPK13 (p38d), MAPK14 (p38a), MAP2K1 (MEK1, MKK1, MAPKK1), MAP2K2 (MEK2, MKK2, MAPKK2), MAP2K4 (MEK4, MKK4, MAPKK4, JNKK1), MAP2K7 (MEK7, MKK7, MAPKK7, JNKK2), MAP3K1 (MEKK1, MAPKKK1), MAP3K7 (MEKK7, TAK1), MAP3K8 (COT), MAP3K14 (NIK), MAP3K17 (TAOK2, PSK1), MAP4K2, MAP4K4 (MEKKK4), MAP4K5, MAP4K6 (MINK), MAPKAPK2 (MK2), MAPKAPK3 (MK3), MAPKAPK5 (MK5), MARK1, MARK2 (EMK1), MARK3 (CTAK1), PKMYT1 (MYT1), NDR1 (STK38), NDR2 (STK38L), NEK1, NEK2, NEK6, NEK7, NLK, NME1 (NM23, NDPK-A), NME2 (NM23B, NDPK-B), NME1-NME2, NUA1, PAK1, PAK2 (p21 (RAC1) activated kinase 2), PAK3, PAK4, PAK7, PDK1 (pyruvate dehydrogenase kinase 1, GeneID: 5163), PDK1 (3-phosphoinositide dependent protein kinase 1, GeneID: 5170), PHKG1, PIM1, PIM2, PIM3, PKN1 (PRK1), PKN2 (PRK2), PRKAA1 (AMPKa1), PRKAA2 (AMPKa2), PRKACA (PKA C-alpha, PKACA = protein kinase cAMP-activated (PKA) catalytic subunit alpha), PRKACB (PKA C-beta, PKACB = protein kinase cAMP-activated (PKA) catalytic subunit beta), PRKACG (PKA C-gamma, PKACG = protein kinase cAMP-activated (PKA) catalytic subunit gamma), PRKCA (PKC alpha, PKCA), PRKCB (PKC beta, PKCB), PRKCD (PKC delta, PKCD), PRKCE (PKC epsilon, PKCE), PRKCG (PKC gamma, PKCG), PRKCH (PKC eta, PKCL), PRKCI (PKC iota, PKCI), PRKCM (PKC mu, PKCM), PRKCN (PKC nu), PRKCQ (PKC theta), PRKCZ (PKC zeta), PRKD1 (PKD1, PKD), PRKD2 (PKD2), PRKD3 (PKD3), PRKG1 (PKG 1, subunits I-alpha & I-beta, PRKGR1, PRKG1 = protein kinase, cGMP-dependent (PKG), type I), PRKG2 (PKG 2, PRKGR2, PRKG2 = protein kinase, cGMP-dependent (PKG), type II), PRKRIR (THAP12, DAP4, P52rPK), ROCK1, ROCK2, RPS6KA1 (RSK1, p90RSK), RPS6KA2 (RSK3, p90RSK2, S6K-alpha-2), RPS6KA3 (RSK2, p90RSK2, S6K-alpha-3), RPS6KA4 (MSK2, S6K-alpha-4), RPS6KA5 (MSK1), RPS6KA6 (RSK4, p90RSK6, S6K-alpha-6), RPS6KB1 (S6K1, p70S6K, S6K-beta-1), RPS6KB2 (S6K2, p70S6Kb, S6K-beta-2), SGK1 (SGK), SGK2, SGK3 (SGKL), SMG1, TAF1, TBK1 (NAK, T2K), TP53RK (PRPK), TSSK1, TSSK2, TSSK4, VRK1, ABL fam. (=ABL1, ABL1(H396P), ABL1(Q252H), ABL1(T315I), BCR-ABL, ABL2), SRC fam. (BRK (PTK6), FGR, FRK, FYN, HCK, LCK, LYN, LYNB, SRC, SRMS, YES1), JAK fam. (JAK1, JAK2, JAK3, TYK2), PTK2/FAK fam. (PTK2/FAK1, PTK2B/FAK2), EGFR fam. (EGFR, ERBB2, ERBB3, ERBB4), FGFR fam. (FGFR1, FGFR2, FGFR2(N549H), FGFR3, FGFR4), FLT/VEGFR fam. (FLT1 (VEGFR1), FLT3, FLT3(D835Y), FLT4 (VEGFR3), KDR (VEGFR2)), IGFR/INSR fam. (IGF1R,

INSRR (IRR), INSR), PDGFR fam. (PDGFRA, PDGFRB), ROR fam. (PDGFRA, PDGFRB), EPHA/B fam. (EPHA1, EPHA2, EPHA3, EPHA4, EPHA5, EPHA8, EPHB1, EPHB2, EPHB3, EPHB4), ALK fam. (ACVR1, ACVR1B, TGFR1, TGFR2, BMPR1B), AKT fam. (AKT1, AKT1(E17K), AKT2, AKT2(E17K), AKT3, AKT3(E17K), AKT3(G171R)), RAF fam. (ARAF, BRAF, RAF1 (cRAF)), AURK fam. (AURKA, AURKB, AURKC), CHEK fam. (CHEK1, CHEK2), PLK fam. (PLK1, PLK3), BRSK fam. (BRSK1, BRSK2), CAMK fam. (CAMK1, CAMK1beta, CAMK1delta, CAMK1gamma, CAMK2, CAMK2A, CAMK2B, CAMK2D, CAMK2G, CAMK4), CDK fam. (CDK1 (CDC2), CDK2, CDK3, CDK4, CDK5, CDK6, CDK7, CDK8, CDK9, CDK11), IKK fam. (CHUK (IKK alpha), IKBKB (IKK beta), IKBKE (IKK-E)), CLK fam. (CLK1, CLK2), CK fam. (CSNK1A1 (CK1A), CSNK1D (CK1D), CSNK2A1 (CK2A1), CSNK2A2 (CK2A2), CSNK2B (CK2B)), IRAK fam. (IRAK1, IRAK4), EIF2AK fam. (EIF2AK1, EIF2AK2, EIF2AK3 (PEK), EIF2AK4), GRK fam. (GRK3 (ADRBK2), GRK5), GSK3 fam. (GSK3A, GSK3B), ERK fam. (MAPK1 (ERK2), MAPK3 (ERK1)), JNK fam. (MAPK8 (JNK1), MAPK9 (JNK2), MAPK10 (JNK3)), p38 fam. (MAPK11 (p38b), MAPK12 (p38g), MAPK13 (p38d), MAPK14 (p38a)), MEK fam. (MAP2K1 (MEK1, MKK1, MAPKK1), MAP2K2 (MEK2, MKK2, MAPKK2)), MAP3K fam. (MAP3K1 (MEKK1, MAPKKK1), MAP3K7 (MEKK7, TAK1), MAP3K8 (COT), MAP3K14 (NIK), MAP3K17 (TAOK2, PSK1)), MAPKAPK fam. (MAPKAPK2 (MK2), MAPKAPK3 (MK3), MAPKAPK5 (MK5)), MARK fam. (MARK1, MARK2 (EMK1), MARK3 (CTAK1)), NEK fam. (NEK1, NEK2, NEK6, NEK7), NME fam. (NME1 (NM23, NDPK-A), NME2 (NM23B, NDPK-B), NME1-NME2), PAK fam. (PAK1, PAK2 (p21 (RAC1) activated kinase 2), PAK3, PAK4, PAK7), PIM fam. (PIM1, PIM2, PIM3), PKN fam. (PKN1 (PRK1), PKN2 (PRK2)), AMPKa fam. (PRKAA1 (AMPKa1), PRKAA2 (AMPKa2)), PKA fam. (PRKACA (PKA C-alpha), PRKACB (PKA C-beta), PRKACG (PKA C-gamma)), PKC fam. (PRKCA (PKC alpha, PKCA), PRKCB (PKC beta, PKCB), PRKCD (PKC delta, PKCD), PRKCE (PKC epsilon, PKCE), PRKCG (PKC gamma, PKCG), PRKCH (PKC eta, PKCL), PRKCI (PKC iota, PKCI), PRKCM (PKC mu, PKCM), PRKCN (PKC nu), PRKCQ (PKC theta), PRKCZ (PKC zeta)), PKD fam. (PRKD1 (PKD1, PKD), PRKD2 (PKD2), PRKD3 (PKD3)), PKG fam. (PRKG1 (PKG 1, subunits I-alpha & I-beta, PRKGR1, PRKG1 = protein kinase, cGMP-dependent (PKG), type I), PRKG2 (PKG 2, PRKGR2, PRKG2 = protein kinase, cGMP-dependent (PKG), type II)), ROCK fam. (ROCK1, ROCK2), RPS6KA fam. (RPS6KA1 (RSK1, p90RSK), RPS6KA2 (RSK3, p90RSK2, S6K-alpha-2), RPS6KA3 (RSK2, p90RSK2, S6K-alpha-3), RPS6KA4 (MSK2, S6K-alpha-4), RPS6KA5 (MSK1), RPS6KA6

(RSK4, p90RSK6, S6K-alpha-6)), RPS6KB fam. (RPS6KB1 (S6K1, p70S6K, S6K-beta-1), RPS6KB2 (S6K2, p70S6Kb, S6K-beta-2)), SGK fam. (SGK1 (SGK), SGK2, SGK3 (SGKL)), and TSSK fam. (TSSK1, TSSK2, TSSK4).

- [0105] In some embodiments, a kinase activity panel to identify a kinase signature that represents a kinase family or a specific kinase comprises biological sensor peptides that correspond to phosphorylated substrate regions from naturally occurring peptides that are phosphorylated by members of various kinase families, *e.g.*, ABL, AKT, ERK, HER, p38, SFK, and TK kinase families. A kinase activity panel can, however, include a plurality of sensor peptides for any kinase of interest, including pseudo kinases and orphan kinases.
- [0106] In some embodiments, a kinase activity panel in accordance with the invention comprises biological peptides that are phosphorylated by kinases enzymes that can include kinases evaluated in cancer cells and tumor cells, or other disease conditions, including BLK, BRK (PTK6), FGR, FRK, FYN, HCK, LCK, LYN, SRC, SRMS, YES1, ABL1, ABL2, EGFR, ErbB2, ErbB4, JAK2, CSK, AKT1, AKT2, AKT3, MAPK1 (ERK2), and/or MAPK14 (p38a) kinases. In some embodiments, a panel may comprises biological peptides that are phosphorylated by kinases found in cancer and tumor cells and may include biological sensor peptide from one or more of PIMs, RPS6KBs, PAKs, PDPK1, GSK3B, PKCs, PKDs, PKAs and the like. In some embodiments, the panel comprises at least 3, or at least 4, biological peptides for each kinase; or at least 4 peptides for each of the kinase families. In some embodiments, the panel comprises 5, 6, 7, 8, 9, 10, 11, or 12 peptides for each kinase family represented in the panel. In some embodiments, the panel comprises 5, 6, 7, 8, 9, 10, 11, or 12 peptides for each kinase represented in the panel.

- [0107] In some embodiments, a kinase panel in accordance with the invention comprises at least 75, at least 80, at least 90, at least 100, at least 110 at least 120, at least 125, at least 130, at least, or at least 140, or at least 150, or 151 of the biological peptides comprise a sequence as shown in Table 1 of a peptide having a peptide ID as shown in Fig. 4. In some embodiments, the kinase activity panel further comprises at least 10, at least 15, at least 20, at least 25, or 27 mutated peptides having a mutation in the substrate region as shown in Fig. 4. In some embodiments, the mutated peptides have the sequence as shown in Table 1 for the peptide ID shown in Fig. 4. In some embodiments, the panel further comprises pre-phosphorylated reference peptides as mutated peptides.

[0108] In some embodiments, a kinase activity panel in accordance with the invention comprises 228 peptides as described in Fig. 4. The sequences of the peptides are provided in Table 1 for each peptide under the peptide probe ID number shown in Fig. 4. The panel comprises 151 biological peptides, 14 generic positive control peptides and 63 reference peptides, i.e., mutated peptides. The 14 generic positive control peptides are commonly used as industry standards. The 63 reference peptides include 27 mutated peptide sequences, 31 pre-phosphorylated peptides and 5 peptides having random sequences. The phosphoresidue target is shown in Fig. 4, as is the origin of the peptide probe (i.e., the name of the biological substrate protein). The kinase enzyme associated with the biological peptides are also provided.

Activity

[0109] Kinase activity can be evaluated using any number of assays, including, *e.g.*, measuring ATP consumption; measuring consumption of any factor related to catalytic activity of an enzyme (e.g. GTP for GTPases instead of ATP for ATP-dependent kinases), measuring post-translational modification of any substrate molecule (e.g. the physical detection of the catalytic addition of a phosphorylation group onto a peptide; and the like). In some embodiments, ATP consumption is conveniently measured using an ATP consumption assay. An illustrative assay is detailed in the Examples under the **Kinase activity assay** section.

[0110] Activity pattern against various peptides in an assay panel can be assessed to determine kinase signature patterns as described herein. Activity can thus be used to assign the kinase activity of a sample undergoing evaluation to one or more kinase families. In some embodiments, kinase activity is assigned to one or more members of a kinase family. Assigning the activity of the kinase can be performed by comparing the activity to control patterns of activity of known kinases. The range of kinases activity of the test sample can vary from a high level of activity to little activity compared to the various control kinases.

[0111] In some embodiments, activity is evaluated as follows. Although this illustration of activity assessment uses ATP consumption to define activity, one of skill understands that similar analyses can be performed using endpoints other than ATP consumption to assess Activity. In an illustrative assay, ATP consumption is measured across a matrix comprising the kinase sensor peptides, *e.g.*, the matrix may be wells or other containers that will hold the kinase assay contents. Again, although this is illustrated using a “well” as a container, other

container may be used. Each well contains reagents/buffers, including the sensor peptide, ATP, and sample. In some embodiments, a kinase inhibitor may be included as a reagent component. A subset of control wells are also employed in which no peptide is present to assess baseline ATP consumption in the sample. ATP consumption measured across all
5 experimental wells provides an activity signature of a sample. A first activity analysis is performed to determine ATP consumption per peptide. This provides an 'agnostic' view of the results. All ATP consumptions for a sample represent a 'fingerprint'. These fingerprints can be compared between samples to determine how different samples are. One example is to use such fingerprints to compare cancer cell lines or tumor tissues (or even recombinant
10 kinases). The results can then be interpreted to assess ATP consumption per kinase.

[0112] In the kinase activity screening panel, as explained above, some of the peptides are typically biological sensor peptides, i.e., they have naturally occurring amino acid sequences that are phosphorylated by a kinase enzyme. In some embodiments, a variant of such a peptide may be employed where the peptide may a minor change *e.g.*, a conservative
15 substitution, that does not influence kinase activity. Kinase activity obtained using the biological sensor peptides can then be used to deconvolute the agnostic peptide phosphorylation signatures to measure the activity of their respective kinases.

[0113] In some embodiments, some kinases may have only one biological sensor peptide associated with them in the assay. Some kinases may have two, some three, etc. More
20 peptides per kinase typically provide for higher sensitivity, specificity and differentiability to identify kinases and their activity levels (i.e. ATP consumption measured and averaged across peptides that are biologically related to each kinase). In typical embodiments, a threshold of at least four biological peptides per kinase is employed to measure the activity of a given kinase, although alternative values, such as three biological peptides per kinase, may
25 also be used.

[0114] For a given kinase, all ATP consumption per biological peptide of this kinase is included to measure the average ATP consumption (i.e. activity) for this kinase. Thus, all activities (ATP consumptions) per peptide can be useful to measure a kinase activity. Activity levels between knases within a sample can then be compared, as can activity levels
30 between samples. Normalization is performed by creating a scale and using a threshold value. This is illustrated in the descriptions of Figs. 1 and 2. Determination of signature patterns and use of a panel to characterizes a kinase in a sample to be evaluated can be

determined as described in the Figures; and in the description of Figures, *e.g.*, Figs. 5, 7, 9, 12, 16, 18, 21, 24, and 26. For example, a sample obtained from a melanoma patient may be evaluated for phosphorylation signatures using a 228-peptide phosphorylation panel as described herein, see, *e.g.*, 4a and Table 1. Phosphorylation activity can be analyzed using
 5 unsupervised hierarchical clustering, principal component analysis, and dual significance threshold selections. Such an analysis can identify not only phosphosignatures that are indicative of particular kinases, or kinases families, but hyperactivity can be identified; and such associates can be associated with outcome.

[0115] In some embodiments, identification of overly active AKT1, PIM1, and RPS6Kb1
 10 kinases as overly active in a melanoma sample is indicative of a poor prognosis (see, *e.g.*, Fig. 23). In some embodiments, targeting RPS6KB or PIM1 in a patient having a BRAF^{V600E} melanoma in which RPS6KB1 or PIM1 is elevated can provide an improved outcome compared to therapies that target other kinases.

Samples

15 [0116] A "sample" as used herein comprises cells or is derived from cells, *e.g.*, comprises cellular extracts, to be evaluated that can be obtained from any biological source, such as tissues, extracts, or cell cultures, including cells (*e.g.* tumor cells), cell lines, cell lysates. The sample can be obtained from animals, preferably mammals, most preferably humans. Samples can be from a single individual or in some embodiments, can be pooled prior to
 20 analysis. Samples may from any source may be evaluate using a panel of sensor peptides that provide the ability to simultaneously evaluate activities of different activities. Thus, a sample can be obtained from any prokaryote or eukaryote, including plants, yeast, bacteria, cyanobacteria, or any other biological source of interest.

[0117] Samples for analysis employing sensor peptides in accordance with the disclosure
 25 can be from any source for which it is desired to determine a kinase activity profile. In some embodiments, the sample is a mammalian sample, *e.g.*, from primates (such as humans and non-human primates, *e.g.*, apes, monkeys) or from other mammals, *e.g.*, a bovine, ovine, porcine, equine, canine, feline, caprine, a murine, or other mammal.

[0118] In some embodiments, the sample is obtained from a patient that has a disease for
 30 which it is interest to assess kinase activity, where the sample comprises cells from a tissue that is affected by the disease. A "disease" refers to any disorder, disease, condition, syndrome or combination of manifestations or symptoms recognized or diagnosed as a

disorder that may be correlated by a kinase activity profile. Illustrative disease include, but are not limited to, cancer, cardiovascular diseases including heart failure, hypertension and atherosclerosis, respiratory diseases, renal diseases, gastrointestinal diseases including inflammatory bowel diseases such as Crohn's disease and ulcerative colitis, hepatic, gallbladder and bile duct diseases, including hepatitis and cirrhosis, hematologic diseases, metabolic diseases, endocrine and reproductive diseases, including diabetes, bone and bone mineral metabolism diseases, immune system diseases including autoimmune diseases such as rheumatoid arthritis, lupus erythematosus, and other autoimmune diseases, musculoskeletal and connective tissue diseases, including including arthritis, achondroplasia infectious diseases and neurological diseases such as Alzheimer's disease, Huntington's disease and Parkinson's disease.

Cancer

[0119] In some embodiments, a sample evaluated in accordance with the disclosure is a cancer. Cancer for which kinase activity profiles can be obtained include breast cancer, melanoma, colorectal cancer, lung cancer, ovarian cancer, uterine cancer, cervical cancer, prostate cancer, pancreatic cancer, bladder cancer, head and neck cancers, liver cancer, kidney cancer, brain cancer, including glioma and astrocytomas, thyroid cancer, laryngeal cancer, nasopharyngeal cancer, and oropharyngeal cancer; stomach cancer, and testicular cancer.

[0120] In some embodiments, the cancer is a hematological cancer, such as a lymphoma or leukemia, or multiple myeloma. Examples of leukemias include acute lymphoblastic leukemia, acute myeloid leukemia, chronic myeloid leukemia, hairy cell leukemia, acute nonlymphocytic leukemia, chronic lymphocytic leukemia, acute granulocytic leukemia, chronic granulocytic leukemia, and acute promyelocytic leukemia, among others. Examples of lymphomas include cutaneous T-cell lymphoma, Hodgkin's lymphoma, an non-Hodgkin's lymphoma.

[0121] In some embodiments, a sample comprising cancer cells is obtained from a patient who has been treated with a therapeutic agent, such as a chemotherapeutic agent, *e.g.*, a protein kinase inhibitor.

[0122] In some embodiments, the analysis of kinase activity in cancer cells can be used to select a therapeutic agent that targets a kinase identified as being overactive in cancer cells.

[0123] The following examples are intended to illustrate, but not limit, the claimed invention. Thus, a panel of the invention may comprise a multiplicity of sensor peptides to detect activity for any kinases of interest and the reaction conditions and reagents for assessing activity can be modified as appropriate for evaluation of the kinase activity of interest.

EXAMPLES

Example 1. Peptide multiplex platform as a robust sensor of kinases' activity

[0124] Directly measuring the activity of kinases is the most proximal assessment of their functional status. This example describes the development of a high throughput kinase activity-mapping (HT-KAM) assay, whereby a compendium of biological peptide targets of kinase enzymes serves as combinatorial sensors of phospho-catalytic activity.

Description of experimental results

Peptide sensing platform to monitor phospho-signatures

[0125] We sought to develop a high-throughput kinase activity-mapping (HT-KAM) assay, whereby a compendium of peptides serves as combinatorial sensors of the phospho-catalytic activity of kinase enzymes. To demonstrate our strategy, we synthesized a 228-peptide library (Fig. 4) that includes 151 biological 11-mer peptides corresponding to substrate protein regions variously phosphorylated by kinases involved in oncogenic processes³⁴. The library also includes 14 generic 'positive control' peptides commonly used as industry standards, and 63 reference peptides comprising 27 mutated, 31 pre-phosphorylated, and 5 random peptide sequences. A liquid dispensing instrument was programmed to aliquot peptide, sample, ATP, and buffer solutions in 384-well plates (Fig. 1a; **Additional Methods** section). Each well contains one peptide, and each plate simultaneously assesses the phospho-signature of one sample.

[0126] The description section below concentrates on the biological relevance and technical advance offered by our strategy. Analyses demonstrating repeatability, specificity and validation for recombinant kinases (Fig. 1; Figs. 5-17), cell extracts (Fig. 2; Figs. 18-21) and tissue extracts (Fig. 3; Figs. 23-26) are briefly explained in this section, and more fully described in the **Additional Methods** section.

Multi-peptide-derived phospho-catalytic signatures distinguish individual kinases and their enzymatic subfamilies

[0127] The phospho-catalytic activity profile of 25 recombinant kinases was measured in presence of all 228-peptides (**Fig. 1b; Figs. 5-7**). Inspection of kinases' activity across all peptides revealed that each kinase displayed a unique phosphorylation fingerprint (**Fig. 1b**). Particular family members were functionally distinguishable from genetically related kinases (e.g. BRK or SRMS vs. other SFK's; MAPK14 vs. MAPK1), and the activities of kinase isoforms or oncogenic variant were discernable (LYN A vs. LYN B; ABL1 vs. ABL1^{T315I}). Nevertheless, the principal factor clustering kinases was their family of origin (**Fig. 1b**, horizontal clustering; **Fig. 1c**, Pearson correlation grid).

[0128] We then examined how including a multiplicity of peptide sensors impacted the sensitivity and specificity of the assay for predicting the identity of an individual kinase. We computed Area Under the Curve (AUC) from repeated iteration of random peptide sampling. Sensitivity and specificity systematically improved when including an increasing number of peptides, while any single peptide performed poorly overall (example in **Fig. 1d; Figs. 8-9**). For any given number of peptides, specific subsets performed significantly better than others. For instance, for HCK, the AUC derived from the specific combination of its 8 biological peptide targets was higher than most other 8-peptide combinations (**Fig. 1d**).

[0129] Next, we asked whether our system could find peptide sets that best differentiate a kinase from others. We compared all phospho-catalytic profiles of kinases using a dual significance threshold ($p < 0.05$ for FDR-corrected t-test and Wilcoxon rank sum test). This revealed that a unique differential phospho-signature could be systematically assigned to every kinase (**Fig. 1e; Figs. 10-11**). Optimal signatures included every type of peptide (biological, generic, reference) associated with both significantly high and significantly low phosphorylation activities. So, using a large spectrum of phospho-catalytic activity sensors is a highly sensitive reporting system to differentially identify a specific kinase/kinase family.

25 Biological peptides are effective combinatorial activity sensors of their kinase enzymes

[0130] We then asked whether kinases preferentially phosphorylated their respective biological peptides. We found that kinases were significantly more capable of phosphorylating the great majority of their biological peptide targets than control pools of 63-reference, or 5-random, or 16-Y/S/T-free peptides (example of AKT1, MAPK1, JAK2 in **Fig. 1f**, box plots in **Fig. 1g**; complete dataset in **Figs. 12-13**). Unsupervised clustering of kinases' activity using only biological peptides showed that biological peptides distinguished individual kinases, yet grouped them by functional relationships and kinase families (**Fig.**

1h). Biological peptides contributed most to kinases' differential phospho-signature (**Fig. 1e,h; Figs. 10-11**) and highest measurable activity (**Fig. 14**). Computational analysis revealed that the phospho-catalytic signatures of kinases derived from their biological peptides outperformed generic peptides and provided excellent specificity and sensitivity (average
 5 AUC>0.9; **Fig. 1d,i; Fig15**).

[0131] To further evaluate the performance of biological peptides as sensors of kinases, we measured the effect of kinase inhibitors. The activity profiles of ABL1, ABL1^{T315I}, LYN A and AKT1 treated with dasatinib, imatinib and staurosporine showed that biological peptides effectively revealed the distinct drug-sensitivities of kinases (**Fig. 1j-k; Figs. 14b, 16**).
 10 Statistical analyses determined that biological peptide sets systematically and significantly correlated with kinases' activity inhibition (**Figs. 14b, 16**). Similar results were found when measuring the effects of SFK-inhibitor PP2 and SU6656 on FYN A, HCK, LCK, and SRC (**Fig. 17**). Altogether, these results show that the identity and activity of kinases can be measured using their respective biological peptides.

15 Identifying druggable kinases that mediate intrinsic resistance to BRAF^{V600E}-targeted therapy

[0132] Providing a functional assay that identifies hyperactive, druggable kinases in cell culture models would be valuable to investigators. In the case of BRAF^{V600E} CRC, finding targeted therapies has been a biomedical challenge. RNA-interference screens performed in WiDr BRAF^{V600E} CRC cells originally found that parallel feedback activation of EGFR
 20 caused intrinsic resistance to vemurafenib (VEM)^{4,36}, but the limited response of patients treated with BRAF+EGFR combination therapy³⁷ underlines how crosstalk between signaling pathways often confounds genetic screen–drug response relationships. We applied our assay to explore whether other kinases drive such unresponsiveness to BRAF-therapy.

[0133] We used WiDr cells as a model system. WiDr treated with VEM displayed reduced
 25 MEK1 and ERK2 kinase activity, and increased EGFR activity (**Fig. 2a; Fig. 18**). This matched the reduction in phospho-MEK1/2 and ERK1/2 proteins, and increase in phospho-EGFR, which confirmed that the HT-KAM assay could functionally replicate what is anticipated from literature. We then concentrated on finding novel targets. Examination of kinase signatures revealed that AKT1 was overly active, while its downstream tumor
 30 suppressor kinase GSK3B was inactivated (**Fig. 2b, left**). Furthermore, the activity of PDPK1 and its downstream effector kinases SGK1, PRKCA, PKNs were increased (**Fig. 2b, right**). These changes in kinase activity were significant (**Fig. 2b; Figs. 18-19**) and confirmed by

immuno-detection (**Fig. 2c**). Such specific changes are cancer-promoting processes implicated in cell cycle, survival and metabolism, which suggests that drug resistance can be functionally defined by our assay as the coordinated re-programing of multiple signaling pathways.

- 5 [0134] To assess the role of these kinases as mediators of resistance to VEM, we tested the response of WiDr to drug combinations in cell survival assays. Besides AKT1, strong synergy was observed when BRAF^{V600E}-targeting was paired with inhibitors for PDPK1 and PRKCA (**Fig. 2d**). Results were confirmed using other BRAF^{V600E} CRC cell lines and inhibitors (**Fig. 20**). This demonstrates that our strategy can serve as a discovery platform to
10 predict differences in kinase activity and provide a rational design for new combination therapies.

The phospho-catalytic signatures of cancer cells reveal their specific kinase dependencies

- [0135] Next, we asked whether our approach could be used to survey the activity of kinases across different cancer cell lines. **Fig. 2e-f** and **Figs. 21-22a,b** show that cancer cell lines
15 could be systematically distinguished based on either their 228-peptide phosphorylation profiles, or their kinase activity signatures. We then evaluated whether the differential kinase activities of different cells could foretell their drug sensitivity (**Fig. 22c-e**), and specifically asked whether HT-KAM could identify kinases that functionally distinguish BRAF^{V600E} CRC from BRAF^{V600E} MEL cells (WiDr vs. A375; **Fig. 2g-h**). Graphs in **Fig. 2g** compare kinase
20 activity (y-axis) to GI50 concentration for individual drugs (x-axis). Data for the 10 kinases and 14 inhibitors we tested, are compiled in **Fig. 2h**. Results showed that the higher the activity of a kinase, the more a cell line was susceptible to respond to a matching drug. For instance, CDKs and MEKs were significantly more catalytically active in A375 than in WiDr, and A375 cells responded to much lower concentrations of the related drugs dinaciclib
25 and trametinib than WiDr did. So, BRAF^{V600E} cells of different tumor origins inherently relied on distinct kinase dependencies that were predictive of their drug sensitivities. Altogether, results in **Fig. 2f-h** and **Fig. 22** indicate that our platform provides a pragmatic solution to find active, druggable kinases in cell culture models.

Mapping the phospho-catalytic signatures of patients' melanomas

- 30 [0136] Identifying which kinases are overly active in patients' tumors would be of high clinical value. Malignant melanoma is a disease ultimately refractory to most current forms of therapy, including BRAF/MEK/ERK inhibitors used to treat metastatic BRAF^{V600E} tumors³⁸⁻

⁴¹. We thus evaluated the potential utility of our assay to map the phospho-catalytic signatures of patients' melanomas.

[0137] Nine surgically excised, fresh-frozen tumors from melanoma patients (**Fig. 3a**; **Fig. 23**) were tested in four independent HT-KAM technical replicates. The 228-peptide phosphorylation signatures were analyzed using unsupervised hierarchical clustering (**Fig. 3b**), principal component analysis (**Fig. 24**), and dual significance threshold selection (**Fig. 25**). We found that the phospho-fingerprints of tumors were highly robust signatures that strongly associated with outcome (**Fig. 3b**, black squares; **Figs. 24, 25**). The group of signatures that retrospectively predicted poor outcome included BRAF^{V600E} tumors that did not respond to VEM treatment (**Fig. 3b**, red squares; **Figs. 24, 25b-e**).

[0138] Next, we asked whether peptide phosphorylation profiles could reveal the hyperactive kinases of poor outcome tumors. **Fig. 3c** and **Fig. 26a-c** show that kinase activity signatures established from biological peptides and compared across tumors, were associated with outcome. Enrichment analysis using the most significantly and differentially phosphorylated biological peptides in poor outcome patients, determined that PIM, RPS6KB and AKT kinases were most active (**Fig. 3d**; **Figs. 26d-e, 27**), and highest in VEM-resistant melanomas (**Fig. 3c**, arrows; **Fig. 3e**). GSK3B was significantly downregulated in these tumors (**Fig. 3c,e**). These hyperactive kinases suggest new vulnerabilities that may be exploitable in the clinic.

20 Translating kinase hits into new therapeutic opportunities for BRAFV600E melanoma treatment

[0139] Based on these patient tumor kinase activity profiles, we assessed the growth response of A375 and Sk-Mel-28 cells to kinase-targeting drug combinations. **Fig. 3f-h** and **Fig. 28** show that inhibitors of PIM, RPS6KB or AKT significantly potentiated the anti-cancer effects of BRAF inhibition, and outperformed MTOR- or MEK-inhibitors whose effects in patients are variable and transient. Cells exogenously expressing constitutively active AKT1 remained sensitive to RPS6KB- or PIM-targeting (**Fig. 29**), suggesting that these signaling hubs can function independently and represent suitable alternative targets to alleviate therapeutic resistance. GSK3B inhibition antagonized VEM effects, thus mimicking a loss of tumor suppressive function by promoting resistance to BRAF-therapy. These results support observations we made in patient tumors (**Fig. 3c-e**), where unresponsiveness to

BRAF^{V600E}-therapy was accompanied with the coordinated inactivation of GSK3B and activation of PIM, RPS6KB and AKT kinases.

[0140] Finally, we used tumor cells isolated from a patient-derived xenograft (PDX) established from a BRAF^{V600E} melanoma patient refractory to VEM 42, and that maintained high levels of phospho-RPS6KB1 (**Fig. 3i; Fig. 30**). Colony formation showed these tumor cells were particularly sensitive to RPS6KB inhibition (**Fig. 3i**). So, the activation of RPS6KB is a confirmed vulnerability that can be targeted to restore therapeutic sensitivity in BRAF therapy-resistant melanomas.

Discussion of experimental results described above

10 [0141] A key to successful therapy is the identification of critical aberrant signaling networks whose inhibition would result in system failure of diseased cells. This example demonstrates the use of an innovative proteomic approach to identify specific kinase vulnerabilities that lie within the proto-oncogenic phospho-circuits of cancer cells and tumor tissues.

15 [0142] We developed a system that relies on collections of peptides to directly monitor the phospho-catalytic signatures of biological samples. Our strategy provides access to a vast, untapped resource of meaningful measurements, whether readouts are interpreted irrespective of which enzymes phosphorylate which probes, or analyzed to convert global phospho-signatures into functional profiles of kinase activities.

20 [0143] In-depth computational analyses and systematic drug targeting of kinases established the advantages of our system to gain biological insights into phospho-signaling circuits. The combination of single phospho-activities measured across peptide sensors collectively distinguished the activity of different kinases within or between samples of simple or complex composition (purified kinases, cells, tissues). Particular subsets of peptides
25 –especially kinases' biological targets– provided superior sensitivity, specificity and discernibility over any single probe-derived measurement. The identity and activity of kinases across a broad range of kinase families could be simultaneously assessed from their subset of biological peptides in various cancer cells and tumor tissues.

[0144] Based on our analyses, the differential spectrum of low-to-high phospho-catalytic activities measured across peptides predicts that expanding the peptide library beyond the
30 current 228 peptides will enable mapping many more kinases with even finer accuracy. It will

also be possible to build ‘fit-for-purpose kits’ relying on narrow sets of best-predictor peptides customized to monitor the activity of particular kinases for research or diagnostic purposes. The HT-KAM strategy is a versatile platform adaptable to users’ needs (e.g. interchangeable peptide library, or assay conditions), and practical to both laboratory research
 5 and clinical settings.

[0145] Mechanisms of drug resistance are the cornerstone of therapeutic failure. We established that peptide phospho-signatures of cancer cells could be deconvoluted to identify hyperactive, druggable kinases. We showed that the intrinsic and adaptive kinase dependencies of different BRAF^{V600E} cancers were distinctive and predictive of their drug-
 10 sensitivity to single or combinatorial targeted therapies. The newly revealed kinase vulnerabilities of VEM-resistance in BRAF^{V600E} CRC cells included signaling pathways orchestrated by PDPK1, PRKCA, SGK1 and GSK3B. In melanoma, PIM1 and RPS6KB1 were identified as new druggable vulnerabilities predictive of poor outcome in BRAF^{V600E} patients. Some of these susceptibilities are new therapeutic alternatives that could be tested in
 15 the clinic. Since kinase circuits are likely cell/patient specific, it will be important to profile the kinome signatures of individual patients’ tumors to personalize medical treatments.

[0146] Our approach effectively identified kinase targets beyond those previously found by synthetic lethality genetic dropout screens in same model systems. While shRNA approaches focus more on individual genetic dependencies, our assay allows capturing the functional
 20 fingerprint of kinases in their native state, without requiring exogenous interventions that may alter the dynamics of signaling circuits. Furthermore, large scale gene expression or mutation analyses would not identify the kinase targets we found because no evident genetic alteration are reported for these genes in cell lines or patient tumors (e.g. RPS6KB1 or PIM1 in MEL; PRKCA or PDPK1 in CRC)^{18,34,42,43}. As well, the results of considerable genomic
 25 study efforts suggest that, for many cancers including CRC, therapeutic resistance is likely not driven by individual genetic dependencies or caused by some dominant driver mutations. Therefore, functional proteomic platforms designed to detect the activity of kinases –and eventually the functionality of the whole kinome– such as HT-KAM, are needed to start elaborating higher-order functional maps of signaling networks that can directly pinpoint
 30 actionable vulnerabilities of tumors.

[0147] How to choose and pair limitless combinations of drugs is a pressing need for pharmaceutical industries and physicians who grapple with treatment resistance in patients⁴⁴.

46. Our platform provides a new rational design to help prioritize drug combinations and maximize likelihood of success. Furthermore, the phospho-peptide signatures uncovered in our study represent a new parameter that has the potential to be configured into diagnostic tests. For instance, BRAF^{V600E} melanoma patients who retrospectively displayed phospho-
 5 signatures indicative of aggressive disease, may have benefited from targeting BRAF + RPS6KB1 or PIM1, instead of standard therapies on which patients almost inevitably relapse.

[0148] The fact that our assay identifies multiple kinases as a cause of drug-resistance defies the usual scenario of trying to map a highly specific feedback loop that ends up describing a ‘unique’ pathway that mediates ‘all’ of the observed resistance. Instead, our
 10 results argue that resistance can result from a combination of pathways that are upregulated, working in concert, and interdependent on each other, such that it requires their coordinated signaling activities to drive the resistance. As such, a finite number of key cooperative dependencies can be identified, thus offering a highly selective choice of relevant targets to explore. Such herd-like mode of resistance can only be discovered by the kind of mass-scale
 15 functional proteomic approach we developed.

[0149] The combinatorial peptide sensing system described herein is thus a new, effective way to capture the functionality of kinases in cells and tissues. This modular strategy addresses a central issue in the bio-medical field, and could play an integral role in improving research productivity and guiding therapeutic decisions. Mapping the phospho-catalytic
 20 signatures of diseases innovates the molecular exploration of signaling networks, and supports the discovery of new actionable dependencies for precision medicine.

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Additional Methods—Results are summarized in the Descriptions of the Figures

- [0150] **Kinase activity assay.** The phospho-catalytic signature of samples was established from simultaneously occurring ATP-consumption tests measured in presence of peptides that are experimentally isolated from each other in multi-well plates. Assays were run in 384
- 20 well-plates (solid white flat bottom plates; Corning, cat.# 3570), where each experimental well contained one kind of peptide. The final 8 uL reaction mixtures per well contained the following final concentrations of reagents: (i) kinase assay buffer (KaB1x prepared daily and diluted in ddH₂O from a 10x stock solution of 25mM Tris-HCl (pH7.5), 10mM MgCl₂, 0.1mM Na₃VO₄, 5mM β-glycerophosphate, 2mM dithiothreitol (DTT); or purchased from
- 25 Cell Signaling cat.# 9802), (ii) 250 nM ATP (prepared from a 10mM stock solution of adenosine-5'-triphosphate in ddH₂O, and diluted daily with KaB1x; Cell Signaling cat.# 9804), (iii) 200 ug/mL 11-mer peptide (lyophilized stocks originally prepared as 1mg/mL in KaB1x, 5% DMSO), and (iv) samples typically made of either 5ng/uL recombinant kinase enzyme protein or 10 ug/mL protein extract from cell or tissue lysates (see below for
- 30 protocol) that were kept on ice and diluted in KaB1x <30min before experimental testing.

Controls with no-ATP, or no-peptide, or no-sample as well as ATP standards, were run side-by-side within each 384-well plate.

[0151] High-throughput liquid dispensing of all reagents was achieved using the Biomek® FX Laboratory Automation Workstation from Beckman Coulter (hosted by the Center for Advanced Technologies, UCSF), and was programmed to specifically address the dispensing requirements of the assay (timing, sequence, tip-touch location/height/depth, etc). Accurate dispensing was thoroughly and regularly validated. All reagents were kept on ice and plates on cold blocks (VWR/BioCision; COOLRACK XT PCR96 cat.# 89239-498; COOLSINK XT 96F cat.# 89239-504) until enzymatic reactions were started. For all intermediary steps over the course of the assay (i.e. buffer and sample preparation, dispensing, etc), we used micro-centrifuge tubes (Costar; cat.# 3621) and clear 96-well PCR-plates (VWR; cat.# 83007-374).

[0152] Once the dispensing of reaction mixtures was completed, 384-well reaction plates were typically incubated for 30min at 30degC. After enzymatic reactions were completed, the final detection step used Kinase-Glo revealing reagent (Promega; cat.# V3772; dispensed using Biomek automated workstation), which stops the activity of kinase enzymes and produces a luminescent signal directly correlated with the amount of remaining ATP in samples over a broad range of ATP concentrations (repeatability and accuracy of the ATP-dependent luminescence assay measurements were tested and validated over 5 logs of ATP concentration; $R^2 > 0.99$). Luminescence data are inversely correlated with the amount of kinase activity. Luminescence was measured using the Synergy 2 Multi-Mode Microplate Reader from BioTek, and occasionally the Molecular Devices Analyst AD Microplate Reader from McKinley Scientific.

[0153] **Peptide sensors.** 11-mer amino acid sequences were made-to-order and mass synthesized by GenScript at >95% purity. The 228-peptide library included 151 biological peptides, 14 generic positive control peptides, and 63 reference peptides that include 27 mutated (Tyrosine (Y) / Serine (S) / Threonine (T) → Glycine (G)) and 31 pre-phosphorylated (Y / S / T → pY / pS / pT) peptides, and 5 random peptide sequences (Fig. 4 provides peptide sequence details and connectivity between peptides and kinases). Biological peptides correspond to phosphorylatable amino acid regions of substrate protein identified from literature and curated in resources such as PhosphoAtlas¹ (Fig. 4c-d) or PhosphoSitePlus². Each generic positive control peptide corresponds to a kinase activity

reporting probe commonly used in single-peptide assays, as available/advertised from literature. Some of the generic positive control peptides were purchased from SignalChem (e.g. Abltide, cat.# A02-58; Poly (4:1 Glu, Tyr) peptide, cat# P61-58).

- [0154] Recombinant kinases.** The following purified, recombinant kinase enzymes were purchased from SignalChem: ABL1/c-ABL (cat.# A03-18H), ABL1^{T315I} (cat.# A03-12DG), ABL2/ARG (cat.# A04-11H), AKT1/PKB/RAC (cat.# A16-10G), AKT2 (cat.# A17-10G), AKT3 (cat.# A18-10G), BLK (cat.# B02-10G), BRK/PTK6 (cat.# P94-10G), CSK (cat.# C63-10G), EGFR/HER1 (cat.# E10-11G), ErbB2/HER2/NEU (cat.# E27-11G), ErbB4/HER4 (cat.# E29-11G), ERK2/MAPK1/p42 (cat.# M28-10G), FGR (cat.# F10-10G), FRK (cat.# F14-11G), FYN isoforms A (cat.# F15-10G), FYN isoform C (inactive; cat.# F15-14G-20), HCK (cat.# H02-11G), JAK2 (cat.# J02-11H), LCK (cat.# L03-10G), LYN isoform A (cat.# L13-18G), LYN isoform B (cat.# L13-10G), p38a/MAPK14 (cat.# M39-10BG), SRC/c-SRC (cat.# S19-18G), SRM/SRMS (cat.# S20-11G), YES/YES1 (cat.# Y01-10G). The same concentration of every kinase was used in all experiments (**Fig. 1, Figs. 5-17**).
- [0155] Kinase inhibitors.** The following 46 inhibitors were used in biochemical assays or cell culture. Inhibitors purchased from Selleck Chemicals are: AT13148 (cat.# S7563), AZD1208 (cat.# S7104), AZD7762 (cat.# S1532), BAY-61-3606 (cat.# S7006), BI-D1870 (cat.# S2843), bosutinib / SKI-606 (cat.# S1014), CI-1040 / PD184352 (cat.# S1020), CHIR-99021 (cat.# S1263), dasatinib (cat.# S1021), dinaciclib / SCH-727965 (cat.# S2768), everolimus / RAD001 (cat.# S1120), fedratinib / SAR302503 (cat.# S2736), gefitinib / ZD-1839 (cat.# S1025), Go6983 (cat.# S2911), GSK2334470 (cat.# S7087), H89 (cat.# S1582), imatinib (cat.# S1026), IPA-3 (cat.# S7093), JNK Inhibitor VIII (cat.# S4901), LFM-A13 (cat.# S7734), LY2584702 (cat.# S7704), MK2206 (cat.# S1078), nilotinib / AMN-107 (cat.# S1033), NPK76-ii-72-1 (cat.# S), OSU-03012 (cat.# S1106), pelitinib / EKB-569 (cat.# S1392), PLX-4720 (cat.# S1152), PF-4708671 (cat.# S2163), ponatinib / AP24534 (cat.# S1490), RO-3306 (cat.# S7747), ruxolitinib / INCB018424 (cat.# S1378), saracatinib / AZD0530 (cat.# S1006), selumetinib / AZD6244 (cat.# S1008), sotrastaurin (cat.# S2791), TAK-715 (cat.# S2928), trametinib / GSK1120212 (cat.# S2673), vemurafenib / PLX4032 (cat.# S1267), VX-702 (cat.# S6005), 1-Azakenpaullone (cat.# S7193). Inhibitors obtained from other companies are AS601245 (Cayman; cat.# 17542), bryostatin 1 (Sigma-Aldrich; cat.# B7431), PP2 (Invitrogen; cat.# PHZ1223), PP3 (Tocris; cat.# 2794), SL 0101-1 (Tocris; cat.# 2250), staurosporine (Sigma-Aldrich; cat.# S4400), SU6656 (EMD-CalBiochem; cat.# 572635). Conditions of use are indicated in the text.

- [0156] **Cell culture.** The 23 cancer cell lines used in this study were purchased from ATCC or provided by the laboratories of Drs. R. Bernards, S. Ortiz-Urda, M. Bissell, or F. McCormick (colorectal: WiDr, HT29, SK-CO-1, HCT-116, RKO-1; melanoma: A375, Sk-Mel-28, Mel888, MM485, Sk-Mel-2; lung: H1755, H3122, PC9; breast: AU565, HCC70, MCF7, MDA-MB-231, MDA-MB-436, T47D, HMT-3522 S1, HMT-3522 T4; prostate: PC3; thyroid: 8505C). Cells were cultured following ATCC's instructions or as previously described³. Information regarding primary melanoma cell lines derived from therapy-resistant BRAF^{V600E}-melanoma patient-derived xenografts (PDXs) are previously described^{4,5}.
- [0157] To assess the growth/survival response of cell lines to single or combinatorial drug treatments in **Figs. 2-3** and **Figs. 20, 22, 28, 29**, we used CellTiter-Glo cell viability assay (Promega; cat# G7571). Cell culture and luminescence readouts were performed in 96- and 384-well plates after 3-day treatments. Drug treatment conditions are described in the text. The effects of drug combinations on cell growth were assessed by calculating drug interaction (D.I.; two-way ANOVA using Prism or Sigma Plot software) and combination index (C.I.; following either the Loewe Additivity⁶⁻⁹ or Bliss Independence models¹⁰⁻¹²). IC50 and GI50 correspond to the concentration of a given drug that causes 50% inhibition of kinase activity (IC50) or cell growth (GI50). We systematically tested the effects of drugs in both regular and low serum culture conditions (i.e. 5% and 0.25% FBS media). To quantitate cell death in **Fig. 3** and **Figs. 28-29**, FACS analysis of nuclear degradation was performed as previously described¹³. Colony formation assay in **Fig. 3** and **Fig. 30** were performed as previously described⁵.
- [0158] To test the responses of WiDr cells treated with VEM in **Fig. 2** and **Figs. 18-20**, cells were plated in medium containing 10% FBS 24h prior to being washed with serum-free medium, and cultured for 24h in medium containing 0.1% serum³. After low serum incubation, cells were treated with drugs for 30min and stimulated by 10% FBS. After 8h, 85% confluent cells were washed in PBS, then lyzed, and supernatants stored at -80degC. In cases where experiments happened at the Netherlands Cancer Institute, samples were shipped on dry-ice to the University of California at San Francisco.
- [0159] Preparing protein lysates from cell lines for kinase assay. To measure the phospho-catalytic activity of cancer cell lines (**Fig. 2** and **Figs. 18, 19, 21, 22**), cultured cells were lyzed for 5min in ice-cold cell lysis buffer. Freshly prepared lysis buffer (1mL per $5 \cdot 10^6$ cells)

contained non-denaturing Cell Lysis Buffer 1x (diluted in ddH₂O from 10x stock of 20mM Tris-HCl (pH7.5), 150mM NaCl, 1mM Na₂EDTA, 1mM EGTA, 1% Triton, 2.5mM sodium pyrophosphate, 1mM b-glycerophosphate, 1mM Na₃VO₄, 1ug/mL leupeptin; or purchased from Cell Signaling, cat.# 9803), complemented with 1x Halt Protease & Phosphatase (ThermoScientific cat.# 1861281, which contains inhibitors of Ser/Thr-phosphatases and Tyr-phosphatases). Scraped off lysates were then spun down at 14,000rpm for 15min, and supernatants stored at -80degC. Protein and ATP concentrations were quantified. Since every HT-KAM assay plate includes 14 wells with sample alone (i.e. without peptide; see (Fig. 4f), internal controls for ATP levels were systematically available for all assays/samples (see Figs. 18, 21, 22).

[0160] Immuno-detection and antibodies. Cell lysates were resolved by SDS gel electrophoresis (gels from BioRad), followed by immuno-blotting as previously described^{3,5} and following manufacturer's instructions. The following primary antibodies and phospho-antibodies were used. Antibodies to detect ATK1/2/3 (cat.# 4691), AKT1/2 pS473 (cat.# 4060), AKT1/2 pT308 (cat.# 2965), EGFR (cat.# 4267 and 3771), EGFR pY1068 (cat.# 3777 and 2234), ERK1/2 (i.e. MAPK1/ERK2/p44 and MAPK3/ERK1/p42; cat.# 4695), ERK1/2 pT202/Y204 (cat.# 4370), GSK3B (cat.# 9315), GSK3B pS9 (cat.# 9323), MAPK14/p38a (cat.# 8690), MAPK14/p38a pT180/pY182 (cat.# 9215), MEK1/2 (i.e. MAP2K1 and MAP2K2; cat.# 4694), MEK1/2 pS217/221 (cat.# 9121), MTOR (cat.# 2972), MTOR pS2448 (cat.# 2971), PIM1 (cat.# 3247), PDK1/PDK1 (cat.# 3062), PDK1/PDK1 pS241 (cat.# 3438), PKN1/PRK1 pT774 and PKN2/PRK2 pT816 (cat.# 2611), PRKCA/PKCa (cat.# 2056), PRKCA/PKCa pT514 (cat.# 9379), RPS6KA1/p90RSK1 (cat.# 8408), RPS6KA1/p90RSK1 pT353 (cat.# 8753), RPS6KB1/p70S6K1 (cat.# 2708), RPS6KB1 pT389 (cat.# 9234), RPS6KB1/p70S6K1 pT421/pS424 (cat.# 9204), SGK1 (cat.# 3272), SGK1 pS78 (cat.# 5599), were from Cell Signaling. Antibodies to detect ERK1 (C-16), ERK2 (C-14), ERK1/2 pT202/pY204 (E-4), HSP90 (cat.# sc-7947) were from SantaCruz. A mixture of ERK1 and ERK2 antibodies was used for detection of total ERK³. Dilutions followed manufacturers' instructions.

[0161] Tumor specimens from melanoma patients. Clinical details regarding patients are available in Fig. 23a. Patient samples were collected at The Rudolfstiftung Hospital, Vienna Austria, and the University of California San Francisco, California USA, under the IRB# 13-204-VK and 12-09483, respectively. Tumor tissue not needed for diagnostic purposes were collected intraoperatively, macroscopically dissected and flash frozen. A small piece of tumor

tissue was O.C.T.-embedded, sectioned, H&E stained and analyzed to ensure >80% tumor cell content in tumor tissue samples.

[0162] Preparing protein lysates from tumor tissues for kinase assay. Flash frozen melanoma tissue specimens were pulverized using BioSpec 59012MS. Protein extracts from powdered samples were prepared following the lysis protocol used for cultured cells, i.e. 5 lyzed for 5min in ice-cold non-denaturing CLB1x diluted in ddH₂O from 10x stock of 20mM Tris-HCl (pH7.5), 150mM NaCl, 1mM Na₂EDTA, 1mM EGTA, 1% Triton, 2.5mM sodium pyrophosphate, 1mM b-glycerophosphate, 1mM Na₃VO₄, 1ug/mL leupeptin, complemented with 1x Halt Protease & Phosphatase (containing Ser/Thr- and Tyr- phosphatases inhibitors), 10 then spun down at 14,000rpm for 15min, and supernatants collected. All samples were stored at -80degC. Internal controls for ATP levels in peptide-free wells were systematically measured (see Fig. 23b). The total amount of tumor tissue sample necessary to profile the activity of kinases across 384well/plates, ranged from 20ug to 30ug (which is less than the typical 100ug collected per core biopsy).

15 [0163] **TCGA data analysis.** mRNA data from melanoma specimens available from the TCGA resource ¹⁴ were analyzed to identify samples with altered gene expression in comparison to reference samples using a z-score cut off of +/-2. The z-score is defined as (expression in tumor sample - mean expression in reference sample) / (standard deviation of expression in reference sample), where the reference population is either all tumors that are 20 diploid for the gene in question, or, when available, normal adjacent tissue. Other statistical analysis details are described in **Fig. 27**.

[0164] **Analysis of phosphorylation activity profiles.** The computational methods, technical notes, and statistical tools developed to analyze the peptide phosphorylation activity profiles and kinase activity signatures are explained in the main text and supplemental 25 information.

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[0165] It is understood that the examples and embodiments described herein are for illustrative purposes only and that various modifications or changes in light thereof will be suggested to persons skilled in the art and are to be included within the spirit and purview of this application and scope of the appended claims. All publications, patents, accession
10 numbers, and patent applications cited herein are hereby incorporated by reference for the purposes in the context of which they are cited.

Table 1

Peptide ID (total of 516 peptides included in the assay, with 576 unique peptides + 20 peptides in duplicate/triplicate)	Label / Location in 384 well plates	well description	Peptide Group (A=biological, B=consensus, C=mutated, D=modulated, E=consensus, F=random)	color-coded per peptide class (i.e. BIC-grpA vs. CON=grpB vs. PEF-grps CDE)	type of peptide probe	13-mer PEPTIDE probe (11-mer seq. with terminal GD) (NOTES: (1) in <i>italics</i> , (2) is the middle target site (with sometime additional site next to it), (3) in <i>bold</i> , (4) are additional nearby sites within that sequence, (5) in <i>bold</i> black are mutated amino acids in negative control peptides)	SUBSTRATE PROTEIN (NOTES: (1) / are between 'different' substrate proteins with same peptide target sequence, (3) / are between 'near to each other' sites within the same protein substrate sequence), (4) (5) are sites nearby each other within the same protein substrate sequence.	PHOSPHO-RESIDUE target (NOTES: (1) are between 'alternative sites' of the same protein substrate, (2) / are between 'different' substrate proteins with same sequence but different location within the protein sequence, (3) / are between 'near to each other' sites within the same protein substrate sequence), (4) (5) are sites nearby each other within the same protein substrate sequence.	KINASE PROTEIN (NOTES: (1) are in between kinases that target the same substrate/target site/peptide, (2) are between 'different' kinase proteins with the different/alternative substrate/peptide target sequences mentioned in the next columns)
PA_001	1G01	Key_L1_001 - sample + KdB + ATP A			BIO	ENTT ¹ SLIMHSC	FCGR2B	Y292	BLK, LYNA, LYNE,
PA_002	1G01	Key_L1_002 - sample + KdB + ATP A			BIO	DDK1 ¹ LTPFGC	FCGR2A	Y304	BLK, FYN, LYNA, LYNE, SYK
PA_003	1K01	Key_L1_003 - sample + KdB + ATP A			BIO	PKCA ¹ REHYGC	IKHDRBS1 (SAM68)	Y415	BRK (PTK6)
PA_004	1M01	Key_L1_004 - sample + KdB + ATP A			BIO	IKEDV ¹ LSDHGC	BRK (PTK6)	Y342	BRK (PTK6)
PA_005	1O01	Key_L1_005 - sample + KdB + ATP A			BIO	IKDDE ¹ YFCQGC	FGFR	Y412	FGFR
PA_006	1A03	Key_L1_006 - sample + KdB + ATP A			BIO	STEPR ¹ QPGVGC	SRC	Y530	FGFR, CSK
PA_007	1C03	Key_L1_007 - sample + KdB + ATP A			BIO	PDNEA ¹ EMPEGC	SNCA	Y125	FGFR, FYN, LYNA, LYNE, PTK2B (FAK2, PYK2), SYK
PA_008	1EE03	Key_L1_008 - sample + KdB + ATP A			BIO	KVEQL ¹ PEEEGC	FOS	S133	FRK
PA_009	1G03	Key_L1_009 - sample + KdB + ATP A			BIO	LPEVA ¹ PESEGC	FOS	T232	PRK
PA_010	1G03	Key_L1_010 - sample + KdB + ATP A			BIO	DDCSM ¹ YEDSRGC	CD79A	Y199	FYN
PA_011	1N03	Key_L1_011 - sample + KdB + ATP A			BIO	VNLIN ¹ QDAEGC	CTNNB1	Y142	FYN, BRK (PTK6), EGFR, FGFR2, FGFR3, NTRK1
PA_012	1M03	Key_L1_012 - sample + KdB + ATP A			BIO	IEDNE ¹ TARGGC	FYN / SRC / YES1	Y420 / Y419 / Y428	FYN / SRC / YES1
PA_013	1O03	Key_L1_013 - sample + KdB + ATP A			BIO	SDQML ¹ QQLGGC	CSF3R	Y752	HCK
PA_014	1A05	Key_L1_014 - sample + KdB + ATP A			BIO	KEPSN ¹ DFVYGC	ELMO1	Y720	HCK
PA_015	1C05	Key_L1_015 - sample + KdB + ATP A			BIO	IEDNE ¹ TAREGC	HCK / LOK / LYNA, LYNE	Y411 / Y384 / Y287	HCK / LOK / LYNA, LYNE
PA_016	1EE05	Key_L1_016 - sample + KdB + ATP A			BIO	GLHGS ¹ VLVVEGC	DDK1	Y174	LOK
PA_017	1G05	Key_L1_017 - sample + KdB + ATP A			BIO	DNISGT ¹ GKIWGC	PRKCD	Y234	LOK, SRC
PA_018	1G05	Key_L1_018 - sample + KdB + ATP A			BIO	GAIDS ¹ TARSGC	ZAP70	Y492	LOK
PA_019	1N05	Key_L1_019 - sample + KdB + ATP A			BIO	SLPEF ¹ YRPRGC	CDKN1B (p27, KIP1)	Y68	ABL1, LYNA, LYNE, SRC, YES1
PA_020	1M05	Key_L1_020 - sample + KdB + ATP A			BIO	FDAHI ¹ EGRVGC	PRKCD	Y64	LYNA, LYNE,
PA_021	1A12	Key_L1_021 - sample + KdB + ATP A			BIO	NIKTV ¹ GVSPNGC	ABL	Y226	SRC, ABL1
PA_022	1A07	Key_L1_022 - sample + KdB + ATP A			BIO	LEDND ¹ GRAVGC	AKT1	Y226	SRC, BRK (PTK6)
PA_023	1C07	Key_L1_023 - sample + KdB + ATP A			BIO	PEQDE ¹ DPRHGC	BCAR1	Y234	SRC
PA_024	1EE07	Key_L1_024 - sample + KdB + ATP A			BIO	POQFT ¹ VQALFSC	GRB2	Y150	SRC, ALK
PA_025	1C12	Key_L1_025 - sample + KdB + ATP A			BIO	CTDGV ¹ MMVAGC	EGFR	Y844	SRC
PA_026	1G07	Key_L1_026 - sample + KdB + ATP A			BIO	LPEGY ¹ EEAVGC	AFAP1	Y64	SRC
PA_027	1A07	Key_L1_027 - sample + KdB + ATP A			BIO	PPHEL ¹ VKNVDCG	TERT	Y707	SRC
PA_028	1M07	Key_L1_028 - sample + KdB + ATP A			BIO	RDSSY ¹ WEIEAGC	RAF1 (GRAF)	Y341	SRC, JAK2
PA_029	1EE12	Key_L1_029 - sample + KdB + ATP A			BIO	GSAAP ¹ LYTFHGC	STAT3	Y705	SRC, HCK, BRK (PTK6), FGFR3, FGFR4, JAK1, JAK2, ALK, FER, RET
PA_030	1A09	Key_L1_030 - sample + KdB + ATP A			BIO	WVPR ¹ LORRGC	CD46	Y354	SRC, YES1
PA_031	1C09	Key_L1_031 - sample + KdB + ATP A			BIO	QESSE ¹ QPTSGC	MDM2	Y364	ABL1
PA_032	1EE10	Key_L1_032 - sample + KdB + ATP A			BIO	KVVAL ¹ YDMPMGCC	IRK	Y229	ABL1, LYNA, LYNE, ETK, ITK, TEC
PA_033	1G09	Key_L1_033 - sample + KdB + ATP A			BIO	IGEST ¹ GVTVKGC	CDK5	Y15	ABL1, FYN, EPH44

1K09	Key_L1_034 - sample + KdE + ATP + pep1_PA_003	CON - _ random	AAAAAG7AAAAAGC	-	-
1K12	Key_L1_035 - sample + KdE + ATP + pep1_PA_003	CON - _ random	GGGGG7GGGGGC	-	-
1K11	Key_L1_036 - sample + KdE + ATP + pep1_PA_009	CON - _ random	GGGGG7pYGGGGGC	-	-
1K11	Key_L1_037 - sample + KdE + ATP + pep1_PA_009	CON - _ mutated BIO	IEDNE7ARGGGC	FYN / SRC / YES1 _ Y to G	FYN / SRC / YES1
1K11	Key_L1_038 - sample + KdE + ATP + pep1_PA_040	CON - _ mutated BIO	IEDNE7Y7ARGGC	FYN / SRC / YES1 _ CON () _ Y to pY	FYN / SRC / YES1
1K11	Key_L1_039 - sample + KdE + ATP + pep1_PA_041	CON - _ mutated BIO	STEPQQPQRENGC	SRC _ CON () _ Y to G	FGF, CSK
1K11	Key_L1_040 - sample + KdE + ATP + pep1_PA_042	CON - _ mutated BIO	STEPQpY7QFENGc	SRC _ CON () _ Y to pY	FGF, CSK
1K11	Key_L1_041 - sample + KdE + ATP + pep1_PA_043	CON - _ mutated BIO	GSAAPGLKTFGCG	STAT3 _ CON () _ Y to G	Y705 _ Y to G
1K11	Key_L1_042 - sample + KdE + ATP + pep1_PA_044	CON - _ mutated BIO	GSAAPpY7LKTfGCG	STAT3 _ CON () _ Y to pY	Y705 _ Y to G
1K11	Key_L1_043 - sample + KdE + ATP + pep1_PA_045	CON - _ mutated BIO	NKPTVGGVSPNGC	ABL1 _ CON () _ Y to pY	Y225 _ Y to G
1K13	Key_L1_044 - sample + KdE + ATP + pep1_PA_046	CON - _ mutated BIO	NKPTVpY7GVSfNGC	ABL1 _ CON () _ Y to pY	Y225 _ Y to pY
1K12	Key_L1_045 - sample + KdE + ATP + pep1_PA_047	CON +	Y7GEFKKKAAC	consensus _ CON +	SRC _ ABL1
1K13	Key_L1_046 - sample + KdE + ATP + pep1_PA_051	CON - _ mutated CON +	Y7GEFKKKAAC	consensus _ CON - _ Y to T	SRC _ ABL1
1K12	Key_L1_047 - sample + KdE + ATP + pep1_PA_065	BIO	KGA7REHP7GRGC	KHDRB1 (SAM68)	all TKs (p-90 Tyrosine Kinases)
1K12	Key_L1_048 - sample + KdE + ATP + pep1_PA_067	BIO	KAGNL7DISEDGC	GRIN2B	all TKs (p-90 Tyrosine Kinases)
1K12	Key_L1_049 - sample + KdE + ATP + pep1_PA_068	BIO	QEKMM7SLVSfGCG	ITPR1	BRK (FTH2)
1K12	Key_L1_050 - sample + KdE + ATP + pep1_PA_069	BIO	SFES7QQFFEGC	PLCG1	FYN
1K12	Key_L1_051 - sample + KdE + ATP + pep1_PA_070	BIO	IPSPV7APFAAGC	RAPGEF1	FYN
1K12	Key_L1_052 - sample + KdE + ATP + pep1_PA_071	BIO	VLDDQ7TESTG5C	ITTK	HCK
1K12	Key_L1_053 - sample + KdE + ATP + pep1_PA_072	BIO	EPYNL7SSLKEGCG	PIK3R1	HCK SRC, ABL1
1K12	Key_L1_054 - sample + KdE + ATP + pep1_PA_073	BIO	NLIRE7VKQTWGC	SYK	LYN, LYNB, SYK
1K12	Key_L1_055 - sample + KdE + ATP + pep1_PA_074	BIO	ANPL7KEATSGCG	ITGB3	SRC
1K12	Key_L1_056 - sample + KdE + ATP + pep1_PA_075	BIO	SEDPD7YQVNGCG	GIT2I	SRC, BTK, JAK2
1K13	Key_L1_057 - sample + KdE + ATP + pep1_PA_076	BIO _ Rep (B1-4) _ 1	SLES7YSACSMGCG	GREB10	SRC, FYN, TEC
1K13	Key_L1_058 - sample + KdE + ATP + pep1_PA_076	BIO	SLES7SACSMGCG	GRB10 _ CON () _ Y to G	SRC, FYN, TEC
1K13	Key_L1_059 - sample + KdE + ATP + pep1_PA_077	BIO _ Rep (B1-4) _ 1	SLES7SACSMGCG	GREB10	SRC, FYN, TEC
1K14	Key_L1_060 - sample + KdE + ATP + pep1_PA_078	CON - _ mutated BIO	KAGNL7DISEDGC	GRIN2B _ CON () _ Y to G	FYN
1K14	Key_L1_061 - sample + KdE + ATP + pep1_PA_078	CON - _ mutated BIO	EPYNL7GSLKEGCG	PIK3R1 _ CON () _ Y to G	LYN, LYNB, SYK
1K14	Key_L1_062 - sample + KdE + ATP + pep1_PA_079	CON - _ mutated BIO	SLES7SACSMGCG	GRB10 _ CON () _ Y to G	FYN
1K14	Key_L1_063 - sample + KdE + ATP + pep1_PA_081	CON - _ mutated BIO	NLIRE7VKQTWGC	SYK _ CON () _ Y to G	LYN, LYNB, SYK
1K14	Key_L1_064 - sample + KdE + ATP + pep1_PA_082	CON - _ mutated BIO	KAGNL7DISEDGC	GRIN2B _ CON () _ Y to pY	LYN, LYNB, SYK
1K14	Key_L1_065 - sample + KdE + ATP + pep1_PA_083	CON - _ mutated BIO	EPYNL7pYSSLKEGCG	PIK3R1 _ CON () _ Y to pY	LYN, LYNB, SYK
1K14	Key_L1_066 - sample + KdE + ATP + pep1_PA_084	CON - _ mutated BIO	SLES7pYSACSMGCG	GRB10 _ CON () _ Y to pY	ABL1
1K14	Key_L1_067 - sample + KdE + ATP + pep1_PA_085	BIO	MTKSG7OPFRGCG	MAP4K1	ABL1, JAK2
1K14	Key_L1_068 - sample + KdE + ATP + pep1_PA_086	BIO	RQDKE7YKVEGCG	JAK2	ABL1
1K14	Key_L1_069 - sample + KdE + ATP + pep1_PA_087	BIO	SEPPV7AHLSNGCG	JUN	ABL1
1K14	Key_L1_070 - sample + KdE + ATP + pep1_PA_088	BIO	VEAVA7APKVEGCG	PAD51	ABL1

1A005	IPA_095	Key_L1_071 - sample + KaB + ATP A + pept. PA_098	BIO	PTTSP:QAQPSKGC	TF73	Y89	ABL1
1A08	IPA_090	Key_L1_072 - sample + KaB + ATP A + pept. PA_099	BIO	TSKVYDPEKGC	WA5L	Y256	ABL1; PTK2 (FAK, FAK1)
1C08	IPA_091	Key_L1_073 - sample + KaB + ATP C + pept. PA_091	CON - _mutated BIO	VEA VAGAFKKEGC	RAD51 - _CON (-) _ Y to G	Y54 _ Y to G	ABL1
1EE08	IPA_092	Key_L1_074 - sample + KaB + ATP C + pept. PA_092	CON - _mutated BIO	VEA VAIYIAPKKEGC	RAD51 - _CON (-) _ Y to pY	Y54 _ Y to pY	ABL1
1A08	IPA_093	Key_L1_075 - sample + KaB + ATP B + pept. PA_093	CON +	EGYIYGLGC	consensus _ CON+	-	FYAI, HCK, LOK, LYNA, LYNB, SRC BLK, BRK (PTK6), FOR, FRK, FYN, HCK, LOK, LYNA, LYNB, SRC, SRMS, YES1
1A08	IPA_094	Key_L1_076 - sample + KaB + ATP B + pept. PA_094	CON +	KYIVGYTGEGIKVEKGC	consensus _ CON+	-	ABL1; ABL1 (H3989), ABL1 (Q3224) ABL1 (T315), ABL2, BTK, FLT3, FLT3 (D635Y), KIT (SCFR), MET, RET
1A08	IPA_095	Key_L1_077 - sample + KaB + ATP B + pept. PA_095	CON +	EAIYAAPGC	consensus _ CON+	-	ABL1, ABL2 LYNB, SRMS, EPHA1, EPHA2, EPHA3, EPHA4, EPHA5, EPHA6, EPHA7, EPHA8, EPH9, EPH10, EPH11, EPH12, EPH13, EPH14, EPH15, EPH16, EPH17, EPH18, EPH19, EPH20, EPH21, EPH22, EPH23, EPH24, EPH25, EPH26, EPH27, EPH28, EPH29, EPH30, EPH31, EPH32, EPH33, EPH34, EPH35, EPH36, EPH37, EPH38, EPH39, EPH40, EPH41, EPH42, EPH43, EPH44, EPH45, EPH46, EPH47, EPH48, EPH49, EPH50, EPH51, EPH52, EPH53, EPH54, EPH55, EPH56, EPH57, EPH58, EPH59, EPH60, EPH61, EPH62, EPH63, EPH64, EPH65, EPH66, EPH67, EPH68, EPH69, EPH70, EPH71, EPH72, EPH73, EPH74, EPH75, EPH76, EPH77, EPH78, EPH79, EPH80, EPH81, EPH82, EPH83, EPH84, EPH85, EPH86, EPH87, EPH88, EPH89, EPH90, EPH91, EPH92, EPH93, EPH94, EPH95, EPH96, EPH97, EPH98, EPH99, EPH100, EPH101, EPH102, EPH103, EPH104, EPH105, EPH106, EPH107, EPH108, EPH109, EPH110, EPH111, EPH112, EPH113, EPH114, EPH115, EPH116, EPH117, EPH118, EPH119, EPH120, EPH121, EPH122, EPH123, EPH124, EPH125, EPH126, EPH127, EPH128, EPH129, EPH130, EPH131, EPH132, EPH133, EPH134, EPH135, EPH136, EPH137, EPH138, EPH139, EPH140, EPH141, EPH142, EPH143, EPH144, EPH145, EPH146, EPH147, EPH148, EPH149, EPH150, EPH151, EPH152, EPH153, EPH154, EPH155, EPH156, EPH157, EPH158, EPH159, EPH160, EPH161, EPH162, EPH163, EPH164, EPH165, EPH166, EPH167, EPH168, EPH169, EPH170, EPH171, EPH172, EPH173, EPH174, EPH175, EPH176, EPH177, EPH178, EPH179, EPH180, EPH181, EPH182, EPH183, EPH184, EPH185, EPH186, EPH187, EPH188, EPH189, EPH190, EPH191, EPH192, EPH193, EPH194, EPH195, EPH196, EPH197, EPH198, EPH199, EPH200, EPH201, EPH202, EPH203, EPH204, EPH205, EPH206, EPH207, EPH208, EPH209, EPH210, EPH211, EPH212, EPH213, EPH214, EPH215, EPH216, EPH217, EPH218, EPH219, EPH220, EPH221, EPH222, EPH223, EPH224, EPH225, EPH226, EPH227, EPH228, EPH229, EPH230, EPH231, EPH232, EPH233, EPH234, EPH235, EPH236, EPH237, EPH238, EPH239, EPH240, EPH241, EPH242, EPH243, EPH244, EPH245, EPH246, EPH247, EPH248, EPH249, EPH250, EPH251, EPH252, EPH253, EPH254, EPH255, EPH256, EPH257, EPH258, EPH259, EPH260, EPH261, EPH262, EPH263, EPH264, EPH265, EPH266, EPH267, EPH268, EPH269, EPH270, EPH271, EPH272, EPH273, EPH274, EPH275, EPH276, EPH277, EPH278, EPH279, EPH280, EPH281, EPH282, EPH283, EPH284, EPH285, EPH286, EPH287, EPH288, EPH289, EPH290, EPH291, EPH292, EPH293, EPH294, EPH295, EPH296, EPH297, EPH298, EPH299, EPH300, EPH301, EPH302, EPH303, EPH304, EPH305, EPH306, EPH307, EPH308, EPH309, EPH310, EPH311, EPH312, EPH313, EPH314, EPH315, EPH316, EPH317, EPH318, EPH319, EPH320, EPH321, EPH322, EPH323, EPH324, EPH325, EPH326, EPH327, EPH328, EPH329, EPH330, EPH331, EPH332, EPH333, EPH334, EPH335, EPH336, EPH337, EPH338, EPH339, EPH340, EPH341, EPH342, EPH343, EPH344, EPH345, EPH346, EPH347, EPH348, EPH349, EPH350, EPH351, EPH352, EPH353, EPH354, EPH355, EPH356, EPH357, EPH358, EPH359, EPH360, EPH361, EPH362, EPH363, EPH364, EPH365, EPH366, EPH367, EPH368, EPH369, EPH370, EPH371, EPH372, EPH373, EPH374, EPH375, EPH376, EPH377, EPH378, EPH379, EPH380, EPH381, EPH382, EPH383, EPH384, EPH385, EPH386, EPH387, EPH388, EPH389, EPH390, EPH391, EPH392, EPH393, EPH394, EPH395, EPH396, EPH397, EPH398, EPH399, EPH400, EPH401, EPH402, EPH403, EPH404, EPH405, EPH406, EPH407, EPH408, EPH409, EPH410, EPH411, EPH412, EPH413, EPH414, EPH415, EPH416, EPH417, EPH418, EPH419, EPH420, EPH421, EPH422, EPH423, EPH424, EPH425, EPH426, EPH427, EPH428, EPH429, EPH430, EPH431, EPH432, EPH433, EPH434, EPH435, EPH436, EPH437, EPH438, EPH439, EPH440, EPH441, EPH442, EPH443, EPH444, EPH445, EPH446, EPH447, EPH448, EPH449, EPH450, EPH451, EPH452, EPH453, EPH454, EPH455, EPH456, EPH457, EPH458, EPH459, EPH460, EPH461, EPH462, EPH463, EPH464, EPH465, EPH466, EPH467, EPH468, EPH469, EPH470, EPH471, EPH472, EPH473, EPH474, EPH475, EPH476, EPH477, EPH478, EPH479, EPH480, EPH481, EPH482, EPH483, EPH484, EPH485, EPH486, EPH487, EPH488, EPH489, EPH490, EPH491, EPH492, EPH493, EPH494, EPH495, EPH496, EPH497, EPH498, EPH499, EPH500, EPH501, EPH502, EPH503, EPH504, EPH505, EPH506, EPH507, EPH508, EPH509, EPH510, EPH511, EPH512, EPH513, EPH514, EPH515, EPH516, EPH517, EPH518, EPH519, EPH520, EPH521, EPH522, EPH523, EPH524, EPH525, EPH

Table 1

PA_112	Key_L1_097 - sample + KαB + ATP A	BIO	TERGP/TKVSA/GC	MUC1	Y1229	EGFR, ZAP70
PA_113	Key_L1_098 - sample + KαB + ATP A	BIO	AEGSA/VEEPTGC	PLCG1	Y472	EGFR
PA_114	Key_L1_099 - sample + KαB + ATP A	BIO	CFQAH/NMYPQGC	STAT5B	Y740	EGFR
PA_115	Key_L1_100 - sample + KαB + ATP A	BIO	NEDGG/DYKPGC	ICBL	Y744	EGFR, SYK
PA_116	Key_L1_101 - sample + KαB + ATP A	BIO	DVETT/ADFIAGC	PKIA	Y8	EGFR
PA_117	Key_L1_102 - sample + KαB + ATP A	BIO	VDAAE/LVPOGGC	ERBB2	Y1023	ERBB2
PA_118	Key_L1_103 - sample + KαB + ATP A	BIO	RRSLG/YAYNFGC	PABPC1	Y54	ERBB3 (in its dimeric state with another HER fam. member)
PA_119	Key_L1_104 - sample + KαB + ATP A	BIO	TCACL/VEKLPGGC	TEK (TIE2)	Y1048	ERBB4
PA_120	Key_L1_105 - sample + KαB + ATP A	BIO	AEINPE/LGLDVGC	ERBB2	Y1248	EGFR, ERBB2
PA_121	Key_L1_106 - sample + KαB + ATP A	BIO	AEINAE/LRVAPOC	EGFR	Y1197	EGFR, ERBB4
PA_122	Key_L1_107 - sample + KαB + ATP A	BIO	KAVDG/VKPOIGC	STAT5A / STAT5B	Y694 / Y699	ERBB4, JAK2, JAK3, BTK, SYK, LCK, SRC, EGFR, JAK2, BSK (PTK8), HCK, SRC
PA_123	Key_L1_108 - sample + KαB + ATP A	BIO	TVPNP/MT/PAGC	JAB1	Y213	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_124	Key_L1_109 - sample + KαB + ATP A	BIO	GVMED/DYVHLGC	BCAR1	Y654	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1), BRK (PTK6)
PA_125	Key_L1_110 - sample + KαB + ATP A	BIO	IGEGA/GKVKPGC	CDK6	Y24	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_126	Key_L1_111 - sample + KαB + ATP A	BIO	EDELP/DDCVFGC	NFKBIA (IKBA)	Y305	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_127	Key_L1_112 - sample + KαB + ATP A	BIO	RADEN/VKARQGC	SYK	Y525	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_128	Key_L1_113 - sample + KαB + ATP C	CON - _mutated BIO	PKGTG/GKTELGC	STAT1 _CON (-) _Y to G	Y701 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1), BRK (PTK6)
PA_129	Key_L1_114 - sample + KαB + ATP C	CON - _mutated BIO	DKQVE/LDLGLGC	GAB1 _CON (-) _Y to G	Y627 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_130	Key_L1_115 - sample + KαB + ATP C	CON - _mutated BIO	AEGSA/EEVPTGC	PLCG1 _CON (-) _Y to G	Y472 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_131	Key_L1_116 - sample + KαB + ATP C	CON - _mutated BIO	VDAAE/LVPOGGC	ERBB2 _CON (-) _Y to G	Y1023 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_132	Key_L1_117 - sample + KαB + ATP C	CON - _mutated BIO	RRSLG/YAYNFGC	PABPC1 _CON (-) _Y to G	Y54 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_133	Key_L1_118 - sample + KαB + ATP C	CON - _mutated BIO	TCACL/VEKLPGGC	TEK (TIE2) _CON (-) _Y to G	Y1048 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_134	Key_L1_119 - sample + KαB + ATP C	CON - _mutated BIO	GVMED/DYVHLGC	BCAR1 _CON (-) _Y to G	Y654 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1)
PA_135	Key_L1_120 - sample + KαB + ATP C	CON - _mutated BIO	EDELP/DDCVFGC	NFKBIA (IKBA) _CON (-) _Y to G	Y305 _Y to G	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_136	Key_L1_121 - sample + KαB + ATP C	CON - _mutated BIO	PKGTG/VEKTELGC	STAT1 _CON (-) _Y to pY	Y701 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1)
PA_137	Key_L1_122 - sample + KαB + ATP C	CON - _mutated BIO	DKQVE/VEKTELGC	GAB1 _CON (-) _Y to pY	Y627 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_138	Key_L1_123 - sample + KαB + ATP C	CON - _mutated BIO	AEGSA/VEEPTGC	PLCG1 _CON (-) _Y to pY	Y472 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_139	Key_L1_124 - sample + KαB + ATP C	CON - _mutated BIO - Rep (B1-4)_J5	AEGSA/VEEPTGC	PLCG1 _CON (-) _Y to pY	Y472 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_140	Key_L1_125 - sample + KαB + ATP C	CON - _mutated BIO	VDAAE/VEKTELPGGC	ERBB2 _CON (-) _Y to pY	Y1023 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_141	Key_L1_126 - sample + KαB + ATP C	CON - _mutated BIO	RRSLG/YAYNFGC	PABPC1 _CON (-) _Y to pY	Y54 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_142	Key_L1_127 - sample + KαB + ATP C	CON - _mutated BIO	TCACL/VEKTELPGGC	TEK (TIE2) _CON (-) _Y to pY	Y1048 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1)
PA_143	Key_L1_128 - sample + KαB + ATP C	CON - _mutated BIO	GVMED/DYVHLGC	BCAR1 _CON (-) _Y to pY	Y654 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4, PTK2 (FAK, FAK1)
PA_144	Key_L1_130 - sample + KαB + ATP A	CON - _mutated BIO	EDELP/VEKTELPGGC	NFKBIA (IKBA) _CON (-) _Y to pY	Y305 _Y to pY	ERBB3 (in its dimeric state with another HER fam. member), ERBB4
PA_145	Key_L1_131 - sample + KαB + ATP A	BIO	SFLOR/SDDPTGC	EGFR	Y1056	EGFR, JAK2
PA_146	Key_L1_132 - sample + KαB + ATP A	BIO	IVRKR/LRRLOGC	EGFR	T678	PRKCA (PKC alpha), PRKCD (PKC delta, PKCQ), PRKDI (PKD1, PKD), PRKN1 (PRK1)
		BIO	AEEKE/HAESGGC	EGFR	Y659	SRC, EGFR

82

Table 1

PA_177	1009	Key_L1_164 - sample + KaB + ATP A + pept. PA_177	BIO	LSYLQSPITSGC	DUSP1	S59	MAPK1 (ERK2), MAPK3 (ERK1)
PA_178	1006	Key_L1_165 - sample + KaB + ATP A + pept. PA_178	BIO	QDKEIYKKEFGC	JAK2	Y1008	JAK2
PA_179	1009	Key_L1_166 - sample + KaB + ATP A + pept. PA_179	BIO	VLREIYPRVGGC	CASP9	T125	MAP2K1 (MEK1, MEK2), MAPK2, MAPK3 (ERK1, ERK2); MAPK4, MAPK5 (ERK1)
PA_180	1009	Key_L1_167 - sample + KaB + ATP A + pept. PA_180	BIO	GVITITPPIPGGC	JUN / JUNB	T81 / T102	MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_181	1005	Key_L1_168 - sample + KaB + ATP A + pept. PA_181	BIO	IDLFIKSPETLDGC	STAT3	S727	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_182	1009	Key_L1_169 - sample + KaB + ATP A + pept. PA_182	BIO	RPRTTIFAEGGC	GSX3B	S9	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_183	1006	Key_L1_170 - sample + KaB + ATP A + pept. PA_183	BIO	PAARTSIFAEGGC	GSX3A	S21	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_184	1011	Key_L1_171 - sample + KaB + ATP B + pept. PA_184	CON +	KPRRAAIFAEGC	consensus _ CON +	-	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_185	1011	Key_L1_172 - sample + KaB + ATP B + pept. PA_185	CON +	RPRAATFPGC	consensus _ CON +	-	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_186	1011	Key_L1_173 - sample + KaB + ATP B + pept. PA_186	CON +	PRTISFAEGC	consensus _ CON +	-	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_187	1011	Key_L1_174 - sample + KaB + ATP B + pept. PA_187	CON +	IFTSPITTFVGC	consensus _ CON +	-	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_188	1011	Key_L1_175 - sample + KaB + ATP A + pept. PA_188	BIO	ADQTPYPRFEGC	ATF2 (CREB2)	T71	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_189	1011	Key_L1_176 - sample + KaB + ATP A + pept. PA_189	BIO	RSRHSYFAGTGC	BAD	S75	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_190	1011	Key_L1_177 - sample + KaB + ATP A + pept. PA_190	BIO	RGRSFAAPPFGC	BAD	S89	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_191	1011	Key_L1_178 - sample + KaB + ATP A + pept. PA_191	BIO	PKRRPISGLHFGC	BRCA1	T509	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_192	1013	Key_L1_179 - sample + KaB + ATP A + pept. PA_192	BIO	LFPWKIFVKTGCG	BRCA1	S888	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_193	1013	Key_L1_180 - sample + KaB + ATP A + pept. PA_193	BIO	KRRQYIMTDFGCG	CDKN1A (p21, WAF1, CIP1)	S146	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_194	1008	Key_L1_181 - sample + KaB + ATP A + pept. PA_194	BIO	PKRRQYIMTDFGCG	CDKN1A (p21, WAF1, CIP1)	T145	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_194	1013	Key_L1_182 - sample + KaB + ATP A + pept. PA_194	BIO	PKRRQYIMTDFGCG	CDKN1A (p21, WAF1, CIP1)	T145	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_194	2006	Key_L1_183 - sample + KaB + ATP A + pept. PA_194	BIO	PKRRQYIMTDFGCG	CDKN1A (p21, WAF1, CIP1)	T145	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)
PA_195	1013	Key_L1_183 - sample + KaB + ATP A + pept. PA_195	BIO	RPRTTIFAEGGC	CHEK1	S280	MAPK1 (ERK2), MAPK3 (ERK1), MAPK4 (ERK2), MAPK5 (ERK1)

Table 1

PA_196	1,113	Key_L1_184 - sample + K _{AB} + ATP + pept_PA_186	BIO	L:RRP:YRKLG	CREB1 / CREM	S113 / S195	AKT1, ATM, CAMK1, CAMK2, CAMK2G, CAMK4, CDK1 (CD2), MAP3K8 (COT), MAPKAPK2 (MK2), PRKACA (PKA C-alpha), PRKACA (PKC alpha, PKCA), PRKCE (PKC epsilon, PKCE), PRKOT (PKOT, PKD), PRKGT (PKG 1), RPS30A1 (RSK1), RPS30A2 (RSK3), RPS30A3 (RSK2), RPS30A4 (MSK2), RPS30A5 (MSK1), SGK (SGK), TSG1, CAMK2G, CAMK2G, CDK1 (CD2), PRKACA (PKA C-alpha), PRKCA (PKC alpha, PKCA)
PA_197	1,113	Key_L1_185 - sample + K _{AB} + ATP + pept_PA_187	BIO	MECRN:PVTKGC	EIF4EBP1 (4EBP1)	S65	AKT1, MAPK1 (ERK2), MAPK14 (p38), MTOR (FRAP1), RPS30A5 (MSK1), DYRK2
PA_198	1,113	Key_L1_186 - sample + K _{AB} + ATP + pept_PA_188	BIO	GDY31:PGAGTLC	EIF4EBP1 (4EBP1)	T37	AKT1, MAPK1 (ERK2), MAPK14 (p38), MTOR (FRAP1), GSK3B
PA_199	1,113	Key_L1_187 - sample + K _{AB} + ATP + pept_PA_189	BIO	RPRRA:IMDNSNGC	FOXO3	S253	AKT1, MAPKAPK3, PIM1, SGK (SGK)
PA_200	1,113	Key_L1_188 - sample + K _{AB} + ATP + pept_PA_200	BIO	RPRSC:WPLQRGC	FOXO3 / FOXO1	T32 / T24	AKT1, AKT2, SGK1 (SGK), PIM1 / AKT1, AKT2, SGK1 (SGK)
PA_201	1,113	Key_L1_189 - sample + K _{AB} + ATP + pept_PA_201	BIO	RTR:DSYSAQDGC	MTOR (FRAP1)	S2443	AKT1, AKT3, RPS30A1 (SGK1), RPS30A2 (SGK2)
PA_202	1,113	Key_L1_190 - sample + K _{AB} + ATP + pept_PA_202	BIO	RSRTR:DSYBAGC	MTOR (FRAP1)	T2446	AKT1, AKT3, RPS30A1 (SGK1)
PA_202	1,113	Key_L1_191 - sample + K _{AB} + ATP + pept_PA_202	BIO	RSRTR:DSYBAGC	MTOR (FRAP1)	T2448	AKT1, AKT3, RPS30A1 (SGK1)
PA_202	1,113	Key_L1_192 - sample + K _{AB} + ATP + pept_PA_202	BIO	RSRTR:DSYBAGC	MTOR (FRAP1)	T2446	AKT1, AKT3, RPS30A1 (SGK1)
PA_203	1,113	Key_L1_193 - sample + K _{AB} + ATP + pept_PA_203	BIO	YDIPP:PGNTYGC	GAB1	T312	MAPK1 (ERK2), MAPK3 (ERK1)
PA_204	1,113	Key_L1_194 - sample + K _{AB} + ATP + pept_PA_204	BIO	N:VFM:PGTFDGC	GAB1	T476	MAPK1 (ERK2), MAPK3 (ERK1)
PA_205	1,113	Key_L1_195 - sample + K _{AB} + ATP + pept_PA_205	BIO	REKRS:APSHSGC	GAB2	S159	AKT1
PA_206	1,113	Key_L1_196 - sample + K _{AB} + ATP + pept_PA_206	BIO	DFQP:SPSPHRC	GAB2	S623	MAPK1 (ERK2), MAPK3 (ERK1)
PA_207	1,113	Key_L1_197 - sample + K _{AB} + ATP + pept_PA_207	BIO	TPGPG:PPVLTGC	GAB2	S150	MAPK1 (ERK2), MAPK3 (ERK1), MTOR (FRAP1)
PA_208	1,113	Key_L1_198 - sample + K _{AB} + ATP + pept_PA_208	BIO	ILGSG:PLHFSGC	GRB10	S476	MAPK1 (ERK2), MAPK3 (ERK1), MTOR (FRAP1)
PA_210	1,113	Key_L1_199 - sample + K _{AB} + ATP + pept_PA_210	BIO	SNFPA:PLQDNGC	GRE10	S58	MAPK1 (ERK2), MAPK3 (ERK1)
PA_211	1,113	Key_L1_200 - sample + K _{AB} + ATP + pept_PA_211	BIO	GSFLP:PPPAEGC	LOCK	T163	MAPK1 (ERK2), MAPK3 (ERK1), GSK3A, GSK3B
PA_212	1,113	Key_L1_201 - sample + K _{AB} + ATP + pept_PA_212	BIO	RRRA:SETEENGC	MLC1	S166	AKT1, CSNK1D (CK1D), DAPK3 (ZFPK), MAPKAPK2, PIM1, SGK (SGK)
PA_212	1,113	Key_L1_202 - sample + K _{AB} + ATP + pept_PA_212	BIO	FELL:PPPLPGC	MDM2	T58	GSK3B, CSNK2A1 (CK2A1), CSNK2A2 (CK2A2)
PA_213	1,113	Key_L1_203 - sample + K _{AB} + ATP + pept_PA_213	BIO	FELL:PPPLPGC	MYC	T58	GSK3B, CSNK2A1 (CK2A1), CSNK2A2 (CK2A2)
PA_213	1,113	Key_L1_204 - sample + K _{AB} + ATP + pept_PA_213	BIO	FELL:PPPLPGC	MYC	T58	GSK3B, CSNK2A1 (CK2A1), CSNK2A2 (CK2A2)
PA_214	1,113	Key_L1_205 - sample + K _{AB} + ATP + pept_PA_214	BIO	HSLPL:SPASTRGC	NFKB1	S807	GSK3B
PA_215	1,113	Key_L1_206 - sample + K _{AB} + ATP + pept_PA_215	BIO	HSLPL:SPASTRGC	NFKB1	S807	GSK3B
PA_216	1,113	Key_L1_207 - sample + K _{AB} + ATP + pept_PA_216	BIO	HSLPL:SPASTRGC	NFKB1	S807	GSK3B
PA_217	1,113	Key_L1_208 - sample + K _{AB} + ATP + pept_PA_217	BIO	YNVTP:PPPGVGC	NOTCH2	S2070	MAPK1 (ERK2), CDK5
PA_218	1,113	Key_L1_209 - sample + K _{AB} + ATP + pept_PA_218	BIO	FELPV:PTRDVGC	PAK1	T212	MAPK1 (ERK2), MAPK3 (ERK1)
PA_219	1,113	Key_L1_210 - sample + K _{AB} + ATP + pept_PA_219	BIO	PLCLP:PLEAGGC	PPARA	S12	MAPK1 (ERK2), MAPK3 (ERK1)
PA_220	1,113	Key_L1_211 - sample + K _{AB} + ATP + pept_PA_220	BIO	AGDLE:PLSEEGC	PPARA	S21	MAPK1 (ERK2), MAPK3 (ERK1)
PA_221	1,113	Key_L1_212 - sample + K _{AB} + ATP + pept_PA_221	BIO	KVEPA:PPYVSGC	PPARG	S112	MAPK1 (ERK2), MAPK3 (ERK1), CDK9
PA_222	1,113	Key_L1_213 - sample + K _{AB} + ATP + pept_PA_222	BIO	OH:YRY:DT:YSGC	PTEN	S880, T382, T383	AKT1, CSNK2A1 (CK2A1), PLK1, PRKC2 (PKC zeta), ROCK1, STRK1 (LKB)
PA_223	1,113	Key_L1_214 - sample + K _{AB} + ATP + pept_PA_223	BIO	GYPNQ:POFTLGC	RPS30A4 (MSK2)	T566	MAP2K1 (MEK1, MEK2), MAPK14 (p38)
				LEPV:PPGSGC	RPS30A4 (MSK2)	S343	MAP2K1 (MEK1, MEK2), MAPK14 (p38)
				PHGPR:AVSSGC	SOS1	S1132	MAPK1 (ERK2)
				TSKAY:PPRYSGC	SOS1	S1193	MAPK1 (ERK2)

Table 1

PA_224	Key_L1_215 - sample + KdB + ATP A + pept. PA_224	1816	VEPPLSQE1F3GC	BIO	TP53	S15	MAPK1 (ERK2), MAPK3 (ERK1), MAPK14 (p38), ATM, ATR, CHEK1, CHEK2, PRKDC (DNA-PKcs), DNAPK1, CDK5, DYRK1A, NUAK1, TP53RK (PRPK)
PA_225	Key_L1_216 - sample + KdB + ATP A + pept. PA_225	1818	IEOWFIEDPGRGC	BIO	TP53	T55	MAPK1 (ERK2), GRK5, TAF1
PA_226	Key_L1_217 - sample + KdB + ATP A + pept. PA_226	1818	COLLMFKEGPDSD	BIO	TP53	S382	MAPK14 (p38), CDK2, CDK7, CDK9, CSNK2A1 (CK2A1), EIF2AK2, NUAK1, STRK11 (LKB)
PA_227	Key_L1_218 - sample + KdB + ATP C + pept. PA_227	1818	AEEKEIYVHAEGGGC	CON - _mutated BIO	EGFR _CON (-) _Y to pY	Y659 _Y to pY	SRC, EGFR
PA_228	Key_L1_219 - sample + KdB + ATP C + pept. PA_228	1818	AEEKEGHAEGGGC	CON - _mutated BIO	EGFR _CON (-) _Y to G	Y659 _Y to G	SRC, EGFR
PA_229	Key_L1_220 - sample + KdB + ATP C + pept. PA_229	1818	SFLGTEIYVATRVGCG	CON - _mutated BIO	MAPK1 (ERK2) / MAPK3 (ERK1) _CON (-) _TY to pTY	Y187 (+T185) / Y204 (+T202) _TY to pTY	MAP2K1 (MEK1, MKK1, MAPKK1), MAP2K2 (MEK2, MKK2, MAPK3), MAPK3 (ERK2), MAPK3 (ERK1), RET, JAK2
PA_230	Key_L1_221 - sample + KdB + ATP C + pept. PA_230	1818	SFLGEGVATRVGCG	CON - _mutated BIO	MAPK1 (ERK2) / MAPK3 (ERK1) _CON (-) _TY to G/G	Y187 (+T185) / Y204 (+T202) _TY to G/G	MAP2K1 (MEK1, MKK1, MAPKK1), MAP2K2 (MEK2, MKK2, MAPK3), MAPK3 (ERK2), MAPK3 (ERK1), RET, JAK2
PA_231	Key_L1_222 - sample + KdB + ATP C + pept. PA_231	1818	HTGRLQTEVVA13C	CON - _mutated BIO	MAPK1 (ERK2) / MAPK3 (ERK1) _CON (-) _T to pT	T185 (+Y187) / T202 (+Y204) _T to pT	MAP2K1 (MEK1, MKK1, MAPKK1), MAP2K2 (MEK2, MKK2, MAPK3), MAPK3 (ERK2), MAPK3 (ERK1), RET, JAK2
PA_232	Key_L1_223 - sample + KdB + ATP C + pept. PA_232	1830	RFRTTQSFIAESGCG	CON - _mutated BIO	GSK3B _CON (-) _S to pS	S9 _S to pS	AKT1, AKT3, AURKA, PRKACA (PKA C-alpha), MAP4K5, PRKCA (PKC alpha, PKCA), PRKCB (PKC beta, PKCB), PRKCD (PKC delta, PRKCD), PRKCG (PKC gamma, PRKCG), PRKCH (PKC eta, PKCL), RPSSA1 (RSK1, p90RSK), RPSSA3 (RSK2, p90RSK2, 56K-alpha-3), RPSSA4 (RSK1, p70S6K, 56K-alpha-1), RPSSA2 (SK2, p70S6K, 56K-alpha-2), SGK1 (SGK), SGK3 (SGKL)
PA_233	Key_L1_224 - sample + KdB + ATP C + pept. PA_233	1820	RGKpSRpG5APPNLECG	CON - _mutated BIO	BAD _CON (-) _S/S to pS/pS	S99 _S/S to pS/pS	AKT1, AKT3, PAK1, PAK2, PAK3, PRKACA (PKA C-alpha), PRKCB (PKC beta, PKCB), PRKCD (PKC delta, PRKCD), PRKCG (PKC gamma, PRKCG), PRKCH (PKC eta, PKCL), RPSSA1 (RSK1, p90RSK), RPSSA3 (RSK2, p90RSK2, 56K-alpha-3), RPSSA4 (RSK1, p70S6K, 56K-alpha-1), RPSSA2 (SK2, p70S6K, 56K-alpha-2), SGK1 (SGK), SGK3 (SGKL)
PA_234	Key_L1_225 - sample + KdB + ATP C + pept. PA_234	1820	RSRTpTDPpSYSA9C	CON - _mutated BIO	MTOR (RAP1) _CON (-) _T/S to pT/pS	T248 _T/S to pT/pS	AKT1, AKT3, PAK1, PAK2, PAK3, PRKACA (PKA C-alpha), PRKCB (PKC beta, PKCB), PRKCD (PKC delta, PRKCD), PRKCG (PKC gamma, PRKCG), PRKCH (PKC eta, PKCL), RPSSA1 (RSK1, p90RSK), RPSSA3 (RSK2, p90RSK2, 56K-alpha-3), RPSSA4 (RSK1, p70S6K, 56K-alpha-1), RPSSA2 (SK2, p70S6K, 56K-alpha-2), SGK1 (SGK), SGK3 (SGKL)
PA_235	Key_L1_226 - sample + KdB + ATP C + pept. PA_235	1830	SNPPA(pS)PQDNGC	CON - _mutated BIO	LOCK _CON (-) _S to pS	S59 _S to pS	MAPK1 (ERK2), MAPK3 (ERK1)
PA_236	Key_L1_227 - sample + KdB + ATP C + pept. PA_236	1820	SNPPA(pLQDNGC	CON - _mutated BIO	LOCK _CON (-) _S to G	S59 _S to G	MAPK1 (ERK2), MAPK3 (ERK1)
PA_237	Key_L1_228 - sample + KdB + ATP C + pept. PA_237	1820	RRRA(pS)E(pTE)ENGCG	CON - _mutated BIO	MDM2 _CON (-) _S/T to pS/pT	S155 _S/T to pS/pT	AKT1, CSNK1D (CK1D), DAPK3 (ZIPK), MAPKAPK2, PIM1, SGK1 (SGK)
PA_238	Key_L1_229 - sample + KdB + ATP C + pept. PA_238	1830	RRRA(pS)E(pTE)ENGCG	CON - _mutated BIO	MDM2 _CON (-) _S/T to G/G	S155 _S/T to G/G	AKT1, CSNK1D (CK1D), DAPK3 (ZIPK), MAPKAPK2, PIM1, SGK1 (SGK)
PA_239	Key_L1_230 - sample + KdB + ATP C + pept. PA_239	1820	ERLP(pT)PDTROVGC	CON - _mutated BIO	PAK1 _CON (-) _T/T to pT/pT	T212 _T/T to pT/pT	MAPK1 (ERK2), CDK5
PA_240	Key_L1_231 - sample + KdB + ATP B + pept. PA_240	1807	KVENGECTGVVYK-amide	CON +	consensus _CON+	-	ABL1, BCR-ABL, ABL2, AXL, AXL496C, FLT2, FLT3(D835Y), LTK (TYK1), all TKs (>80 Tyrosine Kinases, but more commonly used for ALK, ETK, BTK(E41K), EPHA1, EPHB1, PTK2 (FAK, FAK1), PTK2B (FAK2, PYK2), FES, FGFR1, FGFR1(V501M), FGFR2, FGFR3, FGFR4, FLT1 (VEGFR1), FLT4 (VEGFR3), JAK1, JAK2, JAK3, NDR (VEGFR2) KIT (SCFR), MET, NTRK1, NTRK2 (TRKB), PDGFRA, PDGFRB, SYK, TIE2 (TENG), TYK2, ZAP70
PA_241	Key_L1_232 - sample + KdB + ATP B + pept. PA_241	1806	(EEEE - Y)-46	CON +	consensus _CON+	-	ABL1, BCR-ABL, ABL2, AXL, AXL496C, FLT2, FLT3(D835Y), LTK (TYK1), all TKs (>80 Tyrosine Kinases, but more commonly used for ALK, ETK, BTK(E41K), EPHA1, EPHB1, PTK2 (FAK, FAK1), PTK2B (FAK2, PYK2), FES, FGFR1, FGFR1(V501M), FGFR2, FGFR3, FGFR4, FLT1 (VEGFR1), FLT4 (VEGFR3), JAK1, JAK2, JAK3, NDR (VEGFR2) KIT (SCFR), MET, NTRK1, NTRK2 (TRKB), PDGFRA, PDGFRB, SYK, TIE2 (TENG), TYK2, ZAP70
PA_242	Key_L1_233 - sample + KdB + ATP B + pept. PA_242	1812	EALAAFFAKK	CON +	consensus _CON+	-	ABL1, BCR-ABL, ABL2, AXL, AXL496C, FLT2, FLT3(D835Y), LTK (TYK1), all TKs (>80 Tyrosine Kinases, but more commonly used for ALK, ETK, BTK(E41K), EPHA1, EPHB1, PTK2 (FAK, FAK1), PTK2B (FAK2, PYK2), FES, FGFR1, FGFR1(V501M), FGFR2, FGFR3, FGFR4, FLT1 (VEGFR1), FLT4 (VEGFR3), JAK1, JAK2, JAK3, NDR (VEGFR2) KIT (SCFR), MET, NTRK1, NTRK2 (TRKB), PDGFRA, PDGFRB, SYK, TIE2 (TENG), TYK2, ZAP70
PA_245	Key_L1_234 - sample + KdB + ATP A + pept. PA_245	1801	ELRRM2DEFVGGC	BIO	BAD	S118	AKT1, PIM1, PIM2, PMK3, PRKACA (PKA C-alpha), RPSSA1 (RSK1, p90RSK), RPSSA2 (RSK3, p90RSK2, 56K-alpha-2)
PA_246	Key_L1_235 - sample + KdB + ATP B + pept. PA_246	1801	KKVERSGLYRSPSPMPENLGC	CON +	consensus _CON+	-	MAPK1 (ERK2), MAPK3 (ERK1), TSSK1, TSSK2
PA_247	Key_L1_236 - sample + KdB + ATP A + pept. PA_247	1801	PPPLQSPFLQPGC	BIO	ESR1	S118	CDK7, MAPK1 (ERK2), RPSSA1 (RSK1, p90RSK)
PA_248	Key_L1_237 - sample + KdB + ATP A + pept. PA_248	1807	CGGKATQA:QEV	BIO	H2AFX	S140 (C-term end)	ATM
PA_249	Key_L1_238 - sample + KdB + ATP A + pept. PA_249	1809	NDDV:RSLEEGC	BIO	VAI2	Y142	EGFR
PA_250	Key_L1_239 - sample + KdB + ATP A + pept. PA_250	1811	ISEGT:LNDLGG	BIO	TOFBR1	T176	CLK1, TOFBR2

Table 1

PA	Key	CON	PIV	MYC	SE2	MAPK
PA_251	Key_L1_240 - sample + KaB + ATP C + pept. PA_251	CON - mutated BIO	PIVFLAPARRAGC	MYC - CON (-) - ST to A/V	S52	MAPK3 (JNK1)
PA_252	Key_L1_241 - sample + KaB + ATP A + pept. PA_252	BIO	ELDQGR3LCTSGGC	IKKKB (IKK beta)	S177	CHUK (IKK alpha), MAP3K7 (MEKK1/2 TAK1), MAP3K14 (NIK), PRKCG (PKC zeta), PRKCG (PKC zeta)
PA_253	Key_L1_242 - sample + KaB + ATP A + pept. PA_253	BIO	DEDD51LEPDSGGC	SH3BP2	Y183	SYK
PA_254	Key_L1_243 - sample + KaB + ATP A + pept. PA_254	BIO	VOPDM3WSSSLGCG	BFC42	S153	PLK1
PA_255	Key_L1_244 - sample + KaB + ATP A + pept. PA_255	BIO	CGTPKAPGLRRRGCT	CDKN1B (p27, KIP1)	T196 (C-term end)	PRK12 (PRK2)
PA_256	Key_L1_245 - sample + KaB + ATP A + pept. PA_256	BIO	NIKRR5VTFPEGCG	CDK25B	S553	AURKA
PA_257	Key_L1_246 - sample + KaB + ATP C + pept. PA_257	CON - mutated BIO	LELVAVQELFAGCG	CHEK2 - CON (-) - S17Y to A/V/F	T58	ATM, CHEK2, PLK1, PLK3
PA_258	Key_L1_247 - sample + KaB + ATP A + pept. PA_258	BIO	PETQR1PKDSNGCG	RBI	T356	CDK4
PA_259	Key_L1_248 - sample + KaB + ATP A + pept. PA_259	BIO	EFTSR1PKDSPPGCG	RPS9KA1 (RSK1)	T359	MAPK1 (ERK2)
PA_260	Key_L1_249 - sample + KaB + ATP A + pept. PA_260	BIO	MEQMDMDIMSPGCG	NPM1	S4 (N-term end)	PLK1
PA_261	Key_L1_250 - sample + KaB + ATP A + pept. PA_261	BIO	ISSVP1P3PLGGCG	CSNA2A1 (CK2A1)	T360	CDK1 (CDC2)
PA_262	Key_L1_251 - sample + KaB + ATP A + pept. PA_262	BIO	KEDP11DEPEGGCG	DOK1	Y352	INSR
PA_263	Key_L1_252 - sample + KaB + ATP B + pept. PA_263	CON +	CHPRL3RRP3TRK	consensus - CON +	-	AKT1, AKT2, AKT3, AURKA, CAMK1beta, CAMK1delta, CAMK1gamma, CAMK2, CAMK4, IKKKB (IKK beta), IRAK4, KSR1, KSR1(A335F), KSR1(L336F), KSR2, KSR2(R679S), PAK2, PRKACA (PKA C-alpha), PRKACB (PKA C- beta), PRKACG (PKA C-gamma), PRKCA (PKC alpha), PRKCB (PKC beta), PRKCE (PKC epsilon), PRKCD (PKC delta), PRKCE (PKC epsilon), PRKCI (PKC iota), PRKCK (PKC kappa), PRKCK (PKC kappa), PRKCK (PKC kappa), PRKCK (PKC kappa), PRKCK (PKC kappa), RPS9KA5 (MSK1)
PA_264	Key_L1_253 - sample + KaB + ATP A + pept. PA_264	BIO	AHRKAG3SNERGCG	FOS	S522	PRKACA (PKA C-alpha), RPS9KA1 (RSK1), RPS9KA5 (MSK1)
PA_265	Key_L1_254 - sample + KaB + ATP A + pept. PA_265	BIO	SFAPS1PLTGRGCG	RBL1	T369	CDK4
PA_266	Key_L1_255 - sample + KaB + ATP E + pept. PA_266	CON - random	NEVERL4MD2GCG	-	-	-
PA_266	Key_L1_256 - sample + KaB + ATP E + pept. PA_266	CON - random - Rep (lib5-8)_a3	NEVERL4MD2GCG	-	-	-
PA_266	Key_L2_011 - sample + KaB + ATP E + pept. PA_266	CON - random - Rep (lib5-8)_b3	NEVERL4MD2GCG	-	-	-
PA_267	Key_L1_257 - sample + KaB + ATP A + pept. PA_267	BIO	DINSLVDSRMGCG	PLCG2	Y753	FYN, HOK, LOK, LYNA, LYNB
PA_268	Key_L1_258 - sample + KaB + ATP A + pept. PA_268	BIO	CGSSD3LSSFTLLAL	FOS	S374	MAPK1 (ERK2), MAPK3 (JNK2)
PA_269	Key_L1_259 - sample + KaB + ATP A + pept. PA_269	BIO	KEVSKVSDIQRGCG	PDGFRA	Y754	PDGFRA
PA_270	Key_L1_260 - sample + KaB + ATP A + pept. PA_270	BIO	CKEAT5TFNTITRG	ITGB3	T779 (C-term)	AKT1, PDK1, PDK1
PA_271	Key_L1_261 - sample + KaB + ATP B + pept. PA_271	CON +	IFTPTTITTYFRKKGCG	consensus - CON +	-	MAPK1 (ERK2), MAPK3 (JNK2), MAPK10 (JNK3), MAPK11 (p38b), MAPK12 (p38g), MAPK13 (p38c), MAPK14 (p38a)
PA_272	Key_L1_262 - sample + KaB + ATP A + pept. PA_272	BIO	RPF7LSPHIGCG	RBI	S760	CDK4
PA_273	Key_L1_263 - sample + KaB + ATP A + pept. PA_273	BIO	ESDIDVYNNPEGCG	LAT	Y209	ZAP70
PA_274	Key_L1_264 - sample + KaB + ATP A + pept. PA_274	BIO	LITGVVYNNPEGCG	CTL44	Y201	FYN, JAK2, LOK, TYK
PA_275	Key_L1_265 - sample + KaB + ATP A + pept. PA_275	BIO	SMDAG3PMLSPGCG	SMA03	S204	CDK2, CDK4, MAPK1 (ERK2)
PA_276	Key_L1_266 - sample + KaB + ATP A + pept. PA_276	BIO	KLMFVEHCAGCG	TGFB2	S213	CLK1, TGFB2
PA_277	Key_L1_267 - sample + KaB + ATP B + pept. PA_277	CON + - Rep (lib5-8)_a2	ADELIPQGGCG	-	-	EGFR
PA_277	Key_L1_268 - sample + KaB + ATP B + pept. PA_277	CON + - Rep (lib5-8)_b2	ADELIPQGGCG	consensus - CON +	-	EGFR
PA_277	Key_L2_012 - sample + KaB + ATP B + pept. PA_277	CON + - Rep (lib5-8)_b2	ADELIPQGGCG	-	-	EGFR
PA_278	Key_L1_269 - sample + KaB + ATP A + pept. PA_278	BIO	TSLAQ3DSNKGCG	SKA	Y216	BTX, ITR, TEC

[illegible]

[illegible]

Table 1

PA_337	Key_L2_050 - sample + KdB + ATP D + pept_PA_337	CON - _mutates CONH	CON - _mutates CONH	consensus _ CON - _ ST to A/V	Y88	BRK1; BRK2; CHEK1; CHEK2; MARK1; MARK2 (EAK1); MARK3 (TAK1); TSK1; TSSK2
PA_338	Key_L2_051 - sample + KdB + ATP A + pept_PA_338	BIO	LAPPA:POPPASC	BCDL2.11	S98	MAPK3 (JNK1); MAPK3 (JNK2); MAPK10 (JNK3)
PA_339	Key_L2_052 - sample + KdB + ATP A + pept_PA_339	BIO	OGISF:QPTOPGC	CHEK1	S45	ATR
PA_340	Key_L2_053 - sample + KdB + ATP A + pept_PA_340	BIO	SPVTK:PPRDLOG	EIF4EBP1 (4EBP1)	T70	MAPK1 (ERK2); MTOR (FRAP1)
PA_341	Key_L2_054 - sample + KdB + ATP A + pept_PA_341	BIO	RSGLC:SPYVAGC	MYC	S71	MAPK3 (JNK1)
PA_342	Key_L2_055 - sample + KdB + ATP E + pept_PA_342	CON - _random	EFEEAALEENAGC	-	-	-
PA_343	Key_L2_056 - sample + KdB + ATP A + pept_PA_343	BIO	RLRL:LYPQTDGC	RAC1	S71	AKT1
PA_344	Key_L2_057 - sample + KdB + ATP A + pept_PA_344	BIO	LATRL:RRRVVSGC	ONKSR1	S32	NME1-NME2; NME1 (NM23, NDRK-A)
PA_345	Key_L2_058 - sample + KdB + ATP A + pept_PA_345	BIO	QTPVD:SPDSTGC	RPS9B1 (S91)	S384	CDK1 (CDC2); MTOR (FRAP1)
PA_346	Key_L2_059 - sample + KdB + ATP D + pept_PA_346	CON - _mutates CONH	AVFPAPALAH:MAHPHAEDEEGC	consensus _ CON - _ ST to A/V	-	GSK3A; GSK3B
PA_347	Key_L2_060 - sample + KdB + ATP A + pept_PA_347	BIO	PSGLL:TPQSGGC	CCNE1	T395	GSK3B
PA_348	Key_L2_061 - sample + KdB + ATP A + pept_PA_348	BIO	SETDD:AEIDGC	PTK2 (FAK1)	Y397	FGFR1; FGFR2; PTK2 (FAK; FAK1)
PA_349	Key_L2_062 - sample + KdB + ATP A + pept_PA_349	BIO	VKEEG:YELPYNGC	IDCK1	Y388	INSR; RET
PA_350	Key_L2_063 - sample + KdB + ATP A + pept_PA_350	BIO	HVEGL:VEGLPGC	GTF2I	Y388	BTIK
PA_351	Key_L2_064 - sample + KdB + ATP A + pept_PA_351	BIO	LTPPG:GKQSGC	CCNE1	T399	CDK2
PA_352	Key_L2_065 - sample + KdB + ATP 8 + pept_PA_352	CON +	KHKSPGEYVNEFGGC	consensus _ CON +	-	ALK; FGFR1; FLT1 (VEGFR1); IGFR1; JAK2; MET; RET; ROS1
PA_353	Key_L2_066 - sample + KdB + ATP A + pept_PA_353	BIO	TSRMM:PVVVTGC	MAPK8 (JNK1) / MAPK3 (JNK2) / MAPK10 (JNK3)	T155 (Y155); T181 (Y185) / T221 (Y223)	MAPK3 (JNK1); MAPK3 (JNK2); MAP2K4 (MEK4); MKK4; MAPK3 (JNK1); MAP2K7 (MEK7); MKK7; MAPK7 (JNK2)
PA_354	Key_L2_067 - sample + KdB + ATP A + pept_PA_354	BIO	EFISL:SPHEAGC	IE2F1	S403	CDK7
PA_355	Key_L2_068 - sample + KdB + ATP A + pept_PA_355	BIO; rsSNP	KSPVW:GD:SPVHLSN5C	MAPT (TAU) _ R723W variant (ALZJFTD)	S717; T720; S721	GSK3B; CDK5
PA_356	Key_L2_069 - sample + KdB + ATP A + pept_PA_356	BIO	DEEDT:TMPSTGC	PTK2 (FAK1)	Y407	PTK2B (FAK2; PAK2); SRC
PA_357	Key_L2_070 - sample + KdB + ATP A + pept_PA_357	BIO	VFLGF:YVAFSGC	RPS9B1 (S91)	T412	PDHK1
PA_358	Key_L2_071 - sample + KdB + ATP A + pept_PA_358	BIO	CGEAF:LFGEYAPPTDS	SGK1	S422	NEK6; PDHK1
PA_359	Key_L2_072 - sample + KdB + ATP A + pept_PA_359	BIO	CGGSP:SRG:SV3	SMAD3	S422	MAP3K7 (MEK7; TAK1)
PA_360	Key_L2_073 - sample + KdB + ATP A + pept_PA_360	BIO	RNPGE:YVEANFGC	PLCG1	Y783	EGFR; HCK; LYNA; LYNE; RET; SRC; SYK
PA_361	Key_L2_074 - sample + KdB + ATP B + pept_PA_361	CON +	RRADSDSDDDGC	consensus _ CON +	-	CSNK2A1 (CK2A1); CSNK2A2 (CK2A2)
PA_362	Key_L2_075 - sample + KdB + ATP A + pept_PA_362	BIO	CGLAGAPPV:LDVL3	NTRK1 (TRKA)	Y791 (C-term end)	ERBB3 (in its dimeric state with another HER fam. member); ERBB4; NTRK1 (TRKA)
PA_363	Key_L2_076 - sample + KdB + ATP A + pept_PA_363	BIO	YKFFS:PLRIPGC	RBI	S785	CDK4
PA_364	Key_L2_077 - sample + KdB + ATP A + pept_PA_364	BIO	RIRGK:YVQCGC	TERT	S824	AKT1
PA_365	Key_L2_078 - sample + KdB + ATP A + pept_PA_365	BIO	TPTKM:PPSRDGC	RBI	T826	CDK4
PA_366	Key_L2_079 - sample + KdB + ATP A + pept_PA_366	BIO	YNRK:VLDSPVGC	NCOA3	S857	CHUK (IKK alpha); IKKMB (IKK beta)
PA_367	Key_L2_080 - sample + KdB + ATP A + pept_PA_367	BIO	MPDSN:ISKSGGC	PDGFRB	Y657	PDGFRB
PA_368	Key_L2_081 - sample + KdB + ATP D + pept_PA_368	CON - _mutates CONH	LGFGG:VAPGCG	consensus _ CON - _ ST to A/V	-	CHUK (IKK alpha); IKKMB (IKK beta); IRAK1; IRAK4; MAP3K6 (COT); MAP3K17 (TAOK2); PSK1; MAP3K2; MAP3K4 (HGK); MAP3K6 (MINK); STK3 (MST2, KRS1); STK4 (MST1, KRS2); MTOR (FRAP1); NEK1; NEK2; NEK8; NEK7; EIF2AK3 (PEK); PLK1
PA_369	Key_L2_082 - sample + KdB + ATP A + pept_PA_369	BIO	QPRRR:LPAGDGC	AKHGEF2	S886	PAK1
PA_370	Key_L2_083 - sample + KdB + ATP A + pept_PA_370	BIO	VERFE:VMEAGC	ILSST	T890	PRKCD (PKC delta); PRKCD
PA_371	Key_L2_084 - sample + KdB + ATP A + pept_PA_371	BIO	KSPGE:VNEFGC	IRS1	Y886	FGFR1; FGFR2; IGFR1

90

Table 1

PA_402	2A10	Key_L2_117 - sample + KaB + ATP A + pept_PA_402	BIO	FTNGG/FFHFGC	IL2RB	Y564	JAK1 PRKAA1 (AMPA3), PRKAA2 (AMPA32), PRKACA (PKA C-alpha), RPS9KA1 (RSK1, p90RSK), RPS9KB1 (SRK1, p70S6K, S6K-beta-1)
PA_403	2C10	Key_L2_118 - sample + KaB + ATP A + pept_PA_403	BIO	QVRLTSGSRPGC	EEF2K	S866	
PA_404	2EE10	Key_L2_119 - sample + KaB + ATP A + pept_PA_404	BIO	PIDQYTPVSSFGC	PAX6	T500, T504	HIK2
PA_405	2S10	Key_L2_120 - sample + KaB + ATP A + pept_PA_405	BIO	AFERA-VYTRSGC	CCR5	S535, S537	GRK3 (ADRBK2)
PA_406	2I10	Key_L2_121 - sample + KaB + ATP A + pept_PA_406	BIO	NEGFESYGSSEYGGC	RPA2	S6	PRKDC (DNA-PKcs, DNAPK)
PA_407	3N10	Key_L2_122 - sample + KaB + ATP B + pept_PA_407	CON +	CKNSRGDMYVQLG	consensus _ CON +	-	AXL, DDR1 (ATRKA, RTG), PTK3, DDR2 (NTRK3, TYRO10), DNAPK (DNK1), FGFR1, IGF1R, INSR, INSR (PR1, JAK2, RON, (MST1B), STK3 (MST2, NRS1), STK4 (MST1, NRS2), STK10 (LON), TYK2
PA_408	3M10	Key_L2_123 - sample + KaB + ATP A + pept_PA_408	BIO	RGRLPKKSRPGC	NR4A2 / NR4A3	S547 / S376	RPS9KA1 (RSK1, p90RSK)
PA_409	3O10	Key_L2_124 - sample + KaB + ATP A + pept_PA_409	BIO	IESSNVMAPYDGC	PDGFRB	Y771	PDGFRB
PA_410	2A12	Key_L2_125 - sample + KaB + ATP A + pept_PA_410	BIO	PPDHQVYNDFFGGC	SHC1 (p66Shc / p49Shc)	Y249, Y350	FLT3, SRC, SYK, ZAP70
PA_411	2C12	Key_L2_126 - sample + KaB + ATP A + pept_PA_411	BIO	KSKGISTENKGC	MAPT (TAU)	S579	CAMK2A, GSK3B, MARK1, PHKG1, PRKACA (PKA C-alpha), PRKGD (PKC delta, PKC delta)
PA_412	2EE12	Key_L2_127 - sample + KaB + ATP A + pept_PA_412	BIO	IEDEDYKASVTGC	PTK2B (FAK2)	Y579, Y580	PTK2B (FAK2, PAK2)
PA_413	2S12	Key_L2_128 - sample + KaB + ATP C + pept_PA_413	CON - _mutated BIO	KRRQVAMVDFGC	CDKN1A (p21, WAF1, CIP1) _ CON (-) _ SITTY to A/V/F	S146	AKT1, FIM1, FIM2, PRKCA (PKC alpha, PRKCA), NUKA1, LAT52
PA_414	2I12	Key_L2_129 - sample + KaB + ATP A + pept_PA_414	BIO	DFVARVPLQTRGC	BCCL2	T59, S70	MAPK8 (JNK1)
PA_415	3N12	Key_L2_130 - sample + KaB + ATP C + pept_PA_415	CON - _mutated BIO	KSPVAVAGVAPRLANGC	MAPT (TAU) _ CON (-) _ SIT to AV	S717, T720, S721	GSK3B, CDK5
PA_416	2M12	Key_L2_131 - sample + KaB + ATP A + pept_PA_416	BIO	KSPVAVAGVAPRLANGC	MAPT (TAU)	S717, T720, S721	GSK3B, CDK5
PA_417	2C14	Key_L2_132 - sample + KaB + ATP A + pept_PA_417	BIO	QSKRSTWVGTGCG	PAK1	T423	PDPR1
PA_418	2EE14	Key_L2_133 - sample + KaB + ATP A + pept_PA_418	BIO	CGSSNRRLCAKQKQ	STR11 (LKB)	S428 (C-term)	PRKACA (PKA C-alpha), RPS9KA3 (RSK2, p90RSK2, S6K-alpha3), RPS9KA5 (NRSK1), RPS9KB1 (SRK1, p70S6K, S6K-beta-1)
PA_419	2K14	Key_L2_134 - sample + KaB + ATP A + pept_PA_419	BIO	PERKSGSSSEDDGC	BRAF	S429	AKT1
PA_420	2M14	Key_L2_135 - sample + KaB + ATP A + pept_PA_420	BIO	MMNKGESYGSSEGGC	RPA2	S434	PRKDC (DNA-PKcs, DNAPK)
PA_421	2O14	Key_L2_136 - sample + KaB + ATP E + pept_PA_421	CON - _random	KINAPSIEMAPINGC	-	-	-
PA_422	3A16	Key_L2_137 - sample + KaB + ATP A + pept_PA_422	BIO	PKSYTYKDPGRGC	MYT1	S429	PLK1
PA_423	2C16	Key_L2_138 - sample + KaB + ATP A + pept_PA_423	BIO	EPKR3PRRPGC	RPS9KB1 (SRK1)	S434	CDK1 (CDC2)
PA_424	2EE16	Key_L2_139 - sample + KaB + ATP A + pept_PA_424	BIO	VLNAFVQAPSTGCG	SPF	S435	PRKDC (DNA-PKcs, DNAPK)
PA_425	3O16	Key_L2_140 - sample + KaB + ATP D + pept_PA_425	CON - _mutated CONH	LRRRAALGGC	consensus _ CON - _ SIT to AV	-	-
PA_426	2I16	Key_L2_141 - sample + KaB + ATP A + pept_PA_426	BIO	VAFQGLRQTRGC	IL10RA	Y446	JAK1
PA_427	2K16	Key_L2_142 - sample + KaB + ATP A + pept_PA_427	BIO	RPRSLVSPVTVGC	NEDD4L	S448	SGK1 (SGK)
PA_428	3M16	Key_L2_143 - sample + KaB + ATP A + pept_PA_428	BIO	DSDEDYKAVPLGC	SH3BP2	Y448	SYK
PA_429	2O16	Key_L2_144 - sample + KaB + ATP A + pept_PA_429	BIO	GAGGYQSGFGGCG	RPA2	T21	PRKDC (DNA-PKcs, DNAPK)
PA_430	2A18	Key_L2_145 - sample + KaB + ATP A + pept_PA_430	BIO	RSSDSVAVDLGC	FRS3	T455	NTRK2 (TRKB)
PA_431	2C18	Key_L2_146 - sample + KaB + ATP B + pept_PA_431	CON +	AVPPSPSLSRHSHPSHSEDEEEGCG	consensus _ CON +	-	GSK3A, GSK3B
PA_432	2EE18	Key_L2_147 - sample + KaB + ATP A + pept_PA_432	BIO	SSPRFVPSASDGC	MYOCD	S460	GSK3B
PA_433	2S18	Key_L2_148 - sample + KaB + ATP A + pept_PA_433	BIO	GSVEQVFKKFGGCG	CDKN1B (p27, KIP1)	T167	CDK2, PNK2 (PRK2)
PA_434	2I18	Key_L2_149 - sample + KaB + ATP A + pept_PA_434	BIO	CGGSPHNPISVVS	SIMAD1	S465 (C-term)	ACVR1 (ACVR1A, ALK2), TGFBF1 (ALK5)

PA_435	2K18	Key_L2_150 - sample + KaB + ATP B + pept. PA_435	CON +	LRRALGSGC	consensus _ CON +	-	AURKA, AURKB, AURKC, PRKACA (PKA C-alpha), PRKACB (PKA C-beta), PRKACG (PKA C-gamma), PRKGI (PRKG1), PRKG2 (PRKG2),
PA_436	3M18	Key_L2_151 - sample + KaB + ATP A + pept. PA_436	BIO	TGBRYKEPLGC	PRND1 (PKD1)	Y493	AEL1, SRC
PA_437	2O18	Key_L2_152 - sample + KaB + ATP A + pept. PA_437	BIO	GGTHFPQPSYSASG	AKT2	S474	PDPK1
PA_438	2a20	Key_L2_153 - sample + KaB + ATP A + pept. PA_438	BIO	REPLGTGGFGNGC	CHUK (IKK alpha)	T23	AKT1, AKT2
PA_439	3C30	Key_L2_154 - sample + KaB + ATP A + pept. PA_439	BIO	LVRHVAKISDGC	ZAP70	Y474	LCK
PA_440	3EE20	Key_L2_155 - sample + KaB + ATP A + pept. PA_440	BIO	GGEPSEYASVQVPR	SIRPA	Y496 (C-term)	JAK2
PA_441	2O30	Key_L2_156 - sample + KaB + ATP A + pept. PA_441	BIO	HPNDLVVEGLPGC	GTF2I	Y503	ITK
PA_442	2i00	Key_L2_157 - sample + KaB + ATP C + pept. PA_442	CON _ mutated BIO	LARMGARAPVSGC	EEZF1 _ CON (-) _ S1T to AV	S364	CHUK2
PA_443	3I30	Key_L2_158 - sample + KaB + ATP A + pept. PA_443	BIO	RTKSRVWAGEKGC	CDC35A	T507	PRK2 (PRK2)
PA_444	2M20	Key_L2_159 - sample + KaB + ATP A + pept. PA_444	BIO	ESRASITCGTPGC	PRKCD (PKCD)	T507	PDPK1
PA_445	2C20	Key_L2_160 - sample + KaB + ATP A + pept. PA_445	BIO	SSTSVTPVDVSGC	PTEN	T596	CSNK2A1 (CK2A1), CSNK2A2 (CK2A2)
PA_446	2E01	Key_L2_161 - sample + KaB + ATP A + pept. PA_446	BIO	KRRRLSSLRATSGC	RPS6	S235, S238	PAK2
PA_447	3O01	Key_L2_162 - sample + KaB + ATP A + pept. PA_447	BIO	KRKDFSSVDKKKGC	PTPN1	S242, S243	CLK1, CLK2
PA_448	2F01	Key_L2_163 - sample + KaB + ATP A + pept. PA_448	BIO	RPRSCVWPLPRGC	FOXO1 / FOXO4	T24 / T32	AKT1, AKT2
PA_449	2H01	Key_L2_164 - sample + KaB + ATP B + pept. PA_449	CON +	CCKRQEQIANRRRLSSLRATSGSGSQK	consensus _ CON +	-	PAK1, PAK2, PRKACB (PKA C-beta), PRKACG (PKA C-gamma), RPS6KA5 (RSK4, p90RSK4, S9K-alpha-6)
PA_450	3J01	Key_L2_165 - sample + KaB + ATP A + pept. PA_450	BIO	PRKRVYVVGNGFCG	LIHK1	T508	CDC-42BP4 (MRCK alpha), PAK1, ROCK1
PA_451	2L01	Key_L2_166 - sample + KaB + ATP A + pept. PA_451	BIO	QMKRRLIMEEKGCG	NF2	S518	PAK1
PA_452	2H01	Key_L2_167 - sample + KaB + ATP A + pept. PA_452	BIO	PYAGSDEIVTDGCG	XFOC1	T519	CSNK2A1 (CK2A1), CSNK2A2 (CK2A2)
PA_453	3P01	Key_L2_168 - sample + KaB + ATP A + pept. PA_453	BIO	KSYDEVVDVPSGCG	LIHK2	T526	CDC-42BP4 (MRCK alpha), ROCK1
PA_454	2E05	Key_L2_169 - sample + KaB + ATP A + pept. PA_454	BIO	SKRTHSGATGPRGCG	IRS1	S527	NBK5 (IKK beta)
PA_455	2C03	Key_L2_170 - sample + KaB + ATP E + pept. PA_455	CON _ random	PEP(GT)DERIDLEGCG	-	-	-
PA_456	2F03	Key_L2_171 - sample + KaB + ATP A + pept. PA_456	BIO	MPREEYQIDDPGCG	HSP90AA1	T5 (N-term)	PRKDC (DNA-PKcs, DNAPK)
PA_457	3H03	Key_L2_172 - sample + KaB + ATP A + pept. PA_457	BIO	GSRSRSPTSLFTGCG	MAPT (TAU)	T529	CDK5, GSK3B, PRKACA (PKA C-alpha), RPS9KA3 (RSK2, p90RSK2, S9K-alpha-3)
PA_458	2J03	Key_L2_173 - sample + KaB + ATP A + pept. PA_458	BIO	CGTATATGCTDPGEN	FYN	Y531 (C-term)	CSK
PA_459	2L03	Key_L2_174 - sample + KaB + ATP A + pept. PA_459	BIO	HSAGTSPITHGCG	IRS1	S531	IKBK5 (IKK beta), MAPK6 (JNK1)
PA_460	3M03	Key_L2_175 - sample + KaB + ATP A + pept. PA_460	BIO	LIHEDVQEPSGCG	TP53BP1	S25	ATM
PA_461	3P03	Key_L2_176 - sample + KaB + ATP A + pept. PA_461	BIO	SVPTFPPLGRLGCG	CSNK2A1 (CK2A1)	S382	CDK1 (CDK2)
PA_462	2E05	Key_L2_177 - sample + KaB + ATP A + pept. PA_462	BIO	GQSESYGNITGCG	PTPN6	Y536	AEL1, INSR, LCK
PA_463	2O35	Key_L2_176 - sample + KaB + ATP A + pept. PA_463	BIO	DEDDFSADMDGCG	RELA	S536	CHUK (IKK alpha)
PA_464	3P05	Key_L2_179 - sample + KaB + ATP A + pept. PA_464	BIO	KFMQASDILLKGC	NKX1	S44	NME1, NME2, NME1 (NM23, NDPK-A)
PA_465	3H05	Key_L2_180 - sample + KaB + ATP D + pept. PA_465	CON _ mutated CON +	CKRREILARRFAFK	consensus _ CON _ S1T to AV	-	AKT1, AKT2, AKT3, AURKA, CAAMK1beta, CAAMK1gamma, CAAMK2, CAAMK4, IKBK5 (IKK beta), IRAK4, IKK1, IKK2 (IKK3), IKK3 (IKK3), KSR1, KSR2 (KSR2), KSR3 (KSR3), KSR4 (KSR4), KSR5 (KSR5), KSR6 (KSR6), KSR7 (KSR7), KSR8 (KSR8), KSR9 (KSR9), KSR10 (KSR10), KSR11 (KSR11), KSR12 (KSR12), KSR13 (KSR13), KSR14 (KSR14), KSR15 (KSR15), KSR16 (KSR16), KSR17 (KSR17), KSR18 (KSR18), KSR19 (KSR19), KSR20 (KSR20), KSR21 (KSR21), KSR22 (KSR22), KSR23 (KSR23), KSR24 (KSR24), KSR25 (KSR25), KSR26 (KSR26), KSR27 (KSR27), KSR28 (KSR28), KSR29 (KSR29), KSR30 (KSR30), KSR31 (KSR31), KSR32 (KSR32), KSR33 (KSR33), KSR34 (KSR34), KSR35 (KSR35), KSR36 (KSR36), KSR37 (KSR37), KSR38 (KSR38), KSR39 (KSR39), KSR40 (KSR40), KSR41 (KSR41), KSR42 (KSR42), KSR43 (KSR43), KSR44 (KSR44), KSR45 (KSR45), KSR46 (KSR46), KSR47 (KSR47), KSR48 (KSR48), KSR49 (KSR49), KSR50 (KSR50), KSR51 (KSR51), KSR52 (KSR52), KSR53 (KSR53), KSR54 (KSR54), KSR55 (KSR55), KSR56 (KSR56), KSR57 (KSR57), KSR58 (KSR58), KSR59 (KSR59), KSR60 (KSR60), KSR61 (KSR61), KSR62 (KSR62), KSR63 (KSR63), KSR64 (KSR64), KSR65 (KSR65), KSR66 (KSR66), KSR67 (KSR67), KSR68 (KSR68), KSR69 (KSR69), KSR70 (KSR70), KSR71 (KSR71), KSR72 (KSR72), KSR73 (KSR73), KSR74 (KSR74), KSR7

[illegible]

[illegible]

Table 1

PA_536	2809	Key_L2_292 - sample + K ₂ B + ATP A + pept. PA_538	BIO	ESDGG:MDMSKGC	PDGFRB	Y740	PDGFR1, FGFR2, PDGFRB
PA_539	2008	Key_L2_293 - sample + K ₂ B + ATP A + pept. PA_539	BIO	VDAAV:PEERHGC	APP	T744	CDK5, GSK3B
PA_540	2009	Key_L2_294 - sample + K ₂ B + ATP A + pept. PA_540	BIO	MFRSN:SGNLHGC	TLK1	S743	CHUK1
PA_541	2008	Key_L2_295 - sample + K ₂ B + ATP A + pept. PA_541	BIO	RYRDV:PFDSHGC	PTPN1	S60	AKT1, CLK1, CLK2
PA_542	3006	Key_L2_296 - sample + K ₂ B + ATP C + pept. PA_542	CON - _mutated BIO	ELDUGALCVAFGC	IKKKB (IKK beta)	S177	CHUK (IKK alpha), MAP3K14 (NIK), MAP3K7 (MEK1/2, TAK1), PPACQ (PKC beta), PRKCZ (PKC zeta)
PA_543	2602	Key_L2_297 - sample + K ₂ B + ATP A + pept. PA_543	BIO	LLSEL:RRRRHGC	IEF2S1	S62	EIF2AK1, EIF2AK2, EIF2AK3 (PEK), PRKCD (PKC delta, PKCdelta), PRKRIR (THAP12, DAF4, F52IPK)
PA_544	2008	Key_L2_298 - sample + K ₂ B + ATP A + pept. PA_544	BIO	PSEVP:TPRRHGC	HMGAI	T53	CDK1 (CDC2)
PA_545	2008	Key_L2_299 - sample + K ₂ B + ATP A + pept. PA_545	BIO	QVSR:GGAGGCG	DES	S60	AURKB
PA_546	3003	Key_L2_300 - sample + K ₂ B + ATP A + pept. PA_546	BIO	PTPL:SPRRSGC	MYC	S62 (T36)	MAPK3 (JNK1)
PA_547	3006	Key_L2_301 - sample + K ₂ B + ATP A + pept. PA_547	BIO	LARRP:YRKILGC	ATP1	S63	CAMK1, CAMK2, PRKACA (PKA C-alpha), RPS6KA4 (MSK2, S6K-alpha-4), RPS6KA5 (MSK1)
PA_548	2810	Key_L2_292 - sample + K ₂ B + ATP A + pept. PA_548	BIO	MSETVPPAPAGGC	HIST1H1A	S2 or T5 (N-term and)	CDK1 (CDC2)
PA_550	3010	Key_L2_302 - sample + K ₂ B + ATP A + pept. PA_550	BIO	NIHSG:DSVESGC	INME1	S120	NME1-NME2, NME1 (NM23, NDKP(A))
PA_551	3010	Key_L2_304 - sample + K ₂ B + ATP A + pept. PA_551	BIO	EFNPN:SPANSGGC	LUBE2A	S120	CDK1 (CDC2), CDK2
PA_552	2010	Key_L2_305 - sample + K ₂ B + ATP A + pept. PA_552	BIO	SPMET:GCAPAGC	CCNB1	S133	PLK1
PA_553	2812	Key_L2_306 - sample + K ₂ B + ATP A + pept. PA_553	BIO	LTDR:LAIVHGC	CCOR2	Y139	JAK2
PA_554	3012	Key_L2_307 - sample + K ₂ B + ATP A + pept. PA_554	BIO	PKGHE:TNKNGC	PTPN11	Y546	PDGFRB
PA_555	2012	Key_L2_298 - sample + K ₂ B + ATP B + pept. PA_555	CON +	KRRRL:SLRAGC	consensus _ CON+	-	CDK4/28PA (MROCK alpha), CDC42BPB (MROCK beta), CHUK (IKK alpha), IKKB (IKK beta), CLK2, FAK2, PIM1, PIM2, PIM3, PRKG1, (PKG1), PRKG2 (PKG2), ROCK1, ROCK2, RPS6KA1 (RSK1, p90RSK), RPS6KA2 (RSK2, p90RSK2, S6K-alpha-2), RPS6KA3 (RSK2, p90RSK2, S6K-alpha-3), RPS6KA4 (MSK2, S6K-alpha-4), RPS6KA5 (MSK1), RPS6KA6 (RSK4, p90RSK3, S6K-alpha-6), RPS6KB1 (S6K1, p70S6K, S6K-beta-1)
PA_556	3012	Key_L2_299 - sample + K ₂ B + ATP A + pept. PA_556	BIO	KILRN:PHLEKGC	NSUN2	S139	AURKB
PA_557	3012	Key_L2_270 - sample + K ₂ B + ATP A + pept. PA_557	BIO	GEGET:VTGGC	TPT1	S64	PLK1
PA_558	2814	Key_L2_271 - sample + K ₂ B + ATP A + pept. PA_558	BIO	RSGOR:ASVLWGC	IPAK1	T55	AKT1, IRAK1
PA_559	3014	Key_L2_272 - sample + K ₂ B + ATP A + pept. PA_559	BIO	QEDND:INASLGC	PTPN1	S66	EGFR, INSR
PA_560	2014	Key_L2_273 - sample + K ₂ B + ATP A + pept. PA_560	BIO	RSARL:ZAPPGC	HMGNI	S25	PRKACA (PKA C-alpha), PRKCA (PKC alpha, PKCA), RPS6KA3 (RSK2, p90RSK2, S6K-alpha-3)
PA_561	2014	Key_L2_274 - sample + K ₂ B + ATP A + pept. PA_561	BIO	TGSVD:LAIDFGC	GAB2	Y614	ZAP70
PA_563	3014	Key_L2_275 - sample + K ₂ B + ATP A + pept. PA_563	BIO	CGNACT:LT:PRLPV	RAF1 (RAF)	S642 (C-term)	MAPK1 (ERK2)
PA_564	3013	Key_L2_276 - sample + K ₂ B + ATP C + pept. PA_564	CON - _mutated BIO	CGCKKAVCAQAEF	H2AFX _ CON (-) _ S/T to AV	S149	ATM
PA_565	3014	Key_L2_277 - sample + K ₂ B + ATP A + pept. PA_565	BIO	STNAD:QSSSDGC	DCCLRE1C	S645	ATM
PA_566	2816	Key_L2_278 - sample + K ₂ B + ATP A + pept. PA_566	BIO	DERVD:VVDGGC	GAB1	Y659	EGFR, INSR, NET
PA_567	2016	Key_L2_279 - sample + K ₂ B + ATP A + pept. PA_567	BIO	EPKRR:ARLSAGC	HMGNI	S21	PRKACA (PKA C-alpha), PRKCA (PKC alpha, PKCA), RPS6KA3 (RSK2, p90RSK2, S6K-alpha-3)
PA_568	3016	Key_L2_280 - sample + K ₂ B + ATP A + pept. PA_568	BIO	DFGLG:SLPAGC	NFYBIE (IKSB)	S23	CHUK (IKK alpha), IKKB (IKK beta)
PA_569	2013	Key_L2_281 - sample + K ₂ B + ATP D + pept. PA_569	CON - _mutated CON	CONKK:KRLDORHQAOLDAMKDEE	consensus _ CON - _ S/T to AV	-	CHUK (IKK alpha), IKKB (IKK beta)
PA_570	3016	Key_L2_282 - sample + K ₂ B + ATP A + pept. PA_570	BIO	PELL:SPRSKEGC	STMN1	S25	CDK1 (CDC2), MAPK3 (ERK1)

Table 1

PA_571	2L19	Key_L2_293-sample + KaB + ATP A + pept_PA_571	BIO	PGADLQVKKMGCC	RBBP9	S654	ATM
PA_572	2P16	Key_L2_294-sample + KaB + ATP A + pept_PA_572	BIO	EEDTE:MTFSGGC	CBL	Y700	EGFR, FGFR1, FGFR2, STK10 (LOK)
PA_573	3B18	Key_L2_293-sample + KaB + ATP A + pept_PA_573	BIO	NLLPM:PEEFDGC	STAT1	S727	CAMK3G, PRKCD (PKC delta, PKCD), RPS9KA5 (MSK1)
PA_574	2D18	Key_L2_298-sample + KaB + ATP A + pept_PA_574	BIO	GDRTS:FGGTGSC	PKM1 (PRK1) / PKM2 (PRK2)	T774 / T816	POPK1, POK1
PA_575	2F18	Key_L2_297-sample + KaB + ATP A + pept_PA_575	BIO	RYHKR:GSAVWNGGC	IRCEINP	T892, S883, S864	AURKB
PA_576	3M18	Key_L2_299-sample + KaB + ATP D + pept_PA_576	CON - _mutated CON+	KRPRAAFAEGGC	consensus _ CON - _ ST to AV	--	AKT1, AKT2, PAK4, PAK7, SGK1 (SGK), SGK2
PA_577	3L18	Key_L2_289-sample + KaB + ATP A + pept_PA_577	BIO	VQTPP:KPGGGGCG	CDC20	S72	BLUB1
PA_578	2L18	Key_L2_290-sample + KaB + ATP A + pept_PA_578	BIO	YSRAL:SRLLSSGC	HSPB1	S78	MAPKAPK2 (MK2), MAPKAPK3 (MK3)
PA_579	2M18	Key_L2_291-sample + KaB + ATP A + pept_PA_579	BIO	PSPPP:PSQQGCG	SGK1	S78	MAPK7 (ERK5)
PA_580	2F18	Key_L2_292-sample + KaB + ATP C + pept_PA_580	CON - _mutated BIO	CGLMF:VGFQDAD	TF53 _ CON (+) _ ST to AV	S932	CDK2, CDK7, CDK9, CSNK2A1 (CK2A1); EIF2AK2, MAPK14 (p38); NLUK1, STK11 (LKB)
PA_581	2B20	Key_L2_293-sample + KaB + ATP A + pept_PA_581	BIO	LSROL:SGVSEGC	HSPB1	S82	MAPKAPK2 (MK2), MAPKAPK3 (MK3)
PA_582	2L20	Key_L2_294-sample + KaB + ATP A + pept_PA_582	BIO	MKAL:SPVRGCEGC	ID3	S5	CDK2
PA_583	3F20	Key_L2_293-sample + KaB + ATP A + pept_PA_583	BIO	IPGVTS:PSSEGC	EIF4EBP1 (4EBP1)	S83	MAPK1 (ERK2), MTOR (FRAP1)
PA_584	2H20	Key_L2_296-sample + KaB + ATP A + pept_PA_584	BIO	RGRGS:VGGGGGCG	MAP3K5 (ASK1)	S83	AKT1
PA_585	2J20	Key_L2_297-sample + KaB + ATP A + pept_PA_585	BIO	L:SGV:SEIRHTGC	HSPB1	S88	PRK2 (PRK2)
PA_586	2L20	Key_L2_296-sample + KaB + ATP A + pept_PA_586	BIO	GHI:TTT:PTQSC	JUN	T81	MAPK6 (JNK1), MAPK9 (JNK2)
PA_587	1H04	Key_L1_300-sample + KaB + ATP A + pept_PA_587	BIO _ Rep (B5-B)_e5	RGRLP:KPKQPGCG	NR4A1	S951	AKT1, RPS9KA1 (RSK1, p60RSK), RPS9KA3 (RSK2, p60RSK2, SSK-alpha-3), RPS9KA5 (MSK1)
PA_587	3H04	Key_L2_299-sample + KaB + ATP A + pept_PA_587	BIO _ Rep (B5-B)_j6	RGRLP:KPKQPGCG	NR4A1	S951	AKT1, RPS9KA1 (RSK1, p60RSK), RPS9KA3 (RSK2, p60RSK2, SSK-alpha-3), RPS9KA5 (MSK1)
PA_588	2P20	Key_L1_301-sample + KaB + ATP A + pept_PA_588	BIO	RGRLP:KPKQPGCG	NR4A1	S951	AKT1, RPS9KA1 (RSK1, p60RSK), RPS9KA3 (RSK2, p60RSK2, SSK-alpha-3), RPS9KA5 (MSK1)
PA_589	1C13	Key_L1_301-sample + KaB + ATP A + pept_PA_589	BIO _ nsPV	SEGLP:PTNMVGC	R81	T821	CDK2
PA_590	1EE13	Key_L1_302-sample + KaB + ATP A + pept_PA_590	BIO _ nsPV	DFGLA:71:KSRWGC	BRAF	T999 _ p.V500K _ COSMIC prevalence: 472 x	PRK12 (PRK2)
PA_591	1L04	Key_L1_303-sample + KaB + ATP A + pept_PA_591	BIO _ nsPV	DFGLA:71:KSRWGC	BRAF	T999 _ p.K301E _ COSMIC prevalence: 77 x	PRK2 (PRK2)
PA_592	1F12	Key_L1_304-sample + KaB + ATP C + pept_PA_592	CON - _mutated BIO	DFGLA:V5:SRWGC	BRAF _ CON (+) _ T to G	T999	PRK2 (PRK2)
PA_593	1G04	Key_L1_305-sample + KaB + ATP A + pept_PA_593	BIO _ nsPV	DFGLA:V5:SRWGC	CDKN1A (p21, WAF1, CIP1)	S142 (T145) _ p.R143W _ COSMIC prevalence: 2x	AKT1, PIM1, PIM2, PRKCA (PKC alpha, PKC A), NLUK1, LAT52
PA_594	3L17	Key_L2_302-sample + KaB + ATP A + pept_PA_594	BIO _ nsPV	KRP:Q1:MTDFYGC	FOXO1	T24 _ p.R21C _ COSMIC prevalence: 1x	AKT1
PA_595	2P17	Key_L2_303-sample + KaB + ATP A + pept_PA_595	BIO	RP:SC:WLP:RGCG	TP53	T155	CSNK2A1 (CK2A1)
PA_596	2F10	Key_L2_304-sample + KaB + ATP A + pept_PA_596	BIO _ nsPV	TPPPG:TRVRAMGC	TP53	T155 _ p.V157F _ COSMIC prevalence: 164 x	CSNK2A1 (CK2A1)
PA_597	3J24	Key_L2_305-sample + KaB + ATP A + pept_PA_597	BIO _ nsPV	TPPPG:TRVRAMGC	TP53	T155 _ p.R159F _ COSMIC prevalence: 24 x	CSNK2A1 (CK2A1)
PA_598	3D10	Key_L2_306-sample + KaB + ATP A + pept_PA_598	BIO _ nsPV	TPPPG:7:VAMGC	TP53	T55 _ p.E58K _ COSMIC prevalence: 3 x	MAPK1 (ERK2)
PA_599	2K19	Key_L2_307-sample + KaB + ATP A + pept_PA_599	BIO	IEQNF:KDP:SGCG	CTNNB1	T41	GSN3B
PA_600	1K19	Key_L1_306-sample + KaB + ATP A + pept_PA_600	BIO _ nsPV	HSAGATTAPSLGC	CTNNB1	T41 _ p.T41A _ COSMIC prevalence: 828 x	GSN3B
PA_601	3D04	Key_L2_308-sample + KaB + ATP A + pept_PA_601	BIO _ nsPV	HSAGATTAP:LCG	CTNNB1	T41 _ p.S46F _ COSMIC prevalence: 518 x	GSN3B

WHAT IS CLAIMED IS:

- 1 1. A method of determining the phosphorylation activity profile of a
2 sample comprising one or more kinases, the method comprising:
3 incubating the sample with a panel of sensor peptides comprising a diversity
4 of biological sensor peptides, wherein each biological sensor peptide comprises a substrate
5 region phosphorylated by a kinase, and the diversity of biological sensor peptides comprises
6 biological sensor peptides for different kinases; and further, wherein members of the panel of
7 peptides are distributed into separate reaction mixtures to assess phosphorylation activity,
8 such that each reaction mixture represents one biological sensor peptide; and
9 measuring phosphorylation activity for each peptide in the separate reaction
10 mixtures, thereby determining the phosphorylation activity profile of the sample.
- 1 2. The method of claim 1, further comprising a plurality of mutated
2 control sensor peptides.
- 1 3. The method of claim 1 or 2, wherein the step of measuring
2 phosphorylation activity comprises detecting ATP consumption.
- 1 4. The method of claim 1, 2, or 3, further comprising normalizing the
2 phosphorylation activity measured in each reaction mixture and assigning the activity to a
3 subfamily or to a kinase based on the pattern of phosphorylation activity.
- 1 5. The method of any one of claims 1 to 4, wherein the panel comprises
2 biological sensor peptides for kinases that are members of at least two different kinase
3 subfamilies.
- 1 6. The method of any one of claims 1 to 4, wherein the panel of sensor
2 peptides comprises biological sensor peptides for kinases that are members of the ABL,
3 AKT, HER, MAPK and SFK kinase subfamilies.
- 1 7. The method of claim 6, wherein the panel of sensor peptides comprise
2 biological sensor peptides for kinases that are members of the ABL1, AKT1, EGFR,
3 MAPK1/ERK2, MAPK14/p38a, HCK, and SRC subfamilies.
- 1 8. The method of claim 6, wherein the panel of sensor peptides
2 comprises biological sensor peptides for BLK, BRK, FGR, FRK, HCK, LCK, LYN A,

3 SRMS, YES1, ABL1T315I, EGFR, JAK2, CSK, AKT1, AKT2, AKT3, MAPK1/ERK2, and
4 MAPK14/p38a kinases.

1 9. The method of any one of claims 1 to 8, wherein the panel comprises
2 at least three biological sensor peptides for each kinase.

1 10. The method of claim 9, wherein the panel comprises 4, 5, 6, 7, 8, 9, or
2 10 biological sensor peptides for each kinase.

1 11. The method of any one of claims 1 to 10, wherein the individual sensor
2 peptides are 9 to 21 amino acids in length.

1 12. The method of claim 11, wherein the individual sensor peptides are
2 from 9 to 15 amino acids in length.

1 13. The method of claim 12, wherein the individual sensor peptides are 11
2 amino acids in length.

1 14. The method of any one of claims 1 to 13, wherein the panel comprises
2 at least 10, at least 20, at least 30, at least 40, at least 50, at least 60, at least 70, at least 80, or
3 at least 90 biological sensor peptides having a sequence as shown in Table 1 for the peptide
4 id of a biological sensor peptide shown in Fig. 4.

1 15. The method of claim 14, wherein the panel comprises at least 100
2 sensor peptides, or at least 110, at least 120, at least 130, at least 140, or 151 biological sensor
3 peptides having a sequence as shown in Table 1 for the peptide id of a biological sensor
4 peptide shown in Fig. 4.

1 16. The method of claim 15, wherein the panel comprises at least 10, at
2 least 20, at least 30, at least 40, at least 50, at least 60, or 63 reference peptides as identified
3 in Fig. 4.

1 17. The method of claim 15, wherein the panel comprises the 228 peptides
2 as identified in Fig. 4.

1 18. The method of any one of claims 1 to 17, wherein the sample
2 comprises cancer cells.

1 19. The method of claim 18, wherein the cancer cells are from a solid
2 tumor.

1 20. The method of claim 19, wherein the solid tumor is a breast tumor, a
2 colorectal tumor, a melanoma, or a lung tumor.

1 21. The method of claim 19 or 20, wherein the sample comprises
2 metastatic cancer cells.

1 22. The method of claim 18, wherein the cancer cells are from a
2 hematological cancer.

1 23. The method of any one of claims 18 to 22, wherein the cancer cells are
2 from a patient that has been treated with a cancer therapeutic agent.

1 24. The method of claim 23, wherein the therapeutic agent is a kinase
2 inhibitor.

1 25. The method of any one of claims 18 to 22, wherein the cancer cells are
2 from a patient that has not been treated with a cancer therapeutic agent.

1 26. A panel comprising a plurality of biological sensor peptides to evaluate
2 phosphorylation activity of two or more different kinases.

1 27. The panel of claim 26, further comprising a plurality of control
2 mutated sensor peptides.

1 28. The panel of claim 26 or 27, wherein the panel comprises 4, 5, 6, 7, 8,
2 9, or 10 biological sensor peptide for each kinase.

1 29. The panel of any one of claims 26 to 28, wherein the individual
2 sensor peptides are 9 to 21 amino acids in length.

1 30. A panel of biological sensor peptides to evaluate phosphorylation
2 activity, wherein the panel comprises at least four biological sensor peptides for each of
3 kinase families ABL, AKT, HER, MAPK and SFK.

1 31. The panel of claim 30, wherein the panel of sensor peptides comprises
2 at least four biological sensor peptides for each of kinase subfamilies ABL1, AKT1, EGFR,
3 MAPK1/ERK2, MAPK14/p38a, HCK, and SRC.

1 32. The panel of claim 30, wherein the panel of sensor peptides comprises
2 at least three biological sensor peptides for each kinase BLK, BRK, FGR, FRK, HCK, LCK,
3 LYN A, SRMS, YES1, ABL1T315I, EGFR, JAK2, CSK, AKT1, AKT2, AKT3,
4 MAPK1/ERK2, and MAPK14/p38a.

1 33. The panel of claim 32, wherein the panel comprises 4, 5, 6, 7, 8, 9, or
2 10 biological sensor peptide for each kinase.

1 34. The panel of any one of claims 30 to 33, wherein the individual sensor
2 peptides are 9 to 21 amino acids in length.

1 35. The panel of claim 34, wherein the individual sensor peptides are from
2 9 to 15 amino acids in length.

1 36. The panel of claim 35, wherein the individual sensor peptides are 11
2 amino acids in length.

1 37. The panel of any one of claims 30 to 36, wherein the panel comprises
2 at least 10, at least 20, at least 30, at least 40, at least 50, at least 60, at least 70, at least 80, or
3 at least 90 biological sensor peptides having a sequence as shown in Table 1 for the peptide
4 id of a biological sensor peptide shown in Fig. 4.

1 38. The panel of claim 37, wherein the panel comprises at least 100 sensor
2 peptides, or at least 110, at least 120, at least 130, at least 140, or 151 biological sensor
3 peptides having a sequence as shown in Table 1 for the peptide id of a biological sensor
4 peptide shown in Fig. 4.

1 39. The panel of claim 38, wherein the panel comprises at least 10, at least
2 20, at least 30, at least 40, at least 50, at least 60, or 63 reference peptides as identified in Fig.
3 4.

1 40. The panel of claim 38, wherein the panel comprises the 228 peptides
2 shown in Fig. 4.

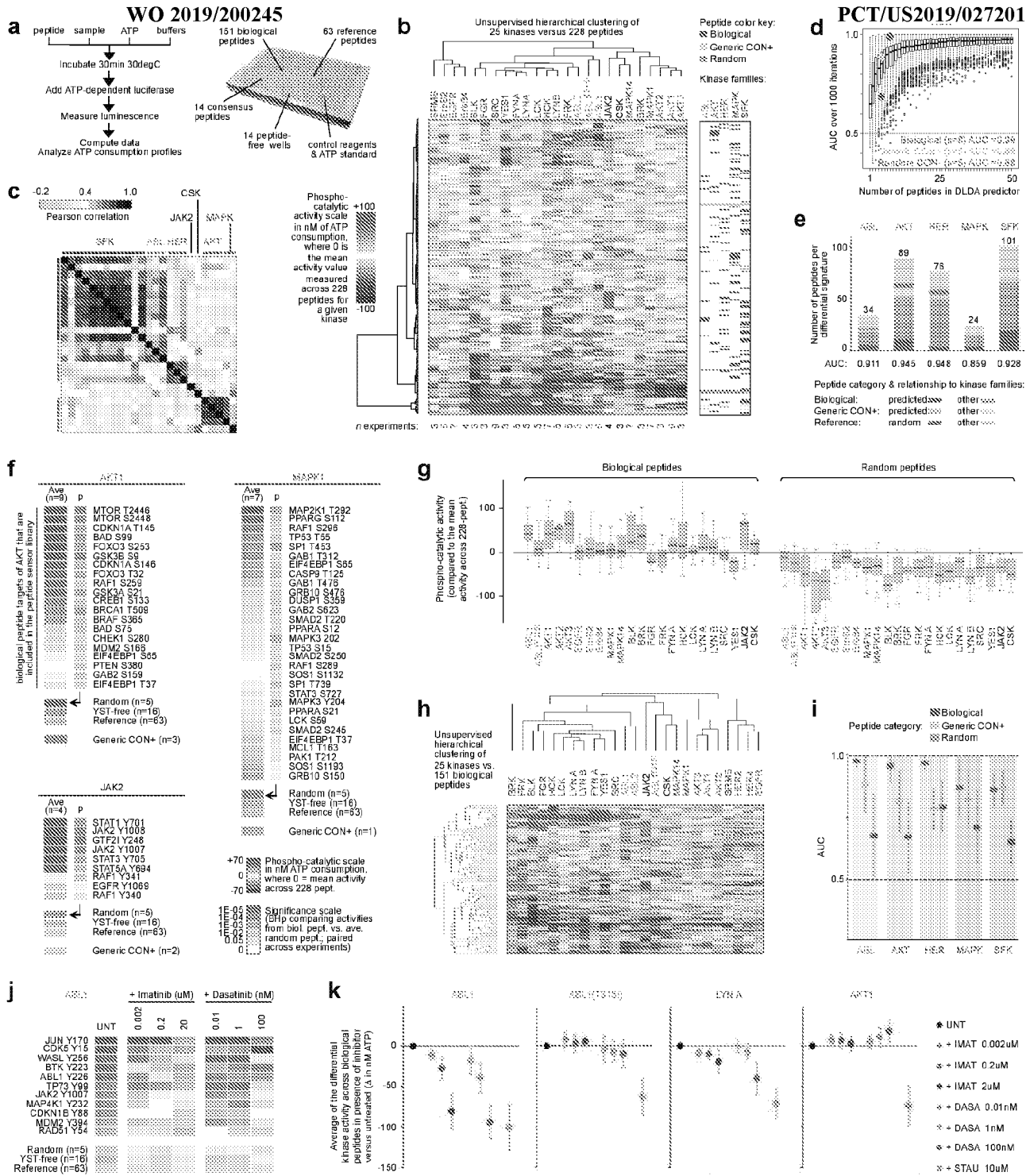


Fig. 1

WO 2019/200245

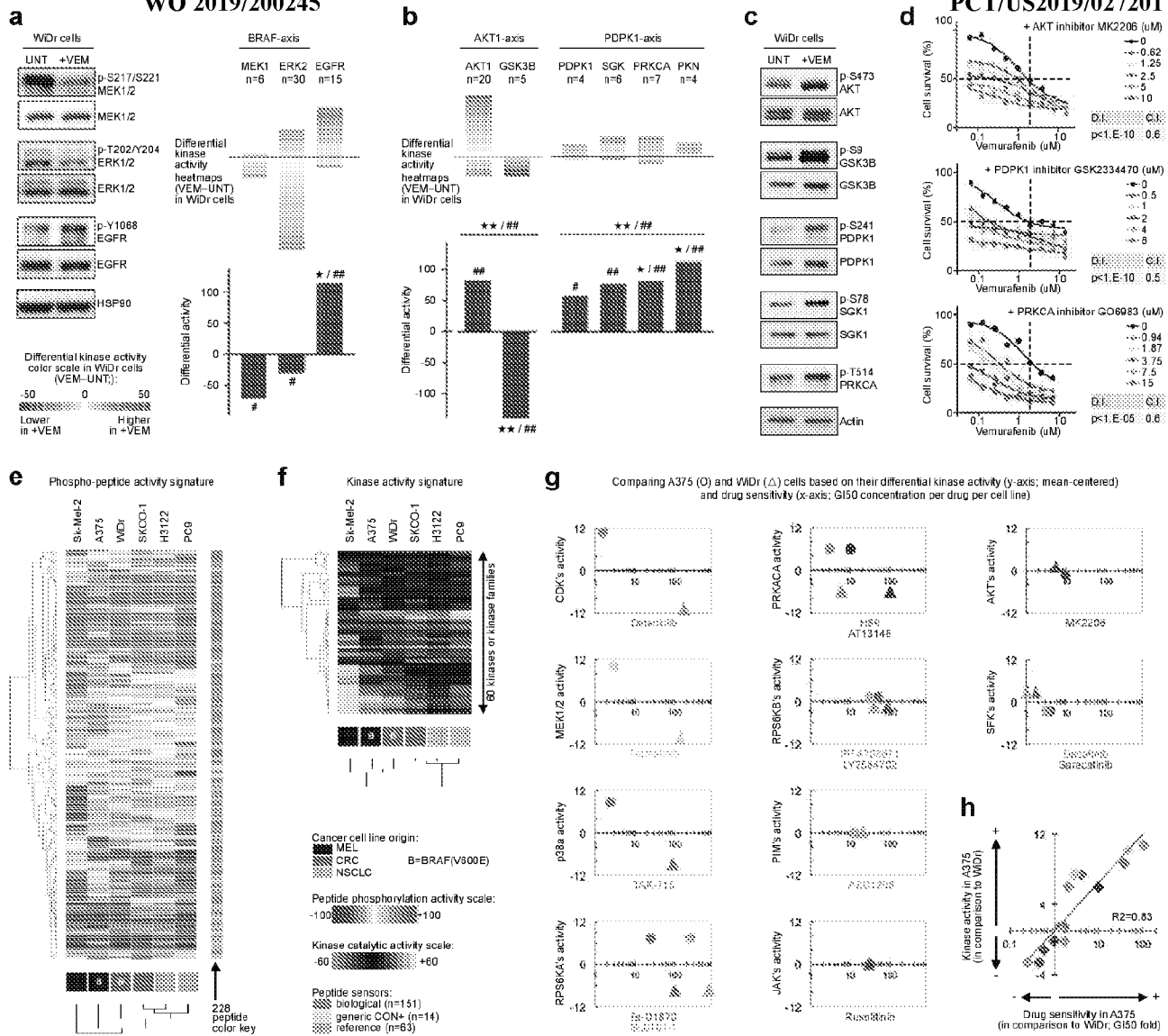


Fig. 2

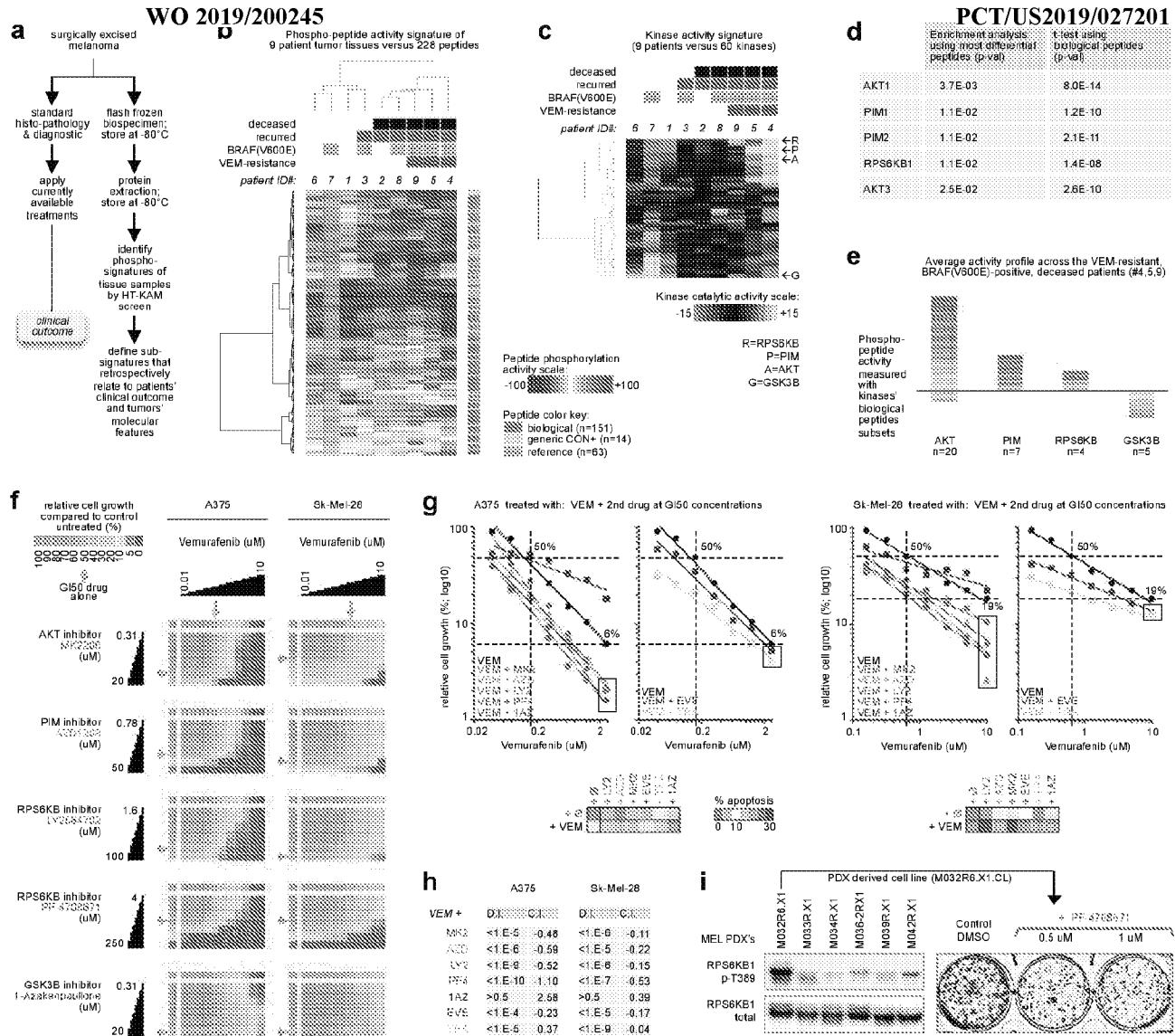


Fig. 3

WO 2019/200245

Tested recombinant kinase enzymes

Fig. 4
(part 1)

PCT/US2019/027201

[illegible]

Generic peptides

Reference peptides

Fig. 4
(part 2)

Biological peptides (n=151)

Generic peptides (n=14)

Reference peptides (n=63)

Section 2

Semi-automated liquid dispensing (Biomek FX)

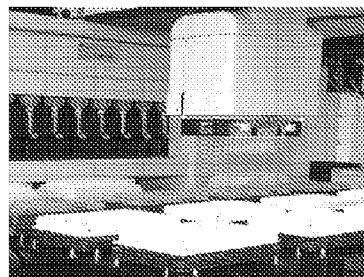


Fig. 4
(part 3)

WO 2019/200245

PCT/US2019/027201

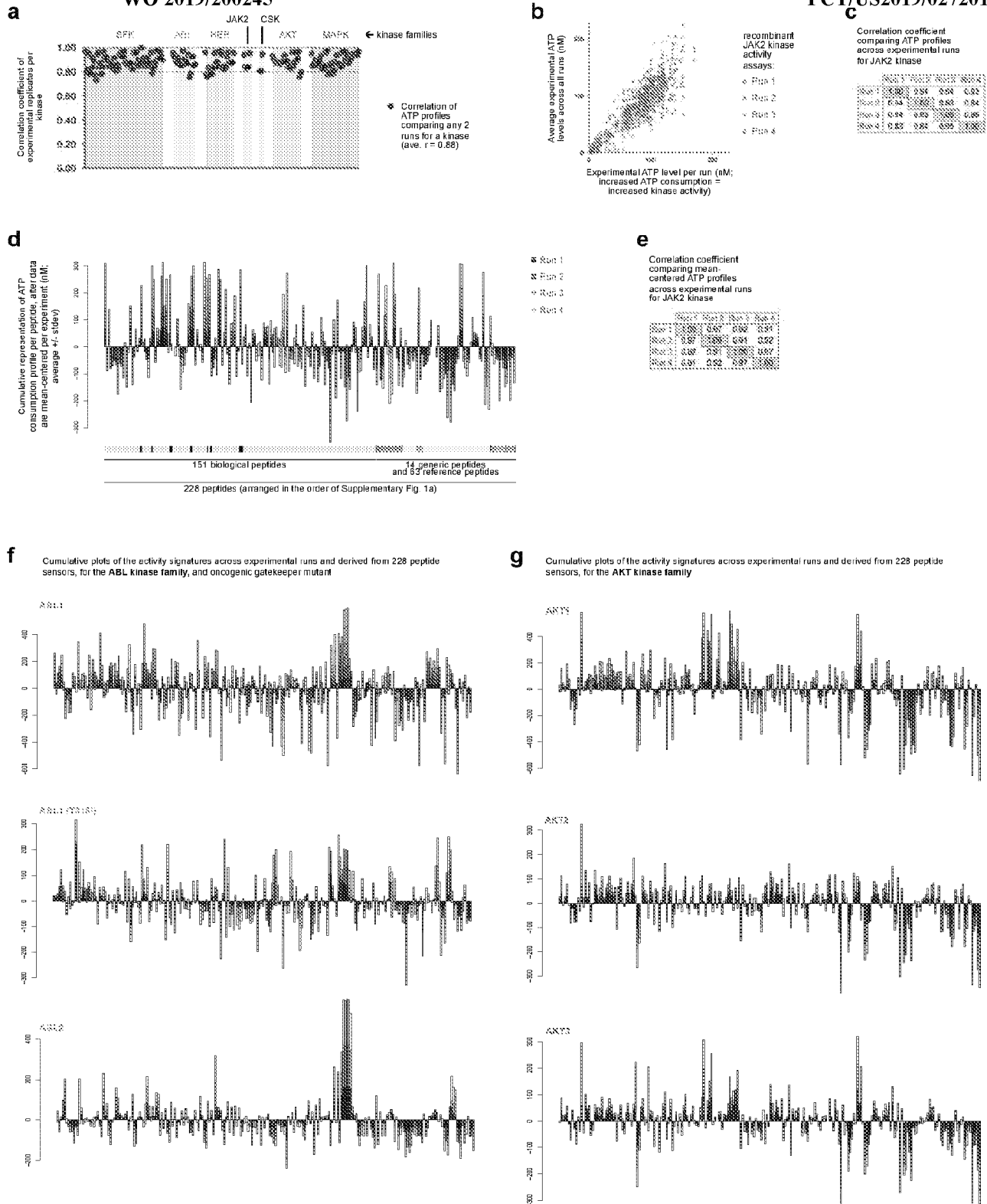
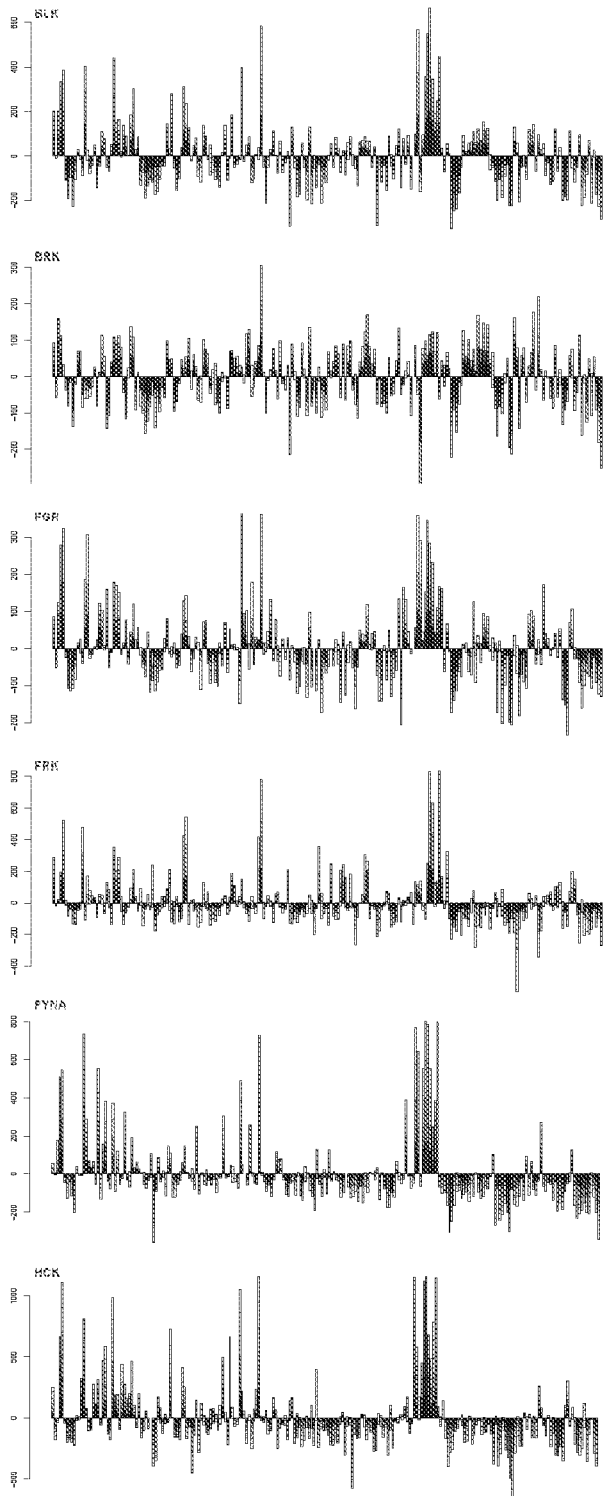


Fig. 5 (part 1)

WO 2019/200245

h

Cumulative plots of the activity signatures across experimental runs and derived from 228 peptide sensors, for the SRC kinase family, and isoforms



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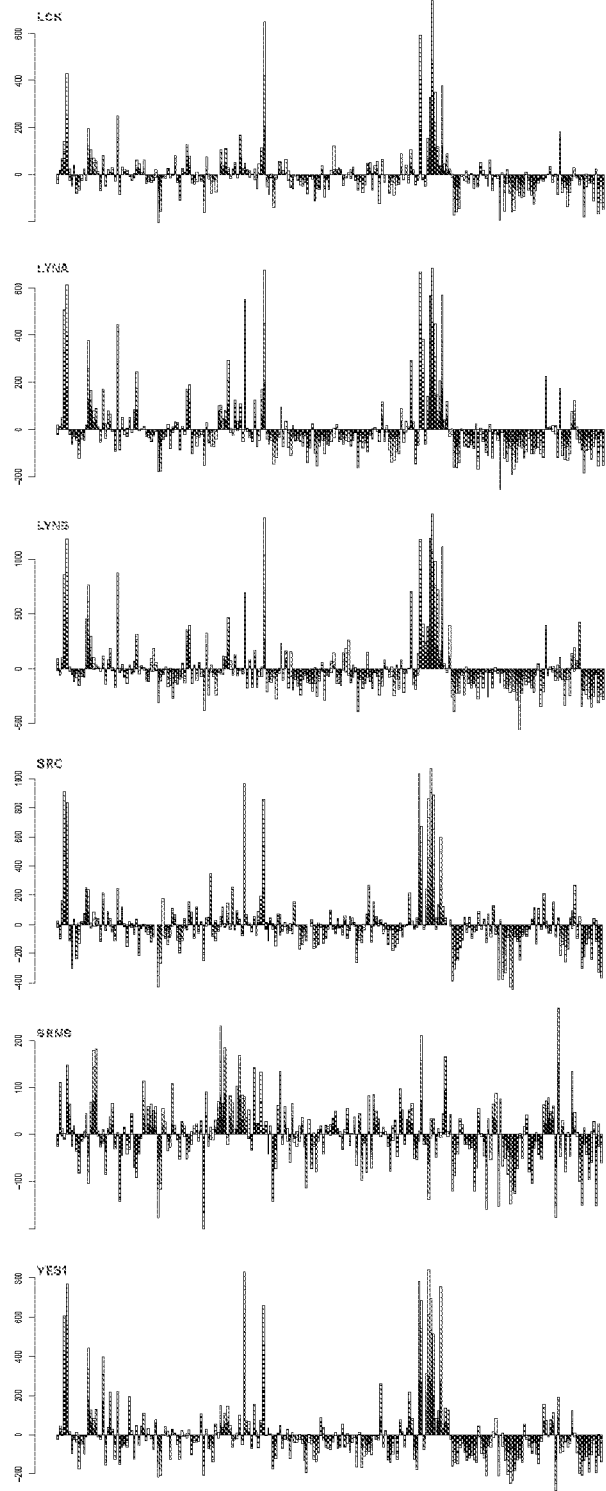
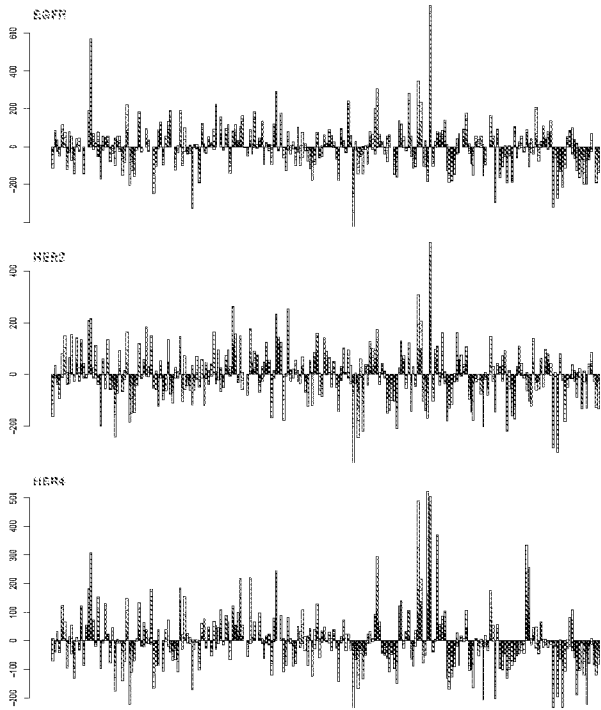


Fig. 5
(part 2)

WO 2019/200245

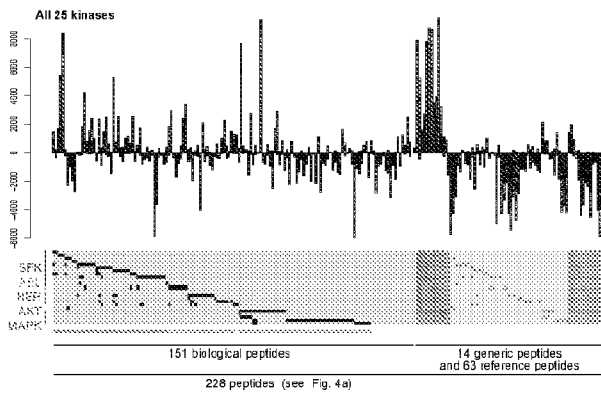
i

Cumulative plots of the activity signatures across experimental runs and derived from 228 peptide sensors, for the HER kinase family



i

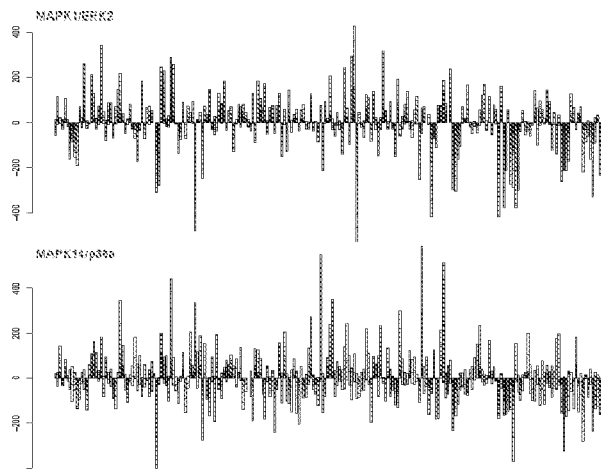
Cumulative plots of the activity signatures across experimental runs and derived from 228 peptide sensors, for all 25 recombinant kinases



PCT/US2019/027201

j

Cumulative plots of the activity signatures across experimental runs and derived from 228 peptide sensors, for the MAPK kinase family



k

Cumulative plots of the activity signatures across experimental runs and derived from 228 peptide sensors, for the CSK kinase

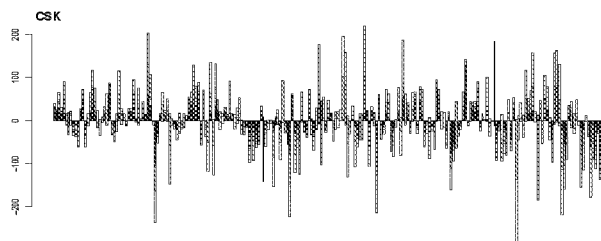


Fig. 5 (part 3)

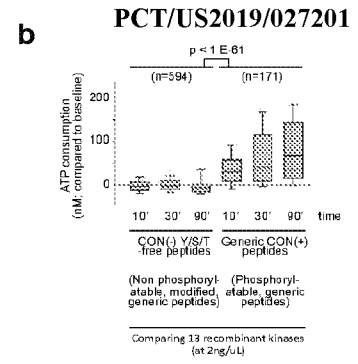
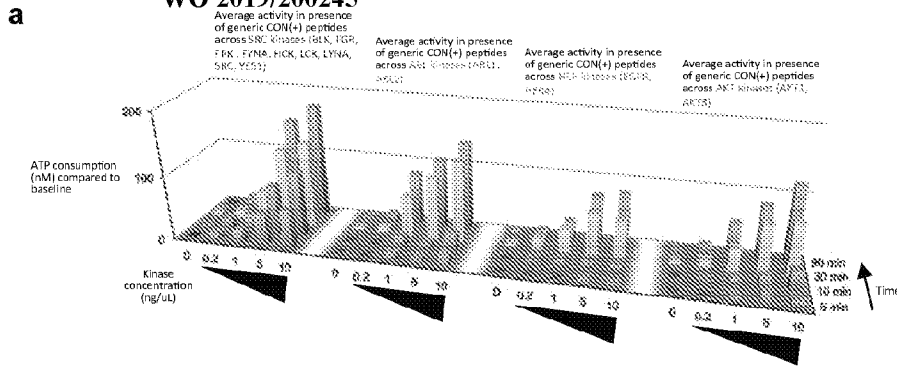


Fig. 6

WO 2019/200245

PCT/US2019/027201

a

Per kinase:

Column #1 = peptide category:

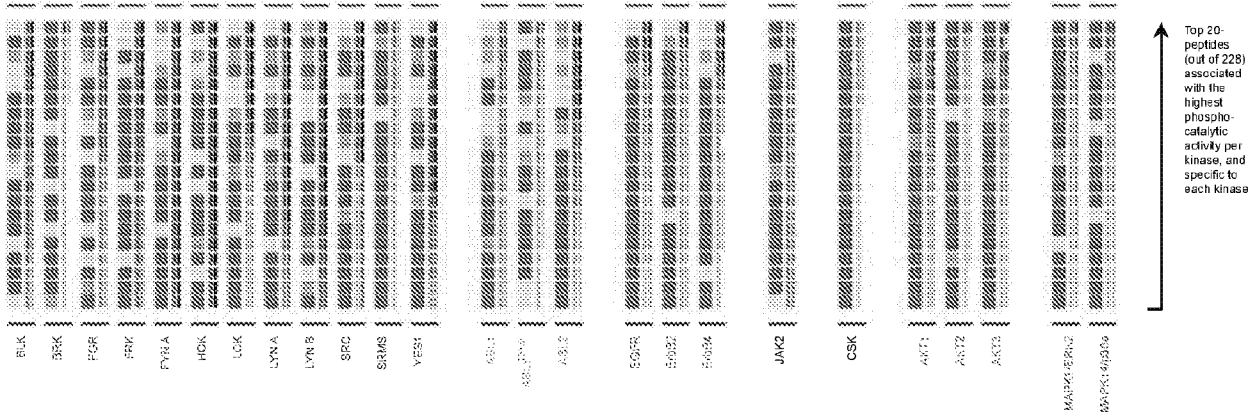
- advised generic CON+ for the indicated kinase
- other/non-advised generic CON+ for the indicated kinase
- biological peptide

Column #2 = level of catalytic activity per peptide (mean-centered across 228 peptides, and averaged across all experiments per kinase; in nM ATP)

0 +100

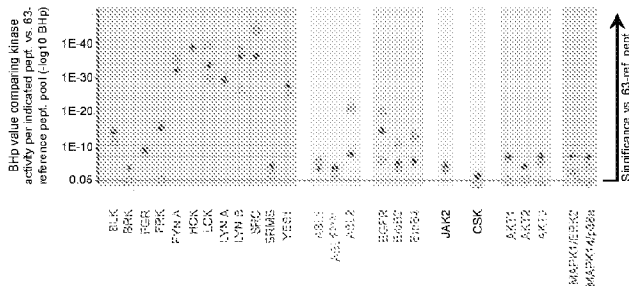
Column #3 = significance (BHP) of activity (comparing activity measured from the indicated peptide versus the pool of 83-reference peptides)

0.05 1E-10

**b**

Top 3-peptides per peptide category:

- best activity-reporting peptide among advised generic CON+ for the indicated kinase
- best activity-reporting peptide among any other/non-advised generic CON+ for the indicated kinase
- best activity-reporting peptide among biological peptides

**c**

Top 3-peptides per peptide category:

- best activity-reporting peptide among advised generic CON+ for the indicated kinase
- best activity-reporting peptide among any other/non-advised generic CON+ for the indicated kinase
- best activity-reporting peptide among biological peptides

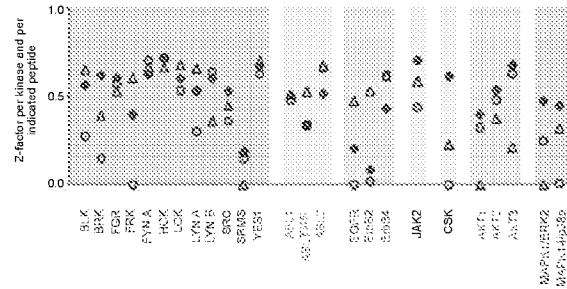


Fig. 7

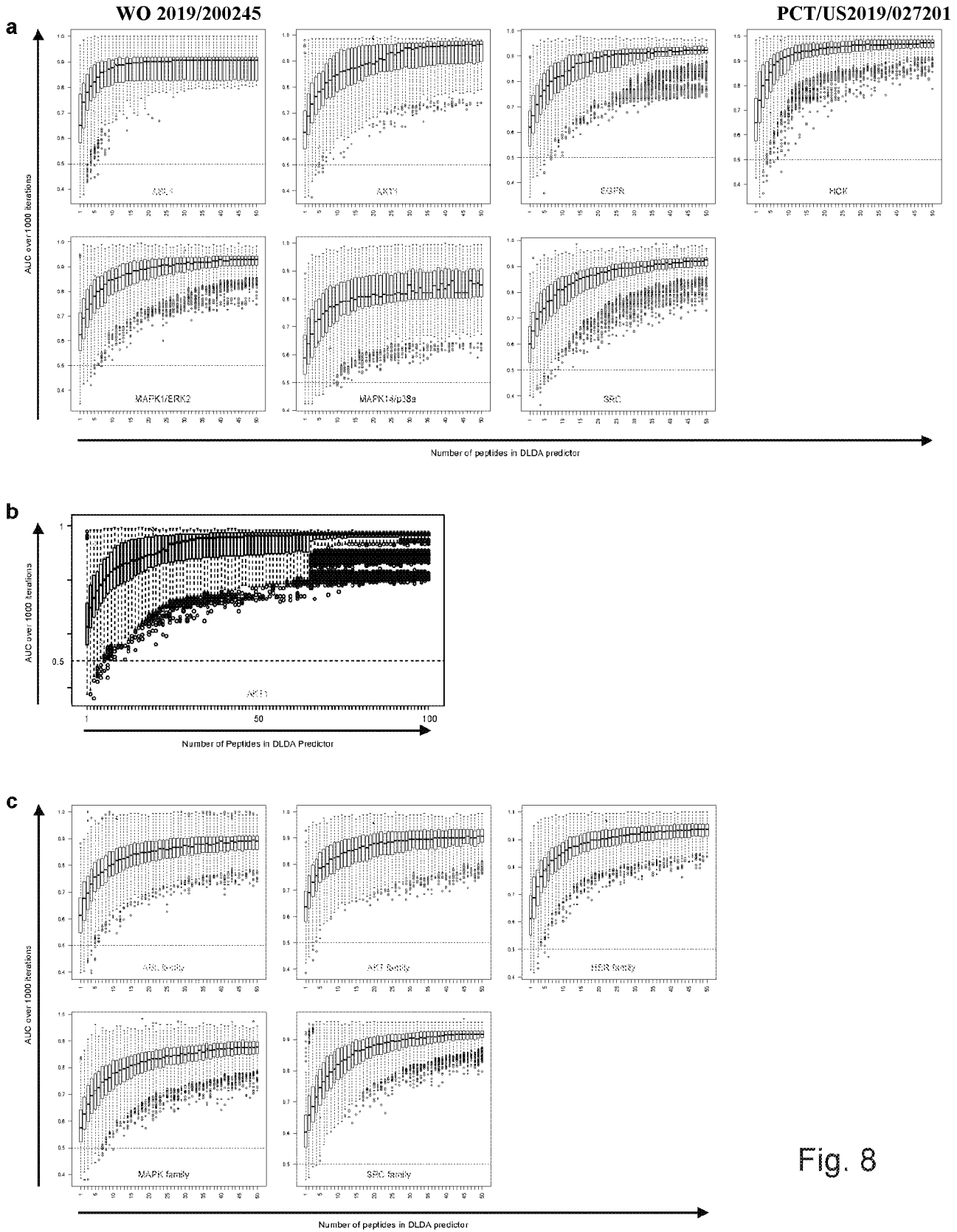


Fig. 8

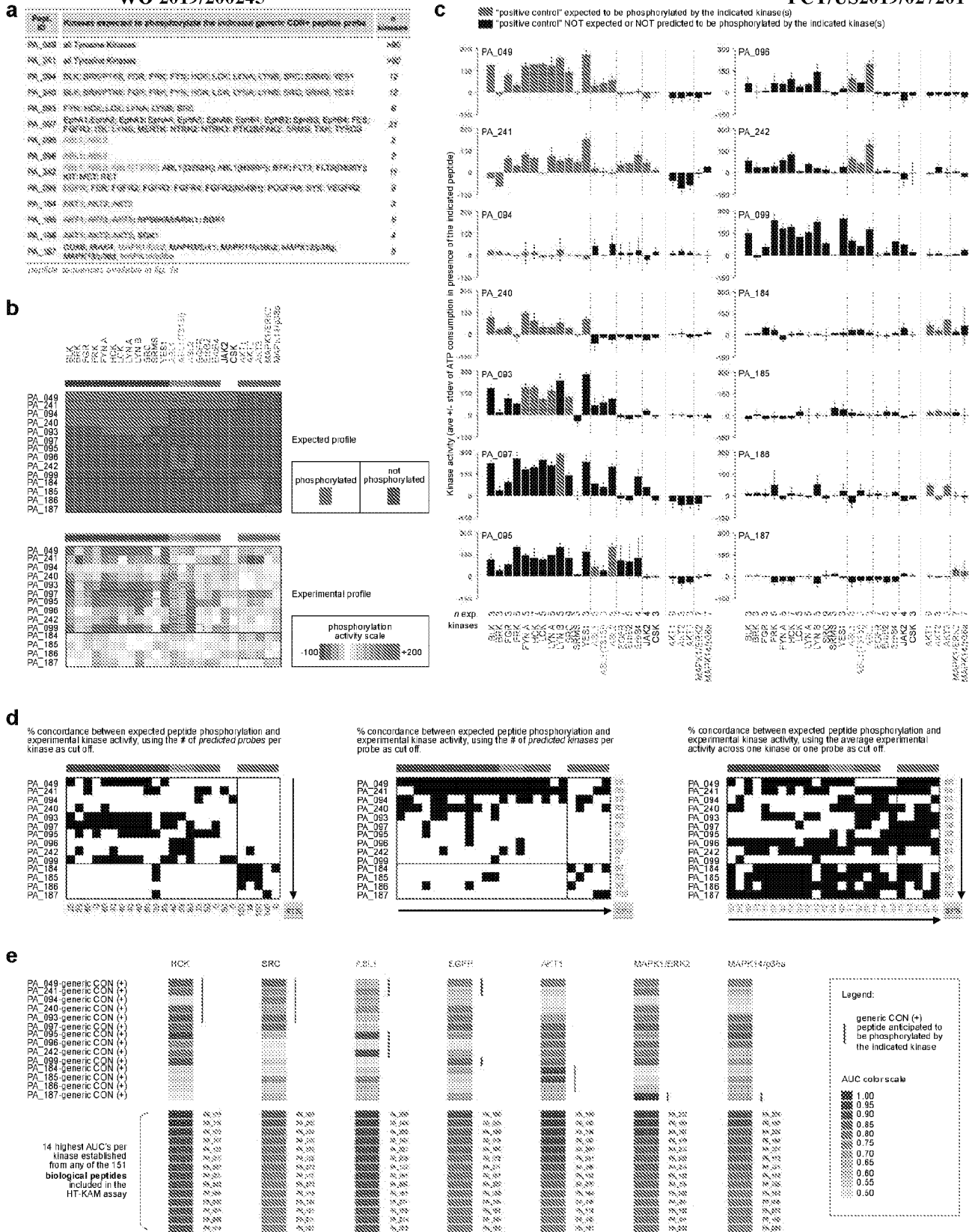


Fig. 9

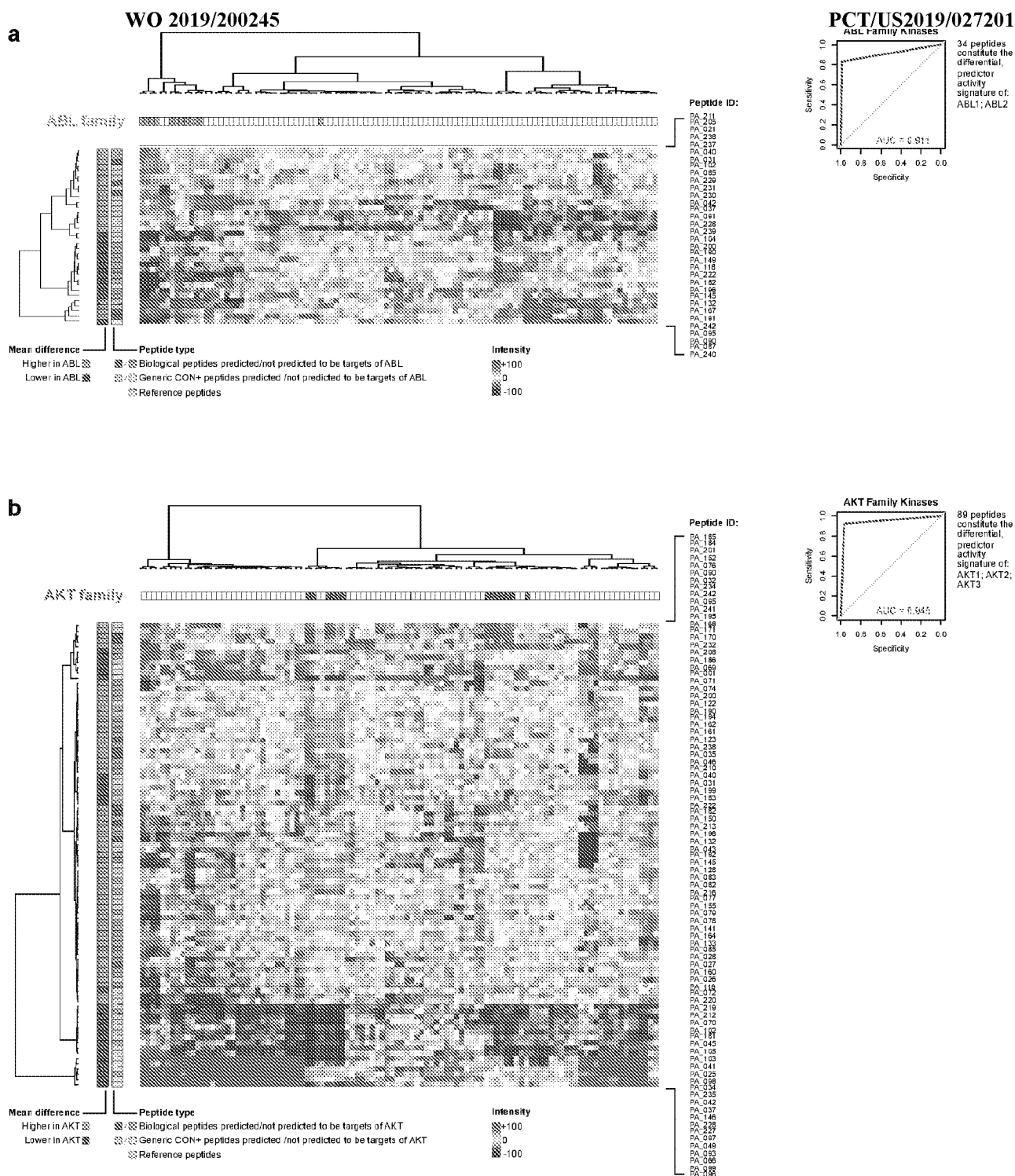


Fig. 10
(part 1)

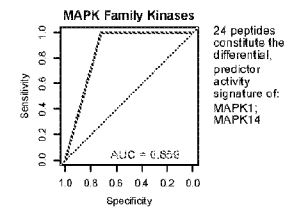
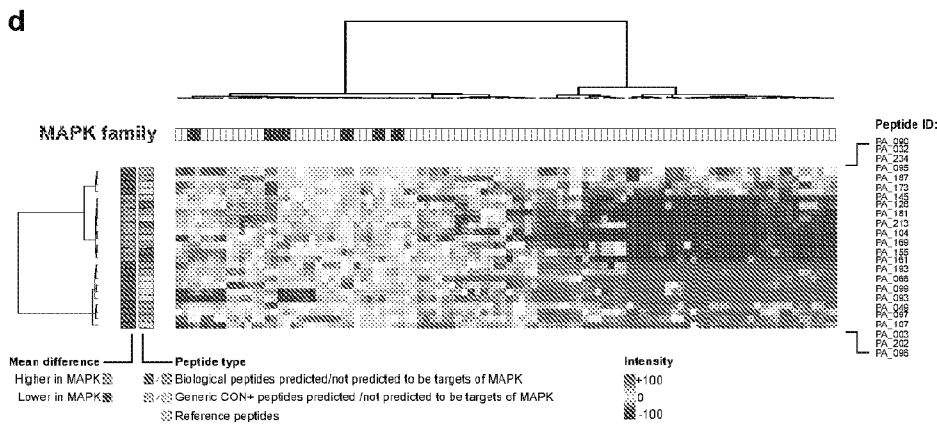
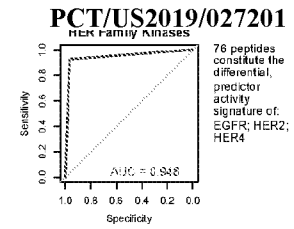
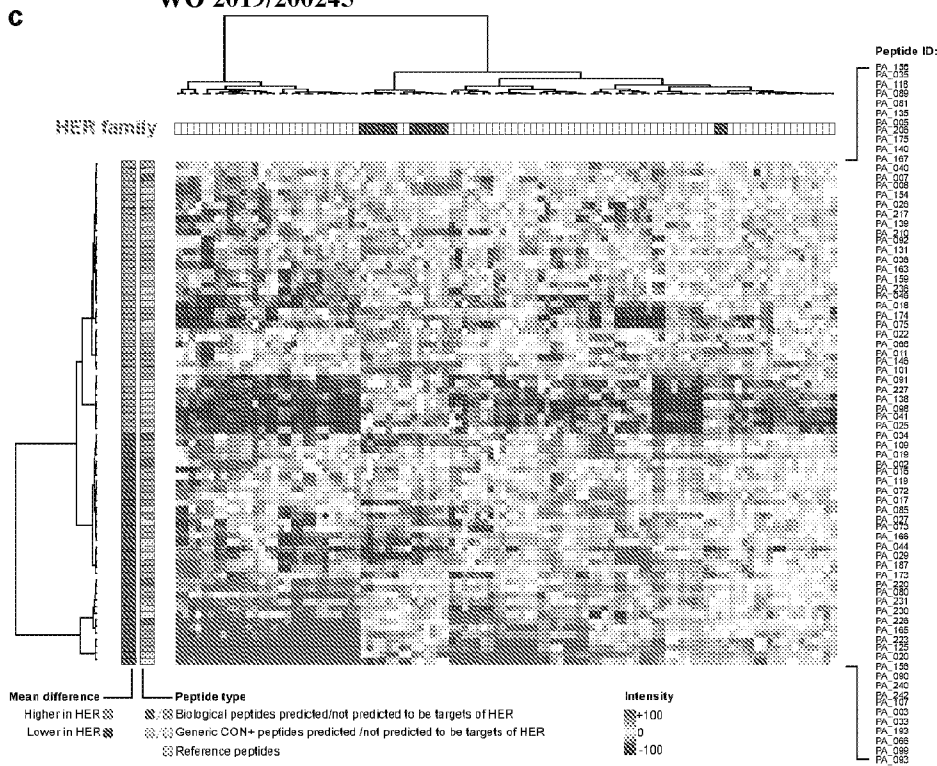


Fig. 10 (part 2)

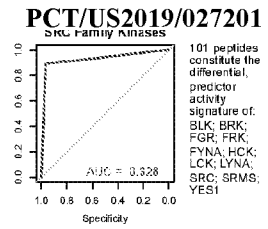
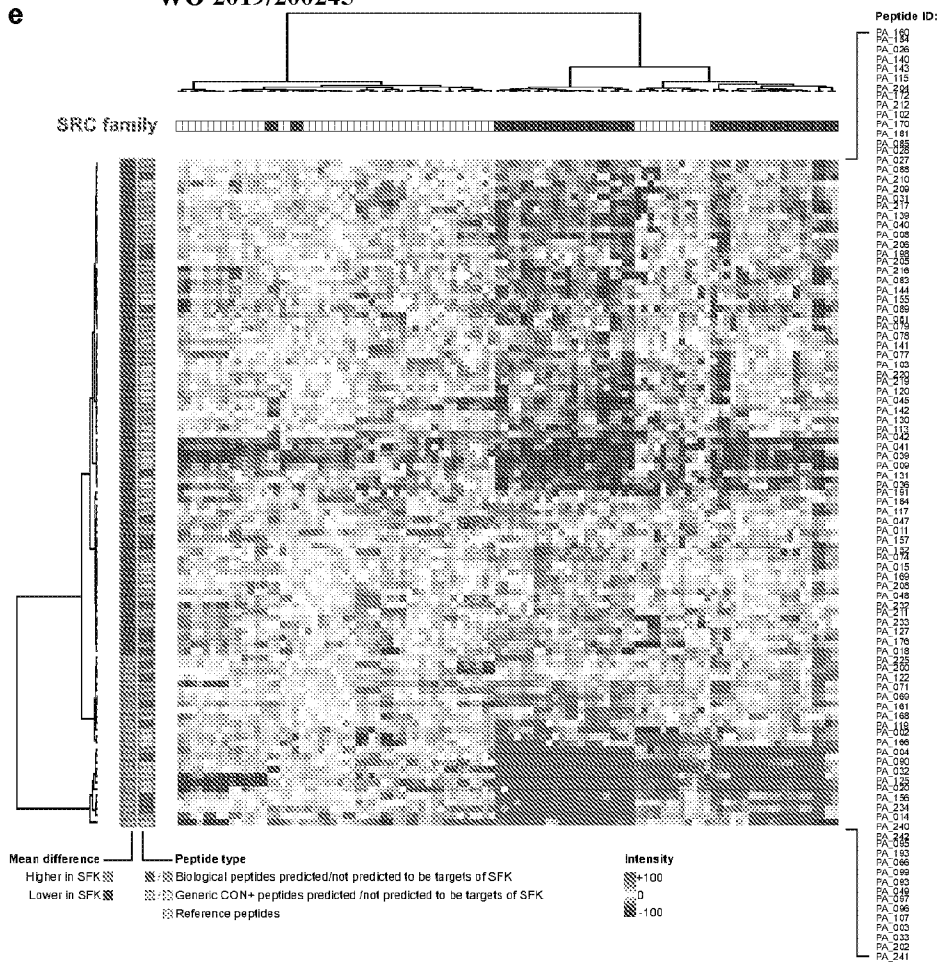


Fig. 10 (part 3)

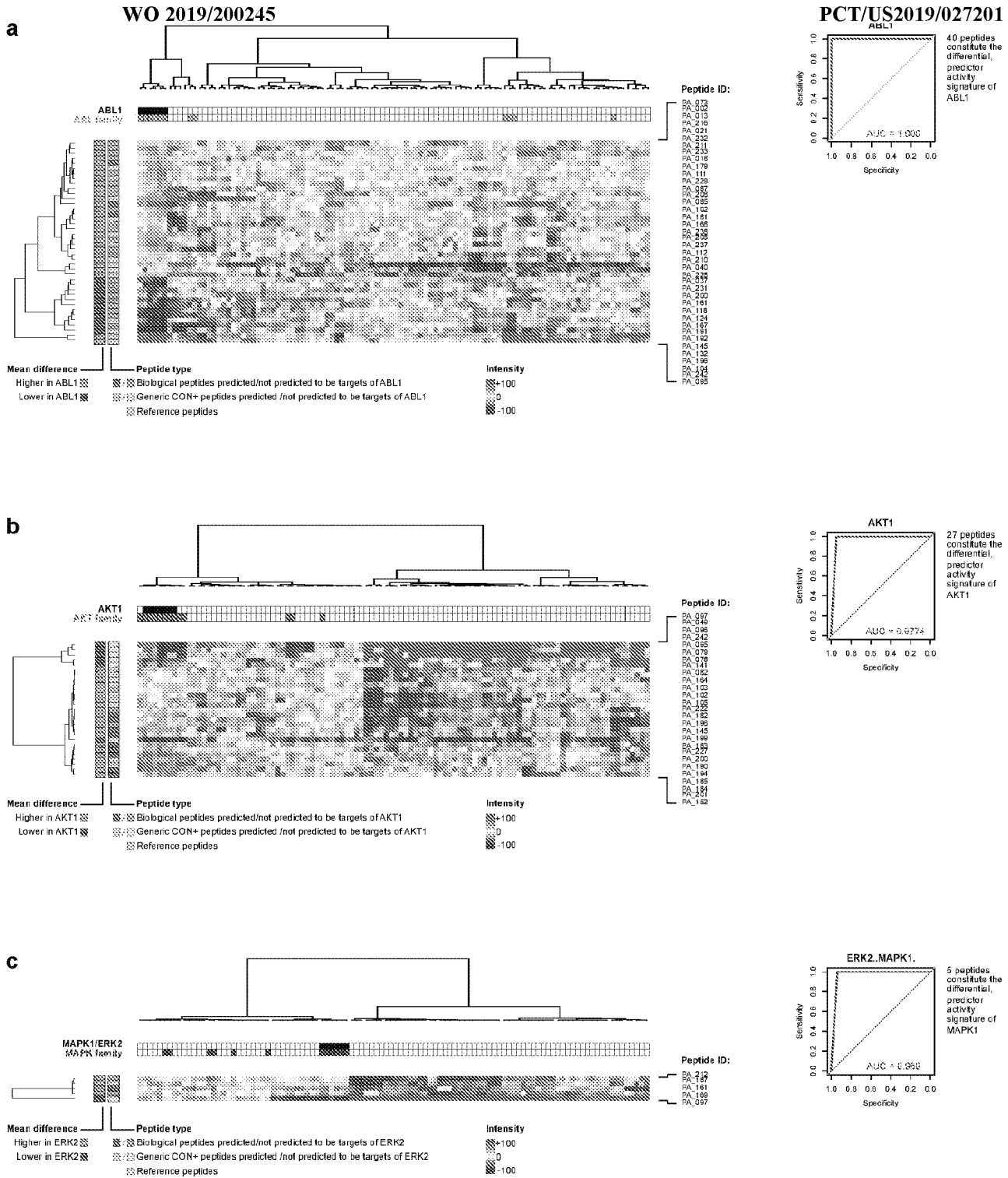


Fig. 11 (part 1)

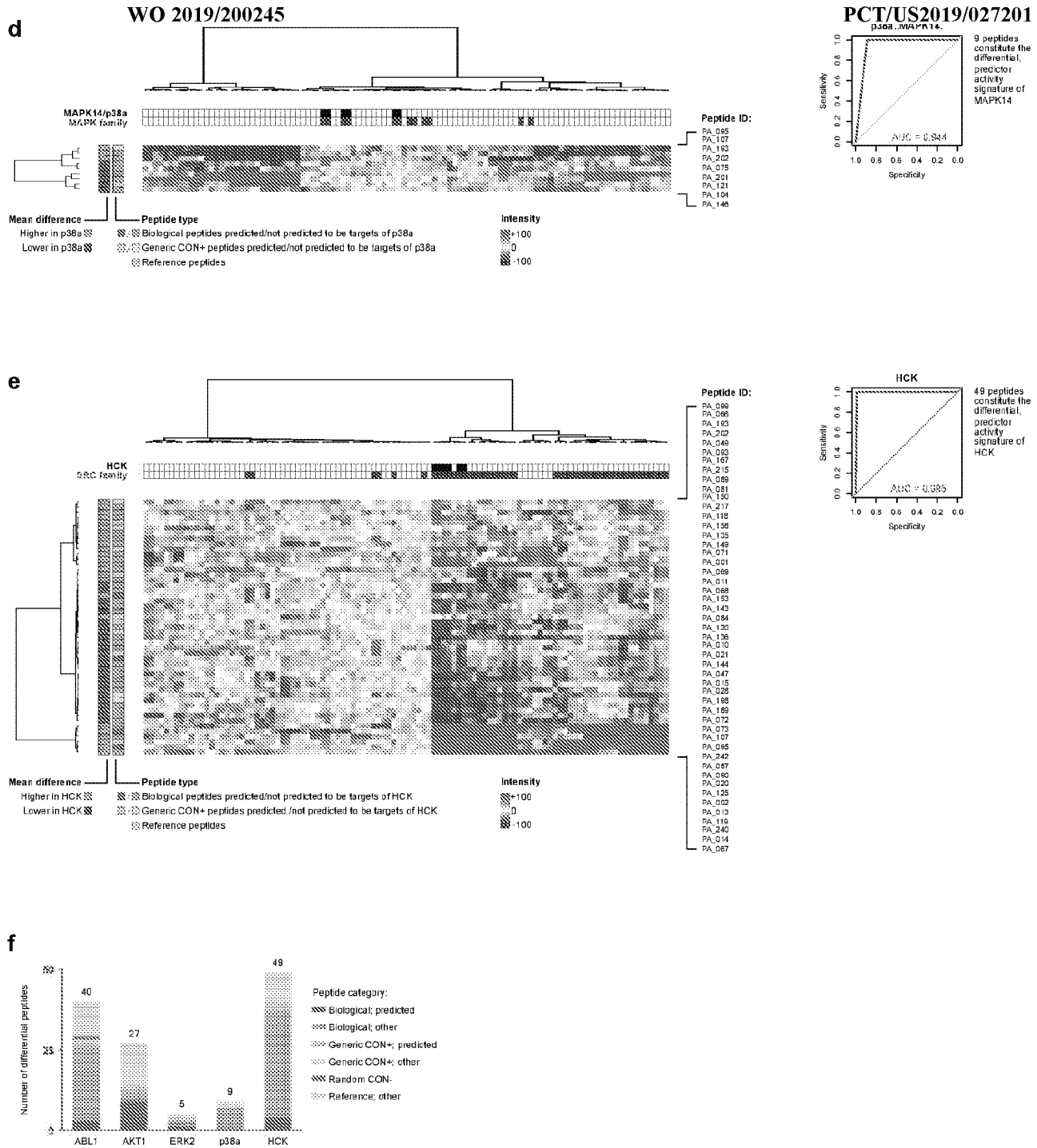


Fig. 11 (part 2)

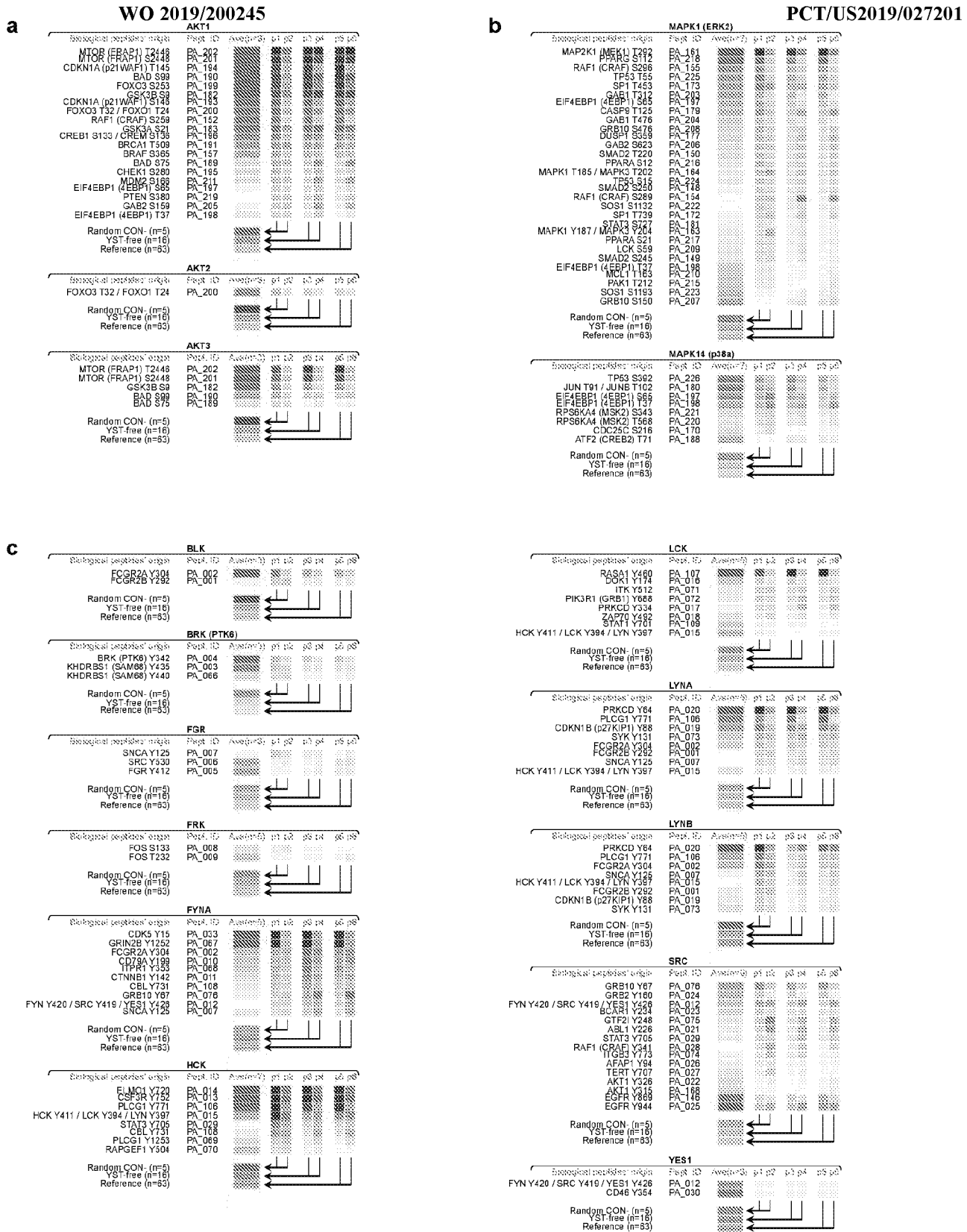


Fig. 12 (part 1)

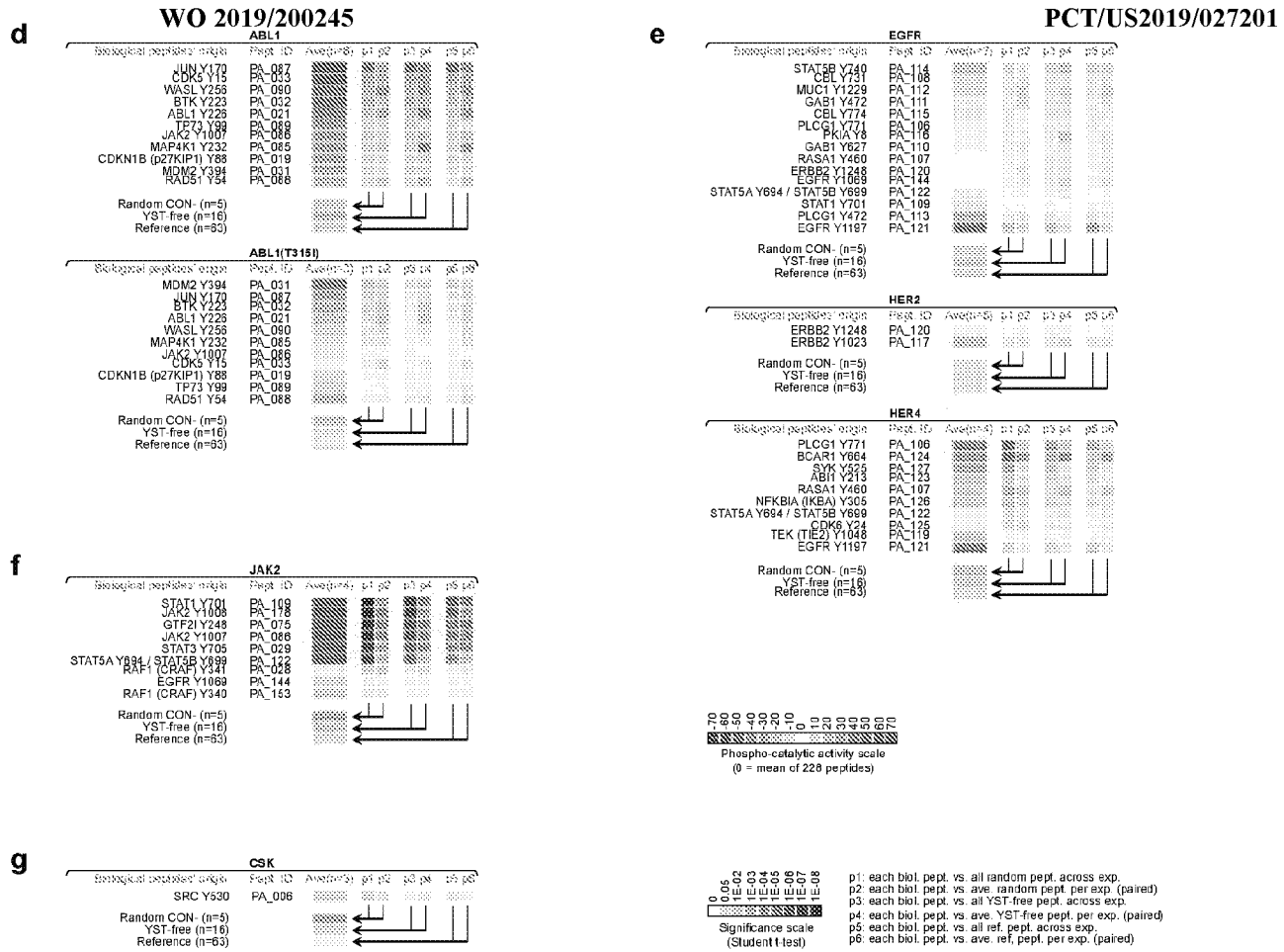


Fig. 12 (part 2)

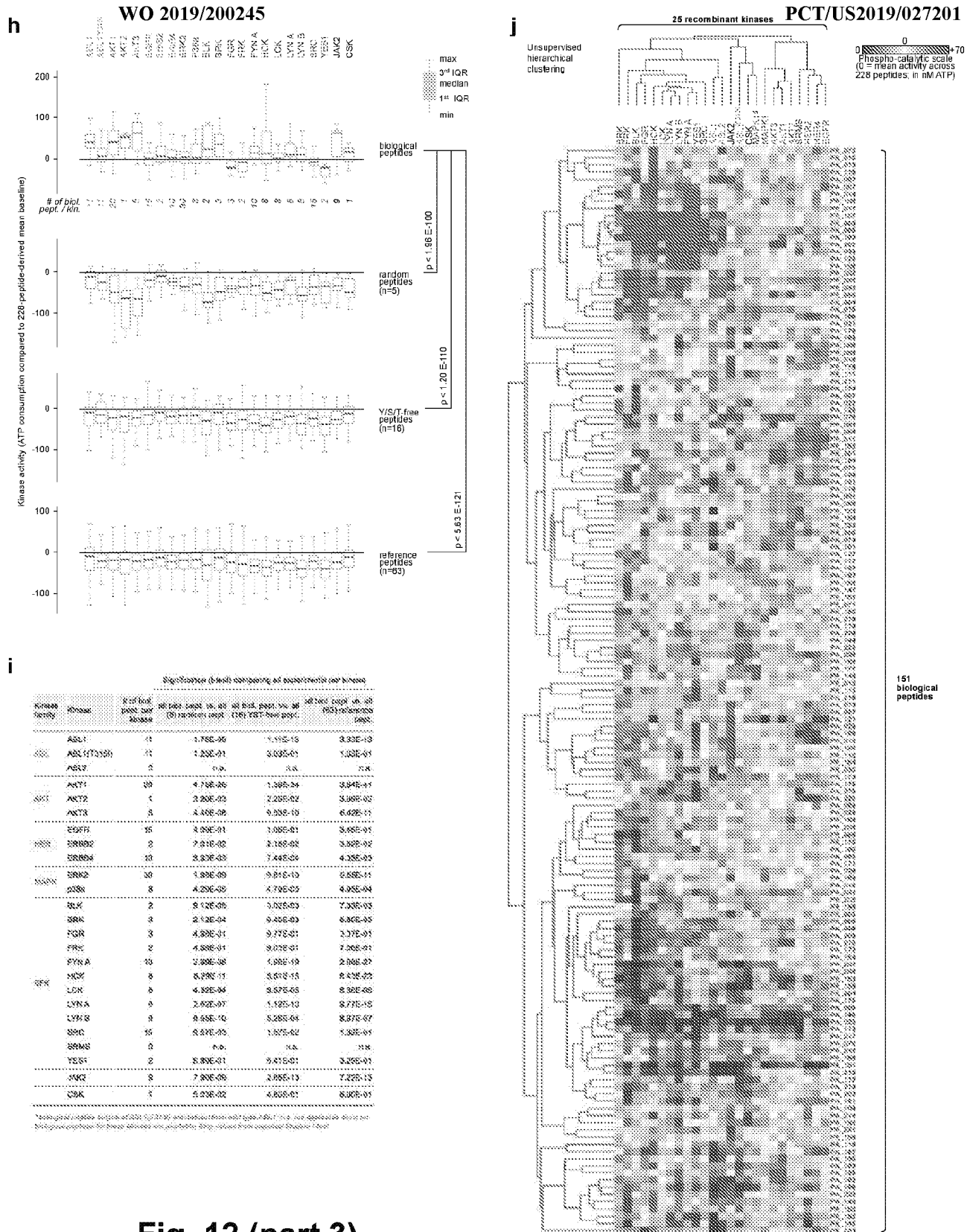


Fig. 12 (part 3)

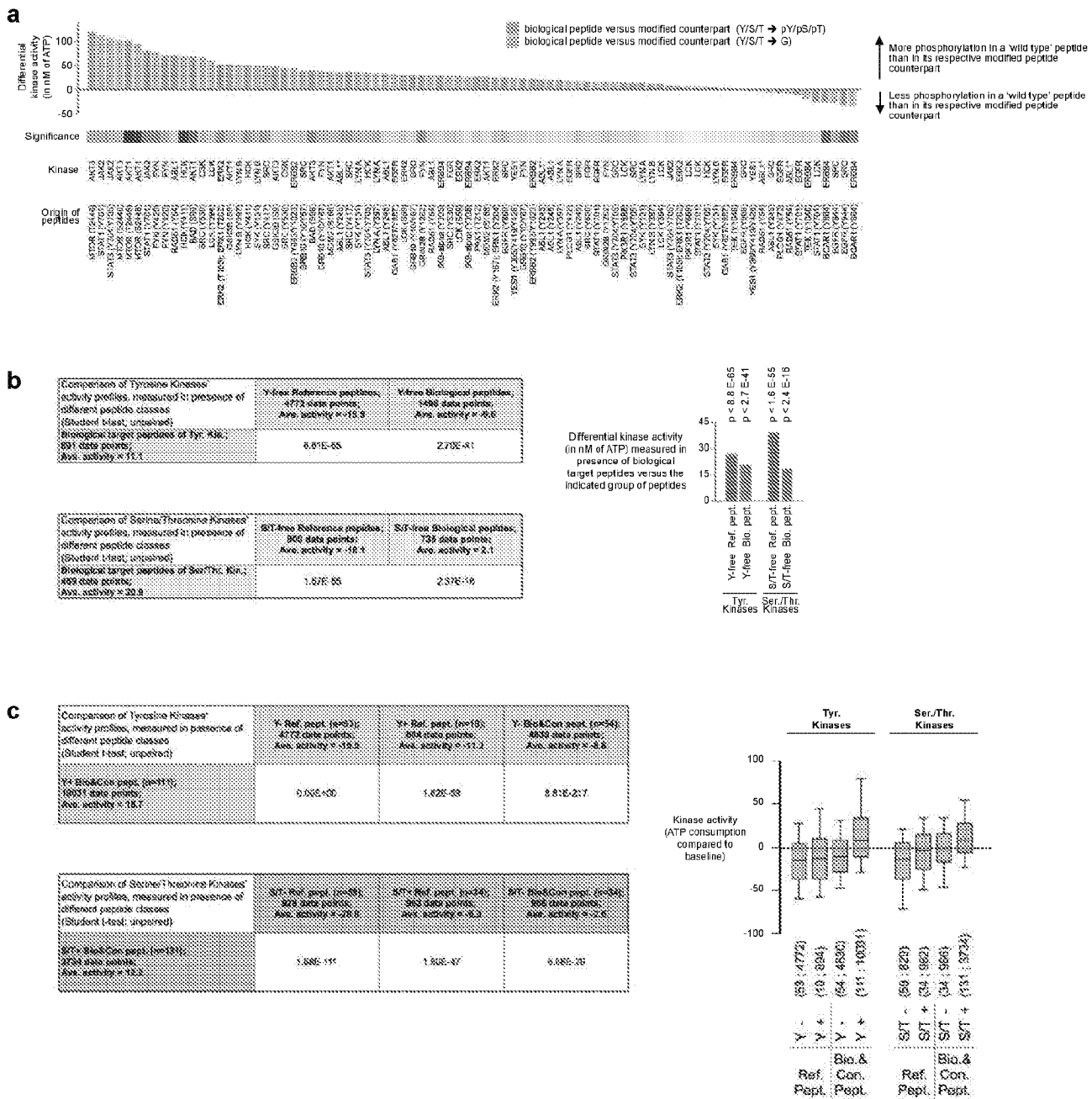
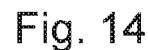


Fig. 13



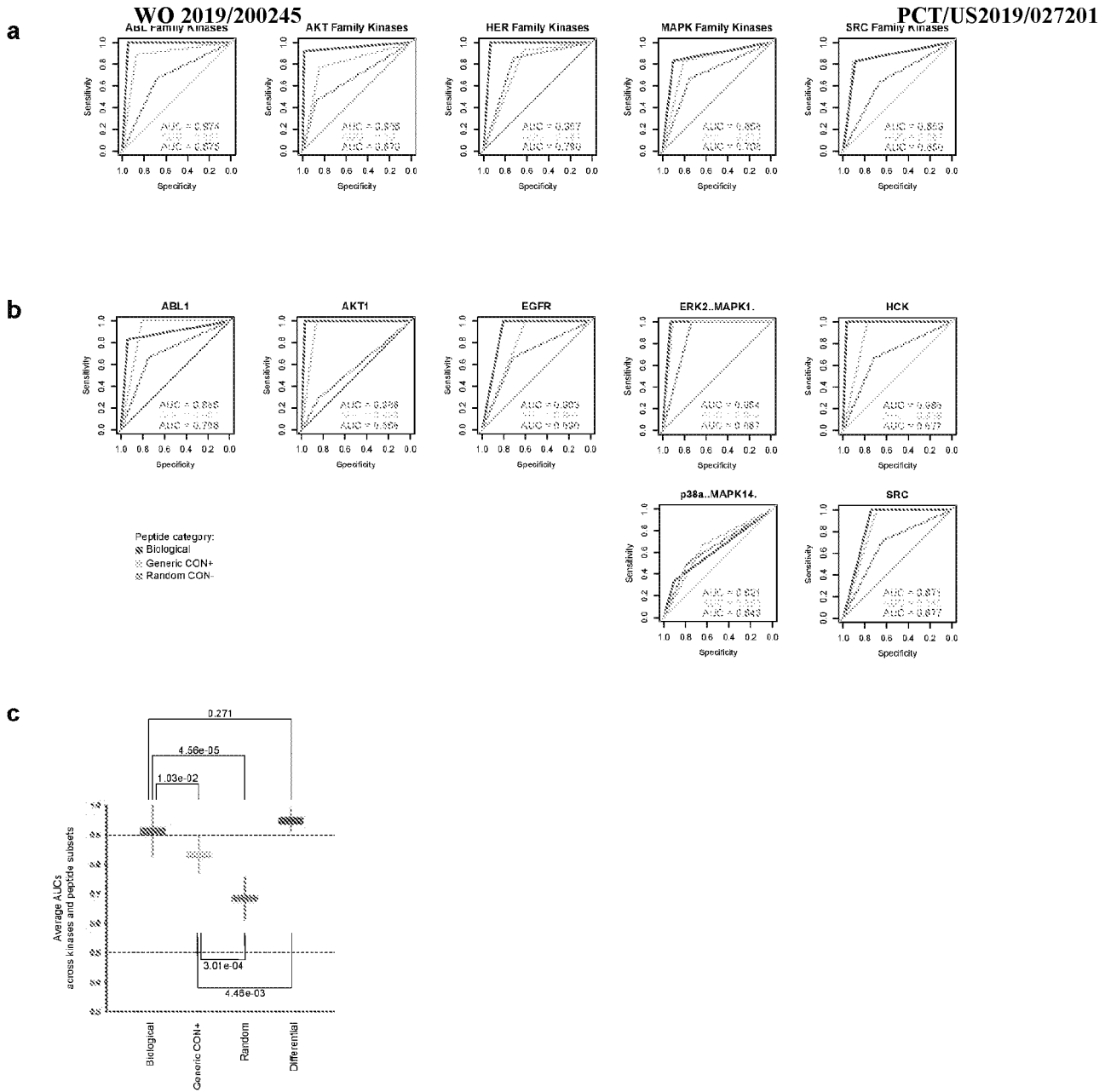


Fig. 15

WO 2019/200245

PCT/US2019/027201

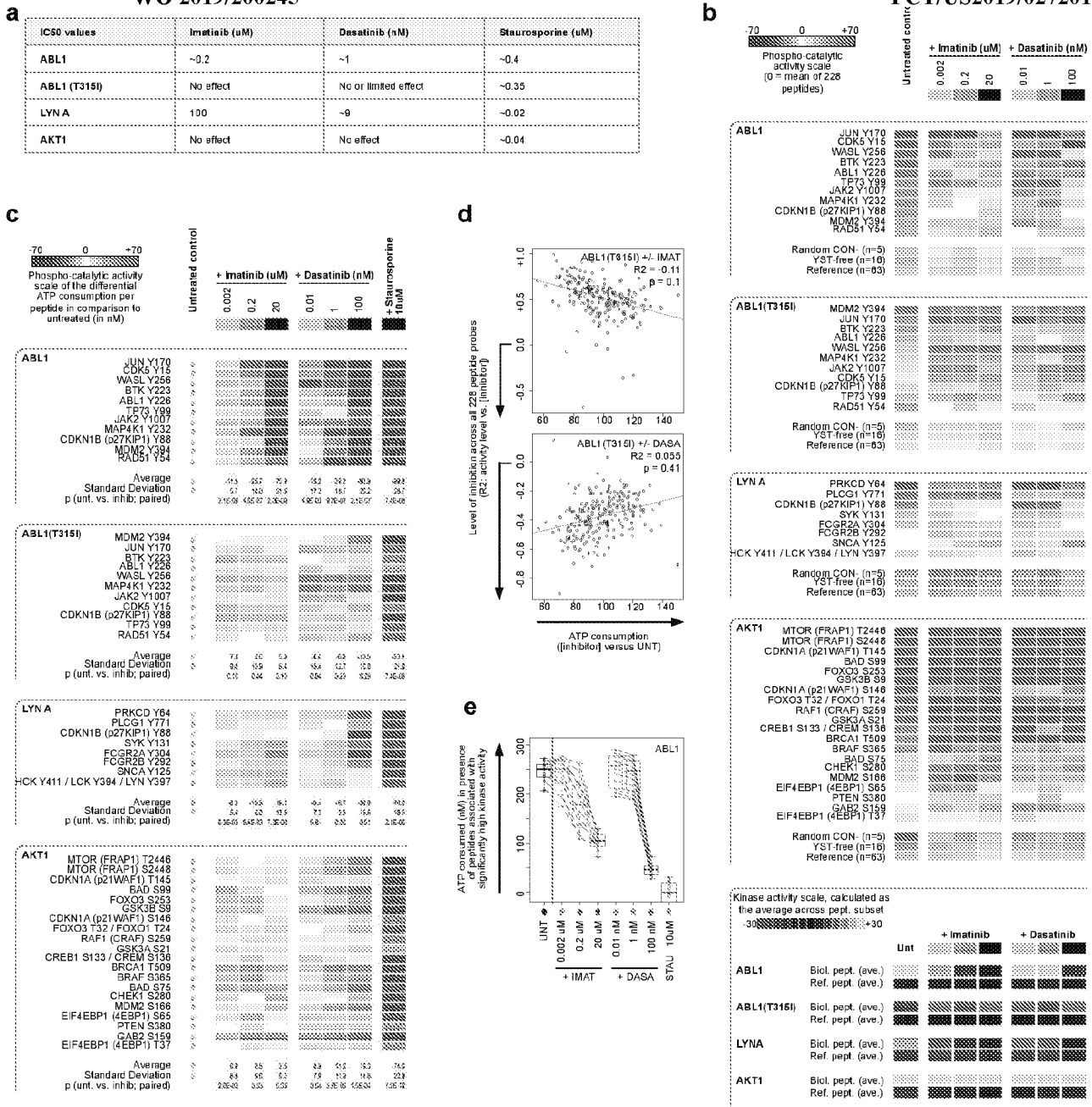


Fig. 16

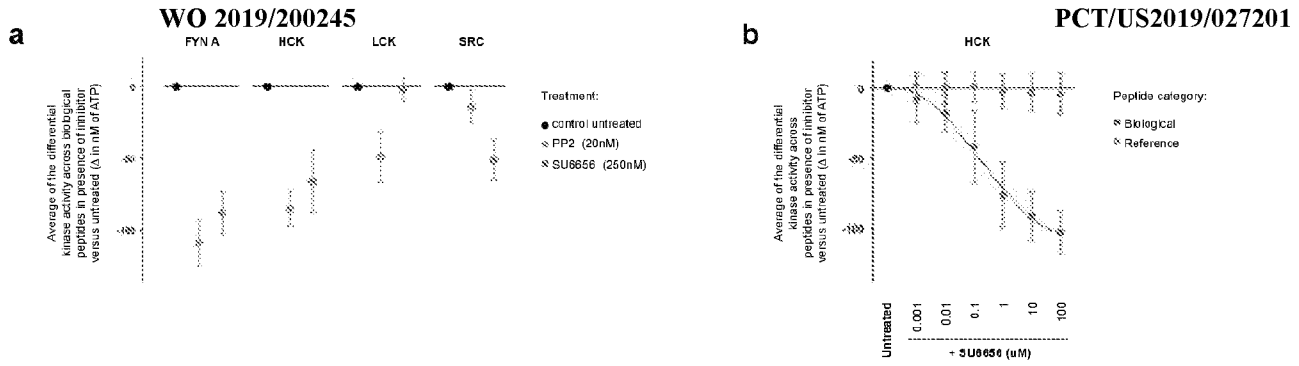


Fig. 17

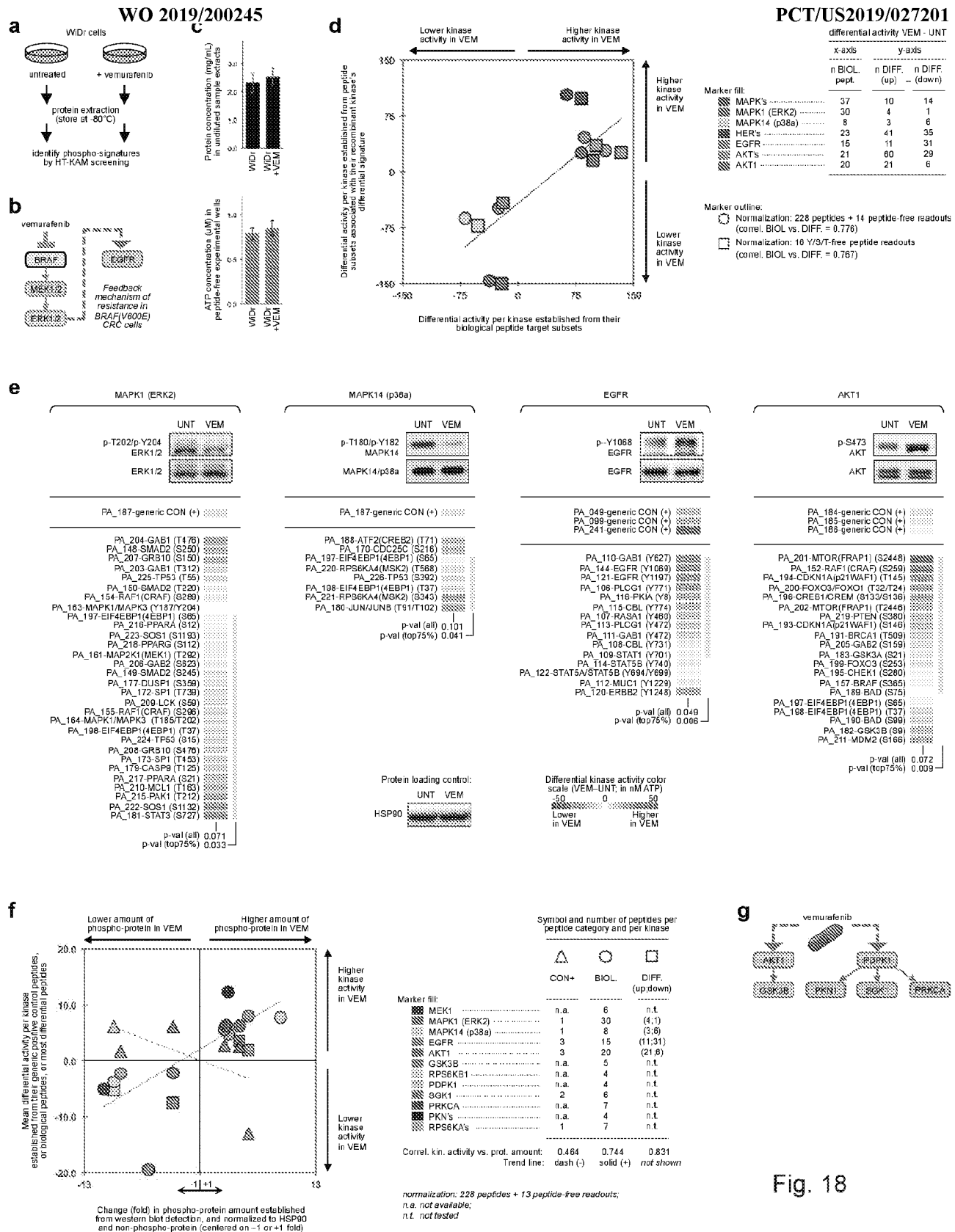


Fig. 18

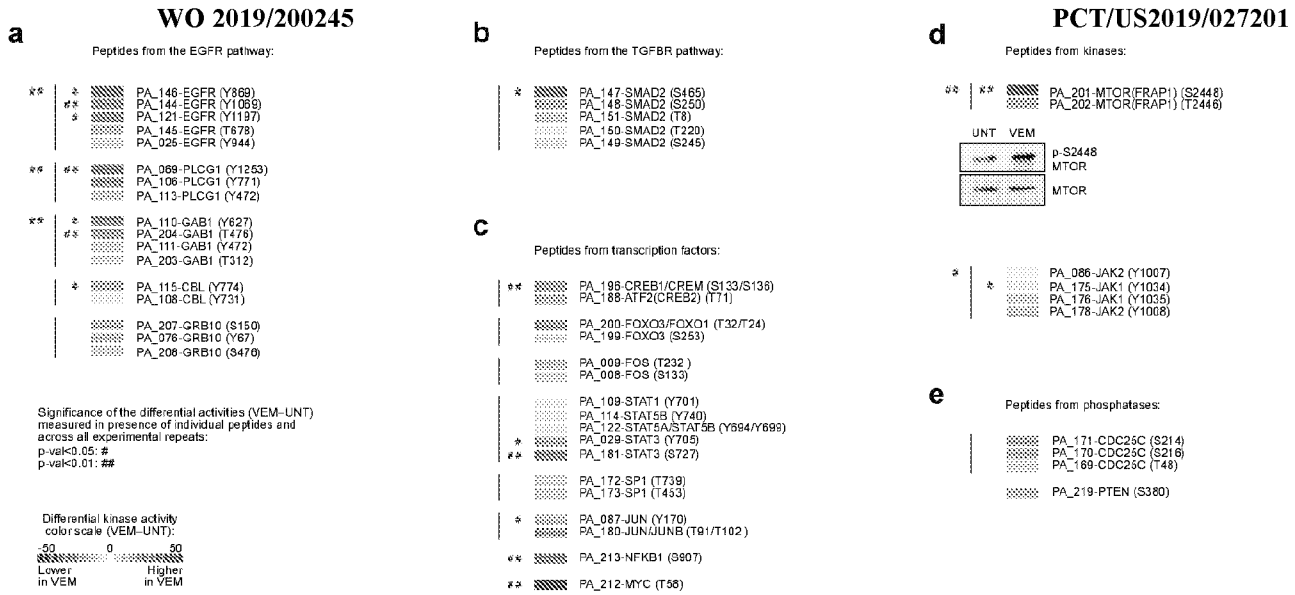
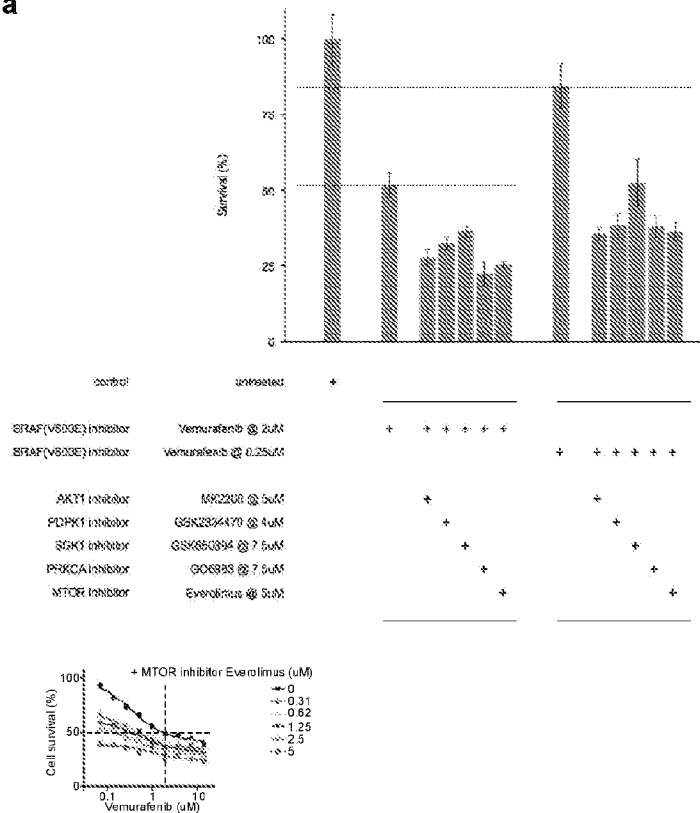


Fig. 19

WO 2019/200245

PCT/US2019/027201

a



b

characterization of WDr cell response to:

Drug name	Vemurafenib combinations with the indicated inhibitor		
	Drug interaction (IC50; uM)	Drug interaction (two-way ANOVA p-val)	Combination index (50% AC50 concentrations)
AKT1 inhibitor	5.5	$p < 1E-10$	0.6
PDPK1 inhibitor	4.5	$p < 1E-10$	0.5
SGK1 inhibitor	12.5	$p < 1E-04$	0.8
PRKCA inhibitor	15	$p < 1E-05$	0.6
MTOR inhibitor	5	$p < 1E-05$	0.7

c

Tested and validated inhibitors and BRAF(V600E) cell lines

Kinase targets	Inhibitors	Cell lines (5)
AKT1	MK2208	WDr, HT29, RKO1
EGFR	Gefitinib (96)	WDr, HT29
BRAF(V600E)	Vemurafenib (PLX4032)	WDr, HT29, RKO1, 8505C
MTOR	Everolimus (RAD001)	WDr, 8505C
PDPK1	GSK2334470, OSU-03012	WDr, HT29
PRKCA	GO6983, Sotrastaurin	WDr
SGK1	GSK850394	WDr, HT29

5 used as control based on Prahallad A et al. 2012 Nature, and Corcoran RB et al. 2012 Cancer Discovery, 5 inhibitor effects we measured in the indicated cell lines were further corroborated using data from Garnett MJ 2012 Nature and Yang W 2013 NAR.

Fig. 20

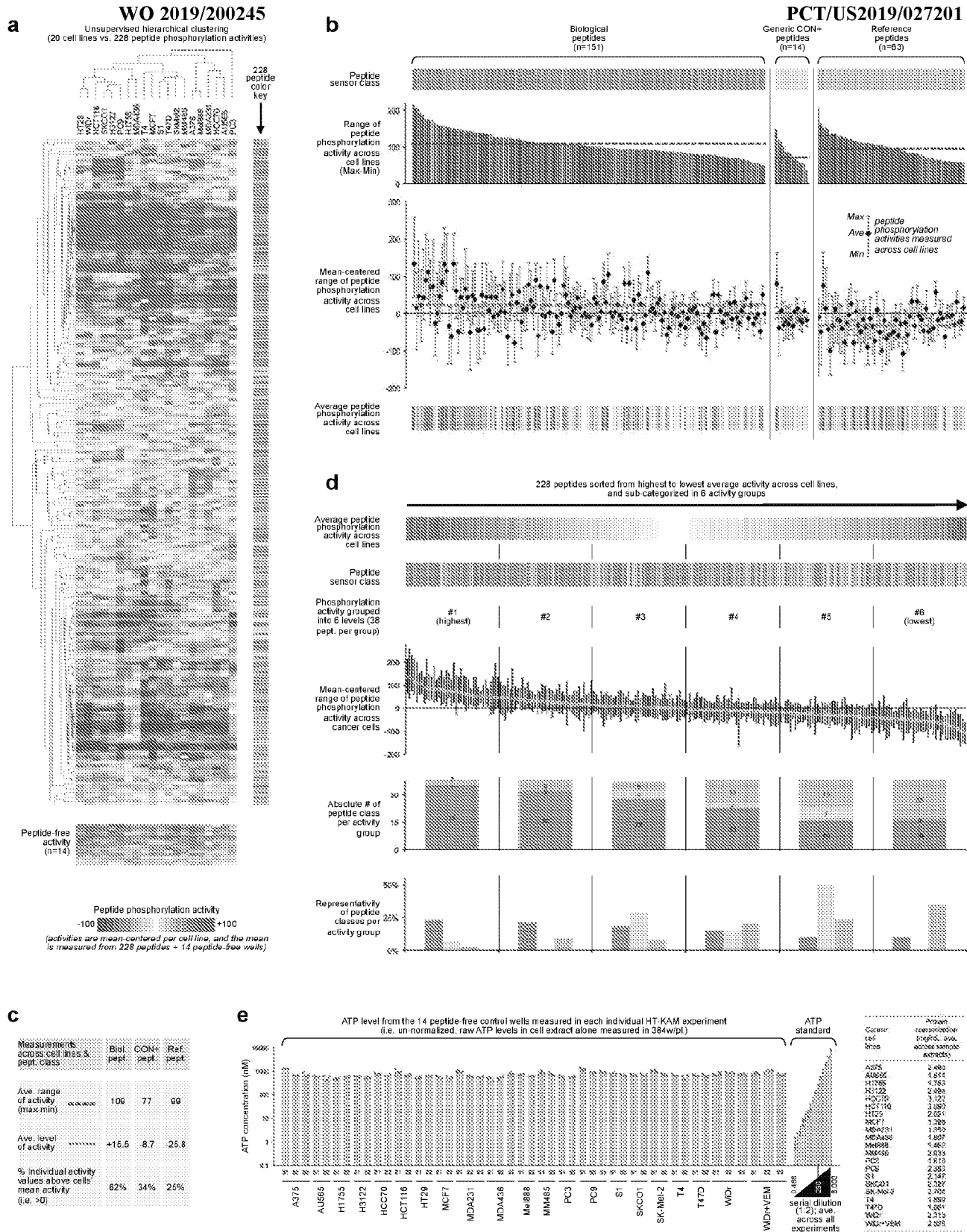


Fig. 21

WO 2019/200245

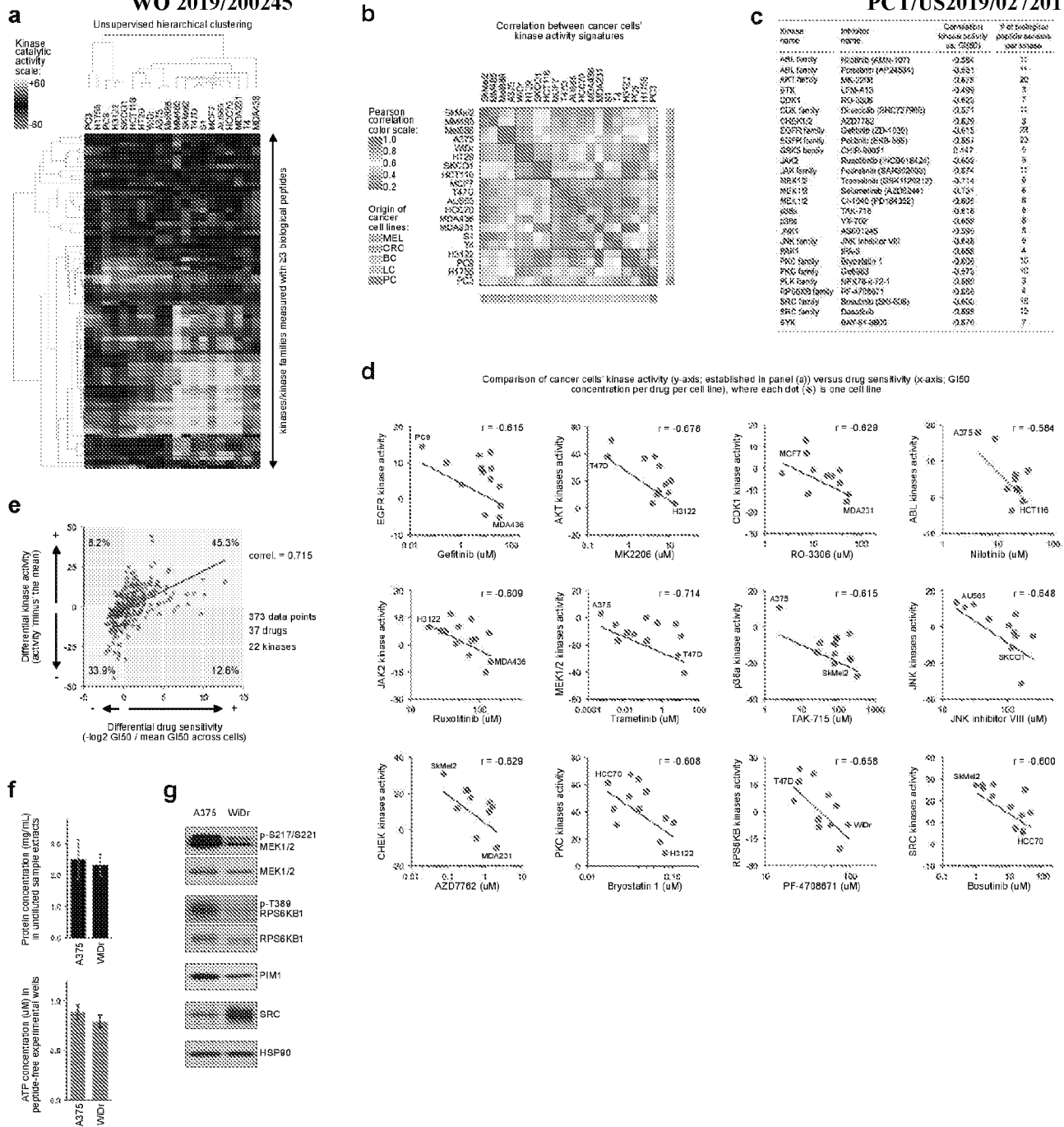


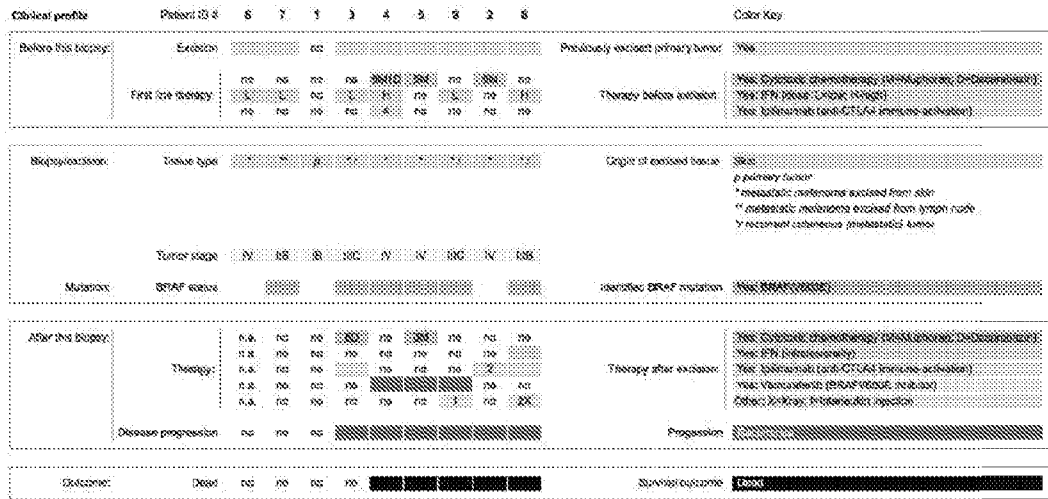
Fig. 22

WO 2019/200245

PCT/US2019/027201

a

 Kinase-targeted therapy
 Immunotherapy
 Chemotherapy



b

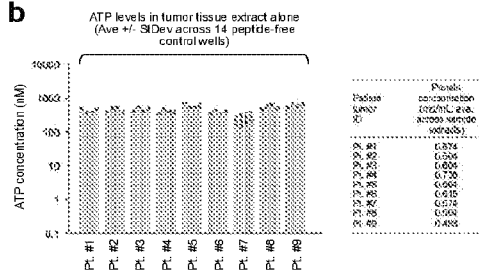


Fig. 23

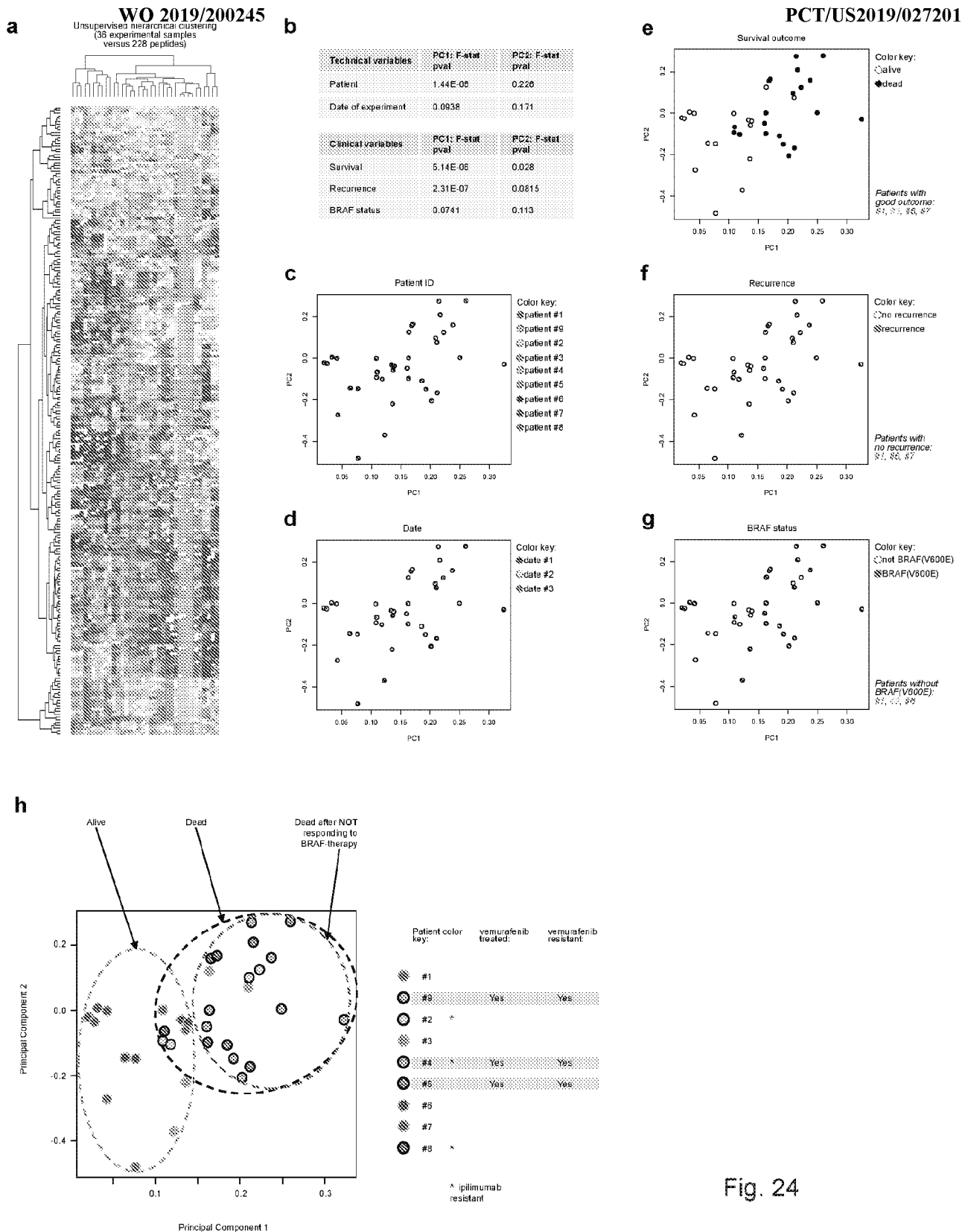


Fig. 24

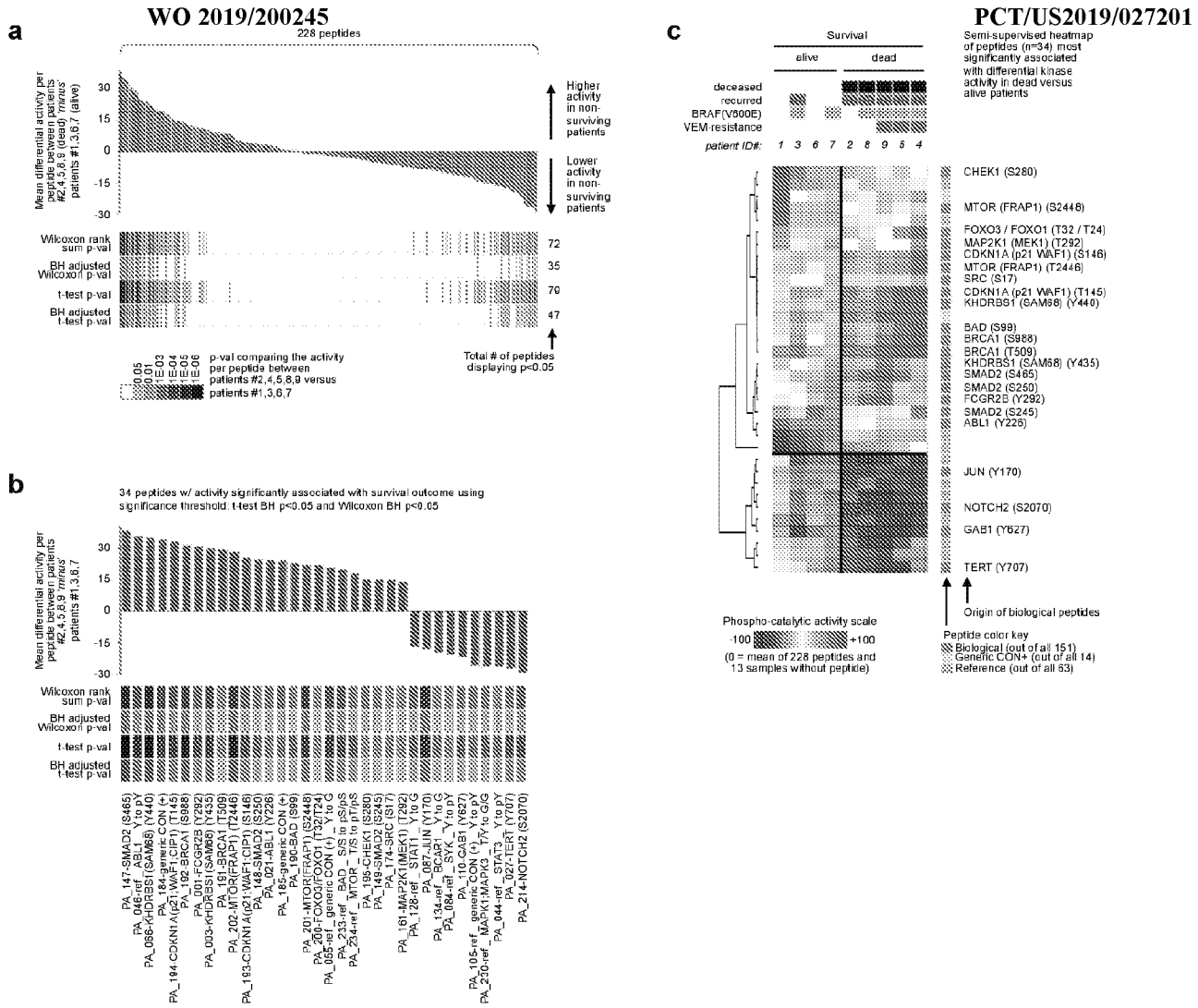


Fig. 25 (part 1)

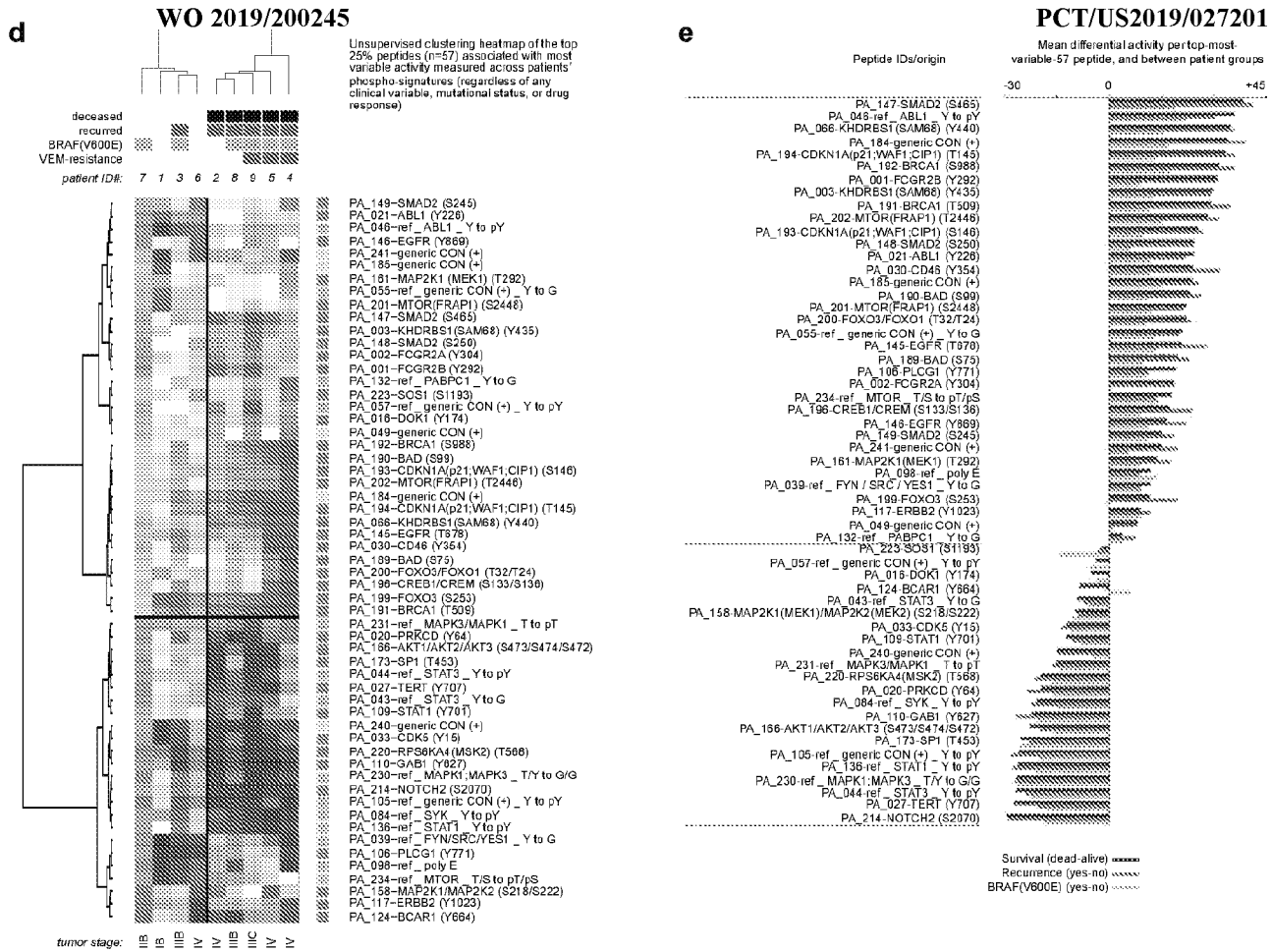
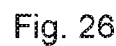


Fig. 25 (part 2)



WO 2019/200245

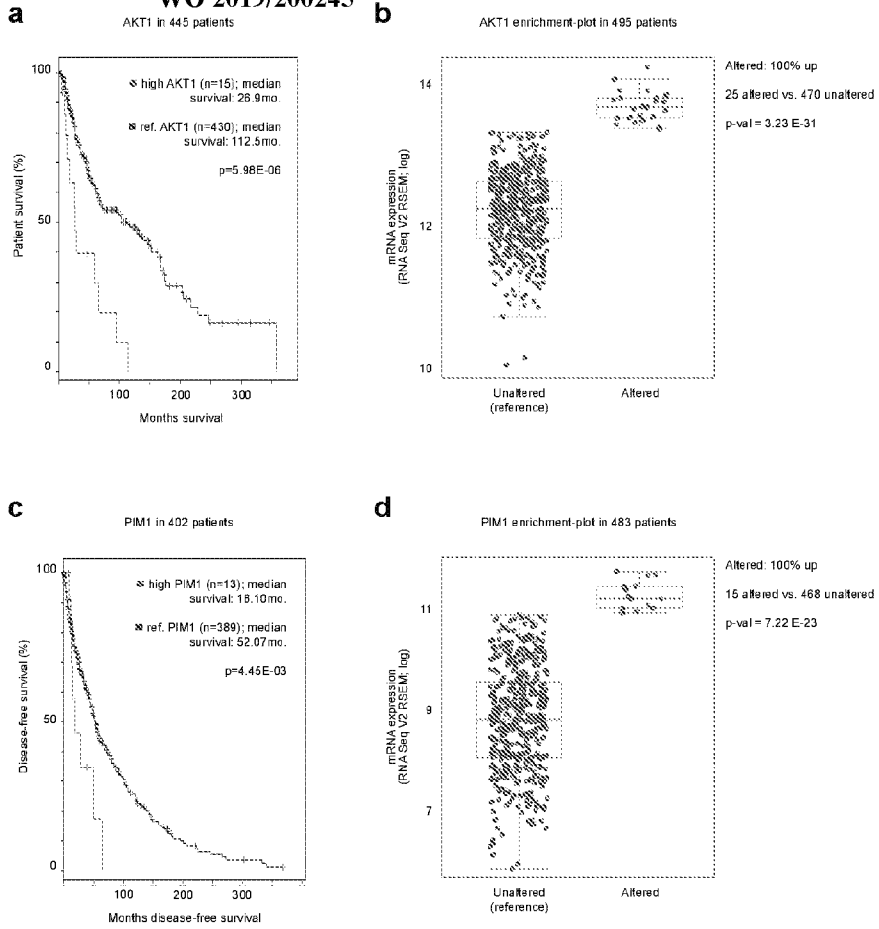
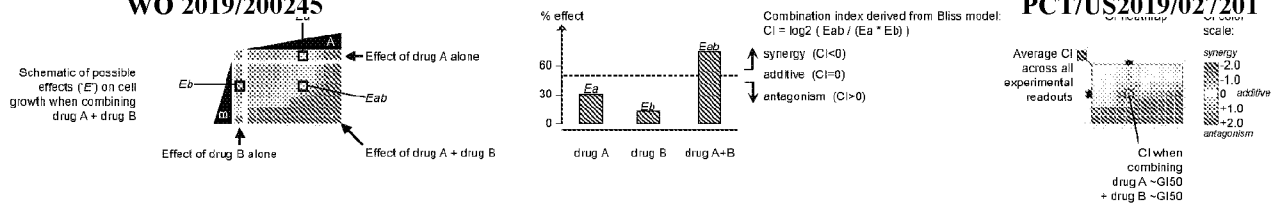


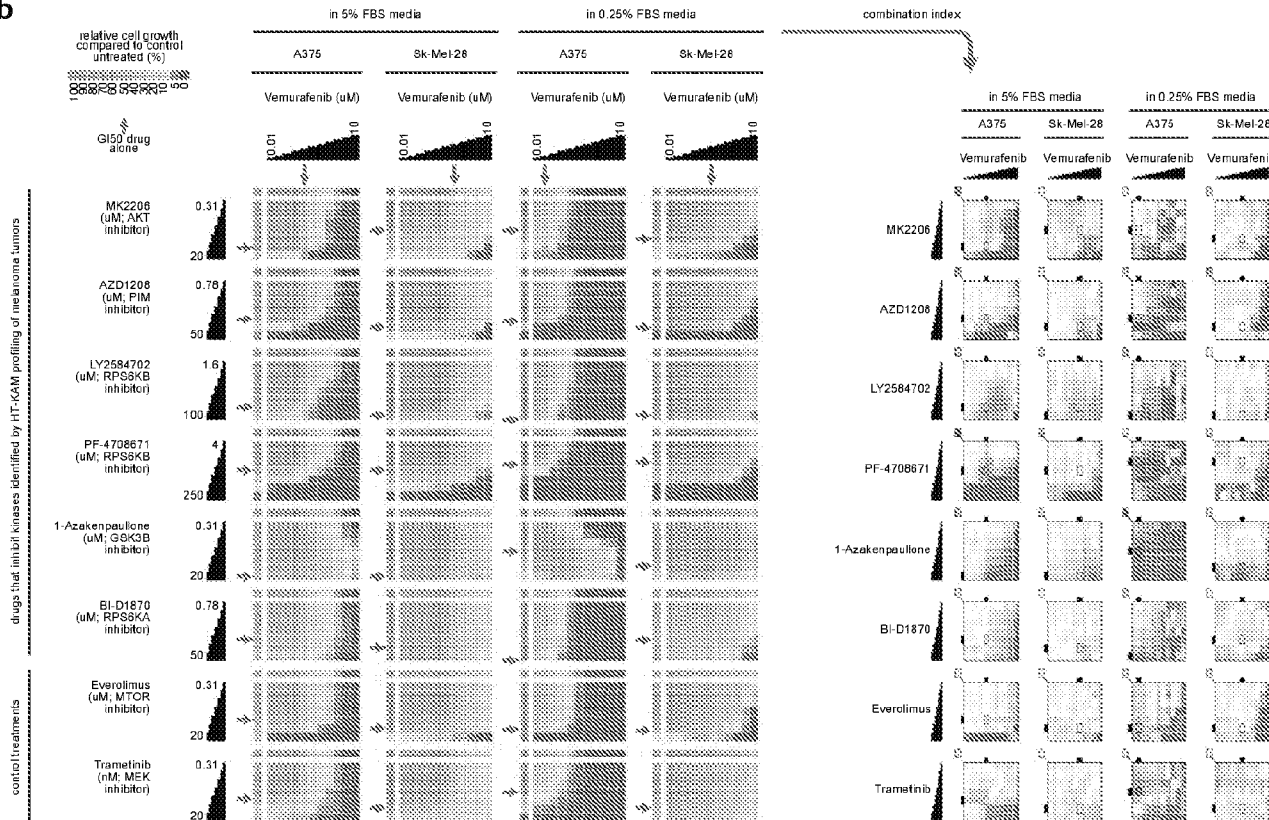
Fig. 27

WO 2019/200245

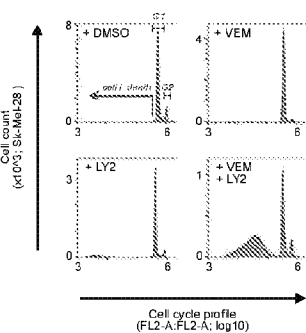
a



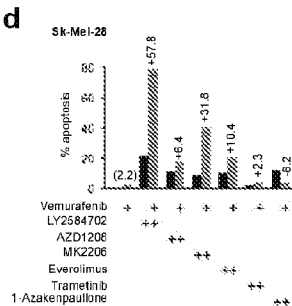
b



c



d



e

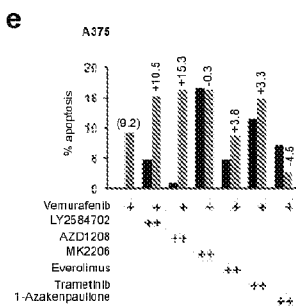


Fig. 28

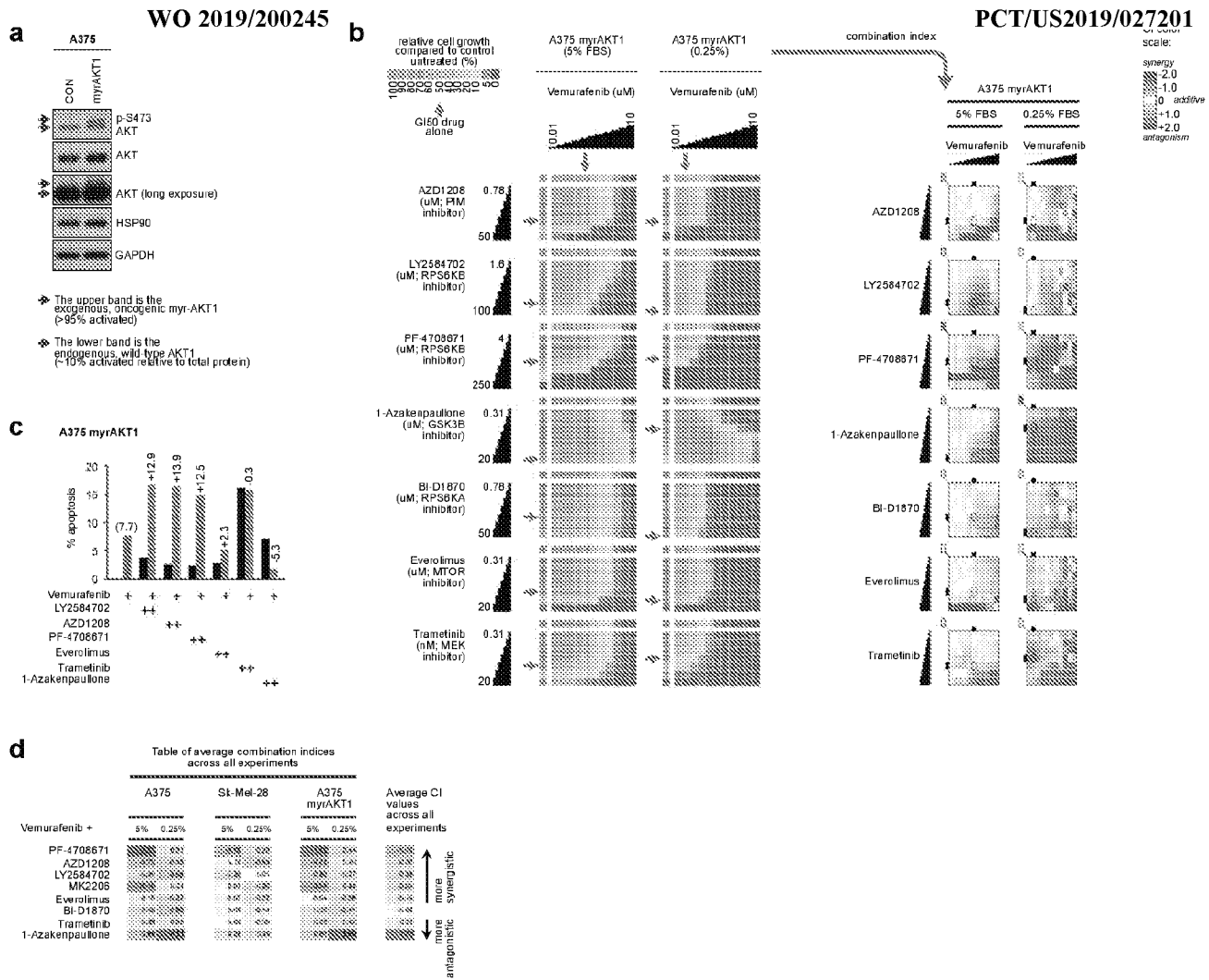


Fig. 29

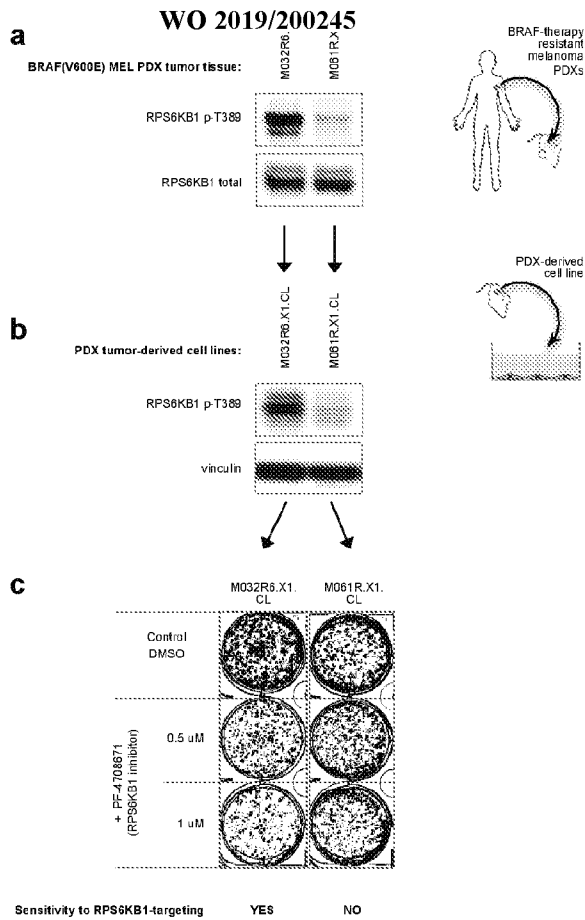


Fig. 30

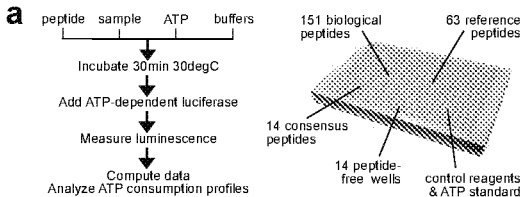


Fig. 1