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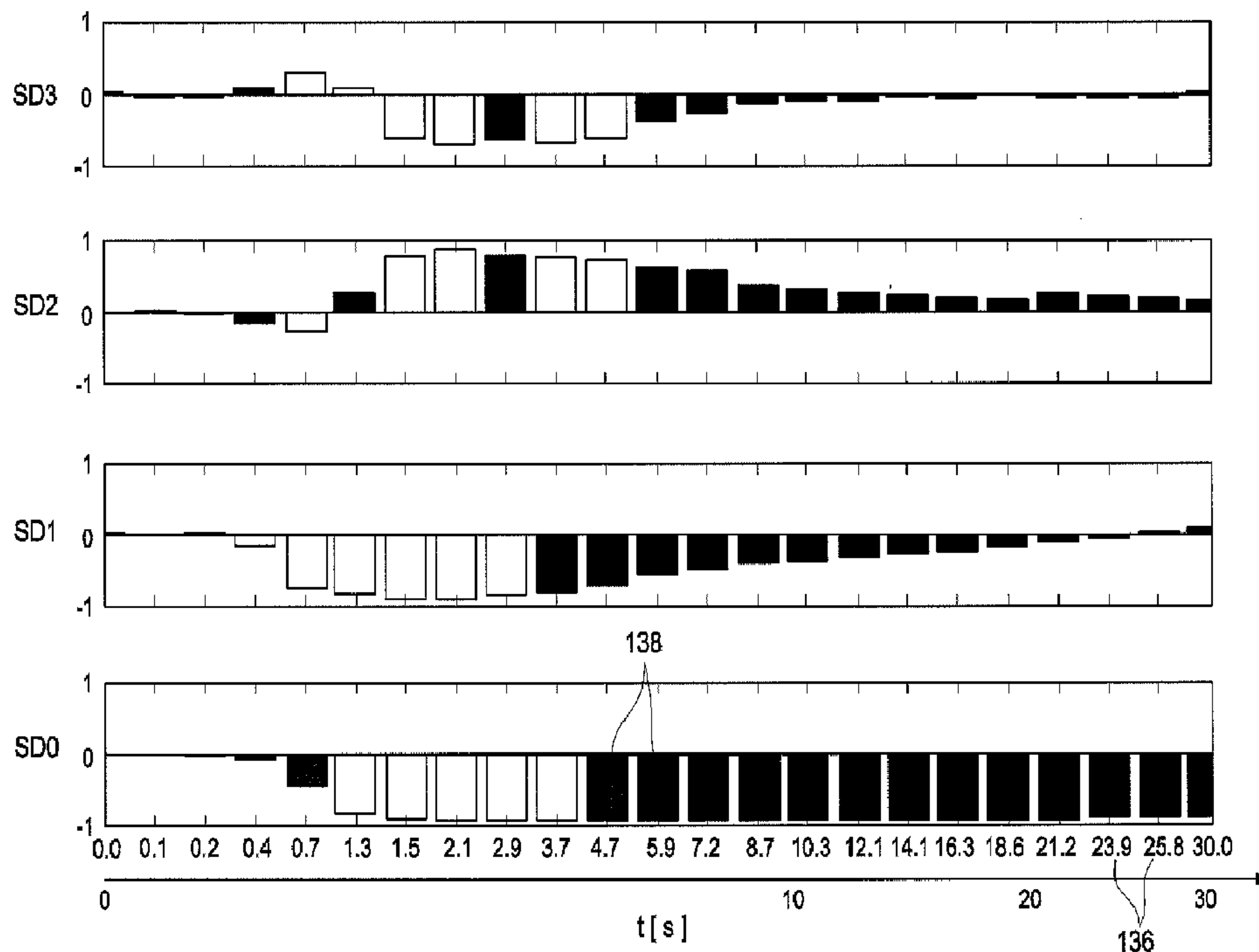
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(54) Titre : PROCEDE D'EVALUATION DE COURBES DE MESURE MEDICALES

(54) Title: METHOD FOR EVALUATING MEDICAL MEASUREMENT CURVES



(57) Abrégé/Abstract:

A method for detecting an analyte in a sample of a body fluid comprising the steps of: a) providing an optical measurement curve, which contains a plurality of measurement values recorded by monitoring a time development of a measurement value indicating a

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(57) Abrégé(suite)/Abstract(continued):

progress of a detection reaction of a test substance and the sample of a body fluid, wherein the detection reaction is known to be influenced by a set of state variables; b) providing a set of at least two different evaluation rules being adapted to derive a characteristic value from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1\dots N}$ from the optical measurement curve; c) performing a multivariate analysis of the first and the second characteristic value by using a predetermined multivariate evaluation algorithm adapted to derive a result from at least two variables, wherein the at least one first and second characteristic value are used as the at least two variables, thereby deriving an estimate value for at least one target variable Y of the state variables; d) determining a concentration of the analyte by using the variable Y.

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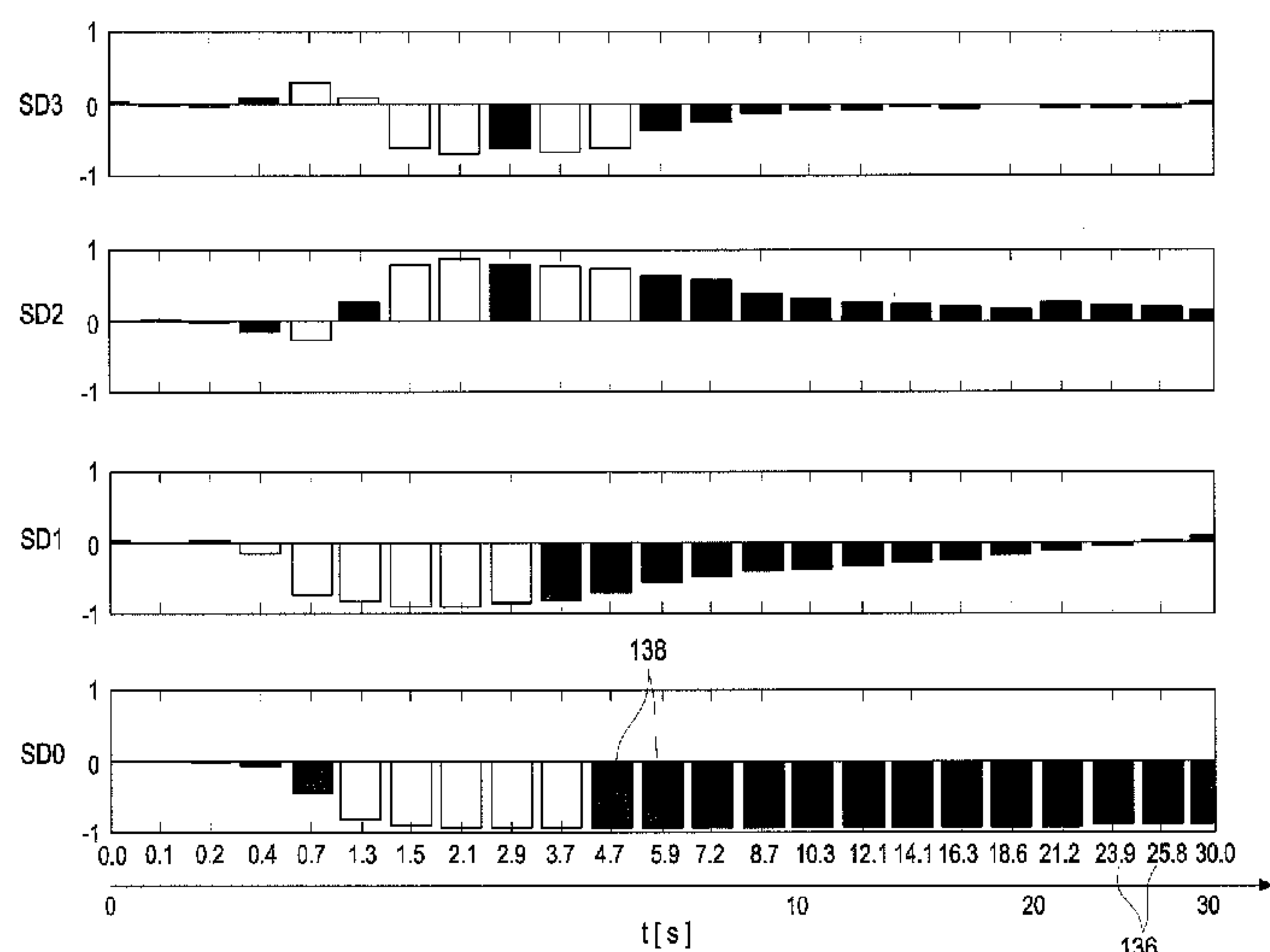


Fig. 3

(57) Abstract: A method for detecting an analyte in a sample of a body fluid comprising the steps of: a) providing an optical measurement curve, which contains a plurality of measurement values recorded by monitoring a time development of a measurement value indicating a progress of a detection reaction of a test substance and the sample of a body fluid, wherein the detection reaction is known to be influenced by a set of state variables; b) providing a set of at least two different evaluation rules being adapted to derive a characteristic value from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1..N}$ from the optical measurement curve; c) performing a multivariate analysis of the first and the second characteristic value by using a predetermined multivariate evaluation algorithm adapted to derive a result from at least two variables, wherein the at least one first and second characteristic value are used as the at least two variables, thereby deriving an estimate value for at least one target variable Y of the state variables; d) determining a concentration of the analyte by using the variable Y.

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Method for evaluating medical measurement curves

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Field of the invention

The present application refers to methods for detecting an analyte in a sample of a body fluid, a computer program including computer-executable instructions for performing the method, an evaluation device and a sample analysis device for characterizing the sample of a body fluid. Methods, computer programs, evaluation devices and sample analysis devices according to the present invention may be used in medical diagnostics, in order to qualitatively and/or quantitatively detect one or more analytes, preferably blood glucose, in one or more body fluids. Additionally or alternatively, however, other types of applications are feasible.

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Related art

In the field of medical diagnostics, in many cases, one or more analytes have to be detected in samples of body fluid, such as blood, interstitial fluid, urine, saliva or other types of body fluids. Examples of analytes to be detected are glucose, triglycerides, lactate, cholesterol or other types of analytes typically present in these body fluids. According to the concentration and/or the presence of the analyte, an appropriate treatment may be chosen, if necessary.

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In devices and methods known from prior art, e.g. in analytical test systems, a time evolution of a measurement signal is observed, e.g. after inserting or positioning of the sample in or on the test system. Usually, devices and methods known to the skilled person make use of test elements comprising one or more test chemistries, which, in presence of the analyte to be detected, are capable of performing one or more detectable detection reactions, such as optically detectable detection reactions.

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In Alexa von Ketteler, Dirk-Peter Herten, and Wolfgang Petrich, *Fluorescence Properties of Carba Nicotinamide Adenine Dinucleotide for Glucose Sensing*, CHEMPHYSICHEM 2012, 13, 1302-1306 spectroscopic properties of highly purified cNADH could be determined for the first time and were compared to those of NADH. Carba Nicotinamide Adenine Dinucleo-

- 2 -

tide (cNAD) may serve as a stable co-factor for an enzyme-based detection of glucose. When changing from NADH to cNADH, an about 50% increase in quantum efficiency could be observed, which – together with a large excitation wavelength and a higher stability – should make cNAD a well suited alternative as coenzyme for robust glucose detection.

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EP 2 006 786 A1 describes a method and a glucose monitoring system for monitoring individual metabolic response and for generating nutritional feedback. The method comprises the step of consecutively performing a plurality of measurements of a glucose level in a qualified subject by a measuring device. In the measuring device first data corresponding to a measured glucose level is generated. This data is further transmitted to an analysis device. There, second data is generated representing at least one measure for variability of a glucose level of the subject from a time-series of glucose measurement represented by the first data. Embodiments are explained by presenting phase space diagrams, e.g. by a phase space diagram representing the glucose value as well as the first derivative of the glucose progressions. Suitable frequencies for taking measurements are from 4 (i.e. a measurement every 15 minutes) to 60 (i.e. a measurement every minute) measurements an hour.

A representation of a function with their first derivative is well known in physics, e.g. in presentations of space and momentum, e.g. described in Gerthsen, Kneser, Vogel, *Physik*, Springer-Verlag, 16. Auflage, Kap. 17.2.7, Seite 868-870.

EP 2 006 786 A1 furthermore suggests to visualize a BG risk index, a measure for an overall risk of extreme blood glucose over time in the form of a phase space diagram.

EP 1 702 559 B2 describes a method and a system for analysis of the glucose metabolism. This European patent is not related to the actual measurement of a glucose concentration but to a processing of data points. The data points may be processed by plotting a graphic phase space representation that may be provided to a physician as diagnostic aid. A function of the blood glucose concentration and/or at least one of its derivatives with respect to time may be used as phase space coordinates. The glucose concentration or a rate of change of the glucose concentration may be one of the phase space coordinates. A trajectory in phase space described by a sequence of data points may be analyzed in order to determine a disturbance parameter. The disturbance parameter may indicate which state of a disease of glucose metabo-

- 3 -

lism is likely to be manifest. With progress of an illness, a regulation mechanism for an adjustment of the blood glucose concentration may be useful.

US 6,448,067 B1 relates to a method and an apparatus for determining a chemical component
5 from a sample, where the sample is measured on a test strip with the help of a color reaction. The method concerns finding the measurement time for measuring the reflection of the test strip. When a function reaches a predetermined value, the method calls for determining a measuring time, and determining from a reflection value measured at the measuring time a content of the chemical component in the sample, preferably the glucose content of a blood
10 sample.

US 2009/0177406 A1 describes a slope-based compensation. A biosensor system determines an analyte concentration from an output signal generated from a light-identifiable species or a redox reaction of the analyte. The biosensor system adjusts a correlation for determining ana-
15 lyte concentrations from output signals with one or more index functions extracted from the output signals. The index functions determine at least one slope deviation value or normalized slope deviation from one or more error parameters. The slope-adjusted correlation between analyte concentrations and output signals may be used to determine analyte concentrations having improved accuracy and/or precision from output signals including components at-
20 tributable to bias. One or more slope deviation values may be determined that are responsive to one or more errors. Slope deviation values may be determined for temperature, hematocrit, and other contributors. In one example, the analyte generates output signals in response to a pulse sequence of a gated amperometry electrochemical analysis. A ratio parameter may represent a relationship between current generated by the analyte in response to two pulses of a
25 gated amperometry pulse sequence. A %-bias of a measured analyte concentration in a biological fluid may be determined from or correlated with the output signals of the analysis, such as the intermediate currents generated by the analyte in response to a gated amperometry sequence. Output signals may be currents or potentials generated from the sample that are responsive to the input signal. Signals that are near or partially linear may be used. The %-
30 bias in the correlation of analyte concentrations with output signals may be represented by one or more slope deviations obtained from one or more error parameters. Index functions may compensate the measured analyte concentration for one or more errors in the analyte concentration analysis. One or more index functions may be used. Index functions may be

experimentally determined as regression equation. The compensation or correction of the analyte concentration value may be started with the error parameter accounting for the largest error in the output signal. After compensating for the largest effect, any error remaining may be compensated or corrected with additional error parameters independent of the parameter responsive to the largest error. Successive corrected analyte concentration values may be determined using additional index functions, the bias in the determined concentration values may decrease until the bias level approaches the random noise level of the analysis.

In WO 2006/138226 A2, an arrangement and an algorithm for calculating the concentration of an analyte contained in a sample are disclosed. Therein, a color change rate of a test chemical is detected, and a hematocrit is derived from the color change rate. An appropriate correction factor indicative of the hematocrit is used for correcting a glucose concentration.

A test strip known from prior art may be the one used with the Accu Chek[®] Active system by Roche Diagnostics Deutschland GmbH.

US 5,420,042 A discloses a method for the analytical determination of the concentration of a component of a medical sample, in which a reaction of the sample with reagents leads to a time-dependent change in a measured quantity. Therein, the concentration correlates according to an evaluation curve with an input variable derived from a time-dependent change, wherein the calibration curve is ambiguous for at least a portion of the possible values of the input variable. In order to obtain an unambiguous correlation to a particular concentration, a training run and an analysis run are separately performed.

WO/2011/131490 describes a method for determining the concentration or activity of an analyte in a sample, particularly an automated algorithm for the quality control of reactions. Therein, plotting the fluorescence intensity of a reporter dye divided by the fluorescence intensity of a passive reference dye against the cycle number leads to a so-called sigmoid function which is characterized by a background phase, an exponential growth phase and a plateau phase. Since the fluorescence intensity as a function of cycles relates to the initial number of template molecules in the sample, the curves can be used to quantify the amount of fragments in the sample by determination of a specific value.

- 5 -

In US 2008/0070234, A1 a method for the quantitative determination of an analyte in a sample is disclosed. Therein, at least two calibration graphs are provided, wherein the calibration graphs have been generated by reacting in each case the same analyte-specific substance with different amounts of in each case the same test analyte for in each case a predetermined reaction time.

WO 2012/084194 A1 describes a method for determining an analyte concentration, wherein a first electrical potential excitation pulse to a body fluid sample in an analyte sensor is applied, and wherein a first current response of the body fluid sample to the first pulse is recorded. Thereafter, a second excitation pulse is applied to the body fluid sample in the analyte sensor, and a second current response of the body fluid sample to the second pulse is recorded. An analyte level in the body fluid sample is determined by compensating for sources of error based on the first current response to the first pulse.

In US 2007/0235346 A1, a method and a device for determining the concentration of an analyte constituent in a physiological sample are provided. For this purpose, an electrochemical signal based on a reaction taking place inside an electrochemical cell is recorded, from which a preliminary concentration of the analyte is derived. The preliminary concentration is then multiplied by a hematocrit correction factor in order to obtain the constituent concentration in the sample, wherein the hematocrit correction factor is a function of the electrochemical signal.

A measurement of glucose may be influenced by different hematocrit values. In methods and devices known from prior art, a point of time development of a remission signal, e.g. a kinetic curve, may be determined wherein at this point the gradient of the kinetic curve after wetting may decrease under a predetermined value, e.g. 2%/s. This criterion is also known as final value criterion or end value criterion. A remission value at this point in time, an end value, may be used in relationship to a value, e.g. a remission value, before the wetting, e.g. a blank value. A ratio between these two values, e.g. a relative remission, may be used, e.g. in combination with a calibration curve, e.g. a code curve, for determining a glucose value, e.g. the concentration of glucose.

These kinds of methods are e.g. described in WO 2008/135128 A1. WO 2008/135128 A1 in particular describes an analysis device with a test carrier for a photometric determination of an analyte in a body fluid. A control value of a control parameter is detected at one detection point during a processing after a calculation step, an error in the measurement and evaluation
5 unit is recognized if the deviation of the control value from the expected value exceeds a pre-defined threshold value.

In methods and devices known from prior art, e.g. in analytical test systems known from prior art, a propagation in time of a measuring signal, which may be observed after an intake or an
10 application of the sample usually does not necessarily only have a dependency on the observed analyte, in particular on a presence and/or on a concentration of the analyte. E.g., it can be experimentally observed that in the context of e.g. a cNAD-developing, e.g. in the context of one-way test stripes for detecting a glucose concentration in blood, a time propagation of a remission may also be dependent on a temperature, e.g. on a temperature being present during
15 a measurement, and probably also on an air humidity during the measurement and/or on a concentration of hematocrit of the sample. For example, the end value of a remission measurement may depend on the temperature. According to prior art, the end value of a remission measurement may be used to derive a glucose concentration by using a calibration curve. A temperature dependency, e.g. for a glucose concentration of 250 mg/dL, may be more than
20 2% per Kelvin.

The measurement signal also may be dependent on different other state variables, e.g. on a storage time of a reagent kit, e.g. on a storage time of the test strip, and/or on a "history" of the test strip and/or on the current state of the test strip. An example may be a change of a dry
25 test chemical caused by or after one or more re-moisturizations. Properties of the measurement signal and/or another signal may be influenced by substances, e.g. substances like acetylsalicylic acid (ASS) and/or citric acid, which may disturb the signal and/or the measurement signal. Alternatively or additionally, similar analytes may change the signal and/or the measurement signal in a noticeable way, wherein the similar analytes may comprise e.g.
30 maltose and/or xylose. This may cause interferences and/or errors.

A main disadvantage of methods known from prior art may be, that the end value may not only be dependent on the glucose concentration but also on at least one more state variable, as

- 7 -

e.g. discussed above. There may be a deviation of a glucose value determined by using the end value criterion from a desired value for different concentrations of hematocrit. A prediction of a glucose concentration according to prior art may comprise deviations being dependent e.g. on hematocrit.

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Problem to be solved

It is therefore an object of the present invention to provide methods and devices which at least partially avoid the disadvantages and shortcoming of the systems and methods known from the prior art. Specifically, it is an object of the present invention to provide methods and devices which enable detecting an analyte in a sample of a body fluid, wherein the detection of the analyte may be less influenced by state variables, e.g. as by at least one temperature and/or at least one air humidity and/or a hematocrit concentration and/or a history of a test strip.

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Summary of the invention

This problem is solved by methods, a computer program, an evaluation device and a sample analysis device with the features of the independent claims. Preferred embodiments, which might be realized in an isolated fashion or in an arbitrary combination, are listed in the dependent claims.

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As used in the following, the terms “have”, “comprise” or “include” or any grammatical variations thereof are used in a non-exclusive way. Thus, these terms may both refer to a situation in which, besides the feature introduced by these terms, no further features are present in the entity described in this context and to a situation in which one or more further features are present. As an example, the expressions “A has B”, “A comprises B” and “A includes B” may both refer to a situation in which, besides B, no other element is present in A (i.e. a situation in which A solely and exclusively consists of B) and to a situation in which, besides B, one or more further elements are present in entity A, such as element C, elements C and D or even further elements.

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In a first aspect of the present invention, a method for detecting an analyte in a sample of a body fluid is disclosed. The term “detecting” may refer to a process of generating information

on the presence of the analyte in the sample of the body fluid. The information preferably may be quantitative information, i.e. information on a concentration of the analyte. Therein, the concentration generally may be given in arbitrary units, referring to an amount of the analyte per amount of the sample of the body fluid, e.g. in absolute units and/or as a percentage. The amount of the analyte may be given in mass units and/or mol and/or volume units. Typical examples of units of concentrations may be: mg/dl (milligrams per deciliters) and/or mg/kg (milligrams per kilograms) and/or % by weight and/or vol.-%. Other units may be possible.

The analyte generally may comprise an arbitrary analyte which may be present in the body fluid. Specifically, the analyte may be a metabolite and/or may be an analyte which may take part in the metabolism of a human or an animal. Preferably, the analyte may be or may comprise glucose. However, additionally or alternatively, other types of analytes may be detected, such as lactate and/or triglycerides. The sample may be an arbitrary sample, e.g. a certain volume and/or a certain percentage and/or a certain part of the body fluid. The sample e.g. may be a drop and/or a distillate and/or a filtered part of a body fluid and/or a separated part of the body fluid. The separation may comprise a separation by mass, e.g. a centrifugation.

Preferably, the body fluid may be selected from the group consisting of blood, such as whole blood, and interstitial fluid. However, generally, one or more other types of body fluids may be used, such as urine or saliva.

The method comprises the following steps. These steps may be method steps. The steps preferably may be performed in the given order. However, other orders of the steps may be feasible. Preferably each of the following steps may just be performed once, e.g. none of the following steps may be repeated. Each step preferably only may be performed once. One or more or even all of the steps may be performed repeatedly, e.g. by repeating one of the steps and/or by repeating more than one of the steps and/or by repeating even all of the steps, e.g. once and/or twice and/or even more than twice. Further, two or more of the steps may overlap in time, e.g. by performing two or more of these steps at least partially simultaneously. One of the steps or a plurality of the steps or even all of the steps may be performed by using a data processing device such as a computer, preferably a microcomputer and/or an application-specific integrated circuit (ASIC).

The steps are as follows:

- a) providing at least one optical measurement curve, wherein the optical measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid. The measurement values contained in the optical measurement curve are acquired at differing points in time. The detection reaction is known to be influenced by a set of state variables. Each state variable characterizes at least one of a state of the sample of the body fluid and a condition of the detection reaction;
- b) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1..N}$ from the optical measurement curve. The set of characteristic values comprises at least one first characteristic value being derived from the optical measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value being derived from the optical measurement curve by using at least one second evaluation rule from the set of evaluation rules. The second evaluation rule is different from the first evaluation rule;
- c) performing at least one multivariate analysis of the at least one first characteristic value and the at least one second characteristic value by using at least one predetermined multivariate evaluation algorithm. The multivariate evaluation algorithm is an algorithm adapted to derive at least one result from at least two variables. The at least one first characteristic value and the at least one second characteristic value are used as the at least two variables, thereby deriving at least one estimate value for at least one target variable Y of the state variables;
- d) determining a concentration of the at least one analyte by using the at least one target variable Y .

The optical measurement curve may be an amount of data characterizing a time development of a detection reaction. The optical measurement curve may contain a plurality of measurement values recorded at differing points in time. The optical measurement curve optionally additionally may contain the respective measurement times of the measurement values, such as by containing data pairs (R_i, t_i) and/or $(t_i, R_i(t_i))$. The optical measurement curve may be a measurement curve preferably comprising measurement values representing an optical physical quantity. The optical measurement curve may comprise a plurality of measurement values selected from the group consisting of: remission values; relative remission values; transmission values; absorption values; fluorescence values; intensity values; frequency values; spectroscopic values; spectral values; coherence values; decoherence values; photon numbers.

The measurement value may be a quantifiable measurement result R_i , recorded by at least one arbitrary measurement method based on at least one of a physical and/or chemical and/or biological measurement principle, preferably an optical measurement principle, such as a reflection measurement, a fluorescence measurement or any other type of optical measurement, as will be explained in further detail below. The measurement method may be selected from the group consisting of: light detection, e.g. by at least one photodiode and/or by at least one CCD camera; spatial light detection, e.g. by at least one CCD camera; light frequency measurements, e.g. by spectroscopy and/or by a light beating method, preferably combined with a Fourier analysis; comparisons of a color of at least a part of a test strip with at least one reference color bar; measurements of an intensity with an optical power meter; measuring a frequency by using a wavemeter; analysis of camera images.

The term “monitoring a time development” may refer to a process of acquiring the plurality of measurement values at different measurement times. During monitoring a time development, the measurement curve, preferably at least the optical measurement curve, may be generated. The monitoring of the time development may solely comprise the acquisition of the measurement values, without acquiring the additional information of the measurement times of the respective measurement values. Thus, as an example, this option may be feasible when using a constant acquisition frequency. As an alternative option, additionally, the monitoring of the time development may comprise the acquisition of the respective measurement times, thereby e.g. generating the measurement curve, preferably the optical measurement curve, comprising data pairs (R_i, t_i) and/or $(t_i, R_i(t_i))$.

- 11 -

Generally, as used herein, the term “detection reaction” may refer to an arbitrary type of chemical reaction of at least one test substance and the sample of the body fluid. As an example, reference may be made to the test substances disclosed in the prior art documents cited above. Additionally or alternatively, other types of test substances may be used. The detection reaction may be adapted to generate a measurement value and/or to generate a change in time of the measurement values and/or to generate analysis information. Preferably, the detection reaction may be a chemical reaction between at least one component of the test substance which may be adapted to indicate a presence and/or a concentration of the at least one analyte in the body fluid. Thus, generally, the test substance may be a chemical compound and/or a chemical mixture which may be adapted to react with the at least one analyte to be detected, preferably in a highly analyte-specific fashion. The detection reaction preferably may be embodied such that a test substance may react with the at least one analyte to be detected and, thereby, may fully or in part change by itself, e.g. may transform into another chemical species and/or may transform its surrounding in a detectable way, which may be measured, thereby deriving the plurality of measurement values and the measurement curve, preferably the optical measurement curve. The progress of the detection reaction may be indicated by at least one physical measurement value and/or a change in at least one physical measurement value, which may be used as the measurement value. Preferably, the detection reaction may be an optically detectable detection reaction, which may be optically observable, such as by using a reflection measurement and/or a transmission measurement. Other types of optical measurements may be feasible.

The term “test substance” generally may refer to a chemical compound or substance or a mixture of two or more chemical compounds or substances adapted for performing the detection reaction, preferably an analyte-specific detection reaction, most preferably an optically observable detection reaction. Preferably, the test substance may comprise one or more enzymes adapted to react with the at least one analyte to be detected. Additionally, the test substance may comprise one or more auxiliary components, such as mediators and/or co-enzymes. For test substances which may also be used within the present invention, reference may be made to the test substances known from prior art, such as the cNAD test substances. The test substance may comprise one or more test chemicals.

As a first example of a test chemical which may be used within the present invention, reference may be made to the so-called "PQQ chemistry". This PQQ test chemistry is disclosed e.g. in EP 0 354 441 A2 and may contain a PQQ-dependent dehydrogenase and a direct electron acceptor which may be an aromatic nitroso compound or an oxim. Further, one or more indicators may be present, such as one or more dyes. Thus, as an example, heteropoly blue indicator may be used, as disclosed in EP 0 431 456 A1.

As a second example of a test substance which may be used within the present invention, reference may be made to the so-called "cNAD chemistry", as disclosed in WO 2007/012494 A1, WO 2009/103540 A1, WO 2011/012269 A2, WO 2011/012270 A1 and WO 2011/012271 A2. Thus, in WO 2007/012494 A1, cNAD derivatives are disclosed. In WO 2009/103540 A1, stabilized enzyme/coenzyme complexes are disclosed. In WO 2011/012269 A2, WO 2011/012270 A1 and WO 2011/012271 A2, the synthesis of cNAD and cNAD/derivatives as well as intermediates/precursors is disclosed.

Additionally or alternatively, other types of test substances may be used.

The term "differing points in time" may refer to the fact that at least two of the measurement values may be recorded at different points in time. The optical measurement curve may comprise discrete measurement values for different points in time. Alternatively, the optical measurement curve may be continuous in time. The optical measurement curve may be generated by extrapolation of discrete measurement values, e.g. by using at least one theoretical model and/or by using at least one fitting method.

The state variable may be a variable characterizing at least one of a state of the sample of the body fluid and/or of the test substance and/or of a test strip and/or a condition of the detection reaction. Examples for a state variable may be the concentration of the analyte in a body fluid, e.g. a concentration of glucose; a content of at least one component of the sample of the body fluid, such as a content of a particulate component, preferably hematocrit and/or a concentration of hematocrit; a temperature of the sample of the body fluid; a humidity of an ambient atmosphere surrounding the sample of the body fluid; a storage time of the test substance; a storage time of the test strip; an illumination of an environment of the detection reaction and/or of the test strip; a volume of the sample of the body fluid. The state of the sample of

- 13 -

the body fluid may be e.g. a temperature of the sample of the body fluid and/or a humidity of the sample of the body fluid and/or a volume of the sample of the body fluid and/or an age of the sample of the body fluid and/or a storage time of the sample of the body fluid and/or a purity of the sample of the body fluid. The condition of the detection reaction may comprise a state of an environment of the test chemistry and/or a property of the detection reaction. The condition of the detection reaction may comprise a temperature during the detection reaction and/or a temperature change during the detection reaction and/or a humidity during the detection reaction and/or a presence of a catalyst during the detection reaction.

At least one of the steps, e.g. one of the steps or even all of the steps, may be performed by using at least one computer.

An evaluation rule may be a rule for directly or indirectly deriving at least one characteristic value from the measurement curve, preferably from the optical measurement curve. According to the present invention, at least two different evaluation rules are used, which, preferably, are applied to one and the same optical measurement curve and/or one and the same measurement curve, e.g. in order to derive the at least two characteristic values from the optical measurement curve and/or from the measurement curve, preferably from one and the same optical measurement curve. The evaluation rule generally may comprise an arbitrary mathematical algorithm or an arbitrary combination of algorithms for deriving one or more numeric values from the optical measurement curve and/or from the measurement curve, e.g. from at least one part of the measurement curve or of the optical measurement curve, e.g. from one time interval of the measurement curve or of the optical measurement curve, or from at least one curve, curve part or data derived from the optical measurement curve and/or from the measurement curve.

At least one of the evaluation rules may be and/or may comprise the end value criterion as described above.

Alternatively, the evaluation rule may be different from the end value criterion. E.g., the first evaluation rule and/or the second evaluation rule may not comprise an end value criterion.

The term “different evaluation rules” as used herein, may refer to the fact that the algorithm of the first evaluation rule may differ from the algorithm of the at least one second evaluation rule in at least one coefficient and/or in at least one parameter and/or in at least one other component defining the algorithm. The term “different evaluation rules” may comprise the fact that the algorithm of the first evaluation rule may differ from the algorithm of the second evaluation rule only in a point in time, e.g. in the parameter t . Thus, two evaluation rules may differ from each other just by the difference of the time parameter t . Alternatively, the two evaluation rules may differ from each other by applying two different components related to the algorithm, such as two different thresholds or two different change rates below a predetermined threshold.

The characteristic value may be a specific numeric value derived from the optical measurement curve and/or from the measurement curve by using at least one evaluation rule, preferably by using one evaluation rule. The characteristic value e.g. may be e.g. a correlation coefficient and/or a concentration.

N may be a natural number, preferably an arbitrary natural number. The second evaluation rule may differ from the first evaluation rule just by a point in time. Additionally or alternatively, the second evaluation rule may differ from the first evaluation rule by at least one parameter being different from the point in time and/or by at least one different algorithm. As a particular example, N may be selected as 3, such that a third evaluation rule may be provided, wherein the third evaluation rule may be different from both, the first evaluation rule and the second evaluation rule. Hereby, it is particularly preferred that the the algorithm of the first evaluation rule may differ from the algorithm of the second evaluation rule only in a point in time, e.g. in the parameter t . Alternatively, the two evaluation rules may differ from each other by any other feature as described above or below. Within this particular embodiment, the at least one first characteristic value may be derived from the first evaluation rule according to step c). However, during step c), depending on the at least one first characteristic value, such as its absolute value, its relative value, and/or a predetermined range wherein the at least one first characteristic value may be found inside or outside, either the second evaluation rule or the third evaluation rule may be used in a further calculation employing the multivariate evaluation algorithm.

- 15 -

The multivariate analysis may comprise at least one mathematical operation. A multivariate analysis may be performed of at least two different numeric values. The multivariate evaluation algorithm may be an arbitrary algorithm for deriving at least one numerical result, also referred to as an output, from at least two variables and/or values, preferably from the at least one first characteristic value and the at least one second characteristic value, also referred to as input variables. The multivariate evaluation algorithm may be an algorithm adapted to derive at least one result from the at least two variables. The algorithm may comprise an arbitrary rule for generating the output by using the at least two input variables. The output may be the target variable Y and the input variables may be the characteristic values. Preferably, the multivariate evaluation algorithm is or may comprise at least one equation. At least two variables, preferably the first characteristic value and the second characteristic value, may be input variables. The input variables may be used or may be combined in this equation, e.g. in order to derive the result. The multivariate evaluation algorithm preferably may comprise one of a linear equation, a quadratic equation, a cubic equation or any other polynomial equation using the at least two variables, preferably the first characteristic value and the second characteristic value, and e.g. a plurality of coefficients, thereby deriving the at least one result.

The multivariate analysis may be a process or a mathematical operation using the multivariate evaluation algorithm and at least two input variables, preferably the first characteristic value and the second characteristic value, specifically the characteristic values, for generating at least one numerical result, specifically the estimate value for the at least one target variable. The multivariate evaluation algorithm may be or may comprise a one-step algorithm in which the first characteristic value and the second characteristic value may be used as input variables for one and the same algorithm, such as using one and the same equation having the first characteristic value and a second characteristic value as input variables. Alternatively, the multivariate evaluation algorithm may be or may comprise multiple steps, wherein, step-by-step, two or more algorithms may be successively applied.

The multivariate analysis may be an analysis and the multivariate evaluation algorithm may be an algorithm as e.g. described in Harald Martens and Tormod Næs, *Multivariate Calibration*, John Wiley and Sons Ltd, 1998, p. 97-165 and/or René Henrion and Günter Henrion, *Multivariate Datenanalyse*, Springer-Verlag, 1995, p. 103-157.

The estimate value may be a value of the at least one target variable Y rendered by a method for evaluating the optical measurement curve and/or the measurement curve. The value may be believed to optimally quantify the target variable. The estimate value may be a numeric value.

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The target variable Y may be a state variable of specific interests. The target variable Y may be derived in a method according to the present invention. Generally, the target variable Y may be or may comprise an arbitrary state variable. The target variable Y even may change during the evaluation. Thus, one and the same optical measurement curve and/or one and the same measurement curve may be evaluated in order to derive different target variables, even by using the same multivariate evaluation algorithm, which may be one of the major advantages of the present invention. As an example, by using the same measurement curve, preferably the same optical measurement curve, both the concentration of glucose and the hematocrit may be derived as target variables. Alternatively, only the concentration of glucose or only the hematocrit may be derived as target variables. The target variable Y may be a scalar or a vector or a matrix.

The concentration of the at least one analyte may be an arbitrary concentration. Preferably, the concentration of the at least one analyte may be the concentration of glucose, preferably the concentration of blood glucose. A typical unit of the concentration of blood glucose may be mmol/l or mg/dl. More than one target variable Y may be combined to a vector.

The term “determining the concentration of the at least one analyte by using the at least one target variable” may comprise different options. The target variable Y may be the concentration of the at least one analyte or may comprise the concentration of the at least one analyte. Additionally or alternatively, the target variable Y may be another target variable Y being different from the concentration of the at least one analyte. The target variable Y may comprise a target variable Y being different from the concentration of the at least one analyte. The target variable Y may be or may comprise a target variable Y being independent from the concentration of the at least one analyte. E.g., the target variable Y being derived in step c) may be the concentration of the at least one analyte or if the target variable Y derived in step c) comprises the concentration of the analyte, nothing may have to be done in step d), or just a simple calculation. If the target variable Y derived in step c) is not the concentration of the at

- 17 -

least one analyte or does not comprise the concentration of the at least one analyte, the concentration of the at least one analyte may be determined in step d) by using the target variable. E.g., a raw value of the concentration of the at least one analyte may be determined, e.g. by using an observable and/or a measurement value. The raw value may be corrected by using a correction algorithm, wherein the correction algorithm may use the target variable Y derived in step c). The correction algorithm may be or may comprise a temperature correction and/or a hematocrit correction. The raw value may be transformed to a corrected value of the concentration of the at least one analyte by using the correction algorithm. E.g., an observable and/or a specific measurement value may be generated out of the optical measurement curve and/or out of the measurement curve, e.g. by using a method known from prior art, preferably by using a remission value at an end time value, most preferably by using a remission end value. The end time value may be the time at which the optical measurement curve, preferably a remission curve, may have a slope being smaller than a specific threshold. The threshold may be 2%/s. The specific measurement value or the remission value may be transformed into the raw value by using a transformation algorithm. The raw value may be transformed into the corrected value, e.g. by using a second transformation algorithm, e.g. the correction algorithm. Alternatively, a method may be used using only one transformation algorithm, e.g. in a one step method, wherein the measurement value and e.g. at least one target variable Y may be transformed by using the transformation algorithm. The transformation may lead directly to the corrected value of the concentration of the at least one analyte. The measurement value here may be e.g. a remission value being determined out of a measurement curve and/or out of the optical measurement curve at a specific point in time, e.g. the remission end value. Other options may be possible.

The state variables may be selected from the group consisting of: a composition of the sample of the body fluid, preferably a content of at least one component of the sample of the body fluid and more preferably a concentration of at least one analyte; a content of at least one particulate component of the sample of the body fluid, preferably a hematocrit; a temperature of the sample of the body fluid; a humidity of an ambient atmosphere surrounding the sample of the body fluid; a storage time of the test substance; a storage time of a test strip comprising the test substance; an interfering substance such as ascorbate; alterations of the sample or of certain properties of the sample caused by pharmacological treatment of a donor of the sample. The composition of the sample of the body fluid may comprise different contents of at

least one component. The composition of the sample of the body fluid may comprise variations of the sample from an average composition of a sample. The composition of the sample may comprise variations of concentrations of blood components, e.g. a hematocrit value being higher or lower as in average blood samples.

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The first evaluation rule may not be transformed into the second evaluation rule by a time transformation. Preferably, the first evaluation rule may not be transformable into the second evaluation rule by a time transformation, preferably not by a linear time transformation. The time transformation may comprise a change of a point in time. The first evaluation rule may not just be different from the second evaluation rule by application of identical algorithms for different times t . The first evaluation rule and the second evaluation rule may have different algorithmic components or may comprise different parameters, wherein the parameter is not the time t or a point in time.

15 The first characteristic value may be determined by using a first time interval of the optical measurement curve. The second characteristic value may be determined by using a second time interval of the optical measurement curve. The first time interval of the optical measurement curve may be different from the second time interval of the optical measurement curve. Alternatively, the first characteristic value may be determined by using a first time interval of a measurement curve, wherein the second characteristic value may be determined by using a second time interval of the measurement curve, wherein the first time interval of the measurement curve may be different from the second time interval of the measurement curve. The first time interval may overlap with the second time interval. Alternatively, the first time interval may be completely separated from the second time interval. A time interval, e.g. the first time interval and/or the second time interval, may be an interval in time, e.g. comprising at least two measurement values.

The target value may be different from the concentration of the at least one analyte. Alternatively, the target value may be the concentration of the at least one analyte or may comprise the concentration of the at least one analyte. E.g., the target value may be a different physical and/or chemical quantity as the concentration of the at least one analyte or the target value may be the same physical or chemical quantity as the concentration of the at least one analyte.

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- 19 -

The at least two evaluation rules may be adapted to derive the characteristic values from at least two derivatives of the optical measurement curve. A derivative may be a derivative of arbitrary order. The derivative e.g. may be a 0-th order derivative. Preferably, at least one of the derivatives may be a higher order derivative, thus, preferably, a first derivative, a second derivative or a higher order derivative. E.g., the first evaluation rule may be used for generating the first characteristic value out of a n-th derivative of the optical measurement curve or of the measurement curve. n may be higher or equal to zero, preferably higher or equal to one. The second evaluation rule preferably may be used to generate the second characteristic value out of a m-th derivative of the optical measurement curve or of the measurement curve. m may be higher or equal to zero, preferably higher or equal to one. Preferably, n may be different from m. The 0-th order derivative may be the identity of the optical measurement curve or of the measurement curve. The first derivative of the optical measurement curve or of the measurement curve may be the slope of the optical measurement curve or the slope of the measurement curve.

The at least two derivatives may be derivatives comprising at least two derivatives of different order. E.g., the first evaluation rule may be a first derivative of the optical measurement curve and the second evaluation rule may be a second derivative of the optical measurement curve.

The derivatives may be generated by using at least one filtering algorithm, preferably a Savitzky-Golay filtering algorithm. The filter algorithm may be a numerical filtering algorithm. The filtering algorithm may be an algorithm being able to derive and/or generate at least one of the derivatives. Other types of algorithms may be possible.

The set of characteristic values may contain 2-20 characteristic values, preferably 3-10 characteristic values. Alternatively, even more characteristic values may be possible. It may be advantageous to minimize the number of characteristic values for saving storage space and/or for simplifying the calculation and/or for saving calculation time.

The target variable Y may comprise the concentration of the at least one analyte in a sample of the body fluid. The target variable Y preferably may comprise the concentration of glucose and/or the concentration of hematocrit.

In step d), in addition to the at least one target variable Y, at least one electrochemical measurement value may be used for determining the concentration of the analyte. The electrochemical measurement value may be determined by using at least one electrochemical measurement. The electrochemical measurement may be a measurement being able to generate electrochemical measurement values. E.g., the measurement curve as described above may be an electrochemical measurement curve. The electrochemical measurement curve may comprise electrochemical measurement values. The electrochemical measurement may be an amperometric measurement. The electrochemical measurement may comprise at least one measurement using at least one electrode. The electrochemical measurement may be a measurement using at least one electrical current measurement and/or at least one electrical voltage measurement and/or at least one impedance measurement. The electrochemical measurement value may be an electrical current and/or an electrical voltage or an impedance, wherein the electrical current and/or the electrical voltage and/or the impedance may correlate with the concentration of the at least one analyte, preferably with the concentration of glucose. E.g., the electric current may be proportional to the concentration of glucose. The electrochemical measurement value may be a raw value as discussed above. The electrochemical measurement value may be corrected by using the target variable Y, preferably for generating a corrected concentration of the analyte, e.g. a corrected concentration of glucose. The concentration of the at least one analyte may be determined by the electrochemical measurement. The optical measurement curve may only be used for a correction, preferably for a calculated correction, of the electrochemical measurement, preferably of the electrochemical measurement value. The correction may comprise the correction algorithm as described above.

By using the electrochemical measurement value, an approximated value of the concentration of the at least one analyte in the sample of the body fluid may be determined. The approximated value of the concentration of the at least one analyte in a sample may be a raw value as discussed above. The target value Y may be used for correcting the approximated value. The target value Y e.g. may comprise an influence of hematocrit and/or a concentration of hematocrit on the concentration of glucose.

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The predetermined multivariate evaluation algorithm may comprise at least one polynomial algorithm selected from:

- 21 -

$$Y = A \cdot X, (1)$$

$$Y = X^T \cdot A \cdot X (2)$$

and

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$$Y = X^T \cdot (X^T \cdot A \cdot X), (3).$$

A may be a one-dimensional, a two-dimensional or a three-dimensional evaluation tensor. A preferably may be a symmetric tensor. E.g., A may be a 3×3 tensor. X and/or Y may be vectors or matrixes. Y being preferably a matrix or a vector may comprise different target variables. X being a matrix or a vector may comprise at least two different characteristic values. Other, preferably polynomial algorithms, may be possible.

The predetermined multivariate evaluation algorithm may comprise at least one algorithm selected from:

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$$Y = \sum_i a_i \cdot X_i, (4)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j, (5)$$

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$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j + \sum_{i,j,k} a_{ijk} \cdot X_i \cdot X_j \cdot X_k (6).$$

a_i , a_{ij} , a_{ijk} may be predetermined coefficients. i , j and k may be, mutually independent, integers from 1 to N .

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The method according to the present invention further may comprise at least one calibration step. In the calibration step, a plurality of calibration measurement curves may be generated by acquiring measurement curves of a plurality of calibration fluids, preferably optical measurement curves of a plurality of calibration fluids, with the respective known target variables Y . The characteristic values may be determined for each calibration measurement curve. E.g., an equation system comprising the coefficients of one or more of equations (4)-(6) may be

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solved. Thereby, numeric values for the coefficients may be determined. The calibration fluid may be a fluid having a known target variable, such as a known concentration of the at least one analyte, preferably a concentration of the analyte determined by a reliable reference method. The calibration fluid may be a fluid most preferably having a known concentration of the analyte, e.g. having a known concentration of glucose. Thus, the calibration fluid may comprise a glucose solution having a known glucose concentration, such as a glucose concentration of 0-500 mg/dl. The calibration measurement curve may be a measurement curve being acquired by using a calibration fluid and/or the calibration measurement curve may be acquired under known conditions, such that at least one target variable Y may be known. Thus, in case the target variable Y may refer to the calibration fluid, the target variable Y may be known via the calibration fluid itself. In case the target variable Y may refer to the measurement conditions, such as a temperature and/or specific properties of the test substance used for the measurement and/or one of the state variables, the target variable Y may be known from the circumstances of the measurement.

Additionally or alternatively, the predetermined multivariate evaluation algorithm may comprise a function which may involve a decision tree. As further used herein, a "decision tree" may comprise at least one decision branch which may allow selecting one out of at least two, preferably two, alternative procedures based on an assessment whether a predetermined condition may be fulfilled or not. The decision branch itself may comprise an additional second-order decision branch which may allow performing one out of at least two, preferably two or three, further alternative procedures depending on the assessment of a further predetermined condition. In addition, the second-order decision branch may comprise at least one further higher-order decision branch. In general, the predetermined condition, which may depend on at least one characteristic value, may assess a (non-)existence of a value or whether a definite value falls within a predetermined range or not. The decision branch may, thus, offer a decision between performing or not performing a specific procedures or performing the specific procedures under a specific parameter, with a specific parameter set, or within a specific parameter range. With regard to the present invention, the specific parameter may comprise the first or the second characteristic value. However, other kinds of predetermined conditions which may especially reflect the specific circumstances of the present method for detecting the analyte in the sample of the body fluid are possible.

- 23 -

As a non-limiting example, the predetermined multivariate evaluation algorithm may comprise the following function involving a first-order decision tree $f(X_1, X_2)$,

$$Y = f(X_1, X_2) = \{g_1(X_2) \text{ for } \textit{cond}(X_1); g_2(X_2) \text{ for NOT } \textit{cond}(X_1)\}, \quad (7)$$

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wherein, depending on the assessment whether the predetermined condition $\textit{cond}(X_1)$, which may depend on the first characteristic value X_1 , may be fulfilled or not, the estimate value for target variable Y may be derived according to Equation (7) by either using a first function $g_1(X_2)$ or an alternative second function $g_2(X_2)$ which both may depend on the second characteristic value X_2 . Other examples for the predetermined multivariate evaluation algorithm may comprise more complex structures of the decision tree, such as a second-order decision tree, wherein, depending on the assessment whether a first predetermined condition, which may depend on one of the characteristic values, may be fulfilled or not, a second predetermined condition, which may depend further on another one of the characteristic values, may be assessed until the estimate value for target variable Y may be derived.

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In step b), the evaluation rules may be adapted such that the characteristic values may be linearly independent, thereby generating unique solutions for the numeric values of the coefficients. Alternatively, the evaluation rules may be adapted such that the characteristic values may be not linearly independent.

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The at least one multivariate evaluation algorithm may comprise at least one algorithm selected from the group consisting of: a partial least squares regression algorithm (PLSR); a principal component regression algorithm (PCR); a support vector machine algorithm (SVM); an artificial neuronal network algorithm (ANN) and/or any genetic algorithm (GA).

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The body fluid may be selected from the group consisting of blood, interstitial fluid, urine, plasma, serum and saliva. Other body fluids may be possible. The body fluid may be prepared by adding at least one additional component. The body fluid may be prepared by heating or cooling or shaking.

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The monitoring of the time development of the at least one measurement value indicating the progress of the detection reaction may be adapted to be an impact-free monitoring of the de-

tection reaction without influencing the detection reaction. The term “impact-free monitoring” may refer to a detection reaction without changes of properties of an environment of the detection reaction. The detection reaction e.g. may preferably be observed without changes of temperature and/or changes of humidity and/or changes of at least one of the state variables.

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At least one of the two different evaluation rules may be selected from the group consisting of:

- 10 a. using a specific measurement value of the optical measurement curve or a derivative of the optical measurement curve at a predetermined point in time as the characteristic value, preferably using one or more specific criteria, particularly using one or more specific conditions, which may comprise at least one end value criterion, more preferably a change rate below a predetermined threshold value;
- 15 b. using a mean value of the optical measurement curve or a derivative of the optical measurement curve over a predetermined period of time as characteristic value;
- 20 c. using a characteristic point in time of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value, preferably a characteristic point in time at which one or more of the following occur: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
- 25 d. using a characteristic parameter of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value, preferably a characteristic parameter at one of: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
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- 25 -

e. using a fit parameter derived by at least one fitting process as the characteristic value, wherein the fitting process may imply a fitting of at least one predetermined fit curve to at least a section of the optical measurement curve or of a derivative of the optical measurement curve; and

5 f. using at least one value derived from a phase plot of at least two derivatives of different order of the optical measurement curve as the characteristic value, wherein the phase plot may comprise at least one phase space curve, wherein the value derived from the phase plot preferably may be selected from the group consisting of: a position
10 of a center of the phase space curve; a length of the phase space curve; a phase space volume; a phase space area; a point with a maximal distance to the center of the phase space curve; a mean squared distance from the origin of the phase space.

The specific measurement value may be a measurement value correlating or anti-correlating
15 with at least one of the target variables. E.g., the specific measurement value may be a measurement value correlating or anti-correlating with at least one of the state variables. The predetermined point in time may be a fixed point in time or a point in time which may be determined by using one or more specific criteria, such as by using one or more specific conditions. The specific criteria or the specific conditions may comprise at least one change rate
20 below a predetermined threshold value, e.g. an end value criterion, such as $\leq 2\%/s$ or $\leq 1\%/s$.

The mean value of the optical measurement curve may be a mean value of the whole optical measurement curve or a mean value of a part of the optical measurement curve. The mean value may be an arithmetic mean or a geometric mean or a harmonic mean or another mean or
25 average. The derivative may be a derivative as defined above. The mean value may be a mean value over a predetermined period of time. The predetermined period of time may comprise at least two measurement values. The predetermined period of time may comprise at least two measurement values lying next to each other.

30 The characteristic point in time may be a fixed point in time or a point in time which may be determined by using one or more specific criteria, such as by using the end value criterion. The fitting process may be an arbitrary fitting process, e.g. a polynomial and/or a linear fitting process and/or an exponential fitting process.

The position of a center of the phase space curve may comprise preferably the coordinates of the center of the phase space curve. The shape of the phase space curve may depend on the state variables, e.g. on the concentration of glucose and/or on the temperature and/or on the concentration of hematocrit.

At least two different evaluation rules may be selected from different numbers of the group a.-f.

Step b) may comprise generating the set of evaluation rules. The generating of the set of evaluation rules may comprise the following sub-steps:

b1) providing a learning set of learning measurement curves, acquired by using a learning set of learning body fluids and by monitoring detection reactions of the test substance and the test body fluids. The test body fluids and the detection reactions may be chosen such that the learning measurement curves may be acquired with different sets of state variables;

b2) identifying a set of candidate evaluation rules and deriving a set of candidate characteristic values from the learning set of learning measurement curves;

b3) determining a correlation between the candidate characteristic values for each candidate evaluation rule and the state variables;

b4) selecting the set of evaluation rules from the set of candidate evaluation rules by accounting for correlations determined in sub-step b3).

As used herein, the term “learning body fluid” may refer to a body fluid having a known set of state variables as long as these state variables refer to properties of the body fluid, such as a known temperature, a known analyte content, such as a known glucose content, and a known hematocrit. Consequently, the term “learning measurement curve” may refer to a measurement curve acquired by using a learning body fluid and by using a known set of state variables. Therein, the known state variables may be defined by the learning body fluid as long as

- 27 -

these state variables refer to properties of the learning body fluid, and may be defined by properties and/or circumstances of the measurement as long as these state variables are defined by the measurement, such as state variables referring to details of the detection reaction and/or the test substance. Further, the term “learning set of learning body fluids” may refer to a set comprising a plurality of learning body fluids having different known sets of state variables. The term “learning set of learning measurement curves” may consequently refer to a set of learning measurement curves acquired by using different sets of known state variables.

The candidate evaluation rule may be an arbitrary evaluation rule which may be arbitrarily chosen out of all potential evaluation rules. Preferably, the candidate evaluation rules may be chosen by using expert knowledge, such as by starting with candidate evaluation rules which have proven to lead to candidate characteristic values which highly, such as by having a correlation coefficient exceeding a predetermined threshold, correlate with one or more of the state variables, preferably with one specific target variable, for similar body fluid and/or similar measurement conditions. Preferably, correlation coefficients for a large number of candidate evaluation rules may be compared visually. E.g., expert knowledge and/or experience may be used regarding candidate evaluation rules which have proven to show a high correlation coefficient with one state variable, such as a target variable, whereas showing a low correlation coefficient, such as the correlation coefficients below a predetermined threshold, for other state variables, preferably for all other state variables for similar body fluids and/or similar measurement conditions. Additionally or alternatively, instead of the correlation coefficients themselves, expert knowledge and/or experience regarding their Merit values may be used.

The candidate characteristic value may be a candidate value derived from a learning measurement curve by using a candidate evaluation rule. The term “correlation” may refer to any of a broad class of statistical relationships. The correlation e.g. may be a dependence. The dependence may be a statistical relationship between two variables and/or between two sets of data.

Sub-step b3) may include determining at least one correlation parameter for each candidate evaluation rule for each state variable. The correlation parameter preferably may be a Pearson correlation coefficient. A correlation parameter may be a correlation coefficient. The correla-

tion parameter and/or the correlation coefficient and/or the Pearson correlation coefficient may be a numerical value for a degree of correlation. The most common correlation coefficient is the Pearson correlation coefficient. The Pearson correlation coefficient preferably may be a correlation coefficient being sensitive only to a linear relationship between the two variables.

In sub-step b4), a Merit value may be calculated for each correlation. The selecting of the set of evaluation rules from the set of candidate evaluation rules may be performed by accounting for the Merit values. The Merit value may be derived out of the correlations and/or the correlation coefficients, e.g. out of the correlation values, preferably out of the Pearson correlation coefficients, e.g. by using the formula:

$$\text{Merit value} = \frac{\text{correlation}^2}{|\text{correlation}_{\text{glucose}}| + |\text{correlation}_{\text{hematocrit}}| + |\text{correlation}_{\text{humidity}}|}$$

The *correlation* may be the correlation coefficient for the concentration of glucose *correlation_{glucose}* or the correlation coefficient for the concentration of hematocrit *correlation_{hematocrit}* or the correlation coefficient for the concentration of the humidity *correlation_{humidity}*. The Merit value may be used for extracting the characteristic values.

In sub-step b4), a candidate evaluation rule may be determined to be an evaluation rule if the corresponding correlation determined in sub-step b3), may fulfill at least one predetermined condition.

In a further aspect (second aspect) of the present invention, a method for detecting an analyte in a sample of a body fluid is disclosed. The analyte, the sample and the body fluid may be defined as discussed above. The method comprises the following steps:

- i) providing at least one measurement curve. The measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid. The measurement values contained in a measurement curve are acquired at differing points in time. The detection reaction is

- 29 -

known to be influenced by a set of state variables. Each state variable characterizes at least one of a state of the sample of the body fluid and a condition of the detection reaction;

- 5 ii) determining at least one target variable Y and/or at least one estimate value for at least one target variable Y by using a first time interval t_1 of the measurement curve, wherein $0 \leq t_1 \leq x \cdot s$, wherein the target variable Y is different from a concentration of the at least one analyte;
- 10 iii) determining the concentration of the at least one analyte by using the at least one target variable.

For further optional details of the method according to this second aspect of the present invention as well as for definitions of the terms used in the method according to the second aspect, 15 reference may be made to the method according to the first aspect of the present invention.

In this second aspect of the present invention, however, the measurement curve not necessarily has to be an optical measurement curve, even though this option still exists. Thus, other types of measurement curves containing a plurality of measurement values recording by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of the body fluid. The detection reaction may be adapted to change at least one measurable property of at least one of the test substance itself and/or the sample of the body fluid. The at least one measurable property not necessarily has to be an optical property and, besides the option of being an optical property, 25 may be or may comprise one or more of an electrical property and/or a chemical property. Thus, as an example, the measurement values of the method according to the second aspect of the present invention may be or may comprise electrical or electrochemical measurement values. Consequently, besides the option of using a test substance which changes at least one optical property due to the detection reaction, other types of test substances may be used additionally or alternatively, such as at least one test substance which changes at least one electrochemical property and/or at least one electrical property due to the detection reaction. Thus, as an example, in the method of the first aspect of the present invention, at least one test element having at least one optical test substance may be used, whereas, in the method according to 30

the second aspect of the present invention and/or in the method according to the third aspect of the present invention, as disclosed in further detail below, at least one test element having at least one of an optical test substance and an electrochemical test substance may be used. Consequently, the measurement curves used and/or evaluated in the method according to the second aspect and/or the third aspect of the present invention, may be selected from the group consisting of optical measurement curves and electrochemical measurement curves. Other options are feasible. Further, consequently, the measurement values in the method according to the second aspect and/or the third aspect of the present invention may be selected from the group consisting of optical measurement values and electrochemical measurement values. Other options are feasible.

Besides the fact that the measurement curve not necessarily has to be an optical measurement curve, the terms used in this method according to the second aspect of the present invention may be defined as the identical terms described above. The first time interval t_1 may be a time interval as described above. The time interval t_1 preferably may comprise at least two measurement values. x may be smaller or equal to 2, preferably x may be smaller or equal to 0.2.

The method further may comprise the following steps:

- iv) providing a set of at least two different evaluation rules. Each evaluation rule may be adapted to derive a characteristic value from the measurement curve, thereby deriving a set of characteristic values $X = \{X_i\}_{i=1..N}$ from the measurement curve. The set of characteristic values comprises at least one first characteristic value being derived from the measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value being derived from the measurement curve by using at least one second evaluation rule from the set of evaluation rules. The second evaluation rule is different from the first evaluation rule.

At least step iv) preferably may be executed by using at least one computer.

The method further may comprise the following step:

- 31 -

- v) performing at least one multivariate analysis of the at least one first characteristic value and the at least one second characteristic value by using at least one predetermined multivariate evaluation algorithm. The multivariate evaluation algorithm is an algorithm adapted to derive at least one result from at least two variables. The at least one first characteristic value and the at least one second characteristic value are used as the at least two variables, thereby deriving the at least one estimate value for at least one target variable Y of the state variables.

The measurement curve may be an optical measurement curve, preferably an optical measurement curve as described above.

The measurement values contained in the measurement curve may be acquired at identical measuring conditions. The term "identical measuring conditions" may comprise e.g. the state of the sample during the measurement. The sample e.g. may be not influenced by an application of an electrical voltage and/or an electrical current, e.g. by using DC and/or AC pulses. The measurement curve may comprise, preferably exclusively, measurement values acquired at identical measuring conditions. Identical measuring conditions preferably may not be given, if an alternating current and/or an alternating voltage is applied on the sample. Most preferably, measurement values measured in step i) may be acquired at identical measuring conditions.

The first evaluation rule may be transformed, preferably may be transformable, into the second evaluation rule by a time transformation.

Alternatively, the first evaluation rule may not be transformed, preferably may not be transformable, into the second evaluation rule by a time transformation.

In a further aspect (third aspect) of the present invention, a method for characterizing a sample of a body fluid is disclosed. As used herein, the term "characterizing" in the context of characterizing the sample of the body fluid refers to a qualitative and/or quantitative determination of at least one property of the sample of the body fluid or a part thereof. The method comprises the following steps:

- 32 -

- 5 A) bringing the sample of the body fluid into contact with at least one test substance, thereby initiating a detection reaction of the test substance and a sample of the body fluid, wherein the detection reaction is known to be influenced by a set of state variables. Each state variable characterizes at least one of a state of the sample of the body fluid and a condition of the detection reaction;
- 10 B) monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at different points in time;
- C) evaluating the optical measurement curve by fully or partially using one of the methods according to the present invention.

15 For further optional details of the method according to this third aspect of the present invention as well as for definitions of the terms used in the method according to the third aspect, reference may be made to the method according to the first and/or the second aspect of the present invention.

Specifically, in method step C), one, more than one or all of the method steps of the methods
20 according to the first aspect and/or the second aspect of the present invention may be performed, with the exemption that the measurement curve not necessarily has to be an optical measurement curve. Thus, other types of measurement curves containing a plurality of measurement values recording by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of
25 the body fluid. The detection reaction may be adapted to change at least one measurable property of at least one of the test substance itself and/or the sample of the body fluid. The at least one measurable property not necessarily has to be an optical property and, besides the option of being an optical property, may be or may comprise one or more of an electrical property and/or a chemical property. Thus, as an example, the measurement values of the method ac-
30 cording to the second aspect of the present invention may be or may comprise electrical or electrochemical measurement values.

- 33 -

Specifically one or more or even all of method steps a), b), c) and d) of the method according to the first aspect of the present invention may be performed, with the exemption that, as outlined above, the measurement curve not necessarily is an optical measurement curve. Thus, as an example, the measurement curve may be an electrochemical measurement curve. For further optional details, reference may be made to the disclosure of the method according to the first aspect as described above or as described in further detail below. Additionally or alternatively, one, more than one or even all of method steps i), ii) and iii) of the method according to the second aspect of the present invention may be performed.

In a further aspect of the present invention, a computer program including computer-executable instructions for performing one or more of the methods according to the present invention, e.g. in one or more of the embodiments enclosed herein, when the program is executed on a computer or a computer network is disclosed. Preferably, at least the steps a)-c) and/or the step b3) may be executed by using the computer program. Additionally or alternatively, one, more than one or even all of method steps i), ii), iii), iv) and v) of the method according to the second aspect of the present invention may be executed by using the computer program. Again, additionally or alternatively, one or both of method steps B) and C) of the method according to the third aspect of the present invention may be executed by using the computer program.

Specifically, the computer program may be stored on a computer-readable data carrier.

The invention further discloses and proposes a computer program product having program code means, in order to perform the methods according to the present invention in one or more of the embodiments enclosed herein when the program is executed on a computer or computer network. Specifically, the program code means may be stored on a computer-readable data carrier.

Further, the invention discloses and proposes a data carrier having a data structure stored thereon, which, after loading into a computer or computer network, such as into a working memory or main memory of the computer or computer network, may execute the methods according to one or more of the embodiments disclosed herein.

- 34 -

The invention further proposes and discloses a computer program product with program code means stored on a machine-readable carrier, in order to perform the methods according to one or more of the embodiments disclosed herein, when the program is executed on a computer or computer network. As used herein, a computer program product refers to the program as a tradable product. The product may generally exist in an arbitrary format, such as in a paper format, or on a computer-readable data carrier. Specifically, the computer program product may be distributed over a data network.

Furthermore, the invention proposes and discloses a modulated data signal containing instructions readable by a computer system or computer network, for performing the methods according to one or more of the embodiments disclosed herein.

Preferably, referring to the computer-implemented aspects of the invention, one or more of the method steps or even all of the method steps of the methods according to one or more of the embodiments disclosed herein may be performed by using a computer or computer network. Thus, generally, any of the method steps including provision and/or manipulation of data may be performed by using a computer or a computer network. Generally, these method steps may include any of the method steps, typically except for method steps requiring manual work, such as providing the samples and/or certain aspects of performing the actual measurements.

Specifically, the present invention further discloses:

- a computer or computer network comprising at least one processor, wherein the processor is adapted to perform the methods according to one of the embodiments described in this description,
- a computer loadable data structure that is adapted to perform the methods according to one of the embodiments described in this description while the data structure is being executed on a computer,
- a computer program, wherein the computer program is adapted to perform the methods according to one of the embodiments described in this description while the program is being executed on a computer,

- 35 -

- a computer program comprising program means for performing the methods according to one of the embodiments described in this description while the computer program is being executed on a computer or on a computer network,
- a computer program comprising program means as described above, wherein the program means are stored on a storage medium readable to a computer,
- a storage medium, wherein a data structure is stored on the storage medium and wherein the data structure is adapted to perform the methods according to one of the embodiments described in this description after having been loaded into a main and/or working storage of a computer or of a computer network, and
- a computer program product having program code means, wherein the program code means can be stored or are stored on a storage medium, for performing the methods according to one of the embodiments described in this description, if the program code means are executed on a computer or on a computer network.

In a further aspect of the present invention, an evaluation device for evaluating an optical measurement curve for analyzing at least one sample of a body fluid is disclosed. The device comprises at least one evaluation unit. The evaluation unit is adapted to perform at least one of the methods for evaluating a measurement according to the present invention. The evaluation device may be an evaluation device being able to evaluate the optical measurement curve.

In a further aspect of the present invention, a sample analysis device for characterizing a sample of a body fluid is disclosed. The device comprises:

- at least one measuring unit for measuring a detection reaction of at least one test substance and at least one sample of a body fluid. The detection reaction is known to be influenced by a set of state variables. Each state variable characterizes at least one of a state of the sample of a body fluid and a condition of the detection reaction. The measuring unit further is adapted for monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at different points in time; and
- at least one evaluation device according to the present invention.

Furthermore, the sample analysis device may comprise at least one test element, preferably at least one test strip. The test element may contain the at least one test substance adapted to perform the detection reaction. The sample analysis device may be adapted such that the sample of the body fluid may be applicable to the test element. The test element may be a part of the sample analysis device being able to contain the test substance and being able to perform the detection reaction. The test element, preferably the test strip, may be arranged such, that the optical measurement curve and/or the measurement curve may be generated. The sample analysis device may be arranged to get the optical measurement curve and/or the electrochemical measurement curve. The sample analysis device may be embodied as a hand-held device. The term "hand-held device" may indicate, that the sample analysis device may be used by a single user and may be carried, e.g. in a hand.

The sample analysis device may comprise at least one detector. The detector may be or may comprise at least one detector array, e.g. a detector offering a spatial resolution.

Alternatively, the detector may be a single detector or may comprise only one detection segment, e.g. a detector having no spatial resolution.

The methods, the computer program, the evaluation device and the sample analysis device according to the present invention may provide a large number of advantages over known methods, devices and computer programs. In particular a detection of an analyte in a sample of a body fluid may be performed more accurately, preferably with smaller errors, most preferably with less influences by different state variables by using the methods and/or the computer program and/or the evaluation device and/or the sample analysis device according to the present invention. E.g., the present invention may be able to distinguish between influences on a measurement curve, preferably on an optical measurement curve, caused by different state variables. In particular, the present invention, preferably the use of the multivariate analysis, may significantly reduce deviations of measured concentrations of glucose caused by, e.g. high, fluctuations of a concentration of hematocrit and/or temperature changes.

The methods and/or the computer program and/or the evaluation device and/or the sample analysis device according to the present invention may enable at least a suppression of influ-

- 37 -

ences caused by at least one state variable when detecting an analyte in a sample of a body fluid, preferably without accepting influences of other state variables. Most preferably, an influence of the concentration of hematocrit when detecting the concentration of glucose may be suppressed without accepting an influence of the temperature.

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The multivariate analysis as used in the present invention may enable an isolation of different influencing parameters, e.g. different state variables, when detecting an analyte in a sample of a body fluid, preferably when determining the concentration of glucose.

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The multivariate analysis as used herein may preferably enable a parallel instead of a sequential analysis of the body fluid. Preferably, according to the present invention, a calculation of a concentration of glucose and a correction of this concentration, e.g. a correction from influences by a concentration of hematocrit, may be executed in one, most preferably in one not separable, step. E.g., a global minimum of the deviations between a prediction and a reference value of the concentration of the analyte, e.g. in forms of chi-squared, which itself is influenced e.g. by the concentration of glucose and/or by the concentration of hematocrit, may be searched in one single step, e.g. by using PLS. According to the present invention, only one single value for the concentration of the analyte may be involved. E.g., the concentration of the analyte may be determined in an iterative multidimensional way, wherein the respective multidimensional space may be spanned by the characteristic values, preferably by the first characteristic value and by the second characteristic value. The characteristic values may be determined by using a remission measurement. According to the present invention, a correction from a concentration of hematocrit may not depend on the concentration of glucose.

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Summarizing the findings of the present invention, the following embodiments are preferred:

Embodiment 1: A method for detecting an analyte in a sample of a body fluid, the method comprising the following steps:

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- a) providing at least one optical measurement curve, wherein the optical measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid, wherein the measurement values contained in the optical measurement curve are acquired at differing points in time, wherein the detection reaction is known to be influenced by a set of

state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;

- b) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1...N}$ from the optical measurement curve, the set of characteristic values comprising at least one first characteristic value being derived from the optical measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value being derived from the optical measurement curve by using at least one second evaluation rule from the set of evaluation rules, the second evaluation rule being different from the first evaluation rule;
- c) performing at least one multivariate analysis of the at least one first characteristic value and the at least one second characteristic value by using at least one predetermined multivariate evaluation algorithm, the multivariate evaluation algorithm being an algorithm adapted to derive at least one result from at least two variables, wherein the at least one first characteristic value and the at least one second characteristic value are used as the at least two variables, thereby deriving at least one estimate value for at least one target variable Y of the state variables;
- d) determining a concentration of the at least one analyte by using the at least one target variable Y .

Embodiment 2: The method according to the preceding embodiment, wherein the state variables are selected from the group consisting of: a composition of the sample of the body fluid, preferably a content of at least one component of the sample of the body fluid and more preferably a concentration of at least one analyte; a content of at least one particulate component of the sample of the body fluid, preferably a haematocrit; a temperature of the sample of the body fluid; a humidity of an ambient atmosphere surrounding the sample of the body fluid; a storage time of the test substance; an interfering substance; alterations of the sample or of certain properties of the sample caused by pharmacological treatment of a donor of the sample.

Embodiment 3: The method according to one of the preceding embodiments, wherein the first evaluation rule may not be transformed into the second evaluation rule by a time transformation..

Embodiment 4: The method according to one of the preceding embodiments, wherein the second evaluation rule differs from the first evaluation rule in at least one coefficient and/or in at least one parameter and/or in at least one component related to the algorithm.

- 39 -

Embodiment 5: The method according to the preceding embodiment, wherein the algorithm of the first evaluation rule differs from the algorithm of the second evaluation rule in a point in time.

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Embodiment 6: The method according to one of the preceding embodiments, wherein a third evaluation rule is provided, wherein, in step c), the at least one first characteristic value is derived from the first evaluation rule, and wherein, in the multivariate evaluation algorithm, the second evaluation rule or the third evaluation rule is used depending on the at least one

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Embodiment 7: The method according to one of the preceding embodiments, wherein the first characteristic value is determined by using a first time interval of the optical measurement curve, wherein the second characteristic value is determined by using a second time interval of the optical measurement curve, wherein the first time interval of the optical measurement curve is different from the second time interval of the optical measurement curve.

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Embodiment 8 The method according to the preceding embodiment, wherein the target value is different from the concentration of the at least one analyte.

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Embodiment 9: The method according to one of the preceding embodiments, wherein the at least two evaluation rules are adapted to derive the characteristic values from at least two derivatives of the optical measurement curve.

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Embodiment 10: The method according to the preceding embodiment, wherein the at least two derivatives are derivatives comprising at least two derivatives of different order.

Embodiment 11: The method according to one of the two preceding embodiments, wherein the derivatives are generated by using at least one filtering algorithm, preferably a Savitzky-Golay filtering algorithm.

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Embodiment 12: The method according to one of the preceding embodiments, wherein the set of characteristic values contains 2-20 characteristic values, preferably 3-10 characteristic values.

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Embodiment 13: The method according to one of the preceding embodiments, wherein the target variable Y comprises the concentration of the at least one analyte in the sample of the body fluid.

Embodiment 14: The method according to one of the preceding embodiments, wherein, in step d), in addition to the at least one target variable Y, at least one electrochemical measurement value is used for determining the concentration of the analyte, wherein the electrochemical measurement value is determined by using at least one electrochemical measurement.

Embodiment 15: The method according to the preceding embodiment, wherein, by using the electrochemical measurement value, an approximated value of the concentration of the at least one analyte in the sample of the body fluid is determined, wherein the target value Y is used for correcting the approximated value.

Embodiment 16: The method according to one of the preceding embodiments, wherein the predetermined multivariate evaluation algorithm comprises at least one polynomial algorithm selected from:

$$Y = A \cdot X, \quad (1)$$

$$Y = X^T \cdot A \cdot X \quad (2)$$

and

$$Y = X^T \cdot (X^T \cdot A \cdot X), \quad (3),$$

wherein A is a one-dimensional, a two-dimensional or a three-dimensional evaluation tensor.

Embodiment 17: The method according to one of the preceding embodiments, wherein the predetermined multivariate evaluation algorithm comprises at least one algorithm selected from:

$$Y = \sum_i a_i \cdot X_i, \quad (4)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j, \quad (5)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j + \sum_{i,j,k} a_{ijk} \cdot X_i \cdot X_j \cdot X_k, \quad (6),$$

wherein a_i , a_{ij} , a_{ijk} are predetermined coefficients and wherein i, j and k are, mutually independently, integers from 1 to N.

Embodiment 18: The method according to the preceding embodiment, further comprising at least one calibration step, wherein, in the calibration step, a plurality of calibration measurement curves is generated by acquiring measurement curves of a plurality of calibration fluids with the respective known target variables Y , wherein the characteristic values are determined for each calibration measurement curve, wherein an equation system comprising the coefficients of one or more of equations (4)-(6) is solved, thereby determining numeric values for the coefficients.

Embodiment 19: The method according to the preceding embodiment, wherein, in step b), the evaluation rules are adapted such that the characteristic values are linearly independent, thereby generating unique solutions for the numeric values of the coefficients.

Embodiment 20: The method according to one of the preceding embodiments, wherein the at least one multivariate evaluation algorithm comprises at least one algorithm selected from the group consisting of: a partial least squares regression algorithm (PLSR); a principal component regression algorithm (PCR); a support vector machine algorithm (SVM); an artificial neuronal network algorithm (ANN); a genetic algorithm (GA).

Embodiment 21: The method according to one of the preceding embodiments, wherein the at least one multivariate evaluation algorithm comprises a function involving at least one decision tree, wherein the decision tree comprises at least one decision branch which allows selecting one out of at least two alternative procedures based on an assessment whether a predetermined condition may be fulfilled.

Embodiment 22: The method according to the preceding embodiment, wherein the decision branch offers a decision between performing or not performing a specific procedure or performing the specific procedure under a specific parameter, with a specific parameter set, or within a specific parameter range.

Embodiment 23: The method according to the preceding embodiment, wherein the specific parameter comprises the characteristic value.

Embodiment 24: The method according to one of the preceding embodiments, wherein the body fluid is selected from the group consisting of blood, interstitial fluid, urine, plasma, serum and saliva.

Embodiment 25: The method according to one of the preceding embodiments, wherein the monitoring of the time development of the at least one measurement value indicating the pro-

gress of the detection reaction is adapted to be an impact-free monitoring of the detection reaction without influencing the detection reaction.

Embodiment 26: The method according to one of the preceding embodiments, wherein at least one of the two different evaluation rules is selected from the group consisting of:

- a. using a specific measurement value of the optical measurement curve or a derivative of the optical measurement curve at a predetermined point in time as the characteristic value, preferably using one or more specific criteria, particularly using one or more specific conditions, which may comprise at least one end value criterion, more preferably a change rate below a predetermined threshold value;
- b. using a mean value of the optical measurement curve or a derivative of the optical measurement curve over a predetermined period of time as the characteristic value;
- c. using a characteristic point in time of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value, preferably a characteristic point in time at which one or more of the following occur: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
- d. using a characteristic parameter of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value, preferably a characteristic parameter at one of: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
- e. using a fit parameter derived by at least one fitting process as the characteristic value, wherein the fitting process implies a fitting of at least one predetermined fit curve to at least a section of the optical measurement curve or of a derivative of the optical measurement curve; and
- f. using at least one value derived from a phase plot of at least two derivatives of different order of the optical measurement curve as the characteristic value, wherein the phase plot comprises at least one phase space curve, wherein the value derived from the phase plot preferably is selected from the group consisting of: a position of a center of the phase space curve; a length of the phase space curve; a phase space volume; a phase space area; a point with a maximal distance to the center of the phase space curve; a mean squared distance from the origin of the phase space.

Embodiment 27: The method according to the preceding embodiment, wherein at least two different evaluation rules selected from different members of the group a.-f. are selected.

Embodiment 28: The method according to one of the preceding embodiments, wherein step b) comprises generating the set of evaluation rules, the generating of the set of evaluation rules comprising the following sub-steps:

- b1) providing a learning set of learning measurement curves, acquired by using a learning set of learning body fluids and by monitoring detection reactions of a test substance and the test body fluids, wherein the test body fluids and the detection reactions are chosen such that the learning measurement curves are acquired with differing sets of state variables;
- b2) identifying a set of candidate evaluation rules and deriving a set of candidate characteristic values from the learning set of learning measurement curves;
- b3) determining a correlation between the candidate characteristic values for each candidate evaluation rule and the state variables;
- b4) selecting the set of evaluation rules from the set of candidate evaluation rules by accounting for the correlations determined in sub-step b3).

Embodiment 29: The method according to the preceding embodiment, wherein sub-step b3) includes determining at least one correlation parameter for each candidate evaluation rule for each state variable, preferably a Pearson correlation coefficient.

Embodiment 30: The method according to one of the two preceding embodiments, wherein, in sub-step b4), a Merit value is calculated for each correlation, wherein the selecting of the set of evaluation rules from the set of candidate evaluation rules is performed by accounting for the Merit values.

Embodiment 31: The method according to one of the three preceding embodiments, wherein, in sub-step b4), a candidate evaluation rule is determined to be an evaluation rule if the corresponding correlation determined in sub-step b3) fulfils at least one predetermined condition.

Embodiment 32: A method for detecting an analyte in a sample of a body fluid, the method comprising the following steps:

- i) providing at least one measurement curve, wherein the measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid, wherein the measurement values contained in the measurement curve are acquired at differ-

ing points in time, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;

ii) determining at least one target variable and/or at least one estimate value for at least one Y by using a first time interval t_1 of the measurement curve, wherein $0 \leq t_1 \leq x$, wherein the target variable Y is different from a concentration of the at least one analyte;

iii) determining the concentration of the at least one analyte by using the at least one target variable;

iv) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value from the measurement curve, thereby deriving a set of characteristic values $X = \{X_i\}_{i=1 \dots N}$ from the measurement curve, the set of characteristic values comprising at least one first characteristic value being derived from the measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value being derived from the measurement curve by using at least one second evaluation rule from the set of evaluation rules, the second evaluation rule being different from the first evaluation rule.

Embodiment 33: The method according to the preceding embodiment, wherein the method further comprises the following step:

v) performing at least one multivariate analysis of the at least one first characteristic value and the at least one second characteristic value by using at least one predetermined multivariate evaluation algorithm, the multivariate evaluation algorithm being an algorithm adapted to derive at least one result from at least two variables, wherein the at least one first characteristic value and the at least one second characteristic value are used as the at least two variables, thereby deriving the at least one estimate value for at least one target variable Y of the state variables.

Embodiment 34: The method according to the preceding embodiment, wherein the second evaluation rule differs from the first evaluation rule in at least one coefficient and/or in at least one parameter and/or in at least one component related to the algorithm.

Embodiment 35: The method according to one of the three preceding embodiments, wherein the measurement curve is an optical measurement curve.

Embodiment 36: The method according to one of the four preceding embodiments, wherein the measurement values contained in the optical measurement curve are acquired at identical measuring conditions.

- 5 Embodiment 37: The method according to one of the five preceding embodiments, wherein the first evaluation rule may not be transformed into the second evaluation rule by a time transformation.

Embodiment 38: A method for characterizing a sample of a body fluid, the method comprising the following steps:

- 10 A) bringing the sample of the body fluid into contact with at least one test substance, thereby initiating a detection reaction of the test substance and the sample of the body fluid, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;
- 15 B) monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at differing points in time;
- 20 C) evaluating the optical measurement curve by using the method according to one of the preceding embodiments.

Embodiment 39: A computer program including computer-executable instructions for performing the method according to one of the preceding embodiments when the program is executed on a computer or computer network.

Embodiment 40: An evaluation device for evaluating an optical measurement curve for analyzing at least one sample of a body fluid, the device comprising at least one evaluation unit, wherein the evaluation unit is adapted to perform a method according to one of the preceding
30 embodiments referring to a method for evaluating a measurement.

Embodiment 41: A sample analysis device for characterizing a sample of a body fluid, the device comprising:

- 35 - at least one measuring unit for measuring a detection reaction of at least one test substance and at least one sample of a body fluid, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction, the measuring unit further being adapted for monitoring a time development of at least

- 46 -

one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at different points in time; and

- at least one evaluation device according to the preceding embodiment.

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Embodiment 42: The sample analysis device according to the preceding embodiment, furthermore comprising at least one test element, preferably at least one test strip, wherein the test element contains the at least one test substance adapted to perform the detection reaction, wherein the sample analysis device is adapted such that the sample of the body fluid is applicable to the test element.

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Embodiment 43: The sample analysis device according to the preceding embodiment, wherein the sample analysis device is embodied as a hand-held device.

15 Short description of the figures

Further optional details and optional features of the present invention may be derived from the subsequent description of preferred embodiments, preferably in conjunction with the dependent claims. In these embodiments, in each case, the optional features may be realized in an isolated way or in an arbitrary combination of several features. The invention is not restricted to the embodiments. The embodiments are schematically depicted in the figures. Identical reference numbers in the figures refer to identical, similar or functionally identical elements.

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In the figures:

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Figure 1 shows a schematic view of an embodiment of a sample analysis device according to the present invention;

Figures 2A-2D show diagrams of an embodiment of a method for detecting an analyte in a sample of a body fluid according to the present invention;

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Figure 3 shows correlations between a remission and a concentration of glucose and correlations between derivatives of the remission and the concentration of glucose of another embodiment of a method according to the present invention;

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- Figure 4 shows correlations between a fluorescence and a concentration of glucose and correlations between derivatives of the fluorescence and the concentration of glucose of another embodiment of a method according to the present invention;
- Figure 5 shows a diagram presenting advantages of embodiments of methods according to the present invention;
- Figures 6A and 6B show phase plots of a first order derivative of a remission with respect to a time over a remission for another embodiment of a method according to the present invention for two different temperatures; ;
- Figure 7 shows a diagram of a further embodiment of a method for detecting an analyte in a sample of a body fluid according to the present invention;
- Figure 8 shows a scheme of a further embodiment of a method for detecting an analyte in a sample of a body fluid according to the present invention involving a first decision tree;
- Figure 9 shows a scheme of a further embodiment of a method for detecting an analyte in a sample of a body fluid according to the present invention involving a second decision tree; and
- Figure 10 shows a scheme of a further embodiment of a method for detecting an analyte in a sample of a body fluid according to the present invention involving a third decision tree.

Detailed description of the embodiments

- In figure 1, an embodiment of an analysis device 110 for characterizing a sample of a body fluid according to the present invention is shown. The device, preferably the sample analysis device 110, comprises at least one measuring unit 112 for measuring a detection reaction of at least one test substance and at least one sample of a body fluid. The detection reaction is known to be influenced by a set of state variables. Each state variable characterizes at least

one of a state of the sample of the body fluid and/or a condition of the detection reaction. The measuring unit 112 further is adapted for monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve and/or a measurement curve containing a plurality of the measurement values acquired at different points in time. The device, preferably the sample analysis device 110 further comprises at least one evaluation device 114. The evaluation device 114 may be an evaluation device for evaluating an optical measurement curve and/or a measurement curve for analyzing at least one sample of a body fluid. The device, preferably the evaluation device 114, comprises at least one evaluation unit 116. The evaluation unit 116 is adapted to perform a method according to the present invention. The sample analysis device 110 furthermore may comprise at least one test element 118, preferably at least one test strip 120. The test element 118 may contain the at least one test substance adapted to perform the detection reaction. The sample analysis device 110 may be adapted such that the sample of the body fluid is applicable to the test element 118.

Additionally or alternatively, the test element 118, preferably the test strip 120, may be reversibly attached to the sample analysis device 110. The test element 118 and preferably the test strip 120, may alternatively be not a part of the sample analysis device 110. The test element 118 and/or the test strip 120 may be able to be inserted into the sample analysis device 110, e.g. after attaching the sample of the body fluid on the test element 118.

The sample analysis device 110 may be embodied as a hand-held device 122. The sample analysis device 110 may comprise at least one monitor 124, e.g. at least one touchscreen and/or at least one screen and/or at least one display. The monitor 124 may be adapted to display e.g. a concentration of the at least one analyte and/or at least one estimate value for the at least one target variable. The sample analysis device 110 further may comprise at least one input panel 126. The input panel 126 may be designed to act as an interface between a user and the sample analysis device 110. The input panel 126 may comprise at least one key and/or at least one keyboard and/or at least one knob and/or at least one touchscreen and/or at least one touchscreen pad. The monitor 124 and the input panel 126 may be separated from each other. The input panel 126 may alternatively be integrated in the monitor 124, e.g. in a touchscreen panel. The measuring unit 112 may comprise at least one light source 128. The light source 128 may comprise at least one light emitting diode (LED) and/or at least one laser

- 49 -

and/or at least one lamp. The measuring unit 112 may comprise at least one detector 130. The detector 130 may be a device being able to detect light being emitted by the sample and/or being reflected by the sample and/or by the test element 118 and/or being emitted by the light source 128. The detector 130 may be able to detect the remission, e.g. a reflection and/or a scattering of the light emitted by the light source 128 and/or by the test element 118 and/or by the sample 110 and/or by the analyte. The detector 130 alternatively or additionally may be able to detect a fluorescence signal emitted by the test element 118 and/or by the sample and/or by the analyte. The sample analysis device 110, preferably the measuring unit 112 may comprise different components for doing spectroscopy, preferably spectroscopy of the sample and/or of the analyte. The measuring unit 112 further may comprise at least one lance and/or at least one filter and/or at least one mirror and/or at least one wavemeter. The detector 130 may comprise at least one photodiode, e.g. an avalanche photodiode, and/or at least one CCD chip and/or at least one camera and/or at least one wavemeter and/or at least one frequency comb and/or at least one spectroscopy cell. The detector 130 may be able to detect light, e.g. by generating at least one signal, preferably by generating the measurement curve and/or the optical measurement curve. The signal may be detected in dependence of a time t and/or may be integrated over a certain time period. The sample analysis device 110 may comprise at least one computer 132 and/or at least one computer network 134. The computer 132 and/or the computer network 134 may be integrated in the evaluation unit 116. The computer 132 and/or the computer network 134 may also be at least partially separated from the sample analysis device 110. The computer 132 and/or the computer network 134 may be able to be connected to the sample analysis device 110, e.g. by at least one interface, e.g. by at least one USB connection.

The computer 132 and/or the computer network 134 and/or the sample analysis device 110 may be adapted to execute at least one computer program. The computer program includes computer-executable instructions for performing the method according to the present invention when the program is executed on the computer 132 and/or on the computer network 134.

The sample analysis device 110 may comprise at least one photometric glucose strip, e.g. the test strip 120. The sample analysis device 110 may be an Accu Chek[®] Active system by Roche Diagnostics Deutschland GmbH. The following embodiments may refer to photometric glucose strips and/or the Accu Chek[®] Active system, but the method according to the present

invention, as well as the computer program and/or the evaluation device and/or the sample analysis device 110, may alternatively be assigned to a plurality of different systems, preferably systems in which at least one analyte concentration should be determined and a accuracy of reading of a measured signal may depend on additional properties of the sample and/or of the measurement and/or of other circumstances, e.g. of at least one state variable. The methods according to the present invention e.g. also may be combined with electrochemical measurements of an analyte in a sample of a body fluid, e.g. referring to glucose test strips and/or absorption kinetics in lab analysis systems.

Another method for detecting an analyte in a sample of a body fluid according to the present invention comprises the following steps:

- i) providing at least one measurement curve, wherein the measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid, wherein the measurement values contained in the measurement curve are acquired at differing points in time 136, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;
- ii) determining at least one target variable and/or at least one estimate value for at least one Y by using a first time interval t_1 of the measurement curve, wherein $0 \leq t_1 \leq x$, wherein the target variable Y is different from a concentration of the at least one analyte.
- iii) determining the concentration of the at least one analyte by using the at least one target variable.

The method further may comprise the following step:

- iv) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value 138 from the measurement curve, thereby deriving a set of characteristic values $X = \{X_i\}_{i=1 \dots N}$ from the measurement curve, the set of characteristic values 138 comprising at least one first characteristic value 138 being derived from the measurement curve by using at least one first evaluation rule

- 51 -

from the set of evaluation rules and at least one second characteristic value 138 being derived from the measurement curve by using at least one second evaluation rule from the set of evaluation rules, the second evaluation rule being different from the first evaluation rule, preferably by using a computer 132.

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The method further may comprise the following step:

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- v) performing at least one multivariate analysis of the at least one first characteristic value 138 and the at least one second characteristic value 138 by using at least one predetermined multivariate evaluation algorithm, the multivariate evaluation algorithm being an algorithm adapted to derive at least one result from at least two variables, wherein the at least one first characteristic value 138 and the at least one second characteristic value 138 are used as the at least two variables, thereby deriving the at least one estimate value for at least one target variable Y of the state variables.

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The measurement curve may be an optical measurement curve. The measurement values contained in the optical measurement curve may be acquired at identical measuring conditions. The first evaluation rule may not be transformed, preferably may not be transformable, into the second evaluation rule by a time transformation.

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In figure 2A to figure 7 diagrams of embodiments according to methods for detecting an analyte in a sample of a body fluid according to the present invention are disclosed.

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A method for detecting an analyte in a sample of a body fluid according to the present invention comprises the following steps:

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- a) providing at least one optical measurement curve. The optical measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid. The measurement values contained in the optical measurement curve are acquired at different points in time 136. The detection reaction is known to be influenced by a set of state variables. Each state varia-

ble characterizes at least one of a state of the sample of the body fluid and a condition of the detection reaction.

b) providing a set of at least two different evaluation rules. Each evaluation rule is adapted to derive a characteristic value 138 from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1...N}$ from the optical measurement curve. The set of characteristic values 138 comprises at least one first characteristic value 138 being derived from the optical measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value 138 being derived from the optical measurement curve by using at least one second evaluation rule from the set of evaluation rules. The second evaluation rule is different from the first evaluation rule.

c) performing at least one multivariate analysis of the at least one first characteristic value 138 and the at least one second characteristic value 138 by using at least one predetermined multivariate evaluation algorithm. The multivariate evaluation algorithm is an algorithm adapted to derive at least one result from at least two variables. The at least one first characteristic value 138 and the at least one second characteristic value 138 are used as the at least two variables, thereby deriving at least one estimate value for at least one target variable Y of the state variables;

d) determining a concentration of the at least one analyte by using the at least one target variable.

For an evaluation, preferably not only one characteristic value 138, e.g. a parameter, like an end value and/or a derivative and/or something similar, is used but a plurality of parameters, e.g. a plurality of characteristic values 138, preferably at least two characteristic values 138, e.g. of a kinetic curve. The methods according to the present invention preferably may not exclusively use a univariate analysis, e.g. with only one parameter, e.g. only one characteristic value 138, but a multivariate analysis, preferably with more parameters, most preferably with at least two characteristic values 138.

- 53 -

This may be the reason, why the use of the predetermined multivariate evaluation algorithm may also be called multivariate analysis. Methods for executing the multivariate analysis and/or multivariate evaluation algorithms are e.g. described e.g. in the standard textbooks Harald Martens and Tormod Næs, Multivariate Calibration, John Wiley and Sons Ltd, 1998, p. 97-165 and/or René Henrion and Günter Henrion, Multivariate Datenanalyse, Springer-Verlag, 1995, p. 103-157.

Although, the following embodiments are predominantly described for an analysis of kinetic curves of blood glucose measurements, the method according to the present invention may alternatively be used in a wider context of analyzations of data, preferably but not exclusively, of time dependent data. The optical measurement curve may be e.g. a measurement curve changing during a fermentation for analyzing the fermentation by using the method according to the present invention, e.g. for making production processes more stable.

The use of the methods according to the present invention may be especially advantageous by using periodic and/or similar developing optical measurement curves. The methods according to the present invention e.g. may be used for avoiding an influence of a state variable, e.g. haematocrit and/or a temperature, on a detection of the analyte in the sample of the body fluid. At least one of the characteristic values 138 may not just be a measurement value, but a characteristic value 138. At least one of the characteristic values 138 may be different from a simple measurement value. The points in time 136 may be distributed over 0 seconds to 5 minutes, e.g. 0 seconds to 101 seconds, preferably 0 seconds to 30 seconds, most preferably 3 seconds to 6 seconds.

The methods according to the present invention preferably may not be part of a closed loop and/or may not comprise a closed loop.

In the methods according to the present invention at least one test strip 120 may be observed. Preferably, the analyte in a sample of the body fluid may be detected only once. E.g., only one sample of a body fluid may be taken from the body once the method is performed. Thus, the optical measurement curve preferably may be a single measurement on the test element 118, e.g. the test strip 120, and not a periodic signal.

Alternatively, the method according to a present invention may be performed periodically, e.g. on different samples of the body fluid and/or may be part of a closed loop.

5 A relationship between a signal, e.g. the optical measurement curve and/or the measurement curve, e.g. at least one integration of at least a part of the optical measurement curve, may be inversed proportional to the concentration of glucose. E.g., a high concentration of glucose may result in a smaller signal. The relationship between the concentration of glucose and the signal may be provided by a calibration curve, e.g. for optical sample analysis devices 110. The calibration curve and/or the relationship between the concentration of glucose and the
10 signal may comprise at least one exponential function.

The method according to the present invention may provide a simultaneous correction of at least two error sources, preferably of all error sources influencing the detection of the analyte. The simultaneous correction may be performed without knowledge of connections and/or
15 correlations and/or relationships between individual error sources, e.g. between individual state variables. The method according to the present invention most preferably may perform a simultaneous correction, wherein an individual correction term may not have to be known and/or may not have to be used in the methods according to the present invention.

20 The method according to the present invention may comprise a simultaneous correction, wherein a simultaneous correction may be defined as an opposite of a sequential correction. In a sequential correction, error sources may be corrected successively, e.g. one error source after another. In the method according to the present invention, preferably no successive methods may be performed. In the method according to the present invention each step preferably only may be performed once, most preferably step d) may only be performed once in
25 the methods according to the present invention.

The state variables may be selected from the group consisting of: a composition of the sample of the body fluid, preferably a content of at least one component of the sample of the body
30 fluid and more preferably a concentration of at least one analyte; a content of at least one particulate component of the sample of the body fluid, preferably haematocrit; a temperature of the sample of the body fluid; a humidity of an ambient atmosphere surrounding the sample of the body fluid; a storage time of the test substance; a storage history, e.g. of the test element

- 55 -

118, which may influence the test signal, e.g. the measurement values. The state variables, preferably the storage history, may change at least one of the measurement values and/or at least one prominent point of the optical measurement curve, e.g. at least one minimum, e.g. a local minimum, and/or at least one maximum and/or at least one turning point and/or at least one other characteristic point and/or at least one value of at least one derivative of the optical measurement curve and/or the time of their appearances. A particulate component may comprise e.g. acetylsalicylic acid and/or citric acid and/or maltose and/or xylose.

The methods according to the present invention may be able to suppress interferences caused by the state variables or even may be able to compensate at least a part of the interferences caused by the state variables. The methods according to the present invention may give the opportunity to determine a state of a reagent kit, e.g. of a test strip 120, e.g. in terms of at least one activity of an enzyme, e.g. of glucose oxidase and/or glucose dehydrogenase and/or glucose deoxyreductase and/or similar components.

The methods and devices according to the present invention may be used in detections of analyte concentrations, preferably in detections of analyte concentrations wherein a accuracy of reading out of the measured signal may be influenced by using methods and devices known from prior art by other properties besides the concentrations of these analytes, e.g. on properties of the sample and/or of the measurement.

The methods according to the present invention, preferably the use of the multivariate analysis, may not be limited to optical detection methods, but may also be applicable in systems being able to execute a different unperturbed measurement.

The methods according to the present invention may also be used in electrochemical detection methods and/or may comprise electrochemical detection methods. In a multivariate analysis, not only one parameter, e.g. not only one characteristic value 138, is used but a plurality of parameters, e.g. a plurality of characteristic values 138. The multivariate analysis, e.g. an analysis with more parameters, may preferably deviate from a univariate analysis, e.g. with only one parameter.

The influences of the state variables, like temperature and/or haematocrit, may influence the optical measurement curve, e.g. the kinetic curve, in such a way that these interference factors and/or state variables may be able to be determined by the at least two characteristic values 138.

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The two characteristic values 138 may differ from each other and/or may be independent parameters. The set of characteristic values 138, preferably the two characteristic values 138, may be determined out of only one optical measurement curve, e.g. out of only one kinetic curve. The influences of the state variables may be quantified by the at least two characteristic values 138. The two characteristic values 138 may comprise the first characteristic value 138 and the second characteristic value 138.

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In the methods according to the present invention, besides the detection of at least one analyte concentration at least one additional state variable, e.g. the temperature and/or the haematocrit and/or another state variable, may be determined, preferably by evaluating one single optical measurement curve, e.g. one single kinetic curve.

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The first evaluation rule may not be transformed, preferably not be transformable, into the second evaluation rule by a time transformation. The first characteristic value 138 may be determined by using a first time interval of the optical measurement curve. The second characteristic value 138 may be determined by using a second time interval of the optical measurement curve. The first time interval of the optical measurement curve may be different from the second time interval of the optical measurement curve. The target value preferably may be different from the concentration of the at least one analyte.

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The at least two evaluation rules may be adapted to derive the characteristic values 138 from at least two derivatives of the optical measurement curve. The characteristic values 138 may be derived by using at least two derivatives of the optical measurement curve. At least one of the derivatives of the optical measurement curve may preferably not only be used for determining a criterion, e.g. an end point value. At least one quantitative value of the derivatives of the optical measurement curve may be used directly for deriving at least one of the characteristic values 138. E.g., the numerical value of the first derivative of the optical measurement curve, e.g. of a remission kinetic curve, may be included in the evaluation. By evaluation of

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- 57 -

only one optical measurement curve, e.g. only one kinetic curve, at least one haematocrit value and/or at least one analyte concentration and/or at least one temperature and/or at least one another state variable may be able to be determined, preferably quantitatively. Alternatively, the two evaluation rules may differ from each other by applying two different components related to the algorithm, such as two different thresholds or two different change rates below a predetermined threshold.

The at least two derivatives may be derivatives comprising at least two derivatives of different order. The order of the derivatives may range from 0 to n, wherein n is a natural number.

The derivatives may be generated by using at least one filtering algorithm, preferably a Savitzky-Golay filtering algorithm. The Savitzky-Golay filter is a filter known from data analysis. The filtering algorithm may be a filter. The filtering algorithm may range up to the said order.

During using the filtering algorithm, e.g. 11 neighbouring measurement values, e.g. data points, e.g. in a distance of 0.1 seconds may be examined.

Additionally or alternatively, at least one spline function may be fitted, preferably segmentally, to the optical measurement curve and/or to the measurement curve, e.g. to the kinetic curve. A spline function may be a sufficiently smooth polynomial function. The spline function may be piecewise-defined. The spline function may possess a high degree of smoothness at places where polynomial pieces may connect. These places may be called knots. In the methods according to the present invention at least one polynomial interpolation may be used. The knots may have a distance of about 0.1 seconds. Additionally or alternatively, knots having a varying distance may be used, such as a small distance in regions in which a high alteration rate of the curves is encountered and a higher distance in regions in which a low alteration rate of the curves is encountered. Thus, the density of the knots may be adapted to the degree of changes or alterations in the curves.

Significant changes of the optical measurement curve often may take place shortly after bringing the sample of the body fluid into contact with the at least one test substance, e.g. with the test element 118, preferably with the test strip 120.

A functional dependency of the optical measurement curve, e.g. of a development of the remission in time on at least one state variable, e.g. on a hematocrit value, may be seen as a perturbation of the detection of the analyte in the sample, preferably of determining the concentration of the at least one analyte, e.g. the determining of glucose, but on the other hand, information about at least one state value, e.g. about a concentration of hematocrit, may be included in the measurement curve and/or in the optical measurement curve, e.g. in the kinetic curve.

The optical measurement curves, e.g. measured kinetic curves, and/or the at least two derivatives of the optical measurement curve, e.g. the first derivative and/or the second derivative and/or the third derivative, may be divided in time intervals, e.g. starting with bringing the sample of the body fluid into contact with the at least one test substance at $t=0$.

As essential changes often may happen shortly after bringing the sample of the body fluid into contact with the at least one test substance, e.g. with the test element 118 and/or with the test strip 120, it may be advantageous to use time intervals with different length instead of using equidistant time intervals.

Additionally or alternatively, the time intervals at least partially may be equidistant. E.g., the time intervals may be short shortly after bringing the sample of the body fluid into contact with the at least one test substance, wherein the short time interval may last 1 ms to 2 s, preferably 10 ms to 1 s, most preferably about 100 ms. For later times, longer time intervals may be useful, e.g. 1 s to 5 s, preferably 1.5 s to 2.5 s, most preferably about 2 s. E.g., an optical measurement curve of about 30 s may be divided in 23 time intervals, wherein the length of the time intervals may increase proportional to the square root of the time.

Additionally or alternatively, the optical measurement curve may be divided in different parts of the optical measurement curve, e.g. in different time intervals, wherein different analysis methods may be applied to the parts of the optical measurement curve, e.g. different evaluation rules. The evaluation rule may comprise the determination of at least one slope, e.g. for a first part of the optical measurement curve, and/or a determination of at least one second derivative for at least one second part of the optical measurement curve or something similar.

- 59 -

The set of characteristic values 138 may contain 2-20 characteristic values 138, preferably 3-10 characteristic values 138. The target variable Y may comprise a concentration of the at least one analyte in the sample of the body fluid, preferably a concentration of glucose.

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In step d), in addition to the at least one target variable Y, at least one electrochemical measurement value may be used for determining the concentration of the analyte. The electrochemical measurement value may be determined by using at least one electrochemical measurement. By using the electrochemical measurement value, an approximated value of the concentration of the at least one analyte in the sample of the body fluid may be determined. The target value Y may be used for correcting the approximated value.

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The predetermined multivariate evaluation algorithm may comprise at least one polynomial algorithm selected from:

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$$Y = A \cdot X, (1)$$

$$Y = X^T \cdot A \cdot X (2)$$

and

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$$Y = X^T \cdot (X^T \cdot A \cdot X). (3).$$

A may be a one-dimensional or a two-dimensional or a three-dimensional evaluation tensor, preferably a symmetric tensor. The predetermined multivariate evaluation algorithm may comprise at least one algorithm selected from:

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$$Y = \sum_i a_i \cdot X_i, (4)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j, (5)$$

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$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j + \sum_{i,j,k} a_{ijk} \cdot X_i \cdot X_j \cdot X_k. (6).$$

a_i , a_{ij} , a_{ijk} may be predetermined coefficients. i , j and k may be mutually independently integer from 1 to N. The method may comprise at least one calibration step. Formula (5) e.g. may show, that the use of squared terms already may be sufficient. Formula (6) may show, that

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cross terms between the characteristic values 138, e.g. products of two characteristic values 138, may be important. It may be convenient to use the characteristic values X_i , e.g. input parameters, not only in linear combinations, similar to the partial least squares (PLS) method, but additionally or alternatively also use products of the characteristic values 138, e.g. of the input parameters, among themselves and/or squares of the characteristic values 138, e.g. of an input parameter, e.g. expressed by cross terms. In other words, it may be advantageous to use formula (5) instead of formula (4). Y may be the predicted glucose values.

The characteristic value 138 may be e.g. an end value. The end value as a single characteristic value 138, e.g. as single parameter, may be used in the PLS and/or may be used in a squared term, e.g. as at least the square of the end value. In such a case, the multivariate analysis, e.g. the multivariate data analysis, may e.g. only comprise two characteristic values 138, preferably two input values, e.g. $X_{EV} = EV$ (end value) and $X_{EV}^2 = EV^2$, e.g. with respective coefficients a_{EV} and $a_{EV\ EV}$. Such a step already may decrease a median of a prediction error from 6.3 to 4.8.

The methods further may comprise at least one calibration step. In the calibration step, a plurality of calibration measurement curves may be generated by acquiring measurement curves, preferably optical measurement curves, of a plurality of calibration fluids with the respective known target variables Y . The characteristic values 138 may be determined for each calibration measurement curve. An equation system comprising the coefficients of one or more of equations (4)-(6) may be solved, thereby determining numeric values for the coefficients.

At least one influence of at least one state variable, e.g. of the hematocrit, may be measured, preferably before the calibration, for taking the influence into account in the calibration step, e.g. in a multivariate calibration. In the calibration step, dependencies on the state variables, e.g. on hematocrit, may be taken into account. A computation and/or application of at least one derivative, e.g. by using the Savitzky-Golay filter, may take place inside the evaluation device 114. In the methods according to the present invention, terms of higher order, e.g. with a plurality of sectors, may be taken into account.

- 61 -

In step b), the evaluation rules may be adapted such that the characteristic values 138 may be linearly independent, thereby generating unique solutions for the numeric values of the coefficients.

- 5 The at least one multivariate evaluation algorithm may comprise at least one algorithm selected from the group consisting of: a partial least squares regression algorithm (PLSR); a principal component regression algorithm (PCR) a support vector machine algorithm (SVM); an artificial neuronal network algorithm (ANN); a genetic algorithm (GA). The characteristic values 138 may be independent, e.g. co-linear in a mathematical sense. The characteristic values 138 may be extracted out of one and the same measurement curve, preferably out of one and the same optical measurement curve, e.g. out of one and the same kinetic curve.

The body fluid may be selected from the group consisting of blood, interstitial fluid, urine, plasma, serum and saliva. The monitoring of the time development of the at least one measurement value indicating the progress of the detection reaction may be adapted to be an impact-free monitoring of the detection reaction without influencing the detection reaction. An impact-free monitoring may be a measurement, wherein no, preferably no significant, impact on the measurement procedure may happen during the actual generation of the measurement signals, preferably of the optical measurement curve.

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In the method according to the present invention, at least one of the two different evaluation rules may be selected from the group consisting of:

- a. using a specific measurement value of the optical measurement curve or a derivative of the optical measurement curve at a predetermined point in time as the characteristic value 138;
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- b. using a mean value of the optical measurement curve or a derivative of the optical measurement curve over a predetermined period of time as the characteristic value 138, preferably using one or more specific criteria, particularly using one or more specific conditions, which may comprise at least one end value criterion, more preferably a change rate below a predetermined threshold value;
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- 62 -

- c. using a characteristic point in time of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value 138, preferably a characteristic point in time at which one or more of the following occur: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
- d. using a characteristic parameter of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value 138, preferably a characteristic parameter at one of: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve;
- e. using a fit parameter derived by at least one fitting process as the characteristic value 138. The fitting process may imply a fitting of at least one predetermined fit curve to at least a section of the optical measurement curve or of a derivative of the optical measurement curve;
- f. using at least one value derived from a phase plot, e.g. shown in figures 6A and/or 6B, of at least two derivatives of different order of the optical measurement curve as the characteristic value 138, wherein the phase plot may comprise at least one phase space curve, wherein the value derived from the phase plot preferably may be selected from the group consisting of: a position of a center of the phase space curve; a length of the phase space curve; a phase space volume; a phase space area; a point with a maximal distance to the center of the phase space curve; a mean squared distance from the origin of the phase space.

The phase plot may be used for detecting the analyte in the sample of the body fluid, preferably the phase plot may be used for determining the concentration of glucose. The phase plot may be used for a calculation of at least one parameter, preferably for a calculation of the characteristic values 138, out of the optical measurement curve for detecting the analyte in a

sample of the body fluid, preferably for calculation of the concentration of glucose. The concentration of the analyte, e.g. the concentration of glucose, may be the result of the phase plot. The phase plot may be an input quantity for determining the concentration of the analyte, preferably for determining the concentration of glucose.

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At least two different evaluation rules may be selected from different members of the group a.-f. may be selected. Step b) may comprise generating the set of evaluation rules. The generation of the set of evaluation rules may comprise the following sub-steps:

10 b1) providing a learning set of learning measurement curves, acquired by using a learning set of learning body fluids and by monitoring detection reactions of a test substance and the test body fluids. The test body fluids and the detection reactions may be chosen such that the learning measurement curves are acquired with deferring sets of state variables;

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b2) identifying a set of candidate evaluation rules and deriving a set of candidate characteristic values 138 from the learning set of learning measurement curves;

20 b3) determining a correlation between the candidate characteristic values 138 for each candidate evaluation rule and the state variables;

b4) selecting the set of evaluation rules from the set of candidate evaluation rules by accounting for the correlations determined in sub-step b3).

25 Sub-step b3) may include determining at least one correlation parameter for each candidate evaluation rule for each state variable, preferably the Pearson correlation coefficient.

At least one of the two different evaluation rules, e.g. for determination and/or choice of optimal input parameters, e.g. optimal characteristic values 138, characteristic points may be used. The characteristic points may comprise e.g. the mentioned end value and/or a minima and/or a maxima of the optical measurement curve, preferably of the kinetic curve. Additionally or alternatively, at least one inflection point and/or at least one zero of a function and/or at least one third deviation may be used. In particular, not only the value of the function, e.g.

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of the optical measurement curve, preferably of the kinetic curve, at the time of such a zero of the function and/or of a derivative may be used, but e.g. also values of the respective derivatives may be used. E.g., the value of the first derivative, e.g. the slope, at a zero of the second derivative, e.g. the inflection point, may be an advantageous input parameter, preferably a characteristic value 138. Additionally or alternatively, the points in time 136 of the characteristic points and/or of distinctive positions of the kinetic curve, e.g. of the optical measurement curve, may be advantageous as input parameter, preferably as characteristic value 138.

In each time interval i at least one correlation parameter, preferably a correlation, for at least one of the characteristic values 138, e.g. an averaged numerical value, preferably an input parameter X_i , of the optical measurement curve, e.g. of a function curve, as well as of the first derivative and/or the second derivative and/or the third derivative of the measurement curve and/or the optical measurement curve with the concentration of glucose and/or with different other parameters, preferably with different state variables, e.g. with the hematocrit value of the used blood, may be determined.

The Pearson correlation coefficient may have arbitrary values between one, over 0, to -1. 1 may indicate a strong correlation, 0 may indicate no correlation and -1 may indicate a strong anticorrelation.

Figure 2A shows correlation coefficients of an optical measurement curve of a test element 118, preferably of a test strip to be used with the Accu Chek[®] Active system, with the concentration of glucose for each input parameter X_i . Figure 2A shows a correlation between the optical measurement curve SD0, e.g. an original curve, and the concentration of glucose in the upper diagram. Below, the correlation of the first derivative SD1, the correlation of the second derivative SD2 and the correlation of the third derivative SD3 with the concentration of glucose are shown. The x-axis may correspond to different time intervals i and the y-axis may correspond to the Pearson correlation coefficient. The data e.g. was taken using a test strip for an Accu Chek[®] Active system as test element 118.

In figure 2A, one may identify clearly a strong anticorrelation of the input parameters X_{85} - X_{101} of the optical measurement curve, preferably of the original curve SD0. Furthermore, one may recognize e.g. an anticorrelation in the first derivative SD1 for the input parameters X_{62} -

X_{66} or a correlation in the second derivative SD12 for the input parameters X_{41} - X_{43} . The correlation in the original curve, in the zero derivative SD0, with the concentration of glucose for high times may not be surprising as this may refer to the determination of the concentration of glucose by using the end value. Furthermore, correlations in time intervals of the first derivative may not be surprising as the gradient of the optical measurement curve for small times may be used for an evaluation. E.g., one may use the first derivative instead of the end value for an evaluation of the optical measurement curves. Interferences, e.g. caused by state variables, may influence the first derivative signal too much, thus, a prediction of the concentration of glucose may be too inexact.

In sub-step b4), a Merit value may be calculated for each correlation. The selecting of the set of evaluation rules from the set of candidate evaluation rules may be performed by accounting for the Merit values. In sub-step b4), a candidate evaluation rule may be determined to be an evaluation rule if the corresponding correlation determined in sub-step b3) may fulfill at least one predetermined condition. A method for use in a multivariate data analysis may be the so-called "partial least squares regression" (PLS).

In the embodiment shown in figure 2A, the end value may be combined with at least one gradient value, e.g. a first derivative. For understanding, which parameters, e.g. which characteristic values 138, may be suiting best for a multivariate analysis, a correlation analysis may be useful. Within the generation of the set of evaluation rules, input parameters and/or characteristic values 138, may be chosen, which may correlate maximally with the concentration of the analyte and probably not with other state variables, respectively.

Figure 2B shows for the identical optical measurement curve, e.g. for the identical data set, as for figure 2A not the correlation with the concentration of glucose, but the correlation coefficient referring to the concentration of hematocrit, e.g. to a hematocrit value. Figure 2B particularly shows, that e.g. the parameter X_{69} , the first derivative of the optical measurement curve for a time interval between 7,2-8,7 seconds, correlates very strong with the concentration of hematocrit, e.g. with the hematocrit value, but basically not, as shown in figure 2A, with the concentration of glucose.

- 66 -

E.g., by using the parameters X_{83} and X_{69} , e.g. as characteristic values 138, an influence of hematocrit may be divided from an influence of the concentration of glucose, e.g. by smart analysis, preferably by using the multivariate analysis. By using these two parameters, e.g. together with the end value, in a PLS analysis, one may get already a significant improvement
5 of the prediction of the concentration of glucose.

Figure 2B shows, similar to figure 2A, correlations between the optical measurement curve SD0, the first derivative SD1, the second derivative SD2 and the third derivative SD3 and the concentration of hematocrit. The x-axis may preferably correspond to the different time inter-
10 vals, indicated by i . The y-axis may correspond to the Pearson correlation coefficient. i preferably may be related to at least one index of a parameter matrix. The data shown in figure 2B preferably may be related to measurements using test strips for an Accu Chek[®] Active system.

Figure 2C shows the deviation d in mg/dL or % against a reference glucose concentration in
15 mg/dL for different hematocrit values. The different hematocrit values are indicated by differently shaped symbols as described in the legend comprised by figure 2C. Figure 2C in particular shows the difference between the concentration of glucose determined by an embodiment of a method according to the present invention using the multivariate analysis and a reference concentration, preferably a reference glucose concentration, for different hematocrit
20 values. The measurements according to figure 2C are e.g. carried out by using a sample analysis device 110 according to the present invention, preferably by using an Accu Chek[®] Active system by Roche Diagnostics Deutschland GmbH. Figure 2C shows the deviation d of the prediction of the concentration of glucose from the reference glucose concentration r as reference value, wherein the predicted concentrations of glucose preferably may be determined by
25 using the PLS analysis based on the end value and the parameter X_{69} and X_{83} . By deriving the median of the absolute deviation and/or of the relative deviation one may get as total error a value of 3.5, compared to a value of 6.3 by using an end value criterion known from prior art.

By using a method according to the present invention, preferably by using the multivariate
30 evaluation algorithm, one may be able to distinguish between changes of the optical measurement curve, preferably of the kinetic curve, e.g. changes of the signal, induced by the concentration of glucose and induced by the concentration of hematocrit.

- 67 -

The methods according to the present invention may provide the possibility for a prediction, preferably at least a rough prediction, of determining the concentration of hematocrit out of the optical measurement curve, e.g. out of the kinetic curve. This is shown in figure 2D. Figure 2D in particular shows a predicted hematocrit h in % against a reference hematocrit r_h in %.

5 %. The different symbols may reflect different concentrations of glucose. Figure 2D in particular shows the predicted hematocrit determined out of the parameters, e.g. input parameters, X_{69} and X_{85} , and the end value compared to an actual value of hematocrit, preferably the reference hematocrit r_h exemplary for a test strip 120 to be used with the Accu Chek® Active system for glucose, a test element 118 specified for analyzing glucose.

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The present invention comprises a method for characterizing a sample of a body fluid. The method comprises the following steps:

- 15 A) bringing the sample of the body fluid into contact with at least one test substance, thereby initiating a detection reaction of the test substance and the sample of the body fluid. The detection reaction is known to be influenced by a set of state variables. Each state variable characterizes at least one of a state of the sample of the body fluid and the condition of the detection reaction;
- 20 B) monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at different points in time 136;
- 25 C) evaluating the optical measurement curve by using a method for detecting an analyte in a sample of a body fluid according to the present invention.

In the methods for detecting an analyte in a sample of a body fluid, evaluation rules may be used according to the present invention. The best parameters X_i may be searched for the multivariate data analysis. According to the present invention, the respective numeric coefficient

30 a_i of a multivariate analysis for a code curve may be found, e.g. as numeric values of the coefficients.

Figure 3 shows correlations for another embodiment of the present invention. Figure 3 in particular shows correlations between an optical measurement curve, e.g. an original remission kinetic curve SD0, preferably original data, the first derivative, e.g. the slope SD1, the second derivative, e.g. the curvature SD2, and the third derivative SD3 and the concentration of glucose, preferably for test strips 120. The y-axis relates to the Pearson correlation coefficient and the x-axis relates to the time t in seconds. The white marked intervals i preferably may indicate the time intervals chosen for the characteristic values 138. Figure 3 e.g. shows 22 white marked intervals, thus 22 characteristic values 138 may be used in the further analysis, e.g. in the multivariate analysis. In this embodiment, a suppression of an influence of the temperature on the optical measurement curve, e.g. on the kinetic curve, may be compensated. In this embodiment, at least one, preferably several, features described above may be used. E.g., the methods for determining and/or identification of the input parameters and/or of the characteristic values 138 may be used. As a result, the white marked time intervals may be advantageous, e.g. due to using the methods according to the present invention for the selection of the evaluation rules and/or of characteristic values 138.

E.g., carba nicotinamide adenine nucleotide (cNAD) may serve as co-factor for an enzyme-based detection, as e.g. described in von Ketteler et al., Fluorescence Properties of Carba Nicotinamide Adenine Dinucleotide for Glucose Sensing, CHEMPHYSCHEM 2012. By using a cNAD test substance cNADH may be generated proportional to the concentration of glucose. cNADH may not only absorb ultraviolet UV light with a wavelength of 360 nm during excitation with light but may also emit fluorescence light, preferably fluorescence light with an emission maximum at 460 nm. The fluorescence light may be detected by using an optical filter, preferably a simple optical filter, in front of a photodiode as detector 130. Alternatively or additionally, the detection of the remission and/or the detection of the fluorescence light may be advantageous for use in analyzing the sample, in particular in the multivariate analysis.

Figure 4 shows in particular correlations between the optical measurement curve, e.g. an original fluorescence kinetic curve SD0, the first derivative SD1, e.g. the slope, the second derivative SD2, e.g. the curvature, the third derivative SD3 and the concentration of glucose for a cNAD strip. The white marked time intervals were chosen for a further analysis, e.g. for the multivariate analysis in this example. Figure 4 shows correlation coefficients against a time t

in s. Preferred input parameters, e.g. characteristic values 138, may be deduced from a correlation analysis. The correlation analysis may be shown in figure 4. These preferred input parameters may alternatively or additionally be used, e.g. in a mixture with remission data, e.g. as shown in figure 3.

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Using the multivariate data analysis according to the present invention for the prediction of the concentration of glucose may reduce a dependency, even a strong dependency, of a prediction of a concentration of glucose on a temperature. The strong dependency of the prediction of the concentration of glucose on the temperature for a method using only the end value, e.g. as known from prior art, may be in average 1.76 %/K. The dependency by using the method according to the present invention, e.g. by using a PLS analysis on basis of the optical measurement curve, e.g. the kinetic curve, of the remission, may be 0.47 %/K. The dependency when using the method according to the present invention, e.g. when using the PLS analysis of the optical measurement curve, e.g. of the kinetic curve, generated by photometry and fluorescence, may be 0.02 %/K. This reduction of the dependency is shown in figure 5.

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Figure 5 shows the dependency of the deviation between a prediction of the concentration of glucose and a real concentration of glucose on the influence of strong temperature changes dc_{glu}/dT in %/K against the real glucose concentration r_g in mg/dl. Curve 140 refers to a measurement using an end point detection, e.g. the end value criterion, known from prior art, wherein the dependency on a temperature may be about 1.76 %/K. Curve 142 shows a measurement using a method according to the present invention, preferably using the multivariate analysis, by using the remission, particularly showing a dependency on a temperature of 0.47 %/K. Curve 144 refers to a measurement using a method according to the present invention and using the remission and the fluorescence, wherein the dependency from a temperature may be 0.02 %/K. The deviations may achieve significantly more than 2%/K for using a method known from prior art with end point value detection. The deviation may be reduced by using the multivariate analysis, in average less than about 1 %/K, preferably by using the remission, and less than about 0.2 %/K, preferably by using the remission and the fluorescence. Thus, by using a method according to the present invention, compensation of deviations due to temperature may be possible without using a temperature sensor, preferably only by using the optical measurement curve, e.g. the kinetic curve. In a method according to the

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- 70 -

present invention, preferably no temperature sensor may be used for detecting the analyte in the sample of the body fluid.

Figures 3 and 4 may show methods for selection of at least one of the evaluation rules. Figures 6A and 6B may show an advantageous method for the determination of the characteristic values 138, e.g. of the parameters, which may be called phase space method. The phase space method may comprise generating a phase plot, preferably comprising at least one phase space curve, of at least two derivatives of different order of the optical measurement curve. Auxiliary quantities may be derived from the phase plot. Most preferably, the auxiliary quantities may comprise at least one input parameter, most preferably at least one characteristic value 138. At every point in time, preferably of the optical measurement curve, a pair of data, preferably comprising at least one point out of the optical measurement curve and/or at least one point of at least one derivative of the optical measurement curve, may be plotted. Figure 6A and 6B show examples of phase plots. The phase plot may also be called phase space diagram. The phase plot may comprise pairs of points. One data point may relate to one point in time in the optical measurement curve, e.g. in the kinetic curve. The pair of points may comprise a remission and the first derivative of the remission with respect to time, as shown in figures 6A and 6B. Figures 6A and 6B show phase plots, wherein remission values S are plotted over the first derivative of the remission with respect to time dS/dt in $1/s$, thus, over the slope of the optical measurement curve, e.g. over the slope of the kinetic curve. Figure 6A shows a phase plot for a temperature of $5 \pm 1^\circ C$ for different concentrations of glucose and figure 6B shows a phase plot for a temperature of $45 \pm 1^\circ C$ for different concentrations of glucose. The curves 146 belong to a concentration of glucose of 0 ± 5 mg/dl, the curves 148 belong to a concentration of glucose of 89 ± 5 mg/dl, the curves 150 belong to a concentration of glucose of 258 ± 5 mg/dl and the curves 152 belong to a concentration of glucose of 554 ± 5 mg/dl. Curve 146, curve 148, curve 150 and curve 152 preferably may be phase space curves. Figures 6A and 6B may show phase plots comprising phase space curves. The curve may start at a point (0-0) and may propagate on loops to a value near the value of $dS/dt \approx 0$. $dS/dt \approx 0$ may be the basis for an evaluation using a remission value by using a stop criterion. The loops before reaching the end point in a phase plot may comprise information about at least one state variable, preferably about the temperature, which may be used for a compensation by the multivariate analysis, e.g. by a multivariate evaluation. E.g., one may ask at which remission value the slope may have its minimum, e.g. at $S = 0.44$ or at $S = 0.40$. $S = 0.44$ may

- 71 -

indicate a low temperature, preferably almost independent on a concentration, e.g. on a glucose concentration. $S = 0.40$ may indicate a high concentration, e.g. a high concentration of glucose, and a high temperature. The point with the smallest, preferably negative, slope, preferably the furthest left point of the phase plot, may be helpful, e.g. in all its information, e.g. in respect to its remission value and/or its first and/or second and/or third derivative and/or its point in time at which this point may be reached.

Additionally or alternatively, values directly generated by using the phase plot may be used, e.g. a position of a center of the phase space curve, preferably coordinates of a center of the phase space curve, e.g. of the phase plot, a length of the phase space curve, e.g. of the phase plot, and/or a length of the phase space curve, e.g. of the phase plot, and/or a phase space volume and/or a phase space area and/or points with a maximal distance to the center of the phase space curve and/or a mean squared distance from the origin of the phase space, e.g. indicating $dS/dt = 0$ and/or other properties of the phase space plot. Figures 6A and 6B show, e.g., that a slope and/or an end value in phase space, e.g. in a phase space curve and/or in a phase space plot, may depend on the concentration of glucose. Figures 6A and 6B in particular show that phase plots and/or phase space curves may look different for different temperatures. Differences of phase plots for different temperatures may be analyzed e.g. by using mathematical methods. E.g., mathematical methods may be used for distinguishing phase plots and/or phase space curves in terms of at least one state variable, preferably in terms of the temperature, e.g. by calculating the value derived from the phase plot as described above.

An advantage of the multivariate analysis may be a simultaneous determination of calibration data and/or of at least one target value and/or of at least one state value. The multivariate analysis may alternatively or additionally comprise a consecutive procedure. A consecutive procedure e.g. may be used in a method using a phase plot, e.g. in the phase space analysis as described above. E.g., an end value may be determined out of the optical measurement curve, e.g. out of the kinetic curve, wherein the end value may lead to a prediction of the concentration of glucose. A value of a prediction of the concentration of glucose may be corrected depending on a presence of other criteria of the optical measurement curve, e.g. of the kinetic curve. The method, e.g. a correction of the predicted concentration of glucose, may comprise at least one weighting. E.g., the weighting may be done in respect to a temperature and/or a hematocrit value and/or a humidity and/or an enzyme activity and/or a date of expiry and/or a

marker of an endurance, wherein at least one of these properties may be determined out of the optical measurement curve, e.g. out of the kinetic curve. Alternatively, the weighting may comprise at least one other property of the method and/or of the device and/or of the optical measurement curve.

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The methods according to the present invention may comprise a determination of parts of the optical measurement curve which may be used for a calculation of different parameters, preferably of the characteristic values 138. The calculation may comprise at least one weighting. The weighting may be used for stronger weighting of selected parameters, preferably of selected characteristic values 138. E.g., a parameter, preferably a characteristic value 138, related to haematocrit may be determined by using a first part of the optical measurement curve. A parameter, preferably a characteristic value 138, being related to a dependency on a temperature and/or on another state variable may be determined by using at least one part of the optical measurement curve being different from the first part of the optical measurement curve.

10 The method according to the present invention may comprise the determination of the parts of the optical measurement curve, e.g. a time interval, and a correlation of the parts of the optical measurement curve with at least one parameter, preferably a correlation of the part of the optical measurement curve with at least one state variable.

20 The PLS analysis described above preferably may be a linear multivariate method. The relation between the concentration of glucose and the remission and/or the fluorescence may not be linear, as may e.g. be seen on calibration curves based on the end value criterion. But using the PLS analysis may already lead to an improvement. The improvement may be e.g. caused by the quadratic term in the multivariate analysis, e.g. in the algorithm. Additionally or alternatively, the optical measurement curve, e.g. the original kinetic curve, may be transformed, e.g. by a term being not proportional to the remission, but e.g. by a term being proportional to $1/R$ and/or $1/R^2$ and/or $\log(R)$ and/or $R^{1/2}$ and/or $R^{3/2}$ and/or a similar term, wherein R may be the remission. Additionally or alternatively, inherent non-linear multivariate methods may be used. Inherent non-linear multivariate methods may comprise e.g. a neural network, e.g. with
25 sigmoid neurons, and/or support-vector machines. Thus, the method according to the present invention may comprise at least one neural network and/or at least one support-vector machine. Alternatively or additionally, genetic algorithms may be used in the method according to the present invention.

- 73 -

The method according to the present invention may comprise at least one correlation analysis and/or at least one determination of characteristic values 138 and/or a search in the phase space and/or a renormalization by non-linear functions and/or the use of PLS algorithms and/or the use of non-linear classification methods and/or regression methods as ANNs and/or SVMs.

The characteristic values 138 preferably may be independent. The correlation analysis and/or the calculation of the merit function may be used for avoiding dependencies of the characteristic values 138.

The optical measurement curve, e.g. at least one kinetic curve, may be linearized. For linearization, a model for a prediction of the concentration of glucose out of the remission value may be derived by using an end value evaluation, e.g. based on a -2% end value, out of the optical measurement curve, e.g. out of a precision set of data. This model may be applied to the whole optical measurement curve, e.g. to the whole kinetic curve, afterwards. Furthermore, the method may comprise an averaging of at least a part of the optical measurement curve, e.g. of the kinetic curve. Furthermore, at least one merit-function, preferably for the linearized optical measurement curve, e.g. for the linearized set of data may be used. E.g., in the linearized optical measurement curve, preferably in the linearized set of data, a correlation analysis may be executed for each temperature for the characteristic values 138, e.g. for characteristic curve parameters. The correlation analysis may comprise a calculation of correlations with a concentration of glucose and/or a concentration of hematocrit and/or a humidity. Merit-functions may be derived out of the correlations, e.g. out of correlation values, preferably out of the Pearson correlation coefficients, e.g. by using the following formula:

$$\text{Merit value} = \frac{\text{correlation}^2}{|\text{correlation}_{\text{glucose}}| + |\text{correlation}_{\text{hematocrit}}| + |\text{correlation}_{\text{humidity}}|}$$

The *correlation* may be the correlation coefficient for the concentration of glucose *correlation_{glucose}* or the correlation coefficient for the concentration of hematocrit *correlation_{hematocrit}* or the correlation coefficient for the concentration of the humidity

correlation_{humidity}. The Merit value may be used for extracting the characteristic values 138.

The methods according to the present invention may comprise a multivariate analysis for evaluation of an optical measurement curve, preferably a kinetic curve, most preferably a kinetic curve for the determination of a glucose value, generated by an optical measurement, but also by other types of measurements. The first characteristic value or the second characteristic value may not be the dry empty value of the measurement, i.e. may be different from a dry empty value.

For a successful use of the multivariate analysis, a correlation being as significant as possible between the target values and/or the state variables, preferably the concentration of glucose, and different state variables and/or disturbing factors may be useful. The characteristic values 138 generated thereby may be used, e.g. together with, preferably statistically generated, concrete number values for coefficients, e.g. the Pearson correlation coefficient, for implementing in a simple algorithm in the device according to the present invention. The precision may be increased by taking into account cross-terms and/or terms of higher order.

In the methods according to the present invention, the characteristic values 138 may be different parameters. The different parameters may be weighted with different weights per parameter. The characteristic values 138 may be extracted from defined parts of the optical measurement curve and/or may be single characteristic points of the optical measurement curves, e.g. of the kinetic curve. The method according to the present invention may comprise a determination of the parts of the optical measurement curve and their assignment for getting the parameters, preferably for getting the characteristic values 138.

The optical measurement curve, e.g. the kinetic curve, may be divided in parts of the optical measurement curve, wherein this may enable an application of individual algorithms on the parts of the optical measurement curve.

E.g., defined time intervals, e.g. time windows, in the optical measurement curve may be attached to an influence of concrete state variables, e.g. concrete disturbance values. The meth-

- 75 -

od according to the present invention may comprise a combination of determined parameters, e.g. a combination of the characteristic values 138.

A selection of the parts of the optical measurement curve, e.g. of the time intervals, may be made by calculating the merit function, e.g. as part of the correlation analysis. The method according to the present invention also may comprise the end value criterion. The characteristic values 138 may be independent from each other. The target variable Y may be the concentration of the analyte, which may be determined. The characteristic values 138 may be determined by using the evaluation rules. The characteristic value 138 may be characteristic for at least one variable, e.g. for at least one state variable, preferably for at least one target variable, most preferably for the concentration of the analyte. The characteristic value 138 preferably may be different from the concentration of the analyte, preferably different from the concentration of glucose. The multivariate analysis may be done with at least two different characteristic values 138 and/or with at least two different variables. At least one variable, e.g. at least one characteristic value 138, may be different from the concentration of the analyte. The variable may be determined out of the characteristic value 138, e.g. by using at least one evaluation rule. The characteristic value 138 being different from the concentration of the analyte may be determined by using the evaluation rule. The multivariate analysis may use at least one variable being different from the concentration of the analyte. The variable being different from the concentration of the analyte may be determined by using the evaluation rule.

Figure 7 shows another embodiment according to the present invention. The state variables may comprise at least one disturbance value. Influences of disturbance values on the measurement values, preferably on values for the concentration of glucose, may be compensated by an analysis of at least one derivative, e.g. of more than two derivatives, of the optical measurement curve and/or of the measurement curve. An analysis of derivatives of the measurement curve, preferably of the optical measurement curve, most preferably of a kinetic curve, may show correlations with the concentration of glucose as well as correlations with disturbance values, e.g. state variables, as the temperature and/or the humidity, preferably an air humidity, and/or the concentration of hematocrit. In particular, at very early times of the kinetic curve, e.g. directly after combining the test substance and the sample of the body fluid, it may be probable that there may be no correlation with the concentration of glucose. But directly starting diffusion processes and/or dissolution processes usually may cause a depend-

ency on the disturbance values. These dependencies on the disturbance values may be visible, analogue to an evaluation of the concentration of glucose, by a structure of the derivatives of the measurement curve, preferably of the optical measurement curve, e.g. in a maxima or a minimum. This correlation with at least one disturbance value may be used for a compensation from the disturbance value and/or from the disturbance effect, preferably for calculating a compensation. E.g., a first minimum of the first derivative may depend e.g. on a temperature and/or on the concentration of hematocrit, but may be independent on the concentration of glucose. This phenomena may be referred to as the wetting dip or wetting step in the measurement curve and/or in a first order or higher order derivative of the measurement curve. A combination of this phenomenon, preferably of the compensation, with the end value criterion discussed above may lead to a value for the concentration of glucose which may be corrected from influences caused by a concentration of hematocrit. This method may comprise a multivariate analysis, preferably a bivariate analysis. A formula for deriving the concentration of glucose may be searched "manually", e.g. by comparison of measurement curves and/or derivatives of measurement curves and/or influences of disturbance values and/or state variables by a skilled person. Figure 7 in particular shows the first derivative 1st d of a measurement curve, preferably of an optical measurement curve, in arbitrary units, against the time t, preferably against the measurement time t, in seconds. Figure 7 shows an example of measurement values depending on the temperature and on the concentration of glucose, which may at early times only depend on the temperature. Similar effects may be present also for other disturbance values, e.g. for the concentration of hematocrit. Figure 7 in particular shows the dip, particularly the wetting dip, visible in the first derivative 1st d. The temperatures may vary between 5 and 45 °C. The data comprises different concentrations of glucose, wherein they may not be distinguishable at these times, preferably at these times of the kinetic curve. Only a dependency on the temperature is clearly visible. By using the depth of the dip a, at least partial, compensation for the temperature dependency may be possible.

Additionally or alternatively, the predetermined multivariate evaluation algorithm may comprise a procedure which may involve a first-order, a second-order and/or a higher-order decision tree 154, 164 which may comprise at least one decision branch which may allow selecting one out of at least two, preferably two, alternative procedures based on an assessment whether a predetermined condition may be fulfilled or not. Hereby, the predetermined condition may assess whether a definite value falls within a predetermined range or not and may,

thus, offer a decision between performing or not performing a specific procedure or performing the specific procedure under a specific parameter, with a specific parameter set, or within a specific parameter range, wherein the specific parameter may comprise the first or the second characteristic value 138.

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As a non-limiting example, the predetermined multivariate evaluation algorithm may comprise the following function involving a first-order decision tree $f(X_1, X_2)$,

$$Y = f(X_1, X_2) = \{g_1(X_2) \text{ for } \textit{cond}(X_1); g_2(X_2) \text{ for NOT } \textit{cond}(X_1)\}, (7)$$

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wherein, depending on the assessment whether the predetermined condition $\textit{cond}(X_1)$, which may depend on the first characteristic value X_1 , may be fulfilled or not, the estimate value for target variable Y may be derived according to Equation (7) by either using a first function $g_1(X_2)$ or an alternative second function $g_2(X_2)$ which both may depend on the second characteristic value X_2 .

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As an example related to Equation (7), Figure 8 shows a first decision tree 154, wherein a hematocrit correction 156 of the glucose concentration may only be applied outside a predetermined hematocrit range 158, i.e. only such glucose values may be corrected for which such a correction may be required. In particular, after a determination 160 of the second characteristic value 138, i.e. the glucose concentration, it may firstly be determined whether the first characteristic value 138, i.e. the hematocrit, may be inside or outside the predetermined hematocrit range 158, preferably covering the range from 35 % to 50 %. However, other values for the predetermined hematocrit range 158 are possible. In this exemplary first decision tree 154, the hematocrit correction 156 of the second characteristic value 138, i.e. the glucose concentration, may only be applied in case the first characteristic value 138, i.e. the hematocrit, may be outside the predetermined hematocrit range 158. Thus, a determination 162 of the target value Y involves both the second characteristic value 138, i.e. the glucose concentration, and the first characteristic value 138, i.e. the hematocrit.

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Consequently, the first decision tree 154 as exemplary depicted in Figure 8 exhibits the positive effect that only such glucose values are submitted to the hematocrit correction 156 where the hematocrit correction 156 may be required for a further processing of the respective glu-

cose values, in particular for rare cases in which a patient may display a very low or a very high hematocrit. Therefore, this kind of discrimination according to the first decision tree 154 may, thus, help to improve both the speed and the quality of the determination 162 of the target value Y.

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As a further example based on Equation (7), the determination 162 of the target value Y may be performed according to a second decision tree 164 as exemplary depicted in Figure 9. According to the second decision tree 164, the method may start with a determination 166 of an end value, from which a preliminary value for the second characteristic value 138, i.e. the glucose concentration, may be derived. According to an assessment whether the preliminary value for the second characteristic value 138, i.e. the glucose concentration, may fall within a predetermined glucose concentration range 168, firstly, respective first and second threshold values 170, 172 for determining the actual glucose concentration may be selected. In this example, in case the preliminary value for the glucose concentration may be estimated to be below 100 mg/dl, first and second threshold values 170 of -2 %/s for the first threshold and of -0.5 %/s for the second threshold may be particularly applied, whereas in case the preliminary value for the glucose concentration may be estimated to be 100 mg/dl or more, the above mentioned values of -5 %/s and of -2 %/s may be selected as first and second threshold values 172. However, other values the first threshold and for the second threshold may be chosen.

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Secondly, in an additional second-order decision branch of the second decision tree 164, a hematocrit correction 156 of the glucose values may only be applied outside the predetermined hematocrit range 158. As already described above in relation to figure 8, the hematocrit correction 156 of the glucose concentration may only be performed in case the hematocrit may take a value outside a range of 35 % to 50 %. However, other values are possible. According to the discrimination as depicted in Figure 9, the determination 162 of a final value for the glucose concentration may be determined here also with or without hematocrit correction 156 depending on the actual value of the first characteristic value 138, i.e. the hematocrit. Hereby, the actual values chosen for the hematocrit correction 156 may be independent from the second-order decision branch of the second decision tree 164. Alternatively, for the hematocrit correction 156 actual values may be chosen which might depend on which second-order

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- 79 -

decision branch of the second decision tree 164 the hematocrit correction 156 may be performed.

Consequently, the second decision tree 164 as exemplary depicted in Figure 9 may exhibit the positive effects that, firstly, very low glucose values even down to 40 mg/dl or below may be correctly determined, and that, secondly, only such glucose values may be submitted to the hematocrit correction 156 where it may be required in particular for rare cases in which a patient may display a very low or a very high hematocrit. Therefore, this kind of discrimination according to the second decision tree 164 may, thus, help to improve both the speed and the quality of the determination 162 of the final value of the glucose concentration over a much larger range of glucose concentrations than before.

As a further example, Figure 10 shows a third decision tree 174, wherein from the determination 166 of the end value the preliminary value for the second characteristic value 138, i.e. the glucose concentration, may be derived. According to an assessment whether the preliminary value for the glucose concentration may fall within the predetermined glucose concentration range 168, the preliminary value for the glucose concentration as acquired by the determination 166 of the end value may be kept or not. In the latter case, a determination 176 of a decay constant Γ or a quantity related to the decay constant Γ , such as a quantity proportional to the decay constant Γ or proportional to the inverse $1/\Gamma$ of the decay constant, may be performed, wherein the decay constant Γ may describe an exponential characteristic within at least an evaluation part of a measurement curve related to the progress of a detection reaction of the glucose concentration. According to a further assessment 178 which might deliver an answer to the question whether the decay constant Γ or the quantity related to the decay constant Γ may be equal to or exceed a predefined constant Γ_0 , the preliminary value for the glucose concentration as acquired by the determination 166 of the end value may still be kept or not. In the latter case, an additional evaluation procedure 180 for determining the glucose concentration may be performed, wherein the additional evaluation procedure 180 may take the hematocrit into account. Herein, the additional evaluation procedure 180 may further comprise another decision branch (not depicted here) which might branch out to different hematocrit evaluation procedures depending on whether the decay constant Γ or the quantity related to the decay constant Γ may be equal to or exceed a further predefined constant Γ_1 .

List of reference numbers

110	sample analysis device	178	further assessment
112	measuring unit	180	additional evaluation procedure
114	evaluation device		
116	evaluation unit		
118	test element		
120	test strip		
122	hand-held device		
124	Monitor		
126	input panel		
128	light source		
130	Detector		
132	Computer		
134	computer network		
136	points in time		
138	characteristic value		
140	Curve		
142	Curve		
144	Curve		
146	Curves		
148	Curves		
150	Curves		
152	Curves		
154	first decision tree		
156	hematocrit correction		
158	predetermined hematocrit range		
160	determination of the glucose concentration		
162	determination of the target value		
164	second decision tree		
166	determination of the end value		
168	predetermined glucose concentration range		
170	first and second threshold values		
172	first and second threshold values		
174	third decision tree		
176	determination of decay constant		

Claims

What is claimed Is:

1. A method for detecting an analyte in a sample of a body fluid, the method comprising the following steps:
 - a) providing at least one optical measurement curve, wherein the optical measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid, wherein the measurement values contained in the optical measurement curve are acquired at differing points in time (136), wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;
 - b) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value (138) from the optical measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1\dots N}$ from the optical measurement curve, the set of characteristic values (138) comprising at least one first characteristic value (138) being derived from the optical measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value (138) being derived from the optical measurement curve by using at least one second evaluation rule from the set of evaluation rules, the second evaluation rule being different from the first evaluation rule;
 - c) performing at least one multivariate analysis of the at least one first characteristic value (138) and the at least one second characteristic value (138) by using at least one predetermined multivariate evaluation algorithm, the multivariate evaluation algorithm being an algorithm adapted to derive at least one result from at least two variables, wherein the at least one first characteristic value (138) and the at least one second characteristic value (138) are used as the at least two variables, thereby deriving at least one estimate value for at least one target variable Y of the state variables;
 - d) determining a concentration of the at least one analyte by using the at least one target variable Y.
2. The method according to claim 1, wherein the state variables are selected from the group consisting of: a composition of the sample of the body fluid; a content of at least one particulate component of the sample of the body fluid; a temperature of the sample of the body fluid; a humidity of an ambient atmosphere surrounding the sample of the body fluid; a storage time of the test substance; an interfering substance; alterations of the sample or of certain properties of the sample caused by pharmacological treatment of a donor of the sample.

3. The method according to one of claims 1 or 2, wherein the first evaluation rule may not be transformed into the second evaluation rule by a time transformation.
- 5 4. The method according to one of claims 1 to 3, wherein the second evaluation rule differs from the first evaluation rule in at least one of: in at least one coefficient; in at least one parameter; in at least one component related to the algorithm.
- 10 5. The method according to claim 4, wherein the algorithm of the first evaluation rule differs from the algorithm of the second evaluation rule in a point in time.
- 15 6. The method according to one of claims 1 to 5, wherein a third evaluation rule is provided, wherein, in step c), the at least one first characteristic value is derived from the first evaluation rule, and wherein, in the multivariate evaluation algorithm, the second evaluation rule or the third evaluation rule is used depending on the at least one first characteristic value.
- 20 7. The method according to one of claims 1 to 6, wherein the first characteristic value (138) is determined by using a first time interval of the optical measurement curve, wherein the second characteristic value (138) is determined by using a second time interval of the optical measurement curve, wherein the first time interval of the optical measurement curve is different from the second time interval of the optical measurement curve.
- 25 8. The method according to claim 7, wherein a target value is different from the concentration of the at least one analyte.
- 30 9. The method according to one of claims 1 to 8, wherein the at least two evaluation rules are adapted to derive the characteristic values (138) from at least two derivatives of the optical measurement curve.
10. The method according to claim 9, wherein the at least two derivatives are derivatives comprising at least two derivatives of different order.
- 35 11. The method according to one of claims 9 or 10, wherein the derivatives are generated by using at least one filtering algorithm.

12. The method according to one of claims 1 to 11, wherein the set of characteristic values (138) contains 2-20 characteristic values (138).
13. The method according to one of claims 1 to 12, wherein the target variable Y comprises the concentration of the at least one analyte in the sample of the body fluid.
14. The method according to one of claims 1 to 13, wherein, in step d), in addition to the at least one target variable Y, at least one electrochemical measurement value is used for determining the concentration of the analyte, wherein the electrochemical measurement value is determined by using at least one electrochemical measurement.
15. The method according to claim 14, wherein, by using the electrochemical measurement value, an approximated value of the concentration of the at least one analyte in the sample of the body fluid is determined, wherein the target value Y is used for correcting the approximated value.
16. The method according to one of claims 1 to 15, wherein the predetermined multivariate evaluation algorithm comprises at least one polynomial algorithm selected from:
- $$Y = A \cdot X, \quad (1)$$
- $$Y = X^T \cdot A \cdot X \quad (2)$$
- and
- $$Y = X^T \cdot (X^T \cdot A \cdot X), \quad (3),$$
- wherein A is a one-dimensional, a two-dimensional or a three-dimensional evaluation tensor.
17. The method according to one of claims 1 to 16, wherein the predetermined multivariate evaluation algorithm comprises at least one algorithm selected from:

$$Y = \sum_i a_i \cdot X_i, \quad (4)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j, \quad (5)$$

$$Y = \sum_i a_i \cdot X_i + \sum_{i,j} a_{ij} \cdot X_i \cdot X_j + \sum_{i,j,k} a_{ijk} \cdot X_i \cdot X_j \cdot X_k, \quad (6),$$

wherein a_i , a_{ij} , a_{ijk} are predetermined coefficients and wherein i , j and k are, mutually independently, integers from 1 to N .

18. The method according to claim 17, further comprising at least one calibration step, wherein, in the calibration step, a plurality of calibration measurement curves is generated by acquiring measurement curves of a plurality of calibration fluids with the respective known target variables Y , wherein the characteristic values (138) are determined for each calibration measurement curve, wherein an equation system comprising the coefficients of one or more of equations (4)-(6) is solved, thereby determining numeric values for the coefficients.
19. The method according to claim 18, wherein, in step b), the evaluation rules are adapted such that the characteristic values (138) are linearly independent, thereby generating unique solutions for the numeric values of the coefficients.
20. The method according to one of claims 1 to 19, wherein the at least one multivariate evaluation algorithm comprises at least one algorithm selected from the group consisting of: a partial least squares regression algorithm (PLSR); a principal component regression algorithm (PCR); a support vector machine algorithm (SVM); an artificial neural network algorithm (ANN).
21. The method according to one of claims 1 to 20, wherein the at least one multivariate evaluation algorithm comprises a function involving at least one decision tree (154, 164), wherein the decision tree (154, 164) comprises at least one decision branch which allows selecting one out of at least two alternative procedures based on an assessment whether a predetermined condition may be fulfilled.
22. The method according to claim 21, wherein the decision branch offers a decision between performing or not performing a specific procedure under a specific parameter, with a specific parameter set, or within a specific parameter range, wherein the specific parameter comprises the characteristic value (138).
23. The method according to one of claims 1 to 22, wherein the body fluid is selected from the group consisting of blood, interstitial fluid, urine, plasma, serum and saliva.
24. The method according to one of claims 1 to 23, wherein the monitoring of the time development of the at least one measurement value indicating the progress of the detection

reaction is adapted to be an impact-free monitoring of the detection reaction without influencing the detection reaction.

25. The method according to one of claims 1 to 24, wherein at least one of the two different evaluation rules is selected from the group consisting of:
 - a. using a specific measurement value of the optical measurement curve or a derivative of the optical measurement curve at a predetermined point in time as the characteristic value (138);
 - b. using a mean value of the optical measurement curve or a derivative of the optical measurement curve over a predetermined period of time as the characteristic value (138);
 - c. using a characteristic point in time of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value (138);
 - d. using a characteristic parameter of the optical measurement curve or of a derivative of the optical measurement curve as the characteristic value (138);
 - e. using a fit parameter derived by at least one fitting process as the characteristic value (138), wherein the fitting process implies a fitting of at least one predetermined fit curve to at least a section of the optical measurement curve or of a derivative of the optical measurement curve; and
 - f. using at least one value derived from a phase plot of at least two derivatives of different order of the optical measurement curve as the characteristic value (138), wherein the phase plot comprises at least one phase space curve.
26. The method according to claim 25, wherein at least two different evaluation rules selected from different members of the group a.-f. are selected.
27. The method according to one of claims 1 to 26, wherein step b) comprises generating the set of evaluation rules, the generating of the set of evaluation rules comprising the following sub-steps:
 - b1) providing a learning set of learning measurement curves, acquired by using a learning set of learning body fluids and by monitoring detection reactions of a test substance and the test body fluids, wherein the test body fluids and the detection reactions are chosen such that the learning measurement curves are acquired with differing sets of state variables;
 - b2) identifying a set of candidate evaluation rules and deriving a set of candidate characteristic values (138) from the learning set of learning measurement curves;
 - b3) determining a correlation between the candidate characteristic values (138) for each candidate evaluation rule and the state variables;

- b4) selecting the set of evaluation rules from the set of candidate evaluation rules by accounting for the correlations determined in sub-step b3).
28. The method according to claim 27, wherein sub-step b3) includes determining at least one correlation parameter for each candidate evaluation rule for each state variable.
29. The method according to one of claims 27 or 28, wherein, in sub-step b4), a Merit value is calculated for each correlation, wherein the selecting of the set of evaluation rules from the set of candidate evaluation rules is performed by accounting for the Merit values.
30. The method according to one of claims 27 to 29, wherein, in sub-step b4), a candidate evaluation rule is determined to be an evaluation rule if the corresponding correlation determined in sub-step b3) fulfils at least one predetermined condition.
31. A method for detecting an analyte in a sample of a body fluid, the method comprising the following steps:
- i) providing at least one measurement curve, wherein the measurement curve contains a plurality of measurement values recorded by monitoring a time development of at least one measurement value indicating a progress of a detection reaction of at least one test substance and the sample of a body fluid, wherein the measurement values contained in the measurement curve are acquired at differing points in time (136), wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;
 - ii) determining at least one of at least one target variable or at least one estimate value for at least one Y by using a first time interval t_1 of the measurement curve, wherein $0 \leq t_1 \leq x$, wherein the target variable Y is different from a concentration of the at least one analyte;
 - iii) determining the concentration of the at least one analyte by using the at least one target variable;
 - iv) providing a set of at least two different evaluation rules, each evaluation rule being adapted to derive a characteristic value (138) from the measurement curve, thereby deriving a set of characteristic values $X=\{X_i\}_{i=1...N}$ from the measurement curve, the set of characteristic values (138) comprising at least one first characteristic value (138) being derived from the measurement curve by using at least one first evaluation rule from the set of evaluation rules and at least one second characteristic value (138) being derived from the measurement curve by

using at least one second evaluation rule from the set of evaluation rules, the second evaluation rule being different from the first evaluation rule.

32. The method according to claim 31, wherein the method further comprises the following step:
 - v) performing at least one multivariate analysis of the at least one first characteristic value (138) and the at least one second characteristic value (138) by using at least one predetermined multivariate evaluation algorithm, the multivariate evaluation algorithm being an algorithm adapted to derive at least one result from at least two variables, wherein the at least one first characteristic value (138) and the at least one second characteristic value (138) are used as the at least two variables, thereby deriving the at least one estimate value for at least one target variable Y of the state variables.
33. The method according to claim 32, wherein the second evaluation rule differs from the first evaluation rule in at least one of in at least one coefficient; in at least one parameter; in at least one component related to the algorithm.
34. The method according to one of claims 31 to 33, wherein the measurement curve is an optical measurement curve.
35. The method according to the claim 34, wherein the measurement values contained in the optical measurement curve are acquired at identical measuring conditions.
36. The method according to one of claims 31 to 35, wherein the first evaluation rule may not be transformed into the second evaluation rule by a time transformation.
37. A method for characterizing a sample of a body fluid, the method comprising the following steps:
 - A) bringing the sample of the body fluid into contact with at least one test substance, thereby initiating a detection reaction of the test substance and the sample of the body fluid, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction;
 - B) monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at differing points in time (136);

- C) evaluating the optical measurement curve by using the method according to any one of claims 1 to 36.
38. A computer readable medium having a computer program stored thereon including computer-executable instructions for performing the method according to one of claims 1 to 37 when the program is executed on a computer (132) or computer network (134).
39. An evaluation device for evaluating an optical measurement curve for analyzing at least one sample of a body fluid, the device comprising at least one evaluation unit (116), wherein the evaluation unit (116) is adapted to perform a method according to one of claims 1 to 37.
40. A sample analysis device (110) for characterizing a sample of a body fluid, the device comprising:
- at least one measuring unit (112) for measuring a detection reaction of at least one test substance and at least one sample of a body fluid, wherein the detection reaction is known to be influenced by a set of state variables, each state variable characterizing at least one of a state of the sample of the body fluid and a condition of the detection reaction, the measuring unit (112) further being adapted for monitoring a time development of at least one measurement value indicating a progress of the detection reaction, thereby recording an optical measurement curve containing a plurality of the measurement values acquired at different points in time (136); and
 - at least one evaluation device (114) according to claim 39.
41. The sample analysis device (110) according to claim 40, furthermore comprising at least one test element (118), wherein the test element (118) contains the at least one test substance adapted to perform the detection reaction, wherein the sample analysis device (110) is adapted such that the sample of the body fluid is applicable to the test element (118).
42. The sample analysis device (110) according to claim 41, wherein the sample analysis device (110) is embodied as a hand-held device (122).
43. The method according to claim 2, wherein the state variables are selected from a content of at least one component of the sample of the body fluid.
44. The method according to claim 43, wherein the state variables are selected from a concentration of at least one analyte.

45. The method according to claim 2, wherein the particulate component of the sample of the body fluid comprises a hematocrite.
- 5 46. The method according to claim 11, wherein the derivatives are generated by using a Savitzky-Golay filtering algorithm.
47. The method according to claim 12, wherein the set of characteristic values (138) contains 3-10 characteristic values (138).
- 10 48. The method according to claim 24, wherein the evaluation rule is selected from one or more specific criteria comprising at least one end value criterion.
49. The method according to claim 48, wherein the end value criterion comprises a change rate below a predetermined threshold value.
- 15 50. The method according to claim 24, wherein the evaluation rule is selected from a characteristic point in time at which one or more of the following occur: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve.
- 20 51. The method according to claim 24, wherein the evaluation rule is selected from a characteristic parameter at one of: a maximum of the optical measurement curve or of a derivative of the optical measurement curve; a minimum of the optical measurement curve or of a derivative of the optical measurement curve; an inflection point of the optical measurement curve or of a derivative of the optical measurement curve.
- 25 52. The method according to claim 24, wherein the value derived from the phase plot is selected from the group consisting of: a position of a center of the phase space curve; a length of the phase space curve; a phase space volume; a phase space area; a point with a maximal distance to the center of the phase space curve; a mean squared distance from the origin of the phase space.
- 30 53. The method according to claim 28, wherein sub-step b3) includes determining a Pearson correlation coefficient.
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-90-

54. The sample analysis device according to claim 41, wherein the test element (118) comprises at least one test strip (120).

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1/10

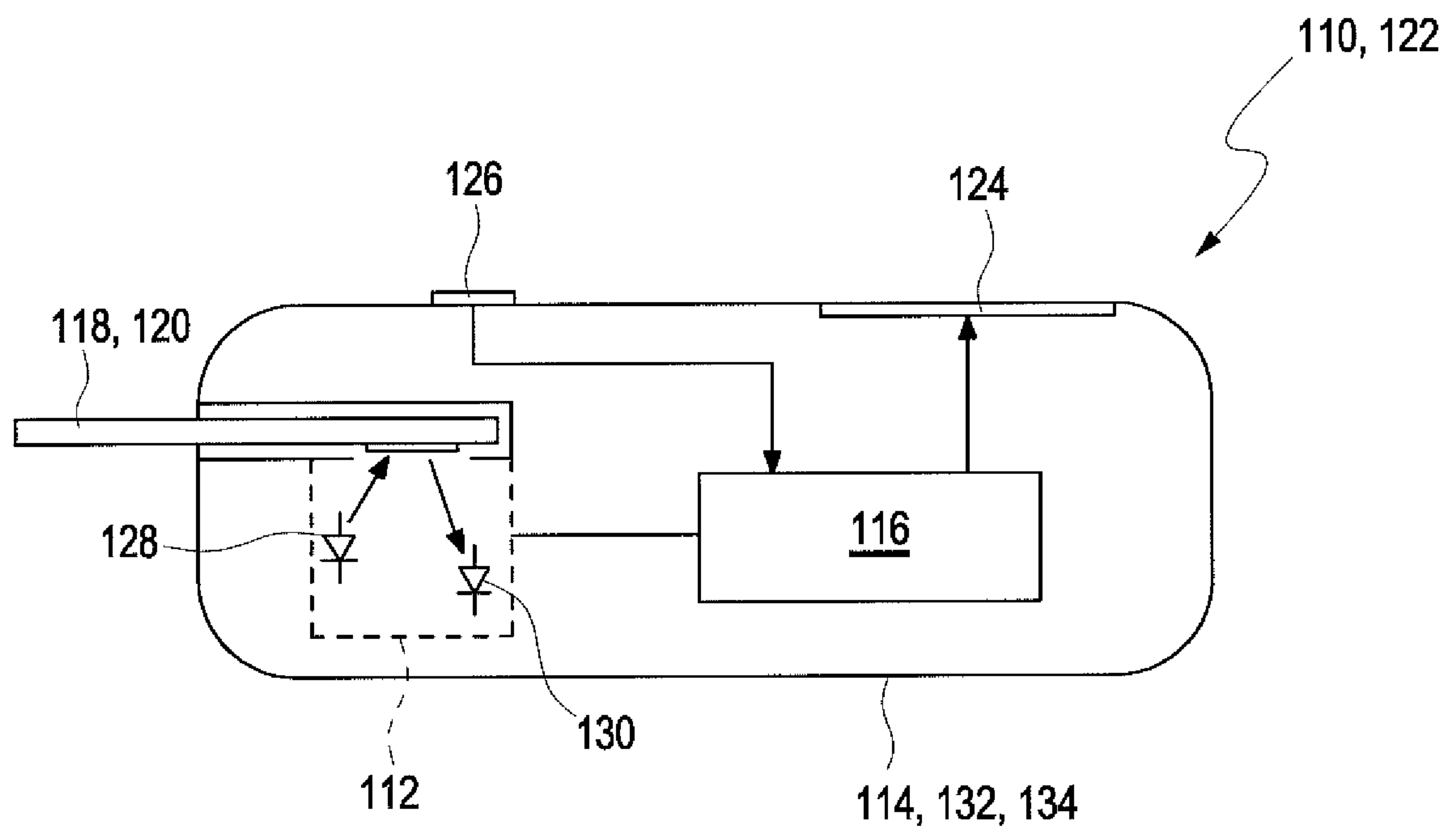


Fig. 1

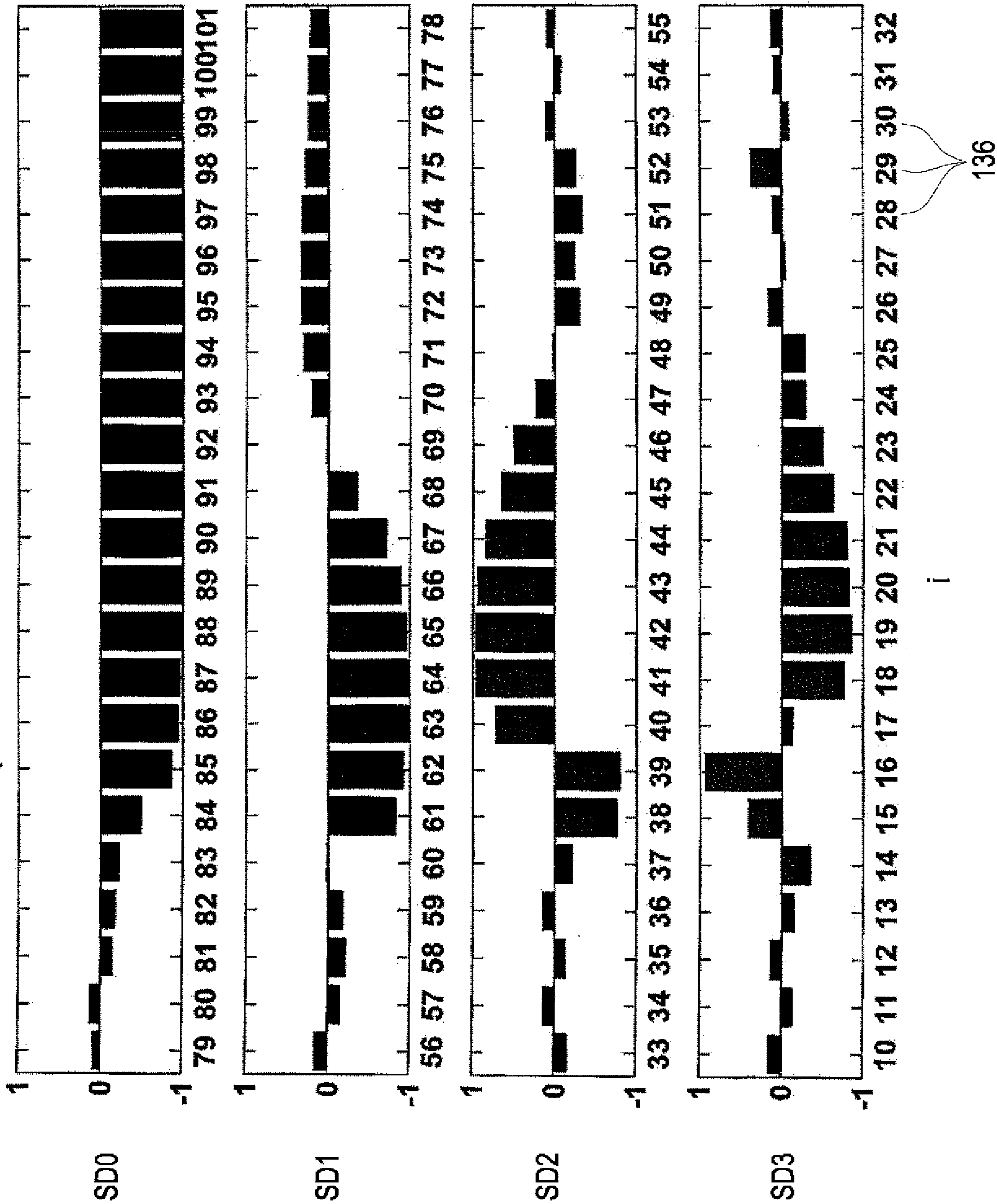


Fig. 2 A

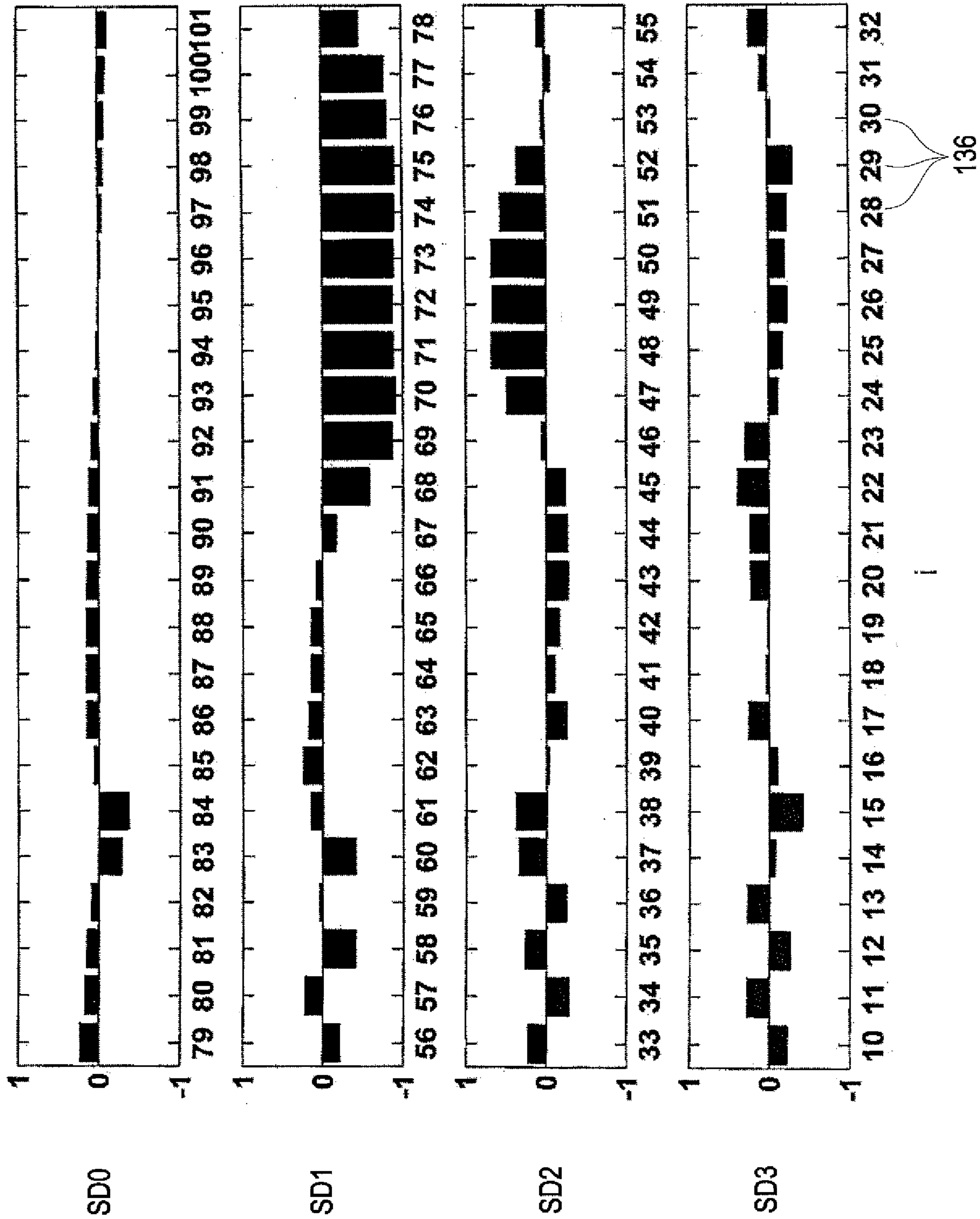


Fig. 2 B

4/10

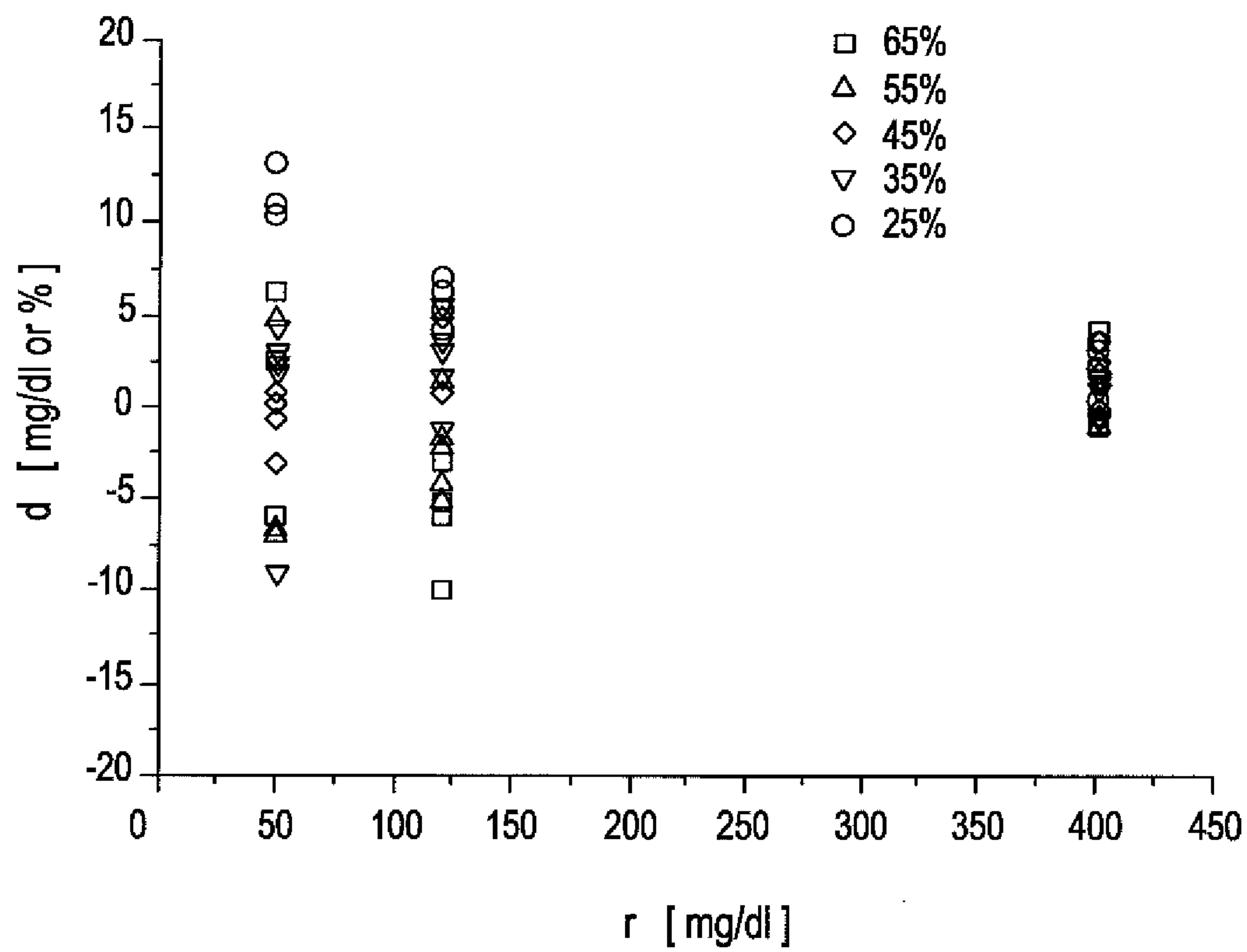


Fig. 2 C

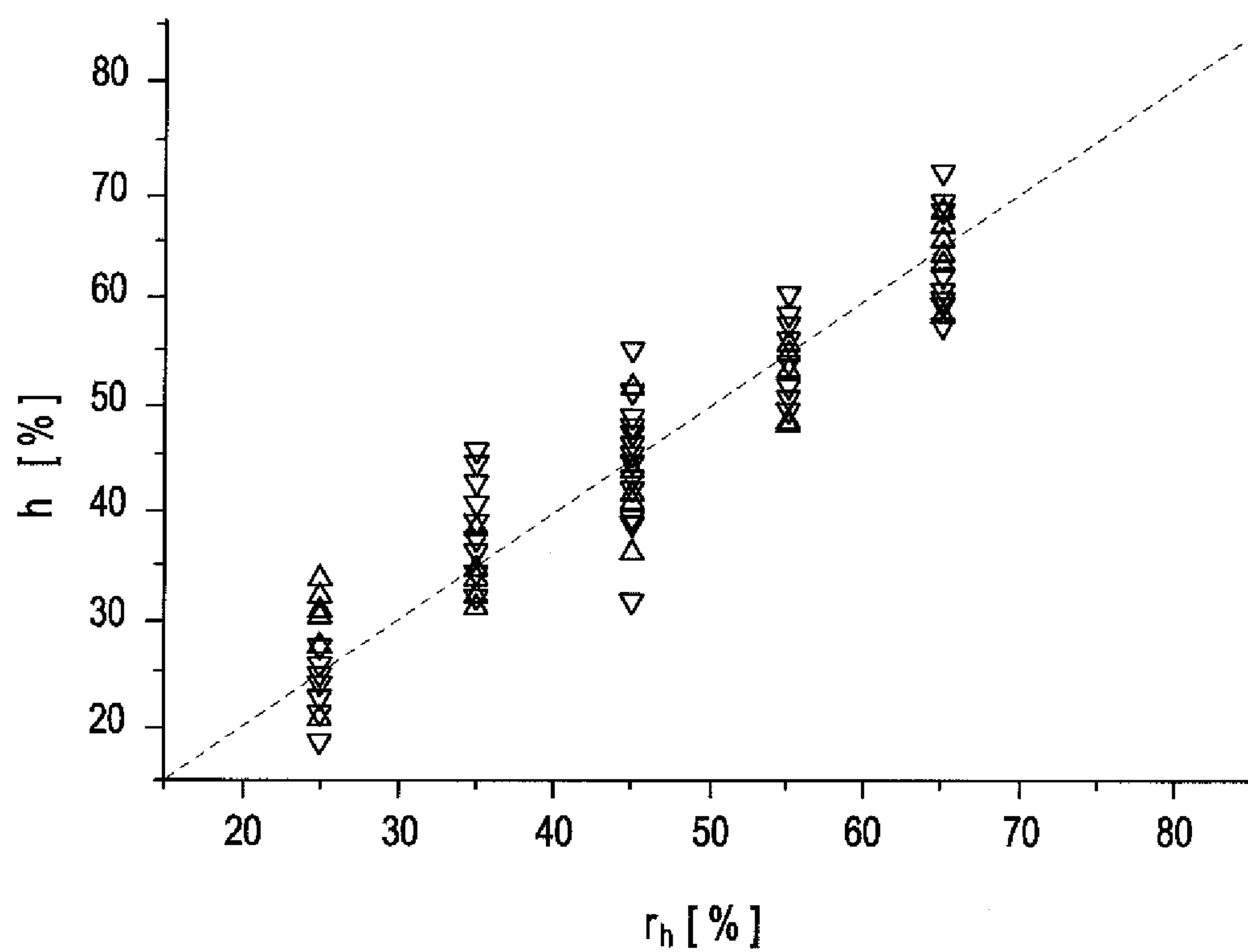


Fig. 2 D

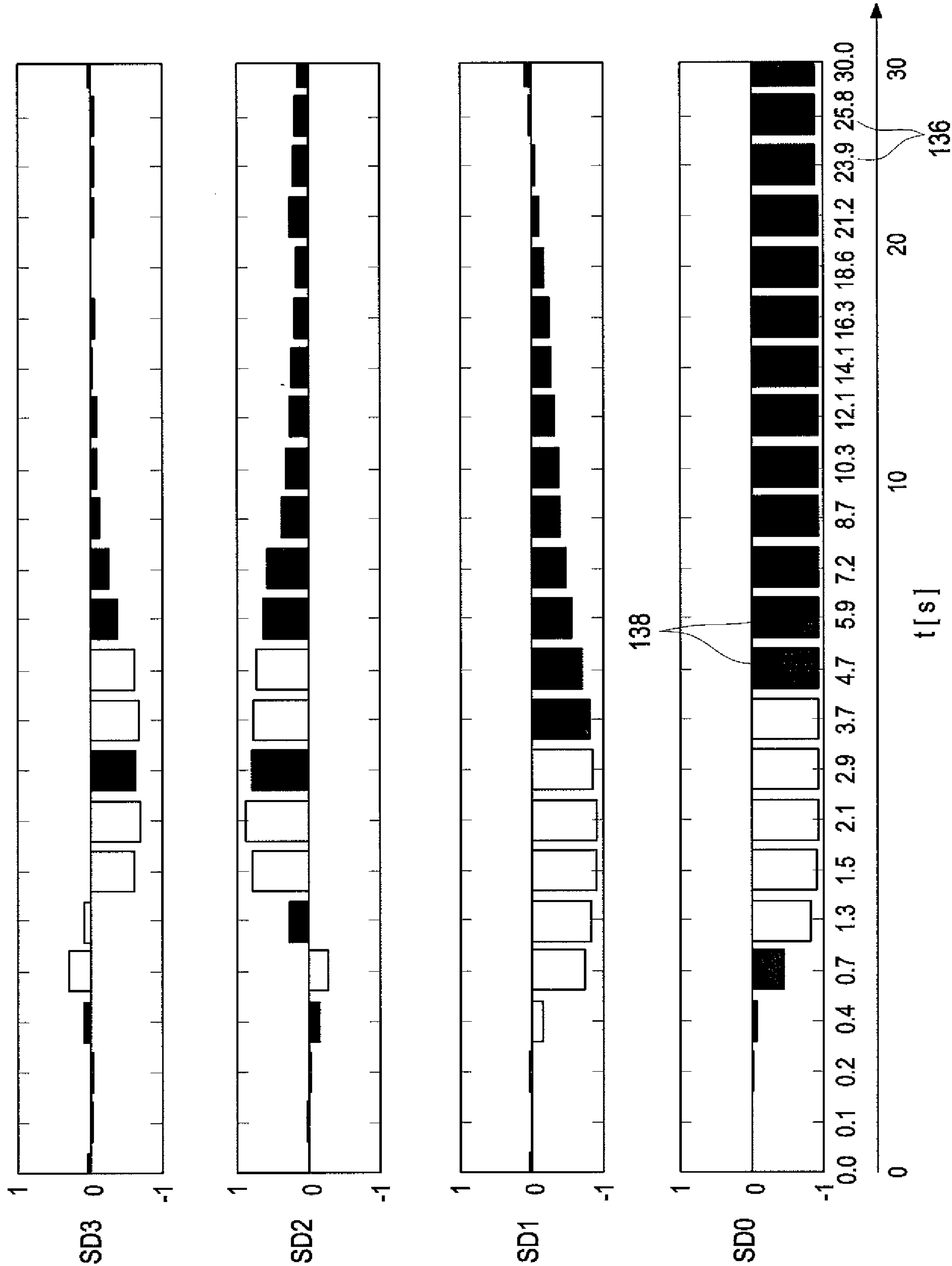


Fig. 3

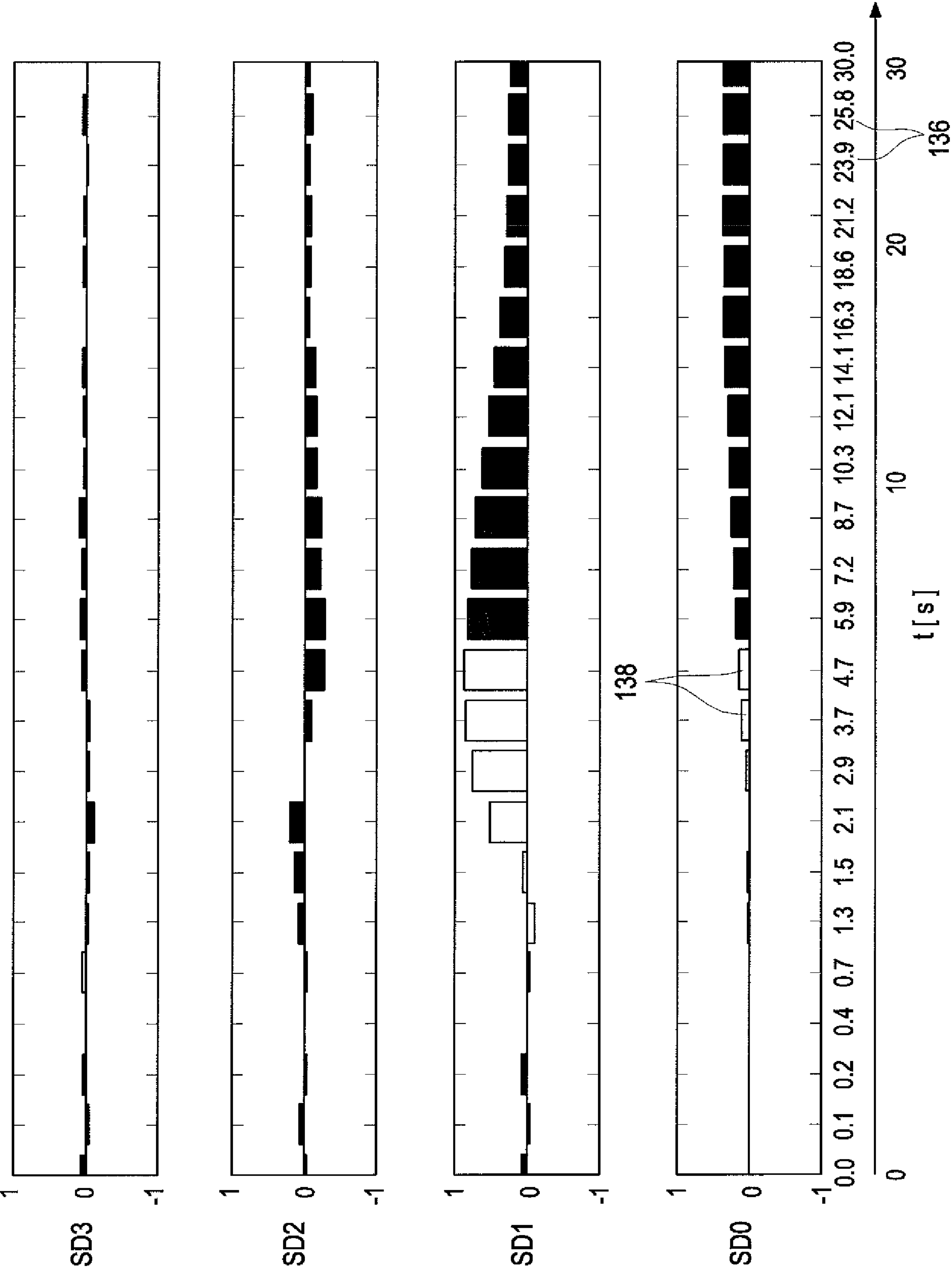


Fig. 4

7/10

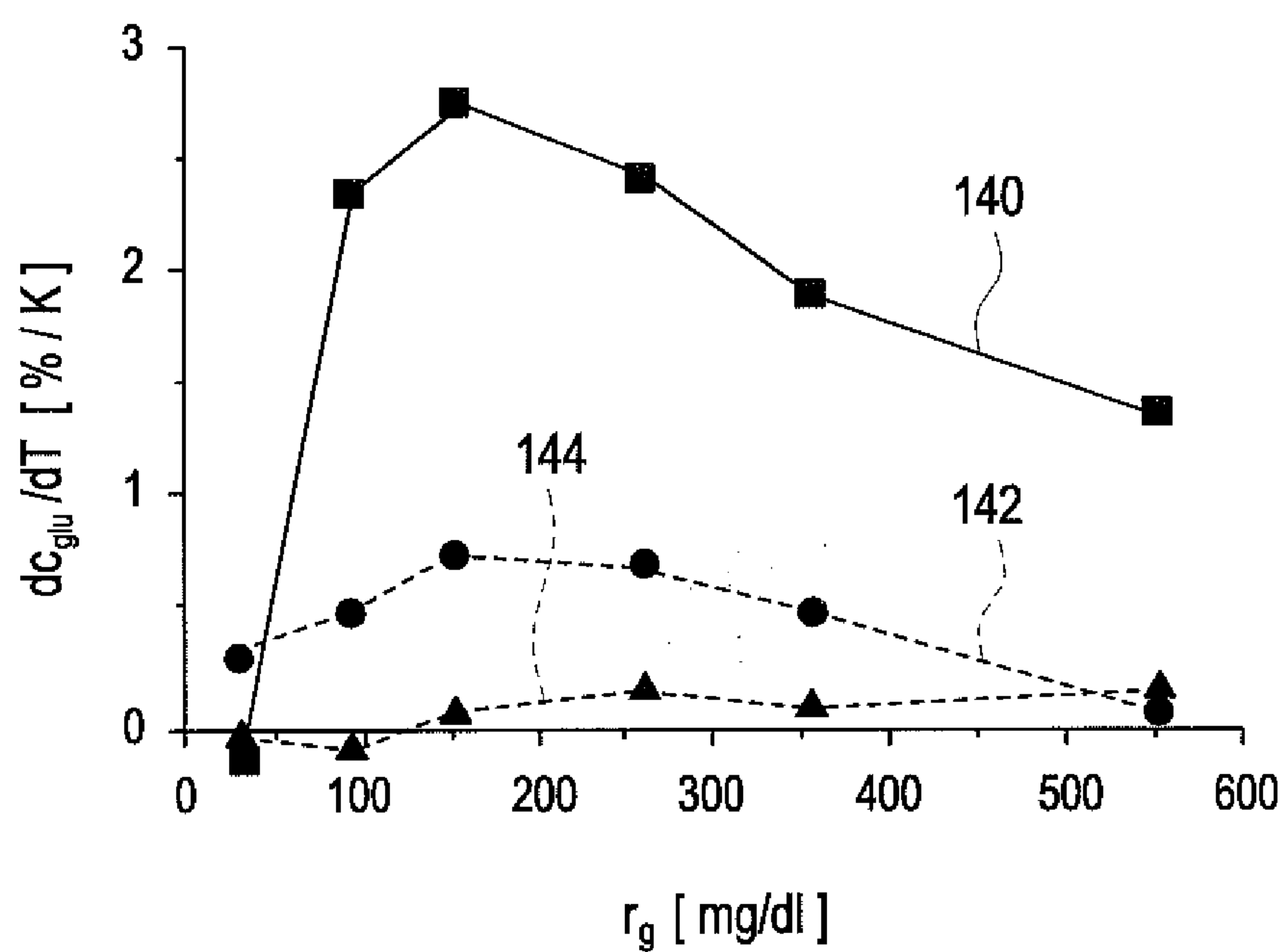


Fig. 5

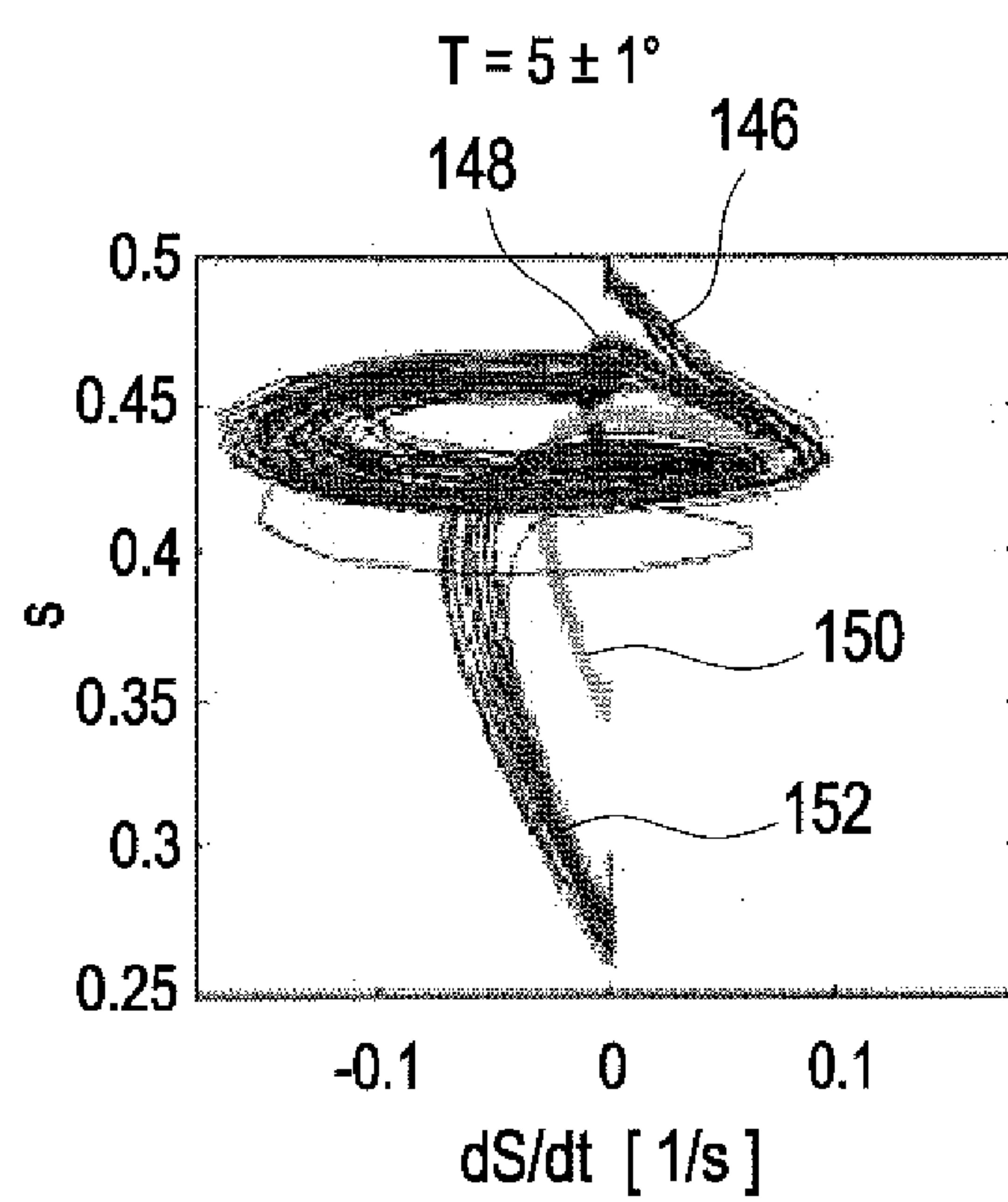


Fig. 6 A

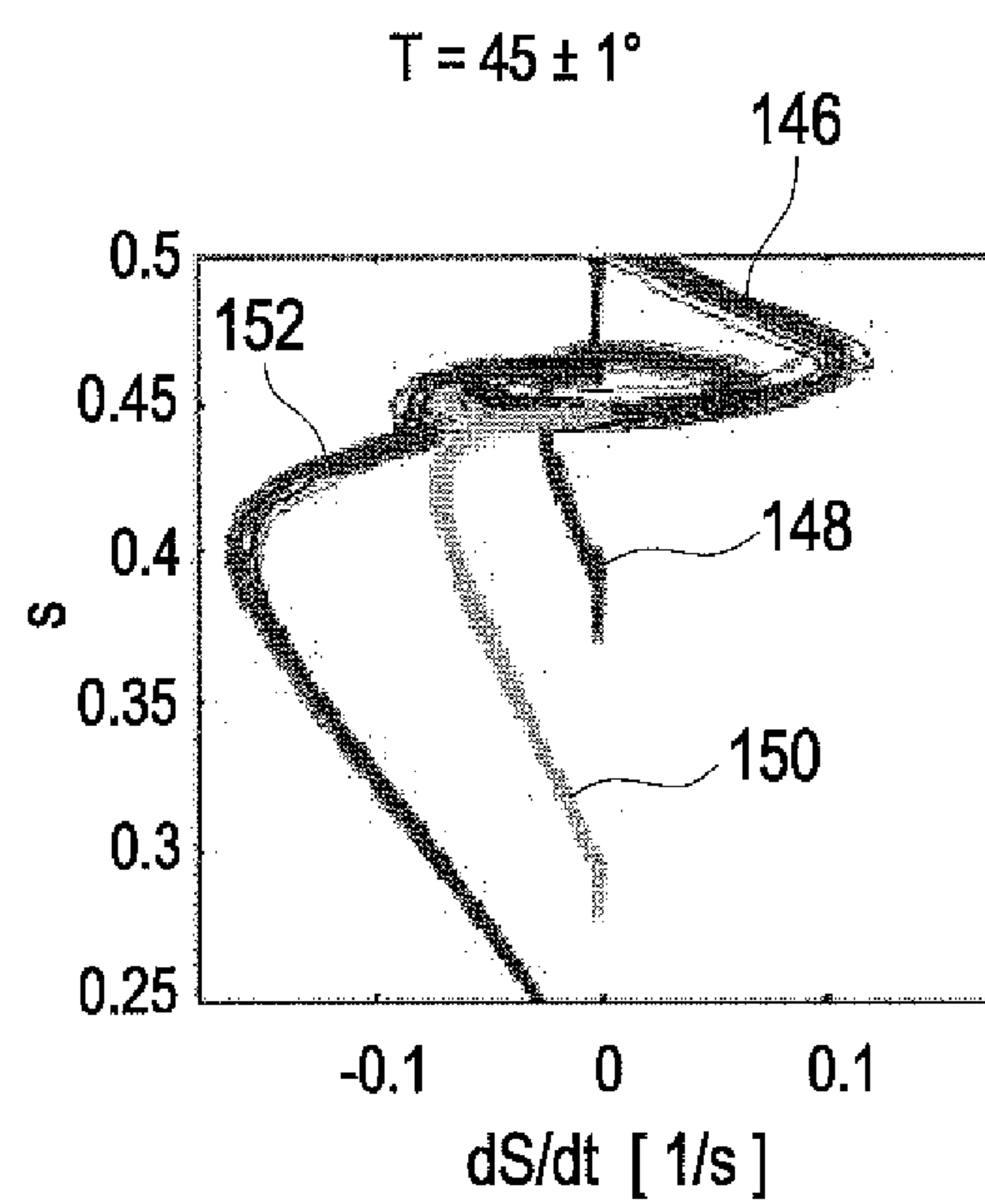


Fig. 6 B

8/10

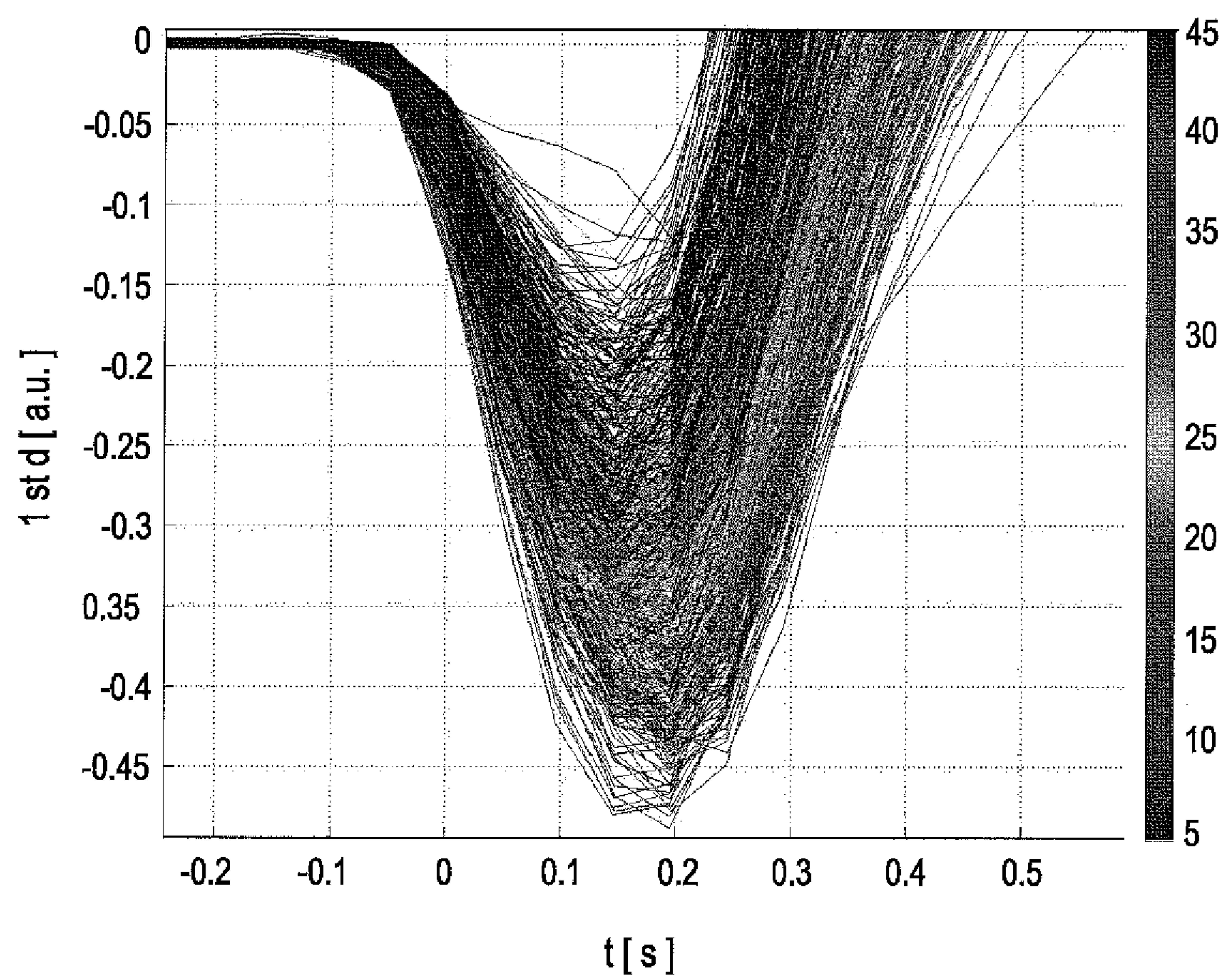


Fig. 7

9/10

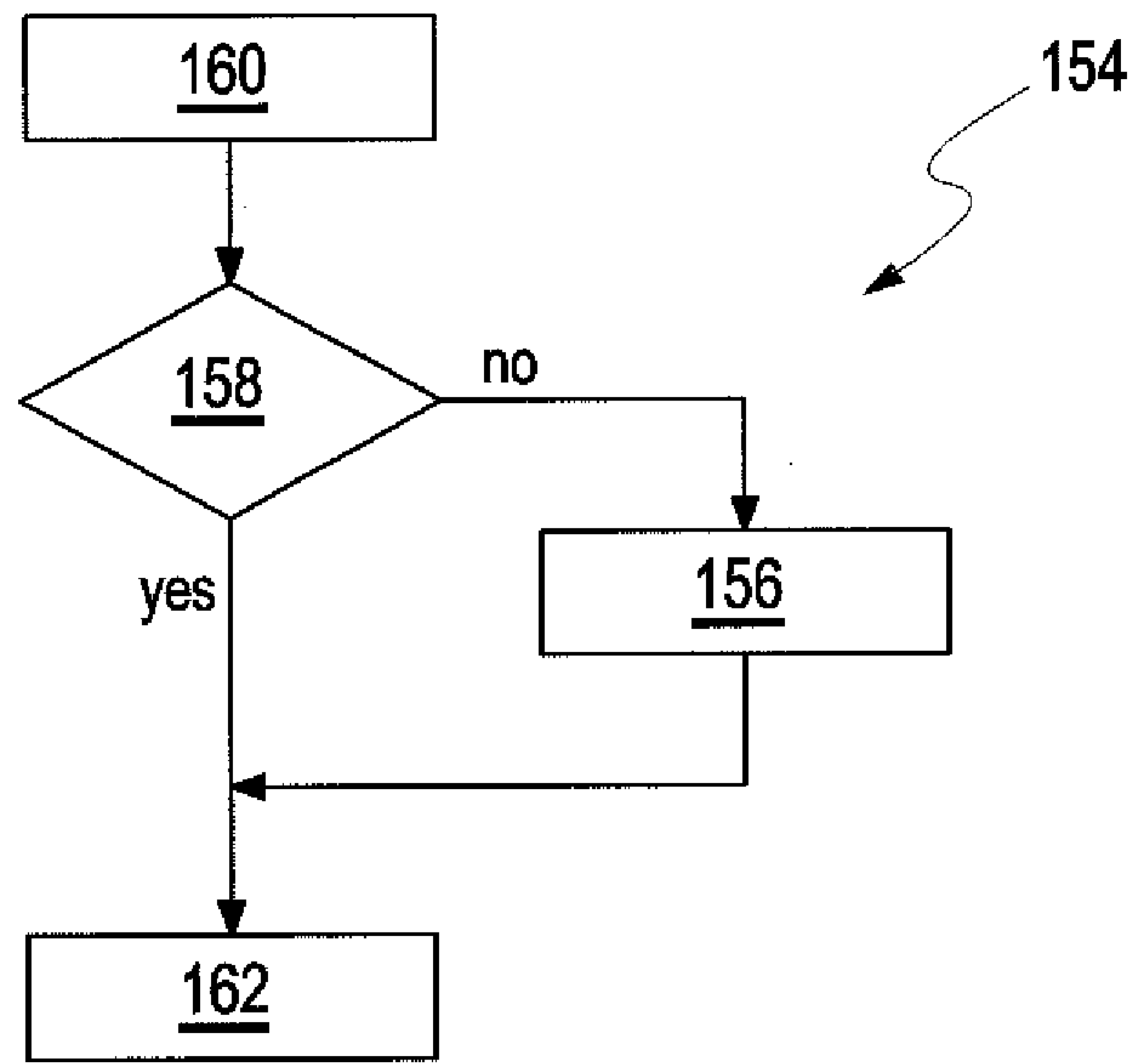


Fig. 8

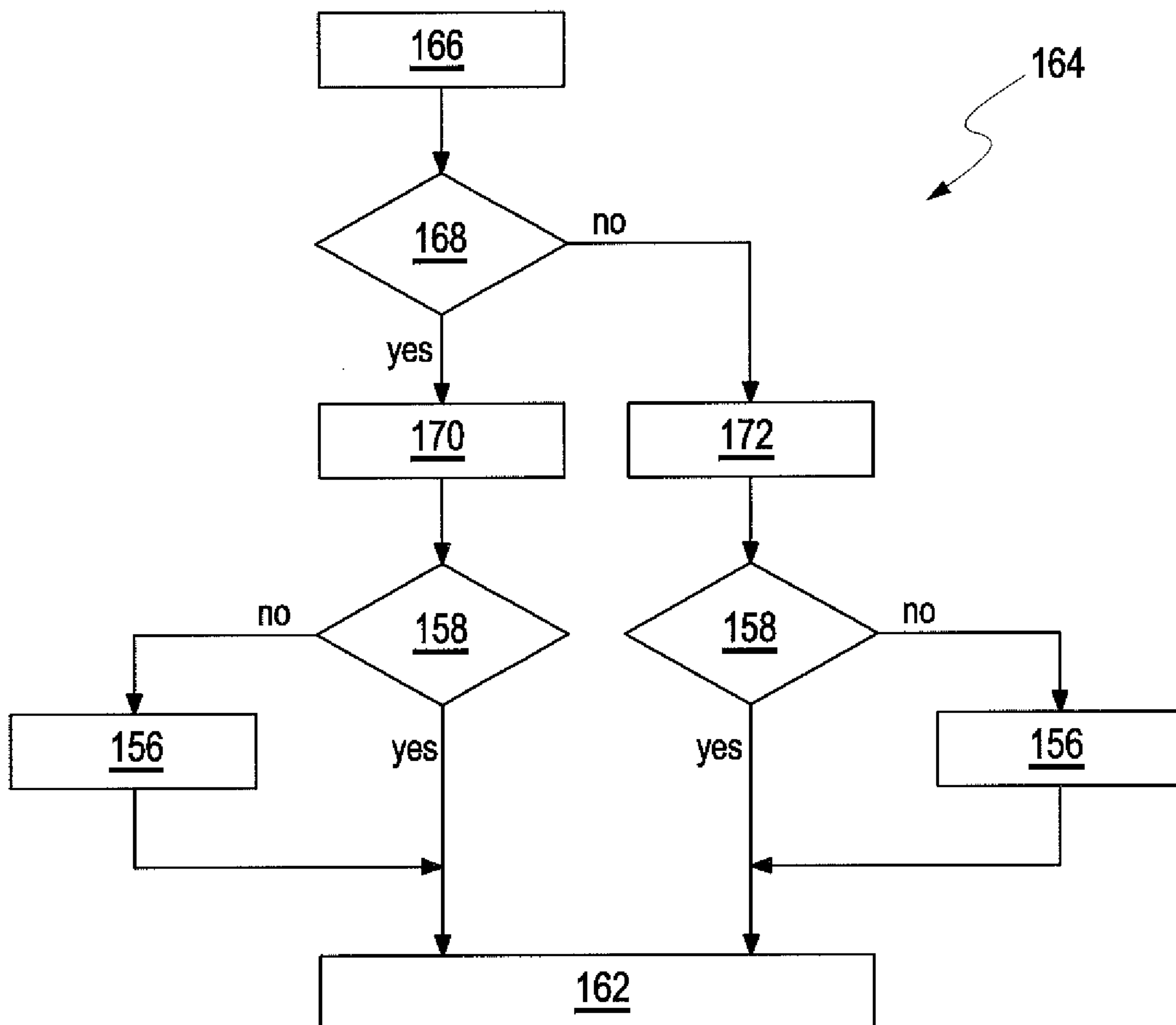


Fig. 9

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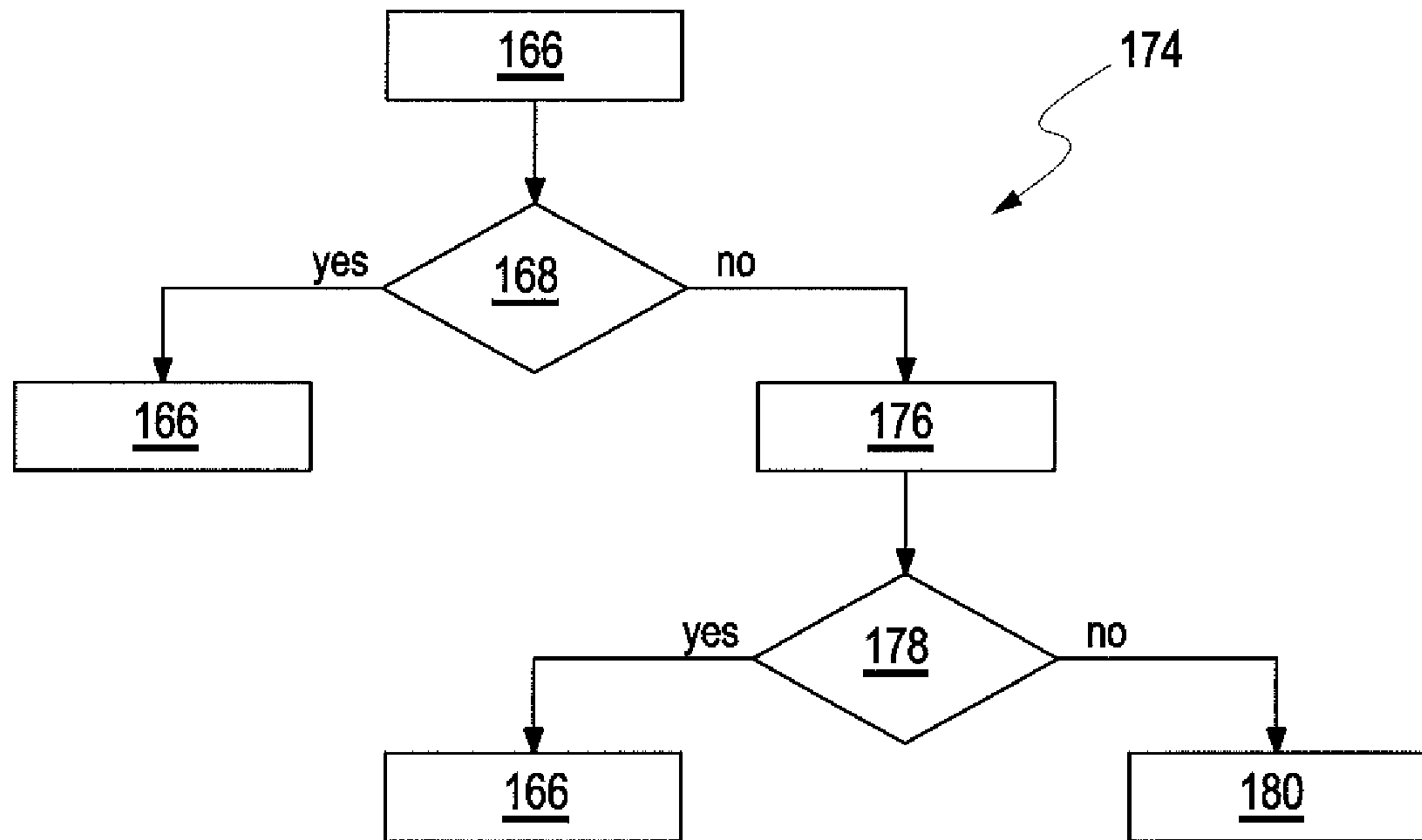


Fig. 10

