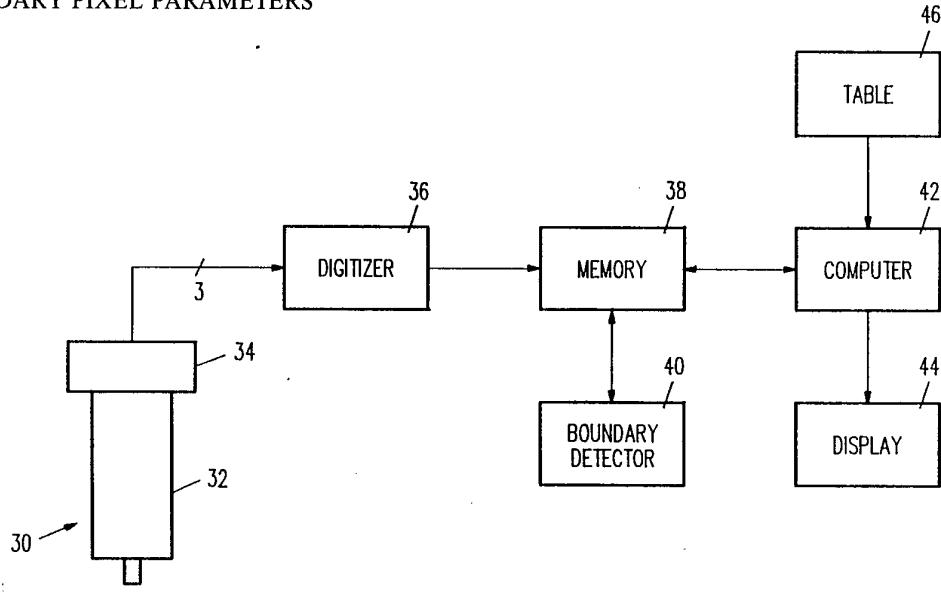




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(54) Title: METHOD AND APPARATUS FOR IDENTIFYING AN OBJECT USING AN ORDERED SEQUENCE OF BOUNDARY PIXEL PARAMETERS



(57) Abstract

33

The present invention relates to a method and apparatus for identifying an examined object having a discernible boundary. An image of the object is formed (34) and segmented into a plurality of pixels (36). The boundary of the object in the image is detected (40) and the parameters of the boundary pixels form an ordered sequence:  $p_1 \dots p_k \dots p_N$ , where  $p_1$  is the parameter value of the first boundary pixel; where  $p_N$  is the parameter value of the last boundary pixel;  $N$  is the total number of boundary pixels and  $k$  is the index to the  $N$  pixels. The  $p_k$  data point is determined (42) such that  $k$  satisfies the relationship (I):  $i-1/M < k/N \leq i/M$ , where  $M$  is the total number of quantile partitions and  $i$  is the  $i$ th quantile partition. The object is identified based upon the  $p_k$  data point value, which is the  $i$ th quantile in  $M$  partitions developed from the above relationship.

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METHOD AND APPARATUS FOR IDENTIFYING  
AN OBJECT USING AN ORDERED SEQUENCE  
OF BOUNDARY PIXEL PARAMETERS  
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Technical Field

5       The present invention relates to a method and an apparatus for identifying an object under examination, wherein an image of the object is formed. More particularly, the present invention relates to a method and an apparatus for identifying an object based upon the object having a discernable boundary and a novel feature termed "quantile".  
10

Background Of The Invention

15       Methods and apparatuses for identifying an object are well-known in the art. The art of identifying biological samples is replete with various techniques for identifying the type of biological samples under examination. See, for example, U.S. Patent No. 4,175,860. Formation of histograms of various parameters are also known. See for example U. S. Patent No. 3,851,156.

20       Heretofore, one of the known methods for identifying an object is to use the size of the object. Other parameters of the object which can be used to identify an object include color, internal optical density, and average intensity within the boundary of the object.  
25

30       In some cases, however, it is not possible to identify an object uniquely or to differentiate two objects if certain of their parameters yield the same result. Thus, for example, if two objects have the same size and have the same internal optical density and have the same average visible light intensity, then it is not possible to identify one of the objects or to differentiate these two objects, based

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upon these parameters. In addition, some of these parameters are computationally intensive requiring considerable amount of time or computer resources. Thus, the present invention contemplates a novel 5 method and apparatus for identifying an object or to differentiate two objects based upon a new parameter which is termed "quantile".

Summary Of The Invention

In the present invention, a method of 10 identifying an object, under examination, which has a discernible boundary is disclosed. The method comprises the steps of forming an image of the object. The image is segmented to form a plurality of pixels. The boundary of the object under 15 examination in the image is then detected. A measurement is made of a parameter of each pixel within the boundary detected. The parameter values of the pixels measured within the boundary detected are formed in an ordered sequence. The particular 20 measured parameter value  $P_k$ , where  $k$  is an index of the ordered sequence, is determined such that the index  $k$  satisfies the following relationship:

$$\frac{i-1}{M} < \frac{k}{N} \leq \frac{i}{M}$$

25 N - is the total number of pixels;  
M - is the total number of quantile partitions  
i - index corresponding to the  $i^{\text{th}}$  fraction of partition or  $i^{\text{th}}$  quantile

30 The object under examination is then identified based upon the  $P_k$  parameter value which is the  $i^{\text{th}}$  quantile within  $M$  partitions developed from the above relationship associated with the object.

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The present invention also contemplates an apparatus to carry out the foregoing method.

Brief Description Of The Drawings

5       Figure 1 is a schematic graphical illustration of two objects "a" and "b" which have the same size and shape and the same average intensity but which are different.

10      Figure 2 is a graph showing a histogram of intensity of light versus the number of pixels in the boundary detected.

15      Figure 3 is a graph showing a cumulative histogram of intensity of light versus the cumulative number of pixels in the boundary detected, and with the quantile feature determined for each object.

20      Figure 4 is a table of pre-determined relationships or quantiles of various known objects.

25      Figure 5 is an ordered sequence of the values of the parameter measured for the pixels in the boundary.

30      Figure 6 is a block diagram of an apparatus suitable for carrying out the method of the present invention.

Detailed Description Of The Drawings

Referring to Figure 1, there is shown a schematic diagram of an image 10 of two objects:  
25      object "A" 12 and "B" 14. Although the description set forth hereinafter is with reference to the method of the present invention to distinguish object "A" from object "B", it can be seen that the method of  
30      the present invention is equally applicable to identifying a single object which is under examination.

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Each of the objects 12 and 14 has a discernible boundary 18 and 20, respectively. As shown in Figure 1, each of the objects 12 and 14 has the same size and shape. Although for illustration purposes, 5 objects 12 and 14 are shown as having the same size and shape, it is not necessary to the practice of the present invention. The purpose of illustrating the method of the present invention using objects 12 and 14 having the same size and shape is to show how the 10 method of the present invention can be used to distinguish objects 12 and 14 under examination when other parameters, such as size and shape, cannot be used.

Objects 12 and 14, of course, represent images 15 of real objects, such as cells, genes or other biological or non-biological objects. In the method of the present invention, an image 10 is taken of the objects 12 and 14. The image 10 can be formed by using an electronic video camera, such as a CCD (the specific details of an apparatus suitable to carry 20 out the method of the present invention will be discussed hereinafter). The image 10 is then filtered in accordance with the different color filters that can be used to distinguish the type of 25 color of the object. Each of the different color images is segmented, to form a plurality of pixels. Each pixel is then digitized. The boundary of each of the objects 12 and 14 is determined. This can be done, for example, by the method disclosed in U.S. 30 Patent No. 4,538,299, which is incorporated herein by reference.

Once the boundary of each of the objects 12 and 35 14 is determined, one of the parameters of each pixel within the boundary is measured. One parameter can be the intensity of visible light. Thus, the

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intensity of visible light at each pixel, positioned within the boundary positions 18 and 20, is measured.

One prior art method is to form a histogram of intensity versus the number of pixels. Referring to 5 Figure 2, there is shown two histograms. The histogram of Figure 2a corresponds to the object 12 whereas the histogram of Figure 2b corresponds to the object 14. As can be seen from Figure 1, object 12 has a substantially uniform intensity (the dark image indicates brightness) throughout the entire region 10 within the boundary 18. Thus, the histogram, as shown in Figure 2a, shows substantially  $N_1$  pixels each having an intensity of  $I_1$ . In contrast, object 14 has a small area 16 that has greater intensity than the 15 uniform intensity  $I_1$  of object 12. The histogram, as shown in Figure 2b, shows that  $N_2$  number of pixels in the spot 16 have substantially an intensity value of  $I_2$ , with  $N_1 > N_2$  and  $I_2 > I_1$ .

Once the histogram of each object 12 and 14 has 20 been formed, a cumulative histogram is then developed. The cumulative histogram is formed by summing the number of pixels that has at least a certain intensity value. If  $N = f(I)$ , where  $N$  is the number of pixels having an intensity level  $I$ , then 25  $\text{Cum } N = F(I)$  where  $\text{Cum } N$  is the cumulative number of pixels having an intensity level  $\leq I$ , and where

$$F(I) = \sum_{j=0}^I f(j)$$

30 The cumulative histograms of the histograms shown in Figure 2a and 2b, are shown in Figure 3.

The particular parameter value, such as 35 intensity of visible light, associated with a cumulative fraction of all pixels, is termed a quantile. Thus, from Figure 3, the 50th quantile is

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I<sub>3</sub> for object 12 and I<sub>4</sub> for object 14. It should be noted that unlike conventional percentile representations, which are even increments of discrete percents, quantile representations are not limited to discrete, even increments. Thus, an object may be identified by its ith, jth, or kth quantile, where i, j, and k do not necessarily occur in discrete, even increments.

Since the quantile relationship, shown in Figure 3, associated with each objects 12 and 14 is different, the objects 12 and 14 can be distinguished from one another, based upon this developed quantile relationship. Alternatively, to uniquely identify a particular object, such as object 14, the quantile relationship associated with the object 14 can be compared to a table of pre-determined relationship of other known objects whose quantile relationships have been pre-determined. The table of pre-determined quantile relationships of other known objects can be based upon experimental results. The comparison of the quantile relationship to the table of pre-determined quantile relationships would then serve to identify uniquely the type of the object that is under examination.

For the purpose of distinguishing the objects 12 and 14 from one another or to identify an unknown object by comparing the quantile relationship, it is not necessary to compare each ordered pair of numbers (Q, I) (where Q is the quantile number and I is intensity associated therewith), to each other (in the case of two particles) or to a table (in the case of attempting to uniquely identify an object). One method is to compare the associated intensity value of the object under examination at a particular quantile number to the associated intensity value of

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another object at the same quantile number, or to a table of associated intensity values of identified objects at the same quantile number. Thus, as shown in Figure 3, the 50th quantile for the object 12 has an intensity value of  $I_3$ , and the same 50th quantile for the object 14 has the value  $I_4$ . These intensity values can be compared to each other to distinguish one object from another. Alternatively, the 50th quantile of one of the objects can be compared to a table of intensity values of identified particles whose 50th quantile have been pre-determined. See Figure 4. By comparing the 50th quantile of the unknown object to the table, the unknown object can be identified.

Of course, if we are dealing with objects of biological particles having statistical variations in intensity, it may not be possible to determine precisely the intensity associated with a particular quantile for the same particle under all cases of examination. Thus, a table relationship may encompass a range of values of intensity for a particular quantile such that the object having that range of intensity values, can be uniquely identified. The range of values can be developed from representative samples identified from experimental results.

The foregoing example illustrates a method of distinguishing objects wherein the average intensity of the two objects 12 and 14 are indistinguishable. However, using the intensity of visible light and the quantiles developed therefor, the objects can be distinguished from one another. Parameters, other than intensity of light can also be used to distinguish the particles. Thus, another parameter suitable for use is differentiating by color

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representation. Color can be represented by three primary colors of red, blue, and green. Alternatively, color can also be represented by hue, intensity, and saturation; or by cyan, magenta and yellow.

In one particular embodiment, the difference in color representation of the primary colors is used. The parameter of  $\log(a) - \log(b)$  may be employed, where a, and b are intensities of red and blue primary colors, respectively. Other possible combinations include: a being green and b being red; or a being green and b being blue.

The problem with the foregoing method of calculating quantiles is that a cumulative histogram must be formed. This necessitates the creation of a plurality of discrete "bins". A bin is defined by a cumulative fraction of all the pixels from a first fraction to a second fraction. As can be seen from Figure 3, the entire value of the cumulative histogram from 0 to  $I_T$  must be divided into a number of discrete bins. The  $i^{\text{th}}$  quantile is the value of  $I$  corresponding to the  $i^{\text{th}}$  cumulative fraction of the pixels. For parameters, such as  $\log(a) - \log(b)$ , as mentioned hereinabove, the range of values can be enormous (due to the nature of the logarithmic operation). Thus, the choice of the boundary of the bins may become difficult.

In the method of the present invention, quantiles can be calculated without calculating a cumulative histogram. In the method of the present invention, the values of the parameters measured for each of the pixels within the boundary detected are placed in an ordered sequence. This is shown in Figure 5. As previously discussed, for the example used in the method of the present invention, the

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objects 12 and 14 are assumed to be the same size and shape. Thus, each of those objects 12 and 14 would be segmented into the same number of total pixels, N. The parameter, for example of intensity of visible light, for each of the pixels within the boundary is then placed in an ordered sequence.  $P_1$  is the value of the pixel having the lowest intensity of visible light.  $P_N$  is the value of the pixel having the largest value of the parameter, such as intensity visible light. The values of the parameter of the pixels measured are then formed in an ordered sequence. A value of  $P_k$  represents the parameter of the pixel between the smallest value and the largest value with k being the index to the N pixels.

Thereafter, k is chosen such that k satisfies the following relationship:

$$\frac{i-1}{M} < \frac{k}{N} \leq \frac{i}{M}$$

where M is the total number of quantile partitions and i is the  $i^{\text{th}}$  fraction of partition or  $i^{\text{th}}$  quantile. For the  $i^{\text{th}}$  chosen quantile, once k is determined, the intensity value of the data point  $P_k$  would be the intensity value associated with the  $i^{\text{th}}$  quantile, or simply the value of the  $i^{\text{th}}$  quantile. Thereafter, the object under examination can be distinguished from other objects based upon the  $i^{\text{th}}$  quantile.

Alternatively, the object under examination can be distinguished from all other objects based upon a table of predetermined values of corresponding  $i^{\text{th}}$  quantile. This can be accomplished even where the two objects do not have the same number of pixels. Thus, for example, object 1 has  $N_1$  pixels and Object 2 has  $N_2$  pixels. The parameters of the pixels are placed in an ordered sequence as follows:

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$P_1 \dots P_{k_1} \dots P_{N_1}$  - object 1 having  $N_1$  pixels

$P_1 \dots P_{k_2} \dots P_{N_2}$  - object 2 having  $N_2$  pixels

where  $k$  is the index. For each object, the  $P_k$  data

point is determined such that  $k$  satisfies the

5 following:

$$\frac{i-1}{M} < \frac{k}{\text{Total No. of pixels for the object}} \leq \frac{i}{M}$$

10 M - total number of quantile partitions;  
i - i<sup>th</sup> fraction of partition or i<sup>th</sup> quantile  
partition

where for Objects 1 and 2 i and M are chosen to be the same. The resultant value of  $P_k$  for each object is compared to one another to differentiate the objects. Of course, as previously discussed, Objects 1 and 2 can be the same type of object even if  $P_{k1}$  does not equal  $P_{k2}$ . There can be a range of values for a particular quantile such that objects having values within that range will still be the same type.

As can be seen from the foregoing, with the method of the present invention, no cumulative histogram calculation is made. As a result, there is no need to set the size of the bin to calculate the cumulative histogram. A simple comparative relationship to determine the index of the data points is determined from which the value of the data point of the parameter measured is then obtained which corresponds to the quantile of interest.

30 An example of the method of the present invention can be seen with reference to the following Table 1, with an example of three particles: Particle 1, Particle 2, and Particle 3, each having 50 pixels of varying intensity of the same parameter.

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TABLE 1

<u>Pixel#</u>	<u>Particle 1</u>	<u>Particle 2</u>	<u>Particle 3</u>
5	1 6.55	3.77	2.64
	2 6.98	5.56	2.60
	3 12.24	6.45	1.92
	4 9.38	4.39	1.82
	5 8.28	5.28	2.25
	6 12.66	6.50	2.46
10	7 8.04	3.14	2.23
	8 6.46	5.78	2.83
	9 10.12	4.18	2.15
	10 7.90	7.58	2.83
	11 8.94	5.45	1.98
15	12 9.99	4.47	2.07
	13 8.78	5.21	2.30
	14 9.16	4.55	1.86
	15 15.12	5.21	2.26
20	16 13.26	3.47	1.90
	17 9.04	4.25	8.44
	18 14.82	4.65	6.72
	19 11.36	4.25	4.98
25	20 8.64	5.96	5.61
	21 30.53	6.16	6.36
	22 33.61	4.61	4.70
	23 30.36	3.42	9.32
	24 30.21	4.38	9.30
30	25 31.12	2.87	5.86
	26 27.71	6.49	6.96
	27 32.21	4.38	2.96
	28 33.27	5.74	4.74
	29 34.31	6.84	8.42
35	30 41.23	3.43	6.32
	31 26.76	6.03	6.72
	32 27.83	4.92	8.90
	33 30.93	5.23	6.64
	34 29.12	6.59	4.02
40	35 26.33	4.19	5.53
	36 30.59	4.87	8.84
	37 33.00	4.01	6.19
	38 29.07	6.34	8.75
	39 26.01	3.96	4.17
45	40 36.96	6.95	6.91
	41 34.02	91.44	91.44
	42 30.15	90.50	90.50
	43 30.13	95.96	95.96
	44 21.58	92.91	92.91
50	45 34.80	88.72	88.72
	46 21.75	97.00	97.00
	47 27.42	94.56	94.56
	48 30.47	94.97	94.97
	49 28.36	94.35	94.35
	50 29.58	93.19	93.19

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The next table, Table 2, shows various statistical values computed for the three particles.

TABLE 2  
Particle 1   Particle 2   Particle 3

5	Mean	22.182	22.682	22.518
	Standard Error	1.504	5.053	5.073
	Median	27.088	5.314	6.025
	Mode	#N/A	#N/A	#N/A
	Standard Deviation	10.637	35.729	35.872
10	Variance	113.141	1276.597	1286.779
	Kurtosis	-1.580	0.421	0.402
	Skewness	-0.219	1.547	1.535
	Range	34.763	94.129	95.341
	Minimum	6.464	2.872	1.660
15	Maximum	41.227	97.001	97.001
	Sum	1109.125	1134.118	1125.918
	Count	50	50	50

The next table, Table 3, shows the creation of a plurality (10) of "bins" with each bin containing a count of the number of pixels having a value between a first count to a second count. Thus, Bin #1 contains all the pixels having intensity values between 0 and 10. For particle #1, there would be 13 such pixels. For particles #2 and #3, there would be 40 and 40 respectively. The 10 bins are chosen to partition the maximum range of pixel intensity values into 10 fractions. Since the maximum range of intensity value is zero to 41.23, 97.00, and 97.00 for particles #1, #2 and #3 respectively, the 10 bins are chosen to delineate pixel intensity values of 0-10-20-... -90-100. The column entitled Cumulative Histogram is a cumulative count (shown as a %) of the total number of pixels having an intensity value between zero and the value associated with the Bin #. As can be seen from Table 3, particles #2 and #3 are indistinguishable based upon cumulative histogram. Further, as

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previously discussed, for intensity values based upon  
log( ), the value can be large. Thus, the range of  
pixel intensity values will be large, rendering  
difficult the division of the intensity values into a  
5 manageable number of bins.

TABLE 3

Bin #	Intensity Value of Pixels	Particle 1		Particle 2		Particle 3	
		Histogram	Cum.Hist.	Histogram	Cum.Hist.	Histogram	Cum.Hist.
1	0-10	13	26.00%	40	80.00%	40	80.00%
2	10-20	7	40.00%	0	80.00%	0	80.00%
3	20-30	12	64.00%	0	80.00%	0	80.00%
4	30-40	17	98.00%	0	80.00%	0	80.00%
5	40-50	1	100.00%	0	80.00%	0	80.00%
6	50-60	0	100.00%	0	80.00%	0	80.00%
7	60-70	0	100.00%	0	80.00%	0	80.00%
8	70-80	0	100.00%	0	80.00%	0	80.00%
9	80-90	0	100.00%	1	82.00%	1	82.00%
10	90-100	0	100.00%	9	100.00%	9	100.00%

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5

The following table, Table 4, shows the table of pixel intensity values (Table 1) rearranged in accordance with the method of the present invention. The pixel intensity values are ordered in a sequential order, from highest to lowest. (Of course, the ordered sequence can be from the lowest to the highest.) In this example,  $P_{50}$  for each of the three particles have the values 6.46, 2.87 and 1.66, respectively, while  $P_1$  is 41.23, 97.00 and 97.00 respectively.

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TABLE 4

	<u>Rank</u>	<u>Percent</u>	Particle 1 <u>Pixel#</u>	Particle 1 <u>Int.</u>	Particle 2 <u>Pixel#</u>	Particle 2 <u>Int.</u>	Particle 3 <u>Pixel#</u>	Particle 3 <u>Int.</u>
5	1	100.00%	30	41.23	46	97.00	46	97.00
	2	97.95%	40	36.96	43	95.96	43	95.96
	3	95.91%	45	34.80	48	94.97	48	94.97
	4	93.87%	29	34.31	47	94.56	47	94.56
	5	91.83%	41	34.02	49	94.35	49	94.35
10	6	89.79%	22	33.51	50	93.19	50	93.19
	7	87.75%	28	33.27	44	92.91	44	92.91
	8	85.71%	37	33.00	41	91.44	41	91.44
	9	83.67%	27	32.31	42	90.50	42	90.50
	10	81.63%	25	31.12	45	88.72	45	88.72
15	11	79.59%	33	30.93	10	7.58	23	9.32
	12	77.55%	36	30.59	40	6.95	24	9.30
	13	75.51%	21	30.53	29	6.84	32	8.90
	14	73.46%	48	30.47	34	6.59	36	8.84
	15	71.42%	23	30.36	6	6.50	38	8.75
20	16	69.38%	24	30.21	26	6.49	17	8.44
	17	67.34%	42	30.15	3	6.45	29	8.42
	18	65.30%	43	30.13	21	6.16	26	6.96
	19	63.26%	50	29.58	31	6.03	40	6.91
	20	61.22%	34	29.12	20	5.96	18	6.72
25	21	59.16%	38	29.07	8	5.78	31	6.72
	22	57.14%	49	28.36	28	5.74	33	6.64
	23	55.10%	32	27.83	2	5.56	21	6.36
	24	53.06%	26	27.71	11	5.45	30	6.32
	25	61.02%	47	27.42	38	5.34	37	6.19
30	26	48.97%	31	26.76	5	5.28	25	5.86
	27	46.93%	25	26.33	33	5.23	20	5.81
	28	44.89%	29	26.01	15	5.21	35	5.53
	29	42.85%	46	21.75	13	5.21	19	4.98
	30	40.81%	44	21.58	32	4.92	28	4.74
35	31	38.77%	15	15.12	36	4.87	22	4.70
	32	36.73%	18	14.82	18	4.65	39	4.17
	33	34.69%	16	13.26	22	4.61	34	4.02
	34	32.65%	6	12.66	14	4.55	27	2.96
	35	30.61%	3	12.24	12	4.47	8	2.83
40	36	28.57%	19	11.36	4	4.39	1	2.64
	37	26.53%	9	10.12	27	4.38	2	2.60
	38	24.48%	12	9.99	24	4.38	6	2.46
	39	22.44%	4	9.38	19	4.25	13	2.30
	40	20.40%	14	9.16	17	4.25	15	2.25
45	41	18.36%	17	9.04	35	4.19	5	2.25
	42	16.32%	11	8.94	9	4.18	7	2.23
	43	14.28%	13	8.78	37	4.01	9	2.15
	44	12.24%	20	8.64	39	3.96	12	2.07
	45	10.20%	1	8.55	1	3.77	11	1.98
50	46	8.16%	5	8.28	16	3.47	3	1.92
	47	6.12%	7	8.04	30	3.43	16	1.90
	48	4.08%	10	7.90	23	3.42	14	1.86
	49	2.04%	2	6.98	7	3.14	4	1.82
	50	0.00%	8	6.46	25	2.87	10	1.66

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where Rank is the order in sequence by the intensity of the pixel, Pixel# is the number of the pixel from Table 1, Intensity is the value of the intensity for the corresponding Pixel# from Table 1, and Percent is the percent of particles that would have intensity value shown under Intensity or less.

Thus, the pixel having the Rank of 3 for Particle 1 is Pixel# 45 from Table 1. Pixel# 45 for Particle 1 has an intensity value of 34.80. Further 95.91% of all pixels in Particle 1 have intensity values of 34.80 or less.

If the number of "bins" or fractions, M is chosen to be 10 (the same as the example based upon a cumulative histogram calculation), and N-number of pixels = 50, then

$$\frac{i-1}{10} < \frac{k_i}{50} \leq \frac{i}{10}$$

where i is the  $i^{\text{th}}$  bin. If  $i=10$

$$\frac{9}{10} < \frac{k_{10}}{50} \leq \frac{1}{10}$$

$k_{10} = 46, 47, 48, 49, 50.$

If we choose  $k_{10} = 50$ , then  $P_{k10} = 6.46, 2.87, 1.66$  for particles #1, #2 and #3, respectively.

Since  $P_{k10} = 2.87$  for particle #2 clearly differs from  $P_{k10} = 1.66$  for particle #3, particles #2 and #3 can be differentiated, while other statistical based analysis such as histogram, and cumulative histogram are unable to differentiate the particles. Furthermore, although a single value was compared, clearly a range of values can be compared. Note that the same holds true for  $P_{k9}$  ( $K_9 = 45$ ) = 8.55, 3.77, 1.98 and  $P_{k8}$  ( $K_8 = 40$ ) = 9.16, 4.25, 2.25 for particles #1, #2 and #3 respectfully.

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An apparatus 30 suitable for carrying out the method of the present invention is shown in Figure 6. The apparatus 30 comprises a microscope 32, which is focused on an examination area 33. The examination area can be a microscopic slide or a flow cell such as that disclosed in U.S. Patent No. 4,338,024. A camera 34 is attached to the microscope 32 and is adapted to image a portion of a suspension having a particle therein. The camera 34 is preferably of a raster scan type and may be a CCD camera model XC-711 manufactured by Sony. The camera 34 can form an electrical image of the field of view as seen through the microscope 32. Further, the camera 34 segments the image into a plurality of pixels, with an electrical signal corresponding to each pixel of the image. The camera 34 also outputs a plurality of signals (3) one for each of the colors (Red, Blue, and Green).

Each of the signals from the camera 34 is supplied to a digitizer 36, which digitizes the image intensity of each pixel into an electrical signal representing the grey scale value. Preferably, the pixel is digitized into a gray scale value between 0 and 255, inclusive.

From the digitizer, the digitized gray scale value is supplied to a memory 38, where the values are stored. The memory 36 can be a RAM. A boundary detector 40, such as that disclosed in U.S. Patent No. 4,538,299 operates on the image in the memory 38 and detects the boundary of the image detected. The result is also stored in the memory 38.

From the memory 38, a computer 42 operates on the image and measures a parameter, such as intensity of light, of each pixel within the boundary detected.

In the event the apparatus 30 seeks to determine the identity of a single particle under examination, then the particular quantile value derived by the computer 42 is compared to the table of stored values 46 of previously

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identified particles. The table 46 can be a ROM or another memory. The result can be displayed on the display 44.

In the event the apparatus 30 seeks to differentiate  
5 a plurality of objects under examination, then a quantile  
is developed for the same parameter for each of the  
pixels in the boundary of each of the images of the  
different objects. From the quantile, the particles  
under examination can be differentiated from one another.  
10 The result can be displayed on the display 44.

The computer 42 carries out its tasks in accordance  
with a computer program. A copy of a program to compute  
quantiles written in the C language for execution under  
the MS-DOS operating system is as follows:

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```
1 /*      cca.h()          Block Diagram Software      cca.h()
2 **      Programmer:      H. L. Kasdan
3 **
4 **      Copyright :    1987 by Harvey L. Kasdan
5 **
6 **
7 ****
8 **
9 **      NAME           cca.h - chain code analysis header file
10 **
11 **      SYNOPSIS
12 **
13 **      DESCRIPTION
14 **
15 **      SEE ALSO
16 **
17 **      DIAGNOSTICS
18 **
19 **      BUGS
20 **
21 **      REVISION HISTORY
22 **
23 **
24 ****
25 */
26 /*
27 ****
28 **      DEFINES
29 ****
30 */
31 #define MAXX 60
32 #define MAXY 60
33 /*
34 ****
35 **      STRUCTURE DEFINITION
36 ****
```

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```

37 */
38 struct ccddef {
39     char ccang,
40     int x,
41     int y;
42 };
43
44 typedef char nBgroup[7];
45
46 typedef struct {
47     int px,
48     int py,
49     j_point;
50 } typedef point nBpoints[7];
51
52 /*
53 /**
54 **** SUBROUTINE DECLARATIONS
55 **
56 ****
57 */
58 int ccabx0(),
59 int ccabrk0(),
60 int ccaconsC(),
61 int ccaconsS(),
62 int ccadconO(),
63 int ccadetpO(),
64 int ccadilate(),
65 int ccadiss1(),
66 int ccadiss2(),
67 int ccaenter(),
68 int ccaerod1(),
69 int ccaerode(),
70 int ccafilin(),
71 int ccafilli1(),
72 int ccafilil2(),
73 void ccafil10(),
74 int ccanBbf0(),
75 int ccanBbf1(),
76
77

```

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```
1 /*          ndap.h()      Block Diagram Software      ndap.h()
2 **          Programmer:    H. L. Kasdan
3 **
4 **          Copyright :    (C) 1988 by
5 **                      International Remote Imaging Systems Inc.
6 **
7 **
8 ****
9 **
10 ** NAME      ndap.h - NIH data analysis header file
11 **
12 **
13 ** SYNOPSIS
14 **
15 ** DESCRIPTION
16 **
17 ** SEE ALSO
18 **
19 ** DIAGNOSTICS
20 **
21 ** BUGS
22 **
23 ** REVISION HISTORY
24 **
25 ****
26 */
27 /*
28 ****
29 ** DEFINES
30 ****
```

-23-

```
31 */
32 #define CAXMAX 50          /* cell array x-maximum */
33 #define CAYMAX 50          /* cell array y-maximum */
34 /*
35 *****
36 ** STRUCTURE DEFINITION
37 *****
38 */
39 struct Image {
40     char fname[14],           /* cell number */
41     int cellno,              /* red iod */
42     long int riod,            /* green iod */
43     long int giod,            /* blue iod */
44     long int biod,            /* red iot */
45     long int riot,             /* green iot */
46     long int giot,            /* blue iot */
47     long int biot;           /* cell area */
48     int carea;               /* box window size */
49     int window;               /* edge trace threshold */
50     char ctype[12],           /* cell type */
51     int thresh;                /* cell trace starting x-position */
52     int cx0;                  /* cell trace starting y-position */
53     int cy0;                  /* cell trace minimum x-position */
54     int cxmin;                /* cell trace minimum y-position */
55     int cymini;                /* cell trace maximum x-position */
56     int cxmax;
```

```

57     int cymax;
58     int cpie;
59     char ccc[256];
60     unsigned char crimg[CAYMAX][CAXMAX], /* cell trace maximum y-position */
61             /* cell points in edge */
62             /* cell chain code */
63             /* cell red image */
64             /* cell green image */
65             /* cell blue image */
66             /* cell normalized red image */
67             /* cell normalized green image */
68             /* cell normalized blue image */
69             /* log green minus log red image */
70             /* log blue minus log green image */
71             /* log red minus log green cell quantiles */
72             /* log red minus log blue cell quantiles */
73             /* log green minus log red cell pixel mean */
74             /* log blue minus log green cell pixel mean */
75             /* log red minus log blue cell pixel mean */
76             /* log green minus log red cell pixel std dev */
77             /* log blue minus log green cell pixel std dev */
78             /* log red minus log blue cell pixel std dev */
79             /* cell chain code image */
80             /* cell chain code struct */
81     struct histo {
82         unsigned int rhisto[256], /* red image histogram */
83         unsigned int ghisto[256], /* green image histogram */
84         unsigned int bhisto[256], /* blue image histogram */
85         unsigned int rhist0[256], /* red image interior histogram */
86         unsigned int ghist0[256], /* green image interior histogram */
87         unsigned int bhist0[256], /* blue image interior histogram */
88         unsigned int rehist0[256], /* red image exterior histogram */
89         unsigned int gehist0[256], /* green image exterior histogram */
90         unsigned int behist0[256], /* blue image exterior histogram */
91     }, /* histo - cell image data structure */
92     struct chisto {
```

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```
93     unsigned int crhisto[256], /* red image cum histogram */
94         int cghisto[256], /* green image cum histogram */
95         int cbhisto[256], /* blue image cum histogram */
96         int crihisto[256], /* red image interior cum histogram */
97         int cghist0[256], /* green image interior cum histogram */
98         int cbhist0[256], /* blue image interior cum histogram */
99         int crehisto[256], /* red image exterior cum histogram */
100        int cgehisto[256], /* green image exterior cum histogram */
101        int cbehisto[256], /* blue image exterior cum histogram */
102        /* chisto - tell image data histogram structure */
103 struct qt10 {
104     unsigned int qr0501, /* red image quantiles */
105     unsigned int qg0501, /* green image quantiles */
106     unsigned int qb0501, /* blue image quantiles */
107     unsigned int qr10501, /* red image interior quantiles */
108     unsigned int qg10501, /* green image interior quantiles */
109     unsigned int qb10501, /* blue image interior quantiles */
110     unsigned int qr01501, /* red image exterior quantiles */
111     unsigned int qg01501, /* green image exterior quantiles */
112     unsigned int qb01501, /* blue image exterior quantiles */
```

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```

113      unsigned int mr0;
114      unsigned int mg0;
115      unsigned int mb0;
116      unsigned int mri0;
117      unsigned int mgi0;
118      unsigned int mbi0;
119      unsigned int mre0;
120      unsigned int mge0;
121      unsigned int mbg0;
122      unsigned int sdr0;
123      unsigned int sdg0;
124      unsigned int sdb0;
125      unsigned int sdr10;
126      unsigned int sdg10;
127      unsigned int sdb10;
128      unsigned int sdre0;
129      unsigned int sdge0;
130      unsigned int sdbe0;
131      /* qt10 - cell image data quantile structure */
132      /*
133 ****
134 ** SUBROUTINE DECLARATIONS
135 ****
136 */
137 char *1b1tbl();

```

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```

1  /*      ndaph0()          Analysis Software           ndaph0()
2  **      Programmer:        H. L. Kasdan
3  **
4  **      Copyright :       1988 by
5  **                      International Remote Imaging Systems, Inc.
6  **
7  **
8  ****
9  **
10 ** NAME
11 **      ndaph0 - image histogram routine
12 **
13 ** SYNOPSIS
14 **      #include <stdio.h>
15 **      #include <string.h>
16 **      #include "cca.h"
17 **      #include "ndap.h"
18 **      int ndaph0( int mode, struct image0 *p_image0,
19 **                  struct histo *p_histo, struct chisto *p_chisto )
20 **      int mode,          histogram formation mode
21 **      struct image0 *p_image0,    pointer to image0 data structure
22 **      struct histo *p_histo,    pointer to histo data structure
23 **      struct chisto *p_chisto,  pointer to chisto data structure
24 **
25 ** DESCRIPTION
26 **      Routine generates histogram to histo data structure
27 **      using data from image0 data structure according to
28 **      specified mode.
29 **      mode
30 **      -----
31 **          0      use un-normalized image data from image0
32 **          1      use normalized image data from image0
33 **
34 ** SEE ALSO

```

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```
35 **      DIAGNOSTICS
36 **      BUGS
37 **
38 **
39 **
40 **      REVISION HISTORY
41 **      900507.hlk      modified for *.URN file precursor to use window size
42 ****
43 ****
44 */
45 #include <stdio.h>
46 #include <string.h>
47 #include "cca.h"
48 #include "ndap.h"
49 #define FALSE 0
50 #define TRUE 1
51 #define PMAXX 50
52 #define PMAXY 50
53 #define PMINX 0
54 #define PMINY 0
55
56
57 int ndaph0( mode, p_image0, p_hist0, p_chisto )
58     int mode;          /* histogram formation mode */
59     struct image0 *p_image0;    /* pointer to image0 data structure */
60     struct hist0 *p_hist0;    /* pointer to hist0 data structure */
61     struct chisto *p_chisto; /* pointer to chisto data structure */
62 {
63
64 **** LOCAL STORAGE
65 ****
66 ****
```

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```

/*
   int err = 0;           /* error return */
   int i;                /* generic index */
   int x;                /* x-coordinate value */
   int y;                /* y-coordinate value */
*/
73 *****
74 *** Do Histogram According to Mode
75 *****
76 */
77 switch (mode)
78 {
79     case 1:             /* normalized image is data source */
80         for ( i = 0; i < 256; i++)
81         {
82             p_hist0->rhisto[i] = 0;
83             p_hist0->ghisto[i] = 0;
84             p_hist0->bhisto[i] = 0;
85             p_hist0->rhistro[i] = 0;
86             p_hist0->ghistro[i] = 0;
87             p_hist0->brhisto[i] = 0;
88             p_hist0->rbhisto[i] = 0;
89             p_hist0->chistogram[i] = 0;
90             p_hist0->gehisto[i] = 0;
91             p_hist0->behisto[i] = 0;
92         } /* end - for ( i = 0, i < 256, i++) */
93         for (y = PMINY, y < p_image0->window, y++)
94         {
95             for (x = PMINX, x < p_image0->window, x++)
96             {
97                 p_hist0->rhistorl(int)p_image0->crimg[y][x]] += 1,
98                 p_hist0->ghistorl(int)p_image0->crimg[y][x]] += 1,
99                 p_hist0->bristorl(int)p_image0->crimg[y][x]] += 1,
100                p_hist0->rbistorl(int)p_image0->crimg[y][x]] += 1,
101                if ((p_image0->ccimg[y][x] == 'x') ||
102                    ((p_image0->ccimg[y][x] >= '0') && (p_image0->ccimg[y][x] < '8'))) {
103                }
104            }

```

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```

105      */
106      /*          Do Interior
107      p_hist0->rhistor0((int)p_image0->cnrlimg[y][x]) += 1;
108      p_hist0->ghistor0((int)p_image0->cnrlimg[y][x]) += 1;
109      p_hist0->bhist0((int)p_image0->cnrlimg[y][x]) += 1;
110  } /* end - if (...) */
111 else
112 {
113     /*
114     **          Do Exterior
115     */
116     p_hist0->rhist0((int)p_image0->cnrlimg[y][x]) += 1;
117     p_hist0->ghist0((int)p_image0->cnrlimg[y][x]) += 1;
118     p_hist0->bhist0((int)p_image0->cnrlimg[y][x]) += 1;
119 } /* end - else (...) */
120 /* end - for (x = PMINX, x < p_image0->window, x++) */
121 /* end - for (y = PMINY, y < p_image0->window, y++) */
122
123 /**
124 **          Compute Cumulative Histogram
125 */
126 p_chist0->rchistor0 = p_hist0->rhistor0;
127 p_ehist0->cgchistor0 = p_hist0->ghistor0;
128 p_ehist0->cbehistor0 = p_hist0->bhist0;
129 p_chist0->crhist0 = p_hist0->rhistor0;
130 p_chist0->cgchistor0 = p_hist0->ghistor0;
131 p_chist0->cbehistor0 = p_hist0->bhist0;
132 p_chist0->crhist0 = p_hist0->rhistor0;
133 p_chist0->cgchistor0 = p_hist0->ghistor0;
134 p_chist0->cbehistor0 = p_hist0->bhist0;
135
136 for ( i = 1, i < 256, i++)
137 {

```

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```

138     p_chisto->crhisto[ij] = p_histo->rhisto[ij] + p_chisto->crhisto[ij] - 1];
139     p_chisto->cghistot[ij] = p_histo->ghistot[ij] + p_chisto->cghistot[ij] - 1];
140     p_chisto->cbhistot[ij] = p_histo->bhistot[ij] + p_chisto->cbhistot[ij] - 1];
141     p_chisto->crihistot[ij] = p_histo->rihistot[ij] + p_chisto->crihistot[ij] - 1];
142     p_chisto->cgihistot[ij] = p_histo->gihistot[ij] + p_chisto->cgihistot[ij] - 1];
143     p_chisto->cbihistot[ij] = p_histo->bihistot[ij] + p_chisto->cbihistot[ij] - 1];
144     p_chisto->crehistot[ij] = p_histo->rehistot[ij] + p_chisto->crehistot[ij] - 1];
145     p_chisto->cgehistot[ij] = p_histo->gehistot[ij] + p_chisto->cgehistot[ij] - 1];
146     p_chisto->cbehistot[ij] = p_histo->behistot[ij] + p_chisto->cbehistot[ij] - 1];
147 } /* end - for { i = 0, i < 256, i++ */ /
148 break;
149
150 case 0: default:
151     for ( i = 0; i < 256; i++)
152 {
153     p_histo->rhisto[ij] = 0;
154     p_histo->ghistot[ij] = 0;
155     p_histo->bhistot[ij] = 0;
156     p_histo->rihistot[ij] = 0;
157     p_histo->gihistot[ij] = 0;
158     p_histo->bihistot[ij] = 0;
159     p_histo->rehistot[ij] = 0;
160     p_histo->gehistot[ij] = 0;
161     p_histo->behistot[ij] = 0;
162 }
163 } /* end - for { i = 0, i < 256, i++ */ /
164
165 for (y = PMINY, y < p_image0->window, y++)
166 {
167     for (x = PMINX, x < p_image0->window, x++)
168 {

```

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```

169     p_hist0->rhist0((int)p_image0->ccimg[y][x]) += 1;
170     p_hist0->ghistor0((int)p_image0->cgimg[y][x]) += 1;
171     p_hist0->bhist0((int)p_image0->cbimg[y][x]) += 1;
172     if ((p_image0->ccimg[y][x] == 'x') ||
173         ((p_image0->ccimg[y][x] >= '0') && (p_image0->ccimg[y][x] < '8'))) {
174
175     /*          */
176     /*          Do Interior
177     */
178     p_hist0->rhist0((int)p_image0->ccimg[y][x]) += 1;
179     p_hist0->ghistor0((int)p_image0->cgimg[y][x]) += 1;
180     p_hist0->bhist0((int)p_image0->cbimg[y][x]) += 1;
181     } /* end - if (...) */
182     else
183     {
184     /*          */
185     /*          Do Exterior
186     */
187     p_hist0->rhist0((int)p_image0->ccimg[y][x]) += 1;
188     p_hist0->ghistor0((int)p_image0->cgimg[y][x]) += 1;
189     p_hist0->bhist0((int)p_image0->cbimg[y][x]) += 1;
190     } /* end - else (...) */
191     } /* end - for (x = PMINX, x < p_image0->window, x++) */
192     } /* end - for (y = PMINY, y < p_image0->window, y++) */
193
194     /*
195     */
196     /*          Compute Cumulative Histogram
197     p_chist0->rhist0[0] = p_hist0->rhist0[0];
198     p_chist0->ghist0[0] = p_hist0->ghist0[0];
199     p_chist0->bhist0[0] = p_hist0->bhist0[0];
200     p_chist0->crihist0[0] = p_hist0->rihist0[0];
201     p_chist0->cgihist0[0] = p_hist0->gihist0[0];
202     p_chist0->cbihist0[0] = p_hist0->bihist0[0];
203     p_chist0->crehist0[0] = p_hist0->rehist0[0];
204     p_chist0->cgehist0[0] = p_hist0->gehist0[0];
205     p_chist0->cbehist0[0] = p_hist0->behist0[0];

```

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```
206     for ( i = 1, i < 256, i++)
207 {
208     p_chisto->rhisto[i] = p_hist0->rhisto[i] + p_rhist0->rhisto[i] - 1];
209     p_chisto->ghisto[i] = p_hist0->ghisto[i] + p_rhist0->ghisto[i] - 1];
210     p_chisto->bhisto[i] = p_hist0->bhisto[i] + p_rhist0->bhisto[i] - 1];
211     p_chisto->crihisto[i] = p_hist0->crihisto[i] + p_rhist0->crihisto[i] - 1];
212     p_chisto->gihistori = p_hist0->gihistori + p_rhist0->gihistori - 1];
213     p_chisto->cbihistori = p_hist0->cbihistori + p_rhist0->cbihistori - 1];
214     p_chisto->crehisto[i] = p_hist0->crehisto[i] + p_rhist0->crehisto[i] - 1];
215     p_chisto->cgehisto[i] = p_hist0->cgehisto[i] + p_rhist0->cgehisto[i] - 1];
216     p_chisto->cbehistori = p_hist0->cbehistori + p_rhist0->cbehistori - 1];
217     }
218     /* end - for ( i = 0, i < 256, i++) */
219     break;
220   /* end - switch (mode) */
221
222   return(err);
223
224 } /* end - int mdapho(char *cc) */
```

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```
1 /*          ndaptq0()      Analysis Software
2 **          Programmer:    H. L. Kasdan
3 **
4 **
5 **          Copyright :   1988 by
6 **                           International Remote Imaging Systems, Inc.
7 **
8 ****
9 **
10 ** NAME          ndaptq0 - produce cum hist quantiles and statistics routine
11 **
12 **
13 ** SYNOPSIS
14 **          #include <stdio.h>
15 **          #include <string.h>
16 **          #include "cca.h"
17 **          #include "ndap.h"
18 **          int ndaptq0( struct histo *p_chisto, struct qt10 *p_qt10 )
19 **          struct histo *p_chisto;
20 **          pointer to chisto data structure
21 **          struct qt10 *p_qt10;
22 **          pointer to qt10 data structure
22 ** DESCRIPTION
23 **          Routine generates quantiles of cell
24 **          image data using data from chisto data structure.
25 ** SEE ALSO
26 **          DIAGNOSTICS
27 **          BUGS
28 **          REVISION HISTORY
29 **
30 **          BUGS
31 **          REVISION HISTORY
32 **
33 **          REVISION HISTORY
34 ****
```

-35-

```

35 */
36 #include <math.h>
37 #include <stdio.h>
38 #include <string.h>
39 #include "cca.h"
40 #include "ndap.h"
41 #define BNPECT 95
42 #define BOFF 28
43 #define GNPCT 95
44 #define GOFF 21
45 #define FALSE 0
46 #define NVALUE 200
47 #define PMAXX 46
48 #define PMAXY 46
49 #define PMINX 2
50 #define PMINY 2
51 #define RNPCT 95
52 #define ROFF 25
53 #define TRUE 1
54
55 void ndapqnt1();
56 void ndapmsd1();
57
58 int ndapctq( p_chisto, p_qt10 )
59 struct chisto *p_chisto, /* pointer to chisto data structure */
60 struct qt10 *p_qt10, /* pointer to qt10 data structure */
61 {
62
63 /* **** LOCAL STORAGE **** */
64 /* **** */
65 /* **** */
66 /* **** */
67 */
68 int err = 0; /* error return */
69 int i; /* generic index */
70 int n; /* number of interior pixels - length of index
array */
71
72 /*
73 **** Compute Statistics for q images
74 **** */
75

```

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-37-

```

113  */
114  */
115  ndapqnt1(&p_chisto->cbihisto[0], &p_qt10->qbi10[0]);
116  ndapmsd1(&p_chisto->cbihisto[0],
117  &p_qt10->mbi0, &p_qt10->sdbi0);
118
119 /*
120 */
121 */
122 ndapqnt1(&p_chisto->crehisto[0], &p_qt10->qrer0[0]);
123 ndapmsd1(&p_chisto->crehisto[0],
124  &p_qt10->mre0, &p_qt10->sdre0);
125
126 /*
127 */
128 */
129 ndapqnt1(&p_chisto->cgehisto[0], &p_qt10->qge0[0]);
130 ndapmsd1(&p_chisto->cgehisto[0],
131  &p_qt10->mgae0, &p_qt10->sdgao);
132
133 /*
134 */
135 */
136 ndapqnt1(&p_chisto->cbehisto[0], &p_qt10->qbe0[0]);
137 ndapmsd1(&p_chisto->cbehisto[0],
138  &p_qt10->mbe0, &p_qt10->sdbae0);
139
140 return(err);
141
142 } /* end - int ndapctq0(char *cs) */
143
144 /* ndapqnt1() Analysis Software
145 */
146 */
147 ** Programmer: H. L. Kasdan
148 */

```

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```
149 ** Copyright : 1988 by  
150 ** International Remote Imaging Systems, Inc.  
151 **  
152 ****  
153 **  
154 ** NAME ndapqnt1 - compute quantiles from cumulative histogram  
155 **  
156 **  
157 ** SYNOPSIS  
158 ** #include <stdio.h>  
159 ** #include <string.h>  
160 ** #include "cca.h"  
161 ** #include "ndap.h"  
162 ** void ndapqnt1( unsigned int chist[256], int quant[50])  
163 **           unsigned int chist[256], /* cumulative histogram array  
164 **           int quant[50], /* array in which 50 quantiles  
165 **           will be returned  
166 **  
167 ** DESCRIPTION  
168 ** Routine orders determines quantiles from cumulative histogram  
169 ** array.  
170 **  
171 ** SEE ALSO  
172 **  
173 ** DIAGNOSTICS  
174 **  
175 ** BUGS  
176 **  
177 ** REVISION HISTORY  
178 **  
179 ****  
180 */  
181 void ndapqnt1( chist, quant)  
182 unsigned int chist[256], /* cumulative histogram array */  
183 int quant[50], /* array in which 50 quantiles  
184 will be returned */  
185
```

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```

186   {
187   /*
188   **** LOCAL STORAGE ****
189   */
190   ****
191   */
192   /*
193   long int cJ;           /* target cum hist value for next quantile */
194   int i;
195   int j0 = 0;            /* quantile index */
196   long int l50 = 50;     /* number of quantiles */
197   long int maxnum = chist[255]; /* total number of values in histogram */
198   int n = 256;          /* size of histogram array */
199   */
200   */
201   */
202   cJ = (((long int)(j0 + 1)) * maxnum) / 150;
203   for (i = 1, (i <= n) && (j0 < 50); i++)
204   {
205     while (((unsigned int)cJ <= chist[i]) && (j0 < 50))
206     {
207       quant[j0++][i];
208       cJ = (((long int)(j0 + 1)) * maxnum) / 150;
209     } /* end - while ((jnext - j0++) > 1) */
210   }
211   /* end - for (i = 1, i <= n; i++) */
212 } /* end - void ndapmsd1( int n, int arrin[], int indx[] ) */
213
214 /* ndapmsd1()      Analysis Software          ndapmsd1()
215 */
216 */
217 ** Programmer: H. L. Kasdan
218 */
219 ** Copyright : 1998 by
220 ** International Remote Imaging Systems, Inc.
221 */
222 ****
223 */
224 ** NAME

```

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```
225 ** ndapmsd1 - compute mean and std deviation
226 **
227 ** SYNOPSIS
228 ** #include <cmath.h>
229 ** #include <stdio.h>
230 ** #include <string.h>
231 ** #include "cca.h"
232 ** #include "ndap.h"
233 ** void ndapmsd1( unsigned int chist[256], int *mean,
234 **                  int *sdev)
235 **          cum histogram array
236 **          mean value of specified
237 **          elements
238 **          standard deviation of
239 **          specified elements
240 **
241 ** DESCRIPTION
242 ** Routine computes mean and standard deviation of
243 ** image from cumulative histogram.
244 **
245 ** SEE ALSO
246 **
247 ** DIAGNOSTICS
248 **
249 ** BUGS
250 **
251 ** REVISION HISTORY
252 **
253 ****
254 */
255 void ndapmsd1( chist, mean, sdev) /* cum histogram array */
256 unsigned int chist[256], /* mean value of specified
257 int *mean, /* elements */
258 /* standard deviation of
259 int *sdev; /* specified elements */
260
```

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```
262      *
263  {
264  /*
265  *****
266  ** LOCAL STORAGE
267  *****
268  */
269  int chvalue;
270  double di,
271  double dmean,
272  int dopixel,
273  double dpixel,
274  double dnt = (double)chist[255], /* total number of points in histogram */
275  double dstddev,
276  double dsm,
277  int i,
278  int n = 256,
279  /*
280  */

  /* current histogram value */
  /* double index */
  /* running first moment */
  /* previous temp pixel value */
  /* temp pixel value */
  /* histogram array size */
  /* Compute Moments
```

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```
281 /*  
282 dmean = 0.;  
283 dsm = 0.;  
284 dopixel = 0.;  
285 for (i = 0; (i < n); i++)  
286 {  
287     chvalue = (int)chist[i];  
288     dopixel = (double)(chvalue - dopixel);  
289     if (dopixel != 0.)  
290     {  
291         di = (double)i;  
292         dmean += dopixel * di;  
293         dsm += dopixel * di * di;  
294     } /* end - if (dopixel != 0.) */  
295     dopixel = chvalue;  
296     } /* end - for (i = 1; i <= n; i++) */  
297     dmean /= dnt;  
298     *mean = (int)dmean;  
299     dsm /= dnt;  
300     dsdev = dsm - (dmean * dmean);  
301     dsdev = (dsdev >= (double)0.) ? dsdev : (double)0.;  
302     dsdev = sqrt(dsdev);  
303     *sdev = (int)dsdev;  
304 } /* end - void ndapmsd1( int n, int arrin[], int index[] ) */  
305
```

```
1 /*      ndapmlq0()          Analysis Software
2 **      Programmer:        H. L. Kasdan
3 **
4 **      Copyright :        1988 by
5 **                          International Remote Imaging Systems, Inc.
6 **
7 **      ****
8 ***** NAME          ndapmlq0 - produce log image quantiles and statistics routine
9 ***
10 ** NAME          ndapmlq0 - produce log image quantiles and statistics routine
11 **
12 **
13 ** SYNOPSIS
14 **      #include <stdio.h>
15 **      #include <string.h>
16 **      #include "cca.h"
17 **      #include "ndsp.h"
18 **      int ndapmlq0( struct image0 *p_image0 )
19 **      struct image0 *p_image0;
20 **
21 ** DESCRIPTION
22 **      Routine generates quantiles of log difference cell
23 **      image data using data from image0 data structure.
24 **      Modified heapsort algorithm is used to order pixel data.
25 **
26 ** SEE ALSO
27 **      DIAGNOSTICS
28 **      BUGS
29 **
30 **
31 **      REVISION HISTORY
32 **      900507 hlk modified for *.URN precursor to use window value
33 **
34 **
35 *****
```

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```

36 */
37 #include <math.h>
38 #include <stdio.h>
39 #include <string.h>
40 #include "cca.h"
41 #include "ndap.h"
42 #define BNPECT 95
43 #define BOFF 28
44 #define ONPECT 95
45 #define GOFF 21
46 #define FALSE 0
47 #define NVALUE 200
48 #define PMAXX 46
49 #define PMAXY 46
50 #define PMINX 2
51 #define PMINY 2
52 #define RNPECT 95
53 #define ROFF 25
54 #define TRUE 1
55
56 void indexx();
57 void ndapqnt0();
58 void ndapmsd0();
59
60 int ndapmlq0( p_image0 )
61 struct image0 *p_image0; /* pointer to image0 data structure */
62 {
63 /*
64 **** GLOBAL STORAGE ****
65 */
66 /*
67 */
68 extern int ltab0[256], /* precomputed log table values */
69
70 /*
71 */
72 */
73 /*
74 */
75 int bvalue; /* blue pixel normalizing value */

```

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```

76     int err = 0;           /* error return */
77     int i;                /* generic index */
78     int index[CAXMAX * CAYMAX + 1]; /* interior pixel index array */
79     int gvalue;            /* temporary pixel normalizing value */
80     long int lptmp;
81     int n;
82     int rvalue;
83     int i;
84     int x;
85     int y;
86     /*
87     *****
88     *** Generate Index Array
89     *****
90     */
91     n = 0;
92     /* Index count */
93     for ( y = 0, y < p_image0->window; y++ )
94     {
95       for ( x = 0, x < p_image0->window; x++ )
96       {
97         if ((p_image0->ccimg[y][x] == 'x') ||
98             ((p_image0->ccimg[y][x] >= '0') && (p_image0->ccimg[y][x] < 'B')) )
99         {
100           index[n] = (y * CAXMAX) + x;
101           /* end - if (.....) && (.....) */
102           /* end - for ( x = 0, x < p_image0->window; x++ ) */
103           /* end - for ( y = 0, y < p_image0->window; y++ ) */
104           /*
105           */
106           /*
107           */
108           indexx(n, &p_image0->lgnl[0][0], index); /* order index array for lgmlr */
109           ndapqnt0(n, &p_image0->lgnl[0][0], index, &p_image0->qclgnl[0]);
110           ndapmsd0(n, &p_image0->lgnl[0][0], index,
111           &p_image0->mc_lgnl, &p_image0->sd_lgnl),
112

```

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```

113 /*
114 */
115 /* indexx(n, &p_image0->lbmlg[0][0], indx); /* order index array for lbmlg */
116 ndapqnt0(n, &p_image0->lbmlg[0][0], indx, &p_image0->qclbmlg[0]);
117 ndapmsd0(n, &p_image0->lbmlg[0][0], indx,
118 &p_image0->mc1bmlg, &p_image0->sdc1bmlg);
119
120
121 /* irmlb Cell Image
122 */
123 /* indexx(n, &p_image0->irmlb[0][0], indx); /* order index array for irmlb */
124 ndapqnt0(n, &p_image0->irmlb[0][0], indx, &p_image0->qclrmlb[0]);
125 ndapmsd0(n, &p_image0->irmlb[0][0], indx,
126 &p_image0->mc1rmlb, &p_image0->sdc1rmlb);
127
128
129 return(err);
130
131 /* end - int ndapmlq0(char *cs) */
132
133 /* indexx() Analysis Software
134 */
135 /* Programmer: H. L. Kasden
136 */
137 */
138 /* Copyright : 1988 by
139 International Remote Imaging Systems, Inc.
140 */
141 ****
142 */
143 /* NAME indexx - order index array for log image data
144 */
145 */
146 /* SYNOPSIS
147 */
148 */
149 */
150 */

```

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```

151 ** void indexx( int n, int arrin[], int indx[])
152 **   int n;
153 **   int arrin[];
154 **   int indx[];
155 **
156 **   Number of elements to be ordered
157 **   array containing elements to
158 **   be ordered
159 **   array containing indecies of
160 **   elements in arrin that will
161 **   be ordered. This index array
162 **   will be ordered in place.

160 ** DESCRIPTION
161 ** Routine orders elements in array arrin specified in
162 ** index array, indx.
163 **

164 ** SEE ALSO
165 **

166 ** DIAGNOSTICS
167 **

168 ** BUGS
169 **

170 ** REVISION HISTORY
171 **

172 ****
173 */
174 void indexx( n, arrin, indx) /* number of elements to be ordered */
175   int n; /* array containing elements to
176   int arrin[]; be ordered */
177   /* array containing indecies of
178   elements in arrin that will
179   be ordered. This index array
180   will be ordered in place. */

182
183 {
184   /*
185   ****
186   LOCAL STORAGE
187   ****

```

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```
188 */
189 int i;
190 int indx[1];
191 int ir;
192 int j;
193 int l;
194 int q;
195 */
196 */
197 */
198 i = (n >> 1) + 1;
199 ir = n;
200
201 for (;;)
202 {
203     if (l > 1)
204         q = arrin[(indx = indx[-1])];
205     else
206         q = arrin[(indx = indx[ir])];
207     if (l > 1)
208         indx[ir] = indx[1];
209     indx[1] = indx[ir];
210     if (--ir == 1)
211         indx[1] = indx[1];
212     return;
213 }
214 /* end - if (-ir == 1) */
215 /* end - else (l > 1) */
216 i = 1;
217 j = 1 << 1;
218 while (j <= ir)
219 {
220     if (j < arrin[indx[j]] < arrin[indx[j + 1]]) j++;
221     if (q < arrin[indx[j]])
222     if (q < arrin[indx[j + 1]])
223     {
224 }
```

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```

225     index[i] = index[j];
226     j += (i=j);
227     } /* end - if (q < arrin[index[j]]) */
228   else j = ir + 1;
229   } /* end - while (j <= ir) */
230
231   index[i] = index[j];
232
233   } /* end - for (i, ) */
234
235   } /* end - void indexx( int n, int arrin[], int index[] ) */
236
237 /* ndapqnto()      Analysis Software
238 **          programmer: H. L. Kasdan
239 **          Copyright : 1988 by
240 **          International Remote Imaging Systems, Inc.
241 **          ****
242 **          ****
243 **          ****
244 **          ****
245 **          ****
246 **          NAME
247 **          ndapqnto - order index array for log image data
248 **          SYNOPSIS
249 **          #include <stdio.h>
250 **          #include <cstring.h>
251 **          #include "ccs.h"
252 **          #include "ndap.h"
253 **          void ndapqnto( int n, int arrin[], int index[], int quant[50] )
254 **          number of elements to be ordered
255 **          array containing elements to
256 **          be ordered
257 **          array containing indexes of
258 **          elements in arrin that will
259 **          be ordered. This index array
260 **          will be ordered in place.
261 **          int quant[50];
262 **          int quant[50];
263 **          */
264

```

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```

265 ** DESCRIPTION
266 ** Routine orders determines quantiles from elements in array
267 ** arrin according to index array, indx.
268 **
269 ** SEE ALSO
270 **
271 ** DIAGNOSTICS
272 **
273 ** BUGS
274 **
275 ** REVISION HISTORY
276 ****
277 ****
278 */
279 void rdapqnto( n, arrin, indx, quant) /* number of elements to be ordered */
280 int n; /* array containing elements to
281 int arrin[], be ordered */
282 int indx[], /* array containing indecies of
283 int index[], elements in arrin that will
284 be ordered. This index array
285 will be ordered in place. */
286 int quant[50], /* array in which 50 quantiles
287 will be returned */
288
289
290 {
291 /*
292 **** LOCAL STORAGE
293 ****
294 */
295 /*
296 */
297 int i, /* quantile index */
298 int j0 = 0, /* target quantile index */
299 int jnext; /* */
300 /*
301 */
302 */

```

Generate Quantiles

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```

3003 quant[0] = arrinIndex[1];
3004 j0 = 1;
3005 for (i = 1; (i <= n) && (j0 < 50); i++)
3006 {
3007     jnext = (i * 50) / n;
3008     while (((jnext - j0) > 0) && (j0 < 50))
3009     {
3010         quant[j0++] = arrinIndex[i];
3011         } /* end - while ((jnext - j0++) > 1) */
3012     } /* end - for (i = 1, i <= n, i++) */
3013     /* end - void ndapmsd0( int n, int errint, int index[], */
3014 }
3015
3016 /* ndapmsd0()           Analysis Software
3017 ** Programmer: H. L. Kasdan
3018 ** Copyright : 1988 by International Remote Imaging Systems, Inc.
3019 ** NAME      ndapmsd0 - order index array for log image data
3020 **
3021 ** SYNOPSIS
3022 ** #include <math.h>
3023 ** #include <stdio.h>
3024 ** #include <string.h>
3025 ** #include "cra.h"
3026 ** #include "ndap.h"
3027 ** void ndapmsd0( int n, int errint, int index[], int *mean,
3028 **                int *sdev)
3029 **
3030 **
3031 **
3032 **
3033 **
3034 **
3035 **
3036 **

```

```

337  ** number of elements to be ordered
338  ** array containing elements to
339  ** be ordered
340  ** array containing indexies of
341  ** elements in arrin that will
342  ** be used.
343  ** mean value of specified
344  ** elements
345  ** standard deviation of
346  ** specified elements
347  **
348  ** DESCRIPTION
349  ** Routine orders computes mean and standard deviation of
350  ** elements in array arrin specified in index array, indx.
351  **
352  ** SEE ALSO
353  **
354  ** DIAGNOSTICS
355  **
356  ** BUGS
357  **
358  ** REVISION HISTORY
359  **
360  ****
361  */
362  void ndapmsdov( n, arrin, indx, mean, sdev)
363  int n; /* number of elements to be ordered */
364  int arrin[]; /* array containing elements to
365  ** be ordered */
366  int indx[], /* array containing indexies of
367  ** elements in arrin that will
368  ** be ordered. This index array
369  ** will be ordered in place. */
370  int *mean; /* mean value of specified
371  ** elements */
372  int *sdev; /* standard deviation of
373  ** specified elements */
374
375
376  /*
377

```

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```

378 ****
379 ** LOCAL STORAGE
380 ****
381 */
382 double dmean,           /* running first moment */
383 double dpixel,          /* temp pixel value */
384 double dsdev,           /* double std dev */
385 double dsm;             /* running second moment */
386 int i;
387 */
388 */
389 */
390 dmean = 0. ;
391 dsm = 0. ;
392 for (i = 1; i <= n; i++)
393 {
394     dpixel = (double)arrinIndex[i];
395     dmean += dpixel;
396     dsm += dpixel * dpixel;
397 } /* end - for (i = 1; i <= n; i++) */
398 dmean /= (double)n;
399 *mean = (int)dmean;
400 dsm /= (double)n;
401 dsdev = (dsdev - (dmean * dmean));
402 dsdev = (dsdev >= (double)0.) ? dsdev : -(double)0. ;
403 dsdev = sqrt(dsdev);
404 *sdev = (int)dsdev;
405 } /* end - void ndapmsd0( int n, int arrin[], int index[] ) */
406

```

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WHAT IS CLAIMED IS:

1. A method of identifying an object under examination, said object has a discernible boundary; said method comprising the steps of:

- 5        a) forming an image of said object;
- b) segmenting said image to form a plurality of pixels;
- c) detecting the boundary of said object under examination in said image;
- 10       d) measuring a parameter of each pixel within the boundary detected;
- e) forming an ordered sequence of the parameter of the pixel measured, within the boundary detected;

15                       $P_1 \dots P_k \dots P_N$

$P_1$  - first value of the parameter of the pixel within the boundary detected;

$P_N$  - last value of the parameter of the pixel within the boundary deducted;

20                      N - total number of pixels within the boundary detected;

k - index to the N pixels;

- f) determining the  $P_k$  data point(s) such that k satisfies the following relationship; and

$$\frac{i-1}{M} < \frac{k}{N} \leq \frac{i}{M}$$

M - total number of quantile partitions;

i - i<sup>th</sup> fraction of partition or i<sup>th</sup> quantile partitions;

- g) identifying said object under examination, based upon the relationship developed from step (f) associated with said object.

2. The method of Claim 1 wherein  $P_1$  is the smallest value and  $P_N$  is the largest value.

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3. The method of Claim 1, wherein said identifying step further comprising:

5 comparing said relationship of step (f) associated with said object under examination, to a table of pre-determined relationships of other identified objects to identify the object under examination.

4. The method of Claim 3, wherein said table of pre-determined relationships of other identified objects comprises a range of values.

10 5. The method of Claim 3, wherein said table of predetermined relationships of other identified object comprises a single value.

15 6. The method of Claim 4, wherein said table of pre-determined relationships are based upon experimental results.

7. The method of Claim 6, wherein said table of predetermined relationships of other identified object comprises a single value.

20 8. The method of Claim 1, wherein said parameter is intensity of visible light.

9. The method of Claim 1, wherein said parameter is a function of color representations at the same position.

25 10. The method of Claim 9, wherein said color representation is based upon three primary colors.

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11. The method of Claim 9, wherein said color representation is based upon hue, intensity, and saturation of colors.

12. The method of Claim 10, wherein said parameter  
5 is

log (a) - log (b)  
where a and b are the intensities of two different colors.

10 13. The method of Claim 12 wherein a is red color  
and b is blue color.

14. The method of Claim 12, wherein a is blue color  
and b is green color.

15 15. The method of Claim 12, wherein a is green color  
and b is red color.

16. A method of differentiating a plurality of  
15 different objects under examination, wherein each object  
has a discernible boundary; said method comprising the  
steps of:

- 20 a) forming a plurality of images, with each image  
of each of said plurality of objects;
  - b) segmenting each of said plurality of images to  
form a plurality of pixels;
  - c) detecting the boundary of each object under  
examination in each of said images;
  - 25 d) measuring a parameter of each pixel within the  
boundary of each object detected;
  - e) forming an ordered sequence of the parameter of  
the pixel measured, within the boundary  
detected for each object;
- 30  $P_1 \dots P_{k1} \dots P_{N1}$  - object 1 having  $N_1$  pixels

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$P_1 \dots P_{k_2} \dots P_{N_2}$  - object 2 having  $N_2$  pixels

- f) determining, for each ordered sequence formed, the  $P_k$  data point(s) such that  $k$  satisfies the following; and

5             $\frac{i-1}{M} < \frac{k}{\text{Total No. of pixels for the object}} \leq \frac{i}{M}$

M - total number of quantile partitions;

10            i -  $i^{\text{th}}$  fraction of partition or  $i^{\text{th}}$  quantile partition;

- g) comparing the  $P_k$  data point(s) of each ordered sequence for each object to one another, for identical variables of  $i$  and  $M$ , to  
15            differentiate the objects.

17. The method of Claim 16, wherein said parameter is intensity of visible light.

18. The method of Claim 16, wherein said parameter is a function of color representations at the same  
20            position.

19. The method of Claim 18, wherein said color representation is based upon three primary colors.

20. The method of Claim 18, wherein said color representation is based upon hue, intensity, and  
25            saturation of colors.

21. The method of Claim 19, wherein said parameter is

$$\log (a) - \log (b)$$

where a and b are the intensities of two different  
30            colors.

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22. The method of Claim 21 wherein a is red color and b is blue color.

23. The method of Claim 21, wherein a is blue color and b is green color.

5 24. The method of Claim 21, wherein a is green color and b is red color.

25. An apparatus for identifying an object under examination, said object has a discernible boundary; said apparatus comprising:

- 10 a) means for forming an image of said object;
- b) means for segmenting said image to form a plurality of pixels;
- c) means for detecting the boundary of said image of said object under examination;
- 15 d) means for measuring a parameter of each pixel within the boundary detected;
- e) means for forming an ordered sequence of the parameter of the pixel measured, within the boundary detected;

20  $P_1 \dots P_k \dots P_N$

$P_1$  - first value of the parameter of the pixel within the boundary detected;

$P_N$  - last value of the parameter of the pixel within the boundary deducted;

25 N - total number of pixels within the boundary detected;

k - index to the N pixels;

- f) means for determining the  $P_k$  data point(s) such that k satisfies the following relationship;

30 and

$$\frac{i-1}{M} < \frac{k}{N} \leq \frac{i}{M}$$

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M - total number of quantile partitions;  
i - i<sup>th</sup> fraction of partition or i<sup>th</sup> quantile  
partition;

- 5 g) means for identifying said object under examination, based upon the relationship of P<sub>k</sub> to i<sup>th</sup> quantile in M partitions developed from step (f) associated with said object.

26. The apparatus of Claim 25 wherein P<sub>1</sub> is the smallest value and P<sub>N</sub> is the largest value.

- 10 27. The apparatus of Claim 25, wherein said means for identifying further comprising:

means for comparing said relationship of step (f) associated with said object under examination, to a table of pre-determined relationships of other identified objects to identify the object under examination.

15 28. The apparatus of Claim 27, wherein said table of pre-determined relationships of other identified objects comprises a range of values.

- 20 29. The apparatus of Claim 26, wherein said parameter is intensity of visible light.

30. The apparatus of Claim 29, wherein said parameter is a difference of color representations at the same position.

- 25 31. The apparatus of Claim 30, wherein said color representation is based upon three primary colors.

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32. The apparatus of Claim 30, wherein said color representation is based upon hue, intensity, and saturation of colors.

5 33. The apparatus of Claim 31, wherein said parameter is

$$\log (a) - \log (b)$$

where a and b are the intensities of two different primary colors.

10 34. The apparatus of Claim 33, wherein a is red color and b is blue color.

35. The apparatus of Claim 33, wherein a is blue color and b is green color.

36. The apparatus of Claim 33, wherein a is green color and b is red color.

15 37. An apparatus for differentiating a plurality of objects under examination, each of said objects has a discernible boundary; said apparatus comprising:

- 20 a) means for forming an image of each of said plurality of objects to form a plurality of images;
- b) means for segmenting each of said plurality of images to form a plurality of pixels;
- c) means for detecting the boundary of each of said plurality of objects under examination in each of said plurality of images;
- d) means for measuring a parameter of each pixel within the boundary of each object detected;
- e) means for forming an ordered sequence of the parameter of the pixel measured, within the boundary detected for each object;

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$P_1 \dots P_{k_1} \dots P_{N_1}$  - object 1 having  $N_1$  pixels

$P_1 \dots P_{k_2} \dots P_{N_2}$  - object 2 having  $N_2$  pixels

- f) means for determining, for each ordered sequence formed, the  $P_k$  data point(s) such that k satisfies the following; and

$$\frac{i-1}{M} < \frac{k}{\text{Total No. of pixels for the object}} \leq \frac{i}{M}$$

10 M - total number of quantile partitions;  
i - i<sup>th</sup> fraction of partition or i<sup>th</sup> quantile partition;

- g) means for comparing the  $P_k$  data point(s) of each ordered sequence for each object to one another, for identical variables of i and M, to differentiate the objects.

38. The apparatus of Claim 37 wherein  $P_1$  is the smallest value and  $P_N$  is the largest value.

- 20 39. The apparatus of Claim 37, wherein said means for identifying further comprising:

means for comparing said relationship of step (f) associated with said object under examination, to a table of pre-determined relationships of other identified objects to identify the object under examination.

- 25 40. The apparatus of Claim 39, wherein said table of pre-determined relationships of other identified objects comprises a range of values.

41. The apparatus of Claim 38, wherein said parameter is intensity of visible light.

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42. The apparatus of Claim 41, wherein said parameter is a difference of color representations at the same position.

5 43. The apparatus of Claim 42, wherein said color representation is based upon three primary colors.

44. The apparatus of Claim 42, wherein said color representation is based upon hue, intensity, and saturation of colors.

10 45. The apparatus of Claim 43, wherein said parameter is

log (a) - log (b)  
where a and b are the intensities of two different primary colors.

15 46. The apparatus of Claim 45, wherein a is red color and b is blue color.

47. The apparatus of Claim 45, wherein a is blue color and b is green color.

48. The apparatus of Claim 45, wherein a is green color and b is red color.

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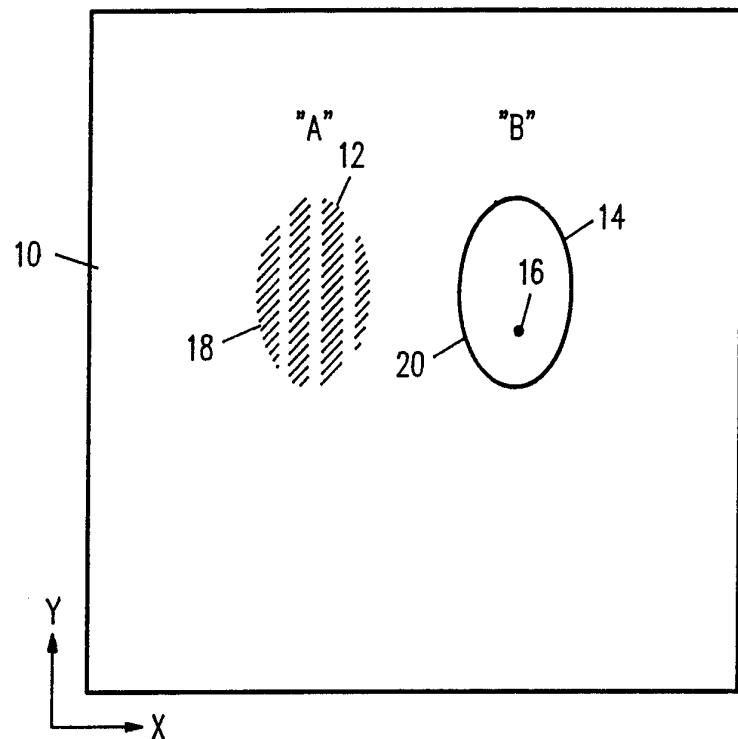


FIG. 1

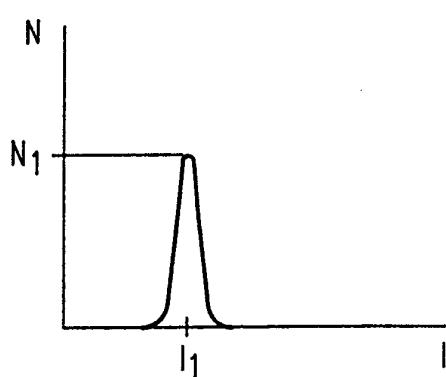


FIG. 2a

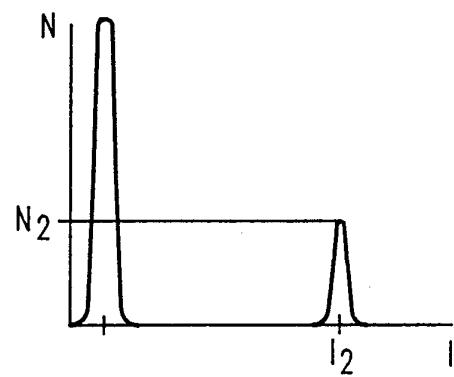


FIG. 2b

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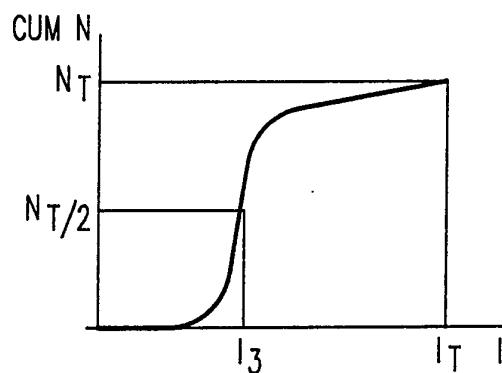


FIG. 3a

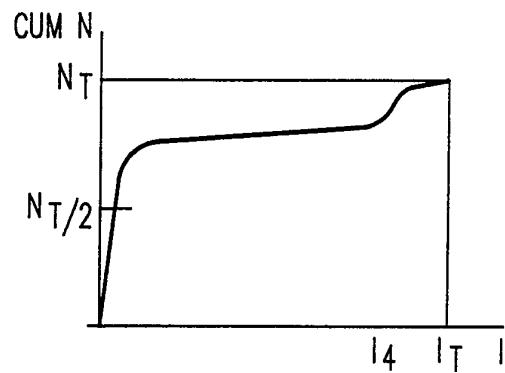


FIG. 3b

50th QUANTILE	<u>INTENSITY</u>	<u>OBJECT</u>
	$ _a$	xxx
	.	.
	.	.
	$ _3$	12
	.	.
	.	.
	$ _4$	14
	.	.
	.	.
	$ _x$	xxx

FIG. 4

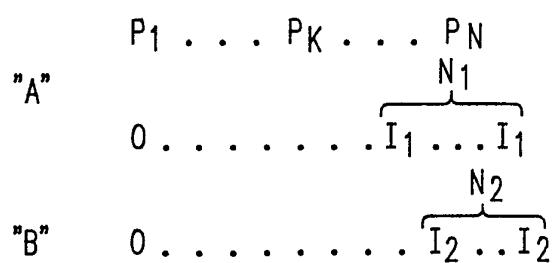


FIG. 5

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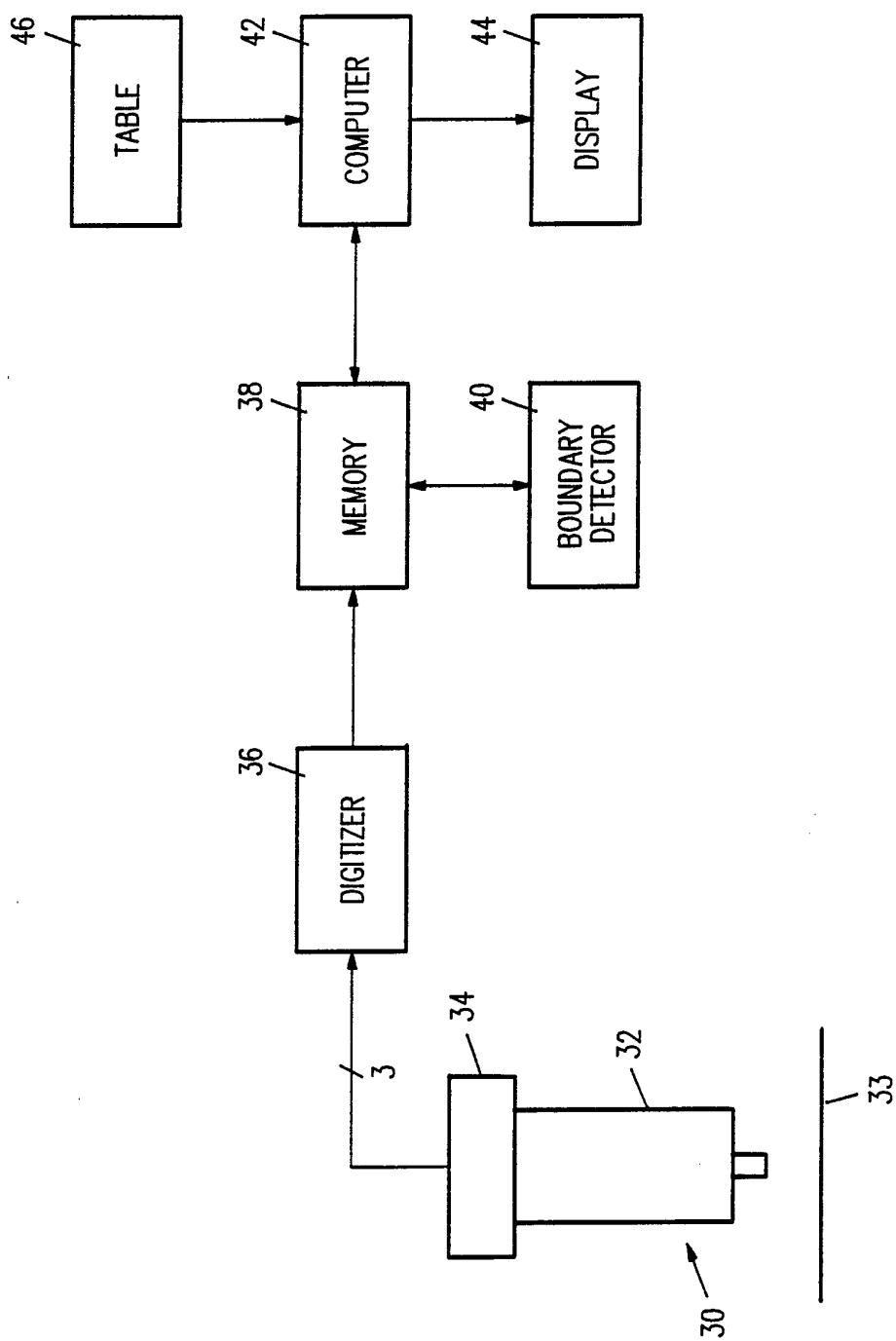


FIG. 6

## INTERNATIONAL SEARCH REPORT

International application No.  
PCT/US93/09159

**A. CLASSIFICATION OF SUBJECT MATTER**

IPC(5) :G06K 09/48

US CL :382/22, 18

According to International Patent Classification (IPC) or to both national classification and IPC

**B. FIELDS SEARCHED**

Minimum documentation searched (classification system followed by classification symbols)

U.S. : 382/22, 18, 6, 17, 21, 54

Documentation searched other than minimum documentation to the extent that such documents are included in the fields searched  
**CHECKED REFERENCES CITED IN US PATENT 5,123,055**

Electronic data base consulted during the international search (name of data base and, where practicable, search terms used)  
Please See Extra Sheet.

**C. DOCUMENTS CONSIDERED TO BE RELEVANT**

Category*	Citation of document, with indication, where appropriate, of the relevant passages	Relevant to claim No.
X	WO, A, WO90/14639 (KASDAN ET AL.) 29 NOVEMBER 1990, whole document.	1-48
A	US, A, 4,075,604 (DEGASPERI) 21 FEBRUARY 1978, see fig. 3 and col. 1, line 63 - col. 2, line 13; col. 4, lines 12-31.	1, 16, 25, 37
A	US, A, 5,121,338 (LODDER) 09 JUNE 1992, see Fig. 2 and col. 6 - col 8, line 25; col. 5, lines 8-29.	1-9, 16, 18, 25-30, 37-42
A	US, A, 4,573,197 (CRIMMINS) 25 FEBRUARY 1986, see Abstract, Figs. 1-3.	1, 16, 25, 37
A	US, A, 3,705,383 (FRAYER) 05 DECEMBER 1972, see col. 20, lines 41-56.	8-10, 17-19, 29-31, 41-43



Further documents are listed in the continuation of Box C.



See patent family annex.

* Special categories of cited documents:	"T"	later document published after the international filing date or priority date and not in conflict with the application but cited to understand the principle or theory underlying the invention
"A" document defining the general state of the art which is not considered to be part of particular relevance	"X"	document of particular relevance; the claimed invention cannot be considered novel or cannot be considered to involve an inventive step when the document is taken alone
"E" earlier document published on or after the international filing date	"Y"	document of particular relevance; the claimed invention cannot be considered to involve an inventive step when the document is combined with one or more other such documents, such combination being obvious to a person skilled in the art
"L" document which may throw doubts on priority claim(s) or which is cited to establish the publication date of another citation or other special reason (as specified)	"&"	document member of the same patent family
"O" document referring to an oral disclosure, use, exhibition or other means		
"P" document published prior to the international filing date but later than the priority date claimed		

Date of the actual completion of the international search

08 November 1993

Date of mailing of the international search report

FEB 07 1994

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## INTERNATIONAL SEARCH REPORT

International application No.

PCT/US93/09159

## C (Continuation). DOCUMENTS CONSIDERED TO BE RELEVANT

Category*	Citation of document, with indication, where appropriate, of the relevant passages	Relevant to claim No.
A	US, A, 4,782,389 (MAYWEATHER III) 01 NOVEMBER 1988, see Abstract, Fig. 1.	1, 16, 25, 37