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1	Navy Case No. 77581
2	SITE AND WORKSPACES LAYOUT PROCESS EMPLOYING
3	
4	MDS AND A PDI FORMULA IN WHICH DENSITY IS CALCULATED USING
5	MEASURED SPAN OF CIRCUMSCRIBING-CONVEX-HULLS
6	
7	STATEMENT OF GOVERNMENT INTEREST
8	The invention described herein may be manufactured and used
9	by or for the Government of the United States of America for
10	governmental purposes without the payment of any royalties
. 11	thereon or therefor.
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13	CROSS-REFERENCE TO RELATED APPLICATIONS
14	The instant application is related to two-pending U.S.
15	Patent Applications entitled SITE AND WORKSPACES LAYOUT PROCESS
16	EMPLOYING MDS AND A PDI FORMULA IN WHICH DENSITY IS BASED ON AREA
17	OF CIRCUMSCRIBING-CONVEX-HULLS (Navy Case No. 77580); and SITE
18	AND WORKSPACES LAYOUT PROCESS EMPLOYING MDS AND A PDI FORMULA IN
19	WHICH DENSITY IS CALCULATED USING A UNIT LATTICE SUPERPOSED OVER
20	CIRCUMSCRIBING-CONVEX-HULLS (Navy Case No. 77585) having same
21	filing date.
22	
23	BACKGROUND OF THE INVENTION
. 24	(1) Field of the Invention
25	The present invention relates to improvements in the art of
26	in functional
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organizations, including the location of building units of 1 multiple building functional facilities and location of personnel 2 work stations and equipments (collectively workplace elements") 3 in functional workspaces. In one of its aspect, the invention 4 relates to a non-metric multidimensional scaling (MDS) matrices 5 process which optimizes these layouts according to patterns of 6 inter-object operational criteria. In another of its aspects it 7 relates to the measurement of crowdedness of the objects 8 ("population density index"). In still another of its aspect it 9 relates to a novel computational geometry technique enabling 10 measurements of crowdedness from data in the form of span 11 measurements (as where working from aerial photographs). 12

13 (2) Description of the Prior Art

A paper, T. Tullis, B.B. Sperling and A.L. Steinberg (1986), 14 "The Use of Multidimensional Scaling for Facilities Layout: An 15 Application to the Design of the Space Station", outlines a 16 process employing the multidimensional scaling (MDS) methodology 17 to modify an experimental layout of a workspace of a naval vessel 18 from at an analytical viewpoint of optimizing performance of a 19 set of operational criteria associated with the function of the 20 workspace on the naval vessel. However, the disclosed process 21 make no provision for any consideration of crowdedness 22 (population density) in connection with the modification. 23

An abandoned U.S. Patent Application Serial No. 07/754,779 filed 30 August 1991 (which is accessible to the public by virtue of it being referred to in. inter alia, a U.S. Patent 5,235,506

to F.J. O'Brien, Jr.) discloses a process for calculating a form 1 of a population density index (PDI). This abandoned application 2 goes on to make the observation that use of PDI data in MDS would 3 However, there is provide additional data for facilities layout. 4 no disclosure or teaching of how to employ PDI data with MDS. 5 Also, the form of density calculation of employed by the PDI 6 equation disclosed therein is based upon the area of the full 7 bounds of the quadrilateral the workspace in which a layout 8 configuration is located. As will become apparent, of in 9 accordance with the present invention a different form density 10 calculation is employed in the PDI equation. 11

Other related references included U.S. Patent 5,402,335 to 12 F.J. O'Brien, Jr. which discloses a process for producing 13 optimized layouts including of calculation of a non-metric PDI 14 MDS matrix (column 24, lines 29-32, therein) which is then 15 representing an MDS matrix of a normative ("best") a non-metric 16 MDS matrix of other inter-object matrices (column 24, lines 35-17 38) are combined. As will become apparent of in accordance with 18 the present invention, one never generates a non-metric PDI 19 matrix and a totally different form interaction between PDI and 20 MDS is involved. Also, the earlier mentioned U.S. Patent 21 5,235,506, and an abandoned U.S. Patent Application 07/756,264, 22 file August 30, 1991 (but publicly available by virtue of a 23 reference thereto, inter alia, in U.S. Patent 5,235,506), each 24 include further observations that used of PDI data in MDS would 25 provide additional data for facilities layout. However, these 26

observations were also made without description of a process of 1 using PDI with MDS. As a further distinction, the PDI formula 2 disclosed in U.S. Patent 5,235,506, and that disclosed in 3 abandoned application 07/756,264 have respective limitations of 4 applicability to (I) a restricted number of objects and (ii) 5 situations where an approximate density index is acceptable. 6 Further each of the U.S. Patent 5,402,335, U.S. Patent 5,235,506 7 and abandoned application 07/756,264 disclose only forms of 8 density calculations in their PDI formula which are based upon 9 area of the full bounds of a quadrilateral workspace form, i.e., 10 different from density calculation in PDI equation of the present 11 invention. 12

The forms of equations supporting the calculation of PDI in all 13 the above disclosures involving PDI have treated the density 14 related term of the equation in ways which readily process data 15 regarding location of the building units and workplace elements 16 when in coordinate measurements. However, there are layout 17 analysis problems where coordinate measurements are not easily 18 obtained, as where working from aerial photographs. It therefore 19 has been a continuing, but heretofore unobtained objective in the 20 development of these layout processes for a feature enabling the 21 process to be applied to measurements in the form of lengths of 22 spans, as can be readily made from aerial photographs. 23

SUMMARY OF THE INVENTION

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The invention provides a process for producing layouts of 2 building units, and layouts of personnel and equipment stations 3 (collectively "workplace elements") within the building units. 4 The building-unit configurations are to be located at a 5 quadrilaterally shaped facility-wide tier ("global area") and the 6 work element configurations are to be located at a 7 quadrilaterally shaped workspace tier ("subarea"). A layout 8 analyst becomes knowledgeable in or acts in concert with some 9 knowledgeable in the function of the facility and workplace 10 (single or in such concert called "expert"). The expert prepares 11 best intuitive experimental configurations of the building units 12 in the global area and workplace elements in their subarea, which 13 are termed "experts normative configuration". The analyst also 14 collects data regarding inter-building-unit and inter-workplace-15 element operational criteria, such as for example inter-building-16 unit-transition-frequency and shared usages of building units. 17 Using the well known non-metric MDS methodology, and separately 18 processing the global area and subarea tiers, the expert's 19 normative configuration is combined with individual matrices sets 20 of the different operational criteria producing a set of 21 normative configuration and operational criterion MDS matrices. 22 Also using well known MDS methodology and again with global area 23 and subarea tier separately processed, all the sets of normative 24 configuration and operational criterion matrices for building 25 units and workplace elements are combined, with uniform weighting 26

assigned to items of the configuration, and during the combining 1 step are interactively non-linearly stressed to produce "least 2 bad fit" configuration for performance of the operating criteria. 3 "Convex hulls" (in the mathematical sense of the term) are 4 circumscribed about the "least bad fit" and experts normative 5 configurations, respectively, and the polygon areas of the 6 interior of the hull are calculated using equation 8a, set forth 7 later in this specification. Population density indices (PDIs) 8 based upon a novel PDI formula (equation 9, later in the 9 specification) are calculated. The equation for the novel PDI 10 calculates a density related term using the above calculated 11 polygon area (rather than area bounded by the quadrilaterally 12 shaped global area and subarea, as was used in the prior art). 13 Further, novel computational geometry is employed in the 14 measurement of crowdedness, which enables obtaining data in the 15 form of measured spans (rather than coordinate measurements), 16 making the process convenient for situtions such as site layouts 17 prepared from aerial photographs. The PDI for the "least bad 18 fit" configuration is compared with the PDI for the "expert's 19 normative" configuration to determine if the "least bad fit" 20 configuration which represents a configuration presenting least 21 stress in meeting the operational criteria is acceptable in light 22 of impact on PDI vis-a-vis the PDI of the expert's experimental 23 configuration. 24

Accordingly, the principal objects of the present invention

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are:

(1) To provide a novel process which aids in laying out the location of objects in a functional facility of objects (e.g., multibuilding facility, workspace) which involves both MDS methodology optimization based upon takes into consideration both inter-object operational criteria and consideration crowdedness (PDI).

(2) To provide a novel process as aforesaid wherein the measure of PDI takes into consideration an observation that human activity tends to clusters away from boundaries of facility sites or workspace areas.

(3) To provide a novel process in accordance with the preceding object which is of special utility in cases where measurements of facility site distances are readily obtainable in the form of span measurements (a where working with aerial photographs).

(4) To provide a novel process in accordance with the first above said object for laying out two tiers of a functional facility, namely a process tier of locating building units in a quadrilateral facility site and a another process tier of locating personnel workstations and equipments (collectively "workplace elements") in quadrilateral workspaces within a building unit.

BRIEF DESCRIPTION OF THE DRAWINGS 1 A more complete understanding of the invention and many of 2 the attendant advantages thereto will be readily appreciated as 3 the same becomes better understood by reference to the following 4 detailed description when considered in conjunction with the 5 accompanying drawings wherein: 6 FIGS. 1A-1F are flow charts of an embodiment of process for 7 laying out multibuilding facilities and workspaces subareas 8 within the building; 9 FIGS. 2, 3, 4, 5A and 5B are diagrammtics useful with 10 respect to a hereinafter presented discussion of mathematical and 11 logical theory underlying the measure of crowdedness in employee 12 in the process of FIGS. 1A-1F; 13 FIGS. 6A and 6B are diagrammatics useful in understanding 14 process steps 26 and 26' in FIGS. 1C and 1E respectively; and 15 FIGS. 7, 8, and 9A-9D are diagrammatics useful with respect 16 a hereinafter presented discussion of mathematical and logical 17 theory related to one of the critical parameters PDINORM employed 18 in calculations performed in the process depicted in FIG. 1. 19 20 DESCRIPTION OF THE PREFERRED EMBODIMENT 21 Reference is now made to FIGS. 1A-1F. In accordance with 22 the present invention there is provided a process 10, FIG. 1A, 23 comprising a first tier of the process consisting of groups of 24

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steps 12a and 12b, FIGS. 1B and 1C, and a second tier of the

process consisting of groups of steps 14a and 14b, FIGS. 1D and

Groups of steps 12a and 12b relate to producing layouts of 1 1E. building units (including temporary structures such as tents of a 2 facility to be located in a designated quadrilateral site 3 ("global area")). The aim of process 10 is to produce layouts 4 which are nearly optimal with respect to the functions being 5 performed in the facility (first tier) and in a workspace (second б In both the cases of the facility and workspace there has 7 tier). been collected data regarding operational criteria as among the 8 building units and workplace elements. The above-identified 9 paper by Tullis et al. discusses and lists operational criteria 10 useful for the first tier (process step groups 12a, 12b). They 11 include: inter-building transition frequency, sequential activity 12 dependencies, inter-building travel distance, shared support 13 functions, privacy requirements of work/living spaces and volume 14 flow among buildings. The Tullis et al paper also discusses and 15 lists individual workspace operational criteria useful for the 16 second tier (process step groups 14a, 14b). These include: 17 human transition frequency, sequential activity dependencies, 18 shared support equipment, visual/auditory interference potential 19 and privacy requirements. The total number of criteria which 20 process 10 can accommodate is unlimited. Among the types of 21 functional facilities for which appropriate facility level 22 operational criteria exist are collectible are: (1) mobile 23 military field hospitals; (2) fixed hospital compounds; (3) 24 aircraft rework depots; (4) industrial facilities for carrying on 25 complex production modes; (5) facilities involved with production 26

of nuclear energy or production of hazardous nuclear materials; 1 (6) college campuses; and (7) prison compounds. The 2 corresponding functional workspace within these functional 3 facilities for which for which appropriate workspace level 4 operational criteria exist are: (1) and (2) in hospitals-triage 5 centers, surgery centers, x-ray centers, etc.; (3) in aircraft 6 rework depots-workspaces for aircraft breakdown, component 7 overhaul, parts and tool storage, reassembly; (6) in college 8 campuses-classrooms, food service areas, labs, libraries, etc.; 9 and (7) in prison compound-cell blocks, food service areas, 10 laundry areas, etc. 11

Final actual construction of a facility, and any appropriate construction and fitting out of the workspace takes place, constitutes a realization tier 16, FIG. 1F of process 10.

Reference is now made to FIGS. 2, 3, 4, 5A and 5B. The present and succeeding eleven (11) paragraphs constitute discussion of the mathematical and logical theory aspects of measuring or modeling (mathematical sense) two-dimensional space. A computational example is also provided.

The conventional formula to measure or model two-dimensional discrete spatial density, i.e., population density or physical crowding is defined as the average number of objects (n) per unit area of space (A):

 $D = \frac{n}{4}$

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(1)

This definition has severe shortcomings since actual spatial 1 orientation within a specified area is disregarded. As an 2 example of this shortcoming, refer to FIG. 1 which displays three 3 different configurations of objects or density points. In each 4 case, the "perceived density" of the four points is obviously 5 different. Since the number of points and area are identical in б each depiction, there is a constant value of .25 for population 7 density. FIG. 2 depicts geometrically the population 8 demographer's model of population density shown for the 9 distributions in FIG. 1. FIG. 2 shows that each point occupies 10 four space units (such as feet); hence, population density or 11 physical crowdedness (D=n/A) equals one object per four square 12 feet. FIG. 2 represents the model for each depiction of FIG. 1. 13 However, large differences in perceived physical crowding clearly 14 exist among the three configurations shown in FIG. 1. 15 See the paper, F. J. O'Brien, "A Crowding Index for Finite 16

Populations", <u>Perceptual and Motor Skills</u>, February 1990, 70, pp.
3-11, by this reference hereby incorporated herein in its
entirety discloses a formula to exploit the difference shown in
FIG. 2 for move accurately representing crowding. This is
accomplished by taking the actual spatial orientation of objects
into account. Also see US Patents 5,235,506 and 5,402,335, each
incorporated herein in its entirety by reference.

This formula, referred to as the Population Density Index (PDI) (and synonymous with "Crowding Index), is as follows:

$$PDI = \frac{1}{d} \sqrt{\frac{n}{A}}$$
(2)
where
n = number of objects
A = the geometric area, and
d = average Euclidean distance among all possible pairs of n
objects.
Basically, the above proposed formula is a generalization of
the bivariate Euclidean distance formula. The derivation of the
proposed density formula is patterned on the well known square-
root law used in the physical sciences. It may be noted in
passing that PDI has the conceptual meaning:
$$\frac{Average distance of one pair of points}{Average distance of all possible pairs of points}$$
(3)
Assume two objects are plotted on an X, Y Cartesian
coordinate system with a fixed origin 0. The mathematical
distance between the two objects is measurable by simple analytic
geometry using the Pythagorean distance formula:
$$d_{12} = [(X_1 - X_2)^2 + (Y_1 - Y_2)^2]^{1/2}$$
(4)
where (X_1, Y_1) ; (X_2, Y_2) represent each object's coordinates.
If, now, we conceive of n objects, each given coordinates
within the same geometric plane such as a room, it is possible to
generalize the above formula to obtain an average Euclidean
distance among the n objects. The average Euclidean distance of
n points, considered pairwise, is given by:

$$\overline{d} = \frac{2\sum d_{ij}}{n(n-1)}$$
(5)

where d_{ij} is the Euclidean distance between any two objects. Note that for n=2 objects, $d;d_{12}$ are equivalent. 3

The last step in deriving a density index is to scale d to 4 adjust for a given number of objects residing within a specific 5 area. A proposed general formula based on the square-root inverse 6 law for distances incorporating size of area and the number of 7 objects is: 8

 $\Delta = \overline{d} \sqrt{\frac{A}{n}}$

(6)

where 10

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A = the geometric area in which objects reside, and 11 n = the number of objects within one area. 12 Dimensional analysis, as well as empirical Monte Carlo 13 simulation investigations, of Δ shows that the units are: 14

$$\frac{A}{\sqrt{n}}; \frac{ft^2}{\sqrt{n}} \tag{7}$$

Essentially, Δ is the average pairwise Euclidean distance among 16 n objects scaled for a given unit area. As will become evident 17 in the following numerical example, Δ is inversely related to 18 the average geometric distances among n points. Calculating the 19 reciprocal of Δ , 1/ Δ will make the relationship monotonically 20 increasing, that is, the more densely packed the objects, the 21

higher the value of the index. This reciprocal of Δ , or $1/\Delta$, 1 is arbitrarily referred to as the population density index, or

PDI. The units for PDI are $\frac{\sqrt{n}}{4}$. 3

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A computational example is provided with the aid of FIG. 3. 4 For four points, there are $4 \times 3/2 = 6$ pairwise distances to 5 The coordinate points for the 4 units are (1,1), calculate. 6 (1,3), (2,4) and (3,2). The area shown is 16 units. Applying 7 Δ , Calculating the reciprocal of Δ and multiplying by 10 to give 8 integer results, PDI=2.3. 9

$$\Delta = \frac{2(d_{12} + d_{13} + d_{14} + d_{23} + d_{24} + d_{34})}{4(3)} \sqrt{\frac{A}{n}} = \sqrt{\frac{16}{4}}(2.22) = 4.4$$
(8)

The Δ index appears to be valid even when areas differ by a 11 large amount. To demonstrate this, consider FIGS. 4A and 4B. 12 The average Euclidean distances are identical (1.6) in each 13 situation depicted. The smaller value of Δ in FIG. 4A (3.7) is 14 in accord with the basic interpretation of Δ , that is, the 15 smaller the value of Δ , the more densely packed are the points 16 relative to the allowed area. The results also correspond to the 17 intuitive notion of density. 18

The proposed crowding index, Δ or PDI, should be 19 interpreted as a relative measure much like a standard deviation 20 The theoretical mathematical minimum value of Δ in statistics. 21

or PDI is always 0, a condition realizable with dimensionless 1 points but not realizable with solid objects such as people. 2 The maximum value depends on the number of objects and the 3 geometric area. Beyond three or four objects, it becomes 4 difficult and perhaps meaningless to attempt calculating a 5 precise maximum value of Δ or PDI. For these reasons, 6 hypothetical minimum and maximum bounds of the PDI formula are 7 derived below and presented as an integral component of the 8 disclosure of the last named application in the section "cross-9 Three additional properties derived from the reference" above. 10 square-root law for average distances of Δ or PDI appear to be 11 critical to the usefulness and interpretability of the index: 1) 12 for constant area, PDI varies directly with the number of 13 objects; 2) for a constant number of objects, PDI varies 14 indirectly with area; and 3) for a constant number of objects and 15 constant area, PDI varies indirectly with distance. Small sample 16 Monte Carlo simulations performed by the inventor have supported 17 these square-root properties for the PDI formula. The values of 18 PDI computed from randomly selected uniform distributions 19 correlated .96 with the conventional formula for population 20 In addition, the PDI formula can be evaluated on density (n/A). 21 three key scientific criteria. First, the model is very simple. 22 It connects population density to three key variables -23 distance, number of points and area - through an equation that 24 can be readily calculated. Second, the formula is justified by 25 mathematical analysis. The inverse square-root properties of the 26

index stated as conjectures are very reasonable and provide a 1 context for prediction and explanation of observed results. 2 Monte Carlo simulations support each conjecture, thereby 3 providing preliminary justification until large scale simulations 4 can be conducted. Third, the formula has been tested and 5 verified by empirical research. The use of the formula in 6 hypothetical military settings has produced results that were 7 readily interpretable and which correlated with qualitative 8 estimates of crowding made by independent expert observers. 9

The population density formula of US Patent No. 5,402,335 10 attempts to express differences such as those shown in FIG. 1 11 more accurately than the conventional population density formula. 12 Since the index can vary widely, as indicated in FIG. 1, it was 13 necessary to develop a new model to predict minimum and maximum 14 bounds of the population density index values. The model was 15 then expanded to accommodate any number of density points. Such 16 models are discussed in US Patent No. 5,402,335. 17

Referring now to FIGS. 6A and 6B, in accordance with the 18 present invention a novel form of population density index (PDI) 19 is employed in the group of facility layout process steps 12b, 20 FIG. 1C, and in the group of subarea process steps 14b, FIG. 1E. 21 This novel form of PDI is calculated using the polygon area of a 22 convex hull (the word "hull" being use in its mathematical sense 23 as a geometric structure formed by linear segments connecting the 24 perimetrical, or outer points of a configuration of points). In 25 the present discussion, wherein it is treated on a generalized 26

basis it is represented by the symbol PDIpoly. Hereinbelow where
 it is discussed relative to specific forms of its in groups of
 process steps 12b and 14b it is variously represented by the
 symbols PDI_{CPOSMDS}, PDI_{NORM}, PDI'_{CPOSMDS} and PDI'_{NORM}.

Reference is now made to FIG. 6A which represents a 5 quadrilateral area containing a configuration of five density б point P_1 through P_5 , representing the centeroid of building units 7 or workplace elements in the context of process 10. Referring 8 now to FIG. 6B, a convex hull having four sides, S, may be 9 visualized as an elastic band wrapped around the outer or 10 perimetrical pins of a configuration of pins occupying the 11 positions of set of point P_1 through P_5 . 12

In accordance with the present invention, the quantity, A_{poly}, of the convex hull calculated by a formula based on span measure interrelationships among the building units or workplace elements represented by points P₁, P₂, P₃ and P₄, FIG. 6B. The following steps are applied:

(1) Connect the points to form a convex hull, as describedabove.

(2) Partition the convex hull into mutually exclusive
 component triangles, such that compositely the partitioned
 triangles overlap the convex hull.

(3) Measure lines representing sides of the component
trianges.

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(4) For each component triangle compute its area, K. (5) Sum the measures of K to get A_{poly} .

FIG. 6B demonstrates the first three steps based upon the 1 configuration of points of FIG. 6A (coordinates now assumed to be 2 Two triangles can be formed from the convex hull unknown). 3 circumscribed over points P_1 , P_2 , P_3 and P_4 . It will be 4 appreciated that the general rule is that n_p -2 component 5 triangles will result from a convex hull consisting of n_1 6 boundary, or perimentrical points. K is calculated for each of 7 the component triangles using the formula 8 $A_{k} = \sqrt{(c - s_{1})(c - s_{2})(c - s_{3})}$ (8a) 9 where s_1 , s_2 , and s_3 are lengths of the triangle's side, and c is 10 the triangle's semiperimeter of the triangle, namely 11 $c = \frac{s_1 + s_2 + s_3}{2}.$ (8b) 12 Then A_{poly} is obtained as follows 13 $A_{poly} = \sum_{i=1}^{n} k_{j}$ (8C) 14 15 PDI_{poly} is calculated as follows: 16 $PDI_{poly} = \frac{1}{\overline{d}} \sqrt{\frac{n}{A_{rob}}}$ (8d) 17 where n = the number of objects in the configuration, and 18 \overline{d} =average Euclidean distance among all possible pairs of 19 points 20 Insofar as the invention is presently understood, the 21 calculation of density aspect of the PDI formula on the basis of 22 the polygon area, A_{poly}, bounded by the convex hull takes into 23 18

account the actual spatial distribution of the objects and 1 thereby provides a more accurate calculation of population 2 density than simply the area of the quadrilateral space which is 3 the basis of the calculation in above identified abandoned, but 4 available to the public, U.S. Patent Application 07/754,779 filed 5 30 August 1991, and other references list in the hereinabove 6 "Description of the Prior Art" section. This conclusion tends to 7 be corroborated by a newer trend of thought in the "human 8 factors" disciplines of engineering, namely that self adjustment 9 responses to crowdedness or congestion by humans at their work 10 stations is more related to the relative position of other 11 workplace elements (personnel/equipment) than to the bounds of 12 the workplace. Stated another way, it is thought that there is 13 an intrinsic human behavior characteristic for individuals to 14 cluster together away from the boundaries of their facility or 15 workspace area, as they adapt to required operational criteria. 16

The primary density metric employed in conjunction with process 10 is a normative population density index, PDI_{norm} or PDI'_{NORM} depending upon whether used in group of process steps 12b or 14b, respectively. It is calculated by the equation:

$$PDI_{NORM}; PDI'_{norm} = \frac{1}{\overline{d}} \sqrt{\frac{n}{A_{poly}}}$$
 (9)

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quadrilateral area (global area/subarea) by the layout analysts, based a human intuitive response to knowledge of the function and human activity occurring in the quadrilateral area. More specifically the planner forms a convex hull out of the

perimetrical or outer points (centeroid of objects), the average 1 inter-object distance, \overline{d} , and polygon area, A_{poly} , are computed, 2 and the value produced by equation (9) is deemed the normative 3 population density index, PDI_{NORM} . Preferably, PDI_{NORM} is derived 4 from an experimental configuration which is the best available 5 human judgment, or combinations of human judgments. Typically, 6 it is selected by the concerted effort of a skilled layout 7 planner and client, or client's representative, who through 8 experience developed knowledge and/or experience involving of the 9 facility/workspace function, and has a keen awareness of inter-10 object requirements. The person, or combination of persons, 11 providing this best human judgment is hereinafter, and in the 12 appended claims sometimes referred to as the "expert". 13

References is now made to FIGS. 7, 8, 9A and 9B. The present and succeeding nineteen (19) paragraphs constitute a discussion of the mathematical and logical theory relating to bounds and other properties of PDI_{NORM}, including computational examples.

As will become apparent in connection with the description of the details of group of product steps 12b and 14b, PDI_{norm} is a critical parameter in producing layouts of building units and workplace-elements in accordance with the present invention.

Reference is now made to the above-identified abandoned, but publicly accessible U.S. Patent Application Serial No. 07/756,264, filed 30 August 1991, which is hereby incorporated by reference. With respect to the method of that invention,

firstly, it is to be understood that it is postulated with a 1 sample size of density points and geometric area of interest, the 2 points are plotted in a uniform lattice. That is, a two 3 dimensional grid, distribution, in a checkerboard arrangement 4 with every consecutive horizontal and vertical point being 5 Secondly, using this plotted distribution, two equidistant. 6 theoretical indices are calculated - a lower bound density index 7 and an upper bound density index. Thirdly, the data regarding 8 personnel and/or equipments which are the subjects of the layout 9 are collected and the population density index values are 10 calculated. Parenthetically, the latter abandoned, but 11 accessible to the public, patent application presents a 12 mathematical proof that a density metric actual PDI 13 (PDI_{act})therein (which correspond to the "norm" PDI (PDI_{norm}) 14 herein) is bounded by minimum and maximum PDIs. Fourthly, the 15 "effective inter-point distance" index is calculated based on the 16 actual population density index values, and the research findings 17 are compared to the model indices. 18

The process disclosed in that abandoned application include 19 at least two process steps involving different "models" of 20 experimental configurations. This feature of a two-step model in 21 that process allows an evaluation of the different layout 22 solutions explored as part of a given layout task, and enables 23 comparisons with solutions provided in connection with other 24 layout tasks. In the second step, the lattice or uniform 25 distribution is an effective visual aid for demonstrating how 26

population density changes with dynamic human or equipment 1 positioning being relocatable at will. 2

In the second step, the lower bound of the population 3 density index can be calculated for any uniform arrangement of 4 The lower bound is based on a lattice of the integers 5 points. called a "unit lattice". In general, a unit lattice means a 6 uniform distribution of n points in area A such that n = A. This 7 implies that the inter-point distance of consecutive horizontal 8 and vertical points is always equal to 1. A non-unit lattice 9 will mean that n and A are not equal. In the special cases of n 10 = 2 and n = 3 objects, "unit lattice" means either a unit line 11 segment (n = 2) or, for n = 3, a Euclidean equilateral triangle 12 (perimeter = 3 units), each constructed in the interior of A. A 13 linear dimension is herein designated in feet. 14

FIG. 7 depicts of a 2 X 2 unit lattice. Note that the area 15 is 4 ft² and that the number of points is 4, or n = A. The 16 horizontal and vertical distance between each of the consecutive 17 points is equal to 1. This is derived from a simple relation 18 that provides inter-point distances of lattices. Namely, if $\boldsymbol{\delta}$ 19 denotes the inter-point distance, then 20

> $\delta = \sqrt{4\frac{ft^2}{4}} = 1$ (10)

where δ is in feet. 22

The above example illustrates the approach that is used for 23 approximating inter-point distances for any lattice of n points 24

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1 uniformly distributed in area A. The general formula to do this 2 is given by:

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$$\delta = \sqrt{\frac{A}{n}} \tag{11}$$

The next step, calculating the average Euclidean distance of all possible pairs of points in FIG. 5, is given by:

$$\overline{\Delta} = \frac{I + I + I + I + \sqrt{2} + \sqrt{2}}{6} = 1.14$$
(12)

The average Euclidean distance for a unit lattice is called $\overline{\Delta}$ to distinguish it from the general Euclidean distance given by \overline{d} in the general population density index formula:

$$PDI = \frac{l}{\overline{d}} \sqrt{\frac{n}{A}}$$
(13)

The reference parameter "lower bound estimate of the index 11 for a non-unit lattice" can now be derived. (Note that this 12 reference parameter is a hypothetical concept). It will be 13 appreciated that this is prior to conducting the density 14 analysis. At that time, an analyst involved with density of a 15 layout will have collected data providing a knowledge of the 16 number of points in the area, but the expected lower bound 17 Euclidean distance will not be known. The known average inter-18 point distance relationship of equation: 19

$$S = \sqrt{\frac{A}{n}} \tag{14}$$

is used to calculate the average Euclidean distance for any nonunit lattice with the same number of points as the unit lattice

1 distribution. The formula to calculate the average Euclidean 2 distance associated with the minimum density (\overline{d}_{\min}) is given in the 3 following equation:

$$\overline{d}_{\min} = \overline{\Delta} \sqrt{\frac{A}{n}}$$
(15)

That is, each coordinate point in the unit lattice is scaled 6 by a constant, equal to δ in equation:

$$\delta = \sqrt{\frac{A}{n}}$$
(16)

8 to calculate the lower bound of the average Euclidean distance.
9 Thus, the lower-bound model of the population density index
10 is obtained by replacing d in the general population density
11 index formula:

$$PDI = \frac{l}{\overline{d}} \sqrt{\frac{n}{A}}$$
(17)

13 || by \overline{d}_{\min} in equation:

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$$\vec{A}_{\min} = \vec{\Delta} \sqrt{\frac{A}{n}} \tag{18}$$

15 || and simplifying. The result is the lower bound (PDI_{min}):

$$PDI_{\min} = \frac{1}{\overline{\Delta}} \frac{n}{A}$$
(19)

To provide an example of calculating the population density index lower-bound (PDI_{min}), assume a study is being conducted on 12 people. Also assume the area involved in the study is a 20 square with area of 25 ft². There are two logical choices in the 21 selection of a unit lattice of 12 points: 4X3 and 6X2 or

equivalently 3X4 and 6X2. A unit lattice of 4X3 or 3X4 will 1 provide an excellent estimate of the lower-bound value of the 2 index for this square. The only other logical choice for a 3 uniform distribution of 12 points is the unit lattice 6X2 or 2X6; 4 however, with this choice, the 12 points cannot be accommodated 5 into a 5-ft.X5-ft. area with inter-point distance δ =1.4. FIG. 8 6 shows the 4X3 unit lattice. $\overline{\Delta}$ can be calculated using equation 7 46, below. Thus, the lower bound population density index based 8 on equation: 9

$$PDI_{\min} = \frac{1}{\overline{\Delta}} \frac{n}{A}$$
(20)

11 is (1/1.90) (12/25), which is approximately equal to 0.25. The 12 population density index can be no smaller than 0.25 units in a 13 lattice distribution consisting of 12(4X3 unit lattice) points 14 and area 25 ft² with inter-point distance of about 1.4.

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The calculation of the upper bound of the index is based on 15 a further assumption. It is assumed that there is a minimum 16 (non-zero) inter-point distance between any two neighboring 17 points in a uniform distribution corresponding to some practical 18 lower limit of elbow room allowable between persons. The 19 distance value selected will correspond to a lattice distribution 20 that produces the maximum population density index value for the 21 given number of objects and area in a layout task. Selecting the 22 minimum inter-point distance is empirical. To exemplify the 23 derivation, assume that 1 ft is the minimum value. As is widely 24 accepted by persons skilled in the art to which the present 25

invention pertains, this distance might correspond to the nose-1 to-nose distance of two persons positioned shoulder-to-shoulder. 2 Other values of minimum inter-point distance, appropriate to the 3 circumstances of the functional organization involved in a given 4 task at hand of laying out objects, are to be selected by the 5 layout analyst employing this invention. Then the task worker 6 computes it as a reasonable value to choose based on the known 7 opinion of experts. Other values could be chosen by this analyst 8 who could derive personally upper bounds for the population 9 density index using the derivation that follows. 10

11 The assumption that 1 ft is the practical minimum inter-12 point distance translates into δ of $\delta = (A/n)^{1/2}$ being set to 1. 13 From this, it follows from the generalized average Euclidean 14 distance equation:

$$\overline{d}_{\min} = \overline{\Delta} \sqrt{\frac{A}{n}}$$
(21)

16 || that the upper bound of the average Euclidean distance (\overline{d}_{max}) is:

$$\overline{d}_{\max} = \overline{\Delta} \sqrt{\frac{A}{n}} = \overline{\Delta}$$
 (22)

18 Substituting equation:

15

17

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21

$$\overline{d}_{\max} = \overline{\Delta} \sqrt{\frac{A}{n}} = \overline{\Delta}$$
(23)

20 into the general population density index formula:

$$PDI = \frac{l}{\overline{d}} \sqrt{\frac{n}{A}}$$
(24)

22 gives the upper bound:

$$PDI_{\max} = \frac{1}{\overline{\Delta}} \sqrt{\frac{n}{A}}.$$
 (25)

2 || Equation

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3

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$$PDI_{\max} = \frac{1}{\overline{\Delta}} \sqrt{\frac{n}{A}}$$
(26)

qives the expected reference parameter "upper bound of the population density index in a lattice" distribution (which is a hypothetical concept) assuming a 1-ft distance as the practical minimum value of inter-point elbow room. From the earlier example where n = 12 objects and area = 25 ft², PDI_{max} is equal to:

$$PDI_{\text{max}} = (1/1.90)(12/25)^{1/2} = 0.36.$$
 (27)

11 That is, density can be no larger than .36 units when the 12 12 density points are distributed uniformly with a one foot distance 13 between each horizontal or vertical point.

In general, if the selected inter-point distance is some arbitrary constant c, then PDI_{max} is:

$$PDI_{max} = \frac{1}{c\overline{\Delta}}\sqrt{\frac{n}{A}}$$
(28)

For example, if 1 in. is the selected value for c, then the maximum population density index value is

$$PDI_{\rm max} = 12(1/1.90)(12/25)^{1/2} = 4.38.$$
 (29)

The final index provided by the model is called the actual effective inter-point distance or δ_{eff} . The effective interpoint distance index translates the clustering of n points

1 observed in an actual study into a lattice distribution for which 2 an hypothetical inter-point distance, or effective inter-point 3 distance can be determined and compared with the theoretical 4 maximum inter-point distance of uniform dispersions in non-unit 5 lattices provided by equation:

$$S = \sqrt{\frac{A}{n}}.$$
 (30)

The utility of this comparison resides in the fact that δ_{eff} 8 varies in accord with the relation $1 \le \delta_{eff} \le \delta$.

The effective uniform - distance index can be derived as 9 A generalized version of equation (30) provides the follows. 10 average Euclidean distance for any lattice. For \overline{d}_{min} equation (22) 11 is calculated directly. For \overline{d}_{max} , $(A/n)^{1/2}$ is set equal to some 12 hypothetical constant (such as 1). Now, a reference parameter 13 "inter-point distance" such as $\delta = (A/n)^{1/2}$ can be conceived 14 theoretically as a variable number for any population density index 15 value in a study with n persons and area A. An inter-point 16 distance such as δ is strictly determined by the average Euclidean 17 distance and vice versa. Hence, the following equation expresses 18 the hypothetical relationship between the actual Euclidean distance 19 d_{act} of an observed population density index value and another 20 reference parameter $\delta_{\mbox{\tiny eff}}$. 21

22

б

 $\overline{\mathbf{d}}_{\mathrm{act}} = \overline{\Delta} \delta_{\mathrm{eff}} \tag{31}$

23 || solving for δ gives:

$$\delta_{\text{eff}} = \frac{\overline{d}_{\text{act}}}{\overline{\Delta}}$$
(32)

From the earlier discussion, it is obvious that δ_{eff} expresses the ratio of the actual PDI relation to the maximum PDI and expresses it in a average inter-point distances. Equation(32) can be stated explicitly in terms of the norm population density index (PDI_{norm}) of the general population density index formula (equation (33) hereinabove):

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$$\delta_{\text{eff}} = \frac{1}{\text{PDI}_{\text{norm}}\overline{\Delta}} \sqrt{\frac{n}{A}}$$
(33)

Either equation (32) or equation (33) provides the effective inter-9 point distance in a population density study. To illustrate the 10 concept of effective inter-point distance, assume that in the 11 example described earlier with 12 persons in an area of 25 ${\sf ft}^2$ 12 (refer to FIG. 9C), that the calculated norm PDI value (PDI_{norm}) is 13 0.30. If a uniform distribution of the 12 points is constructed, 14 the inter-point hypothetical uniform distance that preserves the 15 actual average Euclidean distance δ_{eff} based on equation (33) is 16 equal to $(1/0.30)(1/1.90)(12/25)^{1/2}=1.2$ ft. 17

That is, a population density index value of 0.30 means that the 12 persons can be arranged theoretically in a uniform distribution such that the hypothetical inter-point uniform distance is about 1.2 feet. This value can be compared with

$$\delta_{\text{eff}} = \sqrt{\frac{25}{12}} \approx 1.4 \tag{34}$$

which corresponds to the hypothetical inter point maximum uniform 1 dispersion of 12 persons in a 25 ft² area. When this is done, the 2 percent increase in density associated with a population density 3 index of 0.30 is approximately 20 percent (1.2 vs 1.4ft). In 4 summary, δ_{eff} gives a useful (visual) mathematical summarization of 5 discrete spatial density, translated into terms of uniform interб point distance language. FIGS. 9A-9D are diagrammatic depictions 7 representing the employment of actual effective inter-point 8 distance $\delta_{ extsf{eff}}$ as a model to visualize the translation of 9 mathematical summarizations of discrete density into terms of 10 inter-point distances. To collect all of the information contained 11 in the model, these four figures are presented as a summary for the 12 example employing 12 persons within an area of 25ft². The 1.3 assumptions and findings for this example were as follows. First, 14 n was selected as 12 and the area was 25 ft². Secondly, a unit 15 lattice of 4X3 was determined to be appropriate for the calculation 16 of the unit lattice Euclidean distance $\overline{\Delta}$ (1.90, see FIG. 8). In 17 the third step of the model, lower and upper bounds of the 18 population density index were calculated to be 0.25 and 0.36 units, 19 respectively. The lower and upper bounds of the index were shown 20 to be describable in terms of uniform inter-point distances. In 21 the examples, those values are 1.4 (FIG. 9A) and 1.0(FIG. 9B) feet 22 for the lower and upper bounds, respectively. It was then 23 demonstrated how to translate the PDInorm value into a uniform 24 inter-point distance using equation (11). The data points with a 25 PDInorm value of 0.30 (FIG. 9C) were then translated into a uniform 26

1 distribution of points ("effective distance" which as calculated to 2 be 1.2 feet (FIG. 9D).

The present and succeeding paragraphs describe the functions 3 that a computer program using non-metric multidimensional scaling 4 (MDS) algorithms perform in the practice of process 10. The inputs 5 to an MDS algorithm program are preferably a single matrix 6 representing an initial experimental, two dimensional, Cartesian 7 plot of a configuration of building units located in the 8 quadrilateral global area selected to receive the building units of 9 a facility, and a set matrices representing degrees of association 10 among all possible pairs of these building units for various inter-11 building operational criteria associated with the function the 12 facility performs. (For examples of such operational criteria were 13 refer to the earlier overview description of process 10 which was 14 provided in connection with FIGS. 1A-1F). Briefly, the MDS 15 algorithm computer program performs a transformation of the initial 16 experimental configuration of the building units into an output 17 Cartesian plot of a configuration of the building units which is 18 nearly optimized in its composite performance of the various 19 inputted operational criteria. A figurative analogy (simplified 20 only one set of operational criteria data) would be the MDS 21 algorithm receiving as it inputs someone's best guess of a two-22 dimensional Cartesian plot of places in city which are frequent 23 pairs of termini of taxi trips, and a table of amounts of fare paid 24 for such trips (operational criteria data). Using the experimental 25 plot and the data, the MDS algorithms would convert even a grossly 26

1 distorted initial inaccurate plot into an output plot with accurate 2 relative geographic locations of the places.

As earlier mentioned it is preferred that the initial 3 experimental configuration of building units in the quadrilateral 4 designated facility site be produced based on an intuitive 5 "expert's" (as defined above) response to knowledge of the function 6 served by the facility. It is to be appreciated that optionally 7 the MDS algorithm could function with plural different experiment 8 plots being processed and/or with different set of operational 9 criterion data that are poor or even arbitrary. The data of the 10 experimental configuration and the sets operation criteria data 11 must be conventionally translated into compatible matrix sets. 12 Appropriate inter-building units operational criteria data is 13 identified from an analysis of the facility's function, and then 14 collected by such techniques as extraction from records of facility 15 operation, desk audits, time and motion type studies and time-lapse 16 photography. Each set of data is translated into matrix 17 expressions of degrees of association of the individual operational 18 criteria between all possible pairs of building units of the 19 The set of building units of configuration are configuration. 20 preferably uniformly weighted, and represented in the plot as 21 symbolic points (zero dimensional abstract points). 22

The MDS software algorithms individually combine the experimental configuration matrix of data with each respective matrix representing a set of inter-building unit degrees of association of a operational criteria, producing a set of non-

metric, building-unit and operational criterion MDS matrices. The 1 MDS algorithm program then combines all the matrices of this set 2 into a composite-facility-operational-criteria-Cartesian-3 configuration-of-building units; preferably using a non-linear, 4 least-square stressing function. The algorithm in effect follows 5 a progression of matrices operations starting with an operation 6 upon adjacent matrices, an then operation on compatibility matrices 7 and finally an operation upon similarity or "distance" matrices. 8 What is happening is a measurement of a "badness of fit" interacts 9 with a measure of composite operational criteria, in a way that 10 causes a decrease of the stress function. "Distances" (as 11 mentioned earlier in an abstract senses) are taken to be the 12 measures or metrics of the operational criteria, in such a way that 13 the highest matrix score for any pair of objects represents a 14 strong requirement that they be placed close together, and the 15 lowest score indicates least importance in proximate locations. 16 The MDS algorithm software is applied first for building units in 17 the global area for grouped process steps 12a and 12b, and then for 18 workplace elements in one or more subareas within the building 19 units, groups of process steps, 14a and 14b. 20

Typically, after between 20 and 50 iterations, a suitable stopping criteria related to gradient of stress decrease and/or magnitude of the gradient is reached. At that time the then coordinates of the objects represent a near optimum spatial layout from the viewpoint of operational criteria, and a printed plot of the global area or subarea showing the machine recommended location

1 of each object is presented for consideration. The final solution 2 is one that has a minimum "stress value".

Any of various will known of-the-shelf computer programs for implementing MDS algorithms may be employed. The Matrix Laboratory (MATLAB) software product can be conventionally adapted for MDS algorithms. It is produced by Matworks, of Cambridge Massachusetts is one suitably set of programs. A well known existing package program is "KYST" which is described and discussed in the below identified papers of Kruskal.

Further details regarding the methodology of employing MDS 10 algorithm computer program see A.T. Siegel, J.J. Wolf and J. 11 Pilitis (1982), "A New Method for the Scientific Layout of 12 Workspace". Applied Ergonomics, 18(2) 87-90; Kruskal, Non-Metric 13 "Multidimensional Scaling; A Numerical Method", Psychometrica, Vol. 14 29, No 2, June 1994; and Kruskal "Multidimensional Scaling by 15 Optimizing Goodness of Fit to a Non-metric Hypothesis:, 16 Psychometrica, Vol 29, No. 1 (1964), all of which are by these 17 present references incorporated herein in their entirety. 18

This plot serves as preliminary solution layout so that the 19 layout planner may then, at the planner's discretion, make a series 20 of minor manual adjustments to the computer-generated solution. 21 Such adjustments may be desirable to compensate for special needs 22 such a repair access or for physical constraining features of the 23 area such as the location of posts, stairways, and doorways, in the 24 existing area. Accordingly, the multistage method of the step two 25 in the exemplary description provides for the exercise of 26
workspace-user judgments as well as consideration of work imposed
 conditions. Accordingly, this regard process 10 is computer aided,
 rather then computer generated. The computer offers an initial
 solution followed by a manual adjustment.

In summary, the MDS methodology reveals the underlying structures in data sets and then presents them in a graphic format, i.e., a geometric configuration/mapping, suitable for visual analysis and interpretation.

With this background, the sequence of steps in the practice of 9 process 10 will be described in connection with flow charts of 10 process steps in FIGS. 1A-1F. In steps 18, 20 and 22, FIG 1B, 11 selection of the quadrilateral global area, set of building units 12 generation of an experts normative configuration of the building 13 units in the global area, and the collection of operational 14 criteria data are performed, or obtained, by the layout analyst. 15 In step 24, a set of non-metric, multidimensional scaling (MDS), 16 building units and operational criterion matrices are generated 17 following the MDS algorithm methodology (as hereinbefore described 18 in detail), and in step 24 this set of matrices is combined into 4 19 composite-facility-operational-criterion-Cartesian-configuration-20 of-building units (again following MDS algorithm methodology 21 hereinbefore described in detail). In step 26, convex hulls are 22 circumscribed about both (i) the expert's normative experimental 23 configuration and (ii) the facility-operational-criteria-24 configuration-of-building units and each respective polygon areas, 25 Apoly, of the polygon within the respective convex hulls is 26

35

calculated using Equation 8a. In step 28 calculation of composite-1 facility-operational-criteria-MDS-matrix-population-density-of-2 building units, PDI_{CPOSMDS}, is performed using equation 9. This 3 index constitutes a measure of the crowdedness of the resultant 4 configuration of building units after MDS methodology repositioned 5 the locations of the building units to provide least stress in 6 enabling the facility to compositely performance the operational 7 criteria. Also as part of step 28, a corresponding normative-8 population-density-index PDI_{NORM} of the initial experts 9 configuration of the building units is calculated using equation 10 In step 30 $PDI_{CPOSMDS}$ is compared with PDI_{NORM} , providing the 11 8b. layout analyst with a qualitative indication of the impact upon 12 crowdedness of the readjustment of the experts initial 13 configuration by MDS methodology. The higher the value of PDI 14 index the more crowdedness exist in the configuration. The analyst 15 can thereby quantitatively consider whether the facility-16 operational-criteria-building units-configuration is acceptable 17 relative to the expert's normative experimental configuration by 18 observing the ratio of the values. For example, if PDI_{CPOSMDS} is 19 one-half of the PDI_{NORM} , the resulting ratio 2.0 would indicate the 20 MDS methodology not only reduced stress in the performance of the 21 operational criteria, but also significantly reduced crowdedness. 22 With these circumstances it is likely that MDS methodology induced 23 configuration would be acceptable to the layout analyst's client. 24 If desired, an arbitrary low threshold value of the ratio may be 25 used and the step automated. Facility operational-criteria-26

building units-configurations having unsatisfactory PDI_{CPOSMDS}
 indices result in discontinuing the MDS layout process, step 32.
 Where configuration is acceptable optional manual adjustments are
 made, as for example, to overcome physical constraints, step 34.

Passing to the tier of process 10 for the layout of workplace
elements personnel workstations and equipments) in one or more
quadrilateral subareas of a building units or units, process steps
18', 20', 22', 24', 26', 28', 30', 32' and 34', FIGS. 1D and 1E are
performed utilizing workspace operational criteria data instead of
facility operation criteria.

In the event the MDS methodology induced facility layout and 11 one or more workplace layout are confirmed by PDIpoly comparison, 12 the project effort may proceed to ultimate construction of the 13 building units in locations within the global area and construction 14 and fitting out of workplace elements with one or more subareas, 15 step 36, FIG 1F, in accordance with facility and workspace 16 operational-criteria-building units-and-workplace-elements 17 configuration solutions provided by process 10. 18

19 It will be appreciated by those skilled in the art that 20 various process step or combinations of process step in addition 21 to those identified with the MDS algorithm methodology can be 22 implemented by using a suitably programmed general purpose 23 computer.

Many modifications of the presently disclosed invention will become apparent to those of skill in the art without departing from the inventive concepts.

1	Navy Case No. 77581
2	
3	SITE AND WORKSPACES LAYOUT PROCESS EMPLOYING
4	MDS AND A PDI FORMULA IN WHICH DENSITY IS CALCULATED USING
5	MEASURED SPAN OF CIRCUMSCRIBING-CONVEX-HULLS
6	
7	ABSTRACT OF THE DISCLOSURE
8	A process is provided for producing layouts of building
9	units on a quadrilateral facility site, and layouts of personnel
10	workstations and items of equipment (collectively "workplace
11	elements") in quadrilateral subarea in the building units. There
12	are inter-building-unit, and inter-workplace-element, operational
13	criteria associated with the activity being performed in the
14	facility. The well known multi-dimensional scaling (MDS)
15	methodology is employed in optimizing building unit and workplace
16	element configurations to suit the operational criteria.
17	Measurement of population density index (PDI) is employed to
18	judge whether candidate configurations of building units and
19	workplace elements result in adverse crowdedness conditions. The
20	PDI employed for this purpose is novel. The novelty of the PDI
21	is its use of a "convex hull" (in the mathematical sense of the
22	term) circumscribed about the perimetrical objects in a
23	candidate configuration. The equation for the PDI then employ
24	the polygon interior of the convex hull in calculation of a
25	density related term of the equation.









FIG. 1D









FIG. 4 Y Ø (2,4) 4 ø (1,3) 3-Ø (3,2) 2-0(1,1) 1-Х 0 2 1 3 4





















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