

Collecting Data in Two-Dimensional Euclidean Metric Space

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Both cognitive anthropologists and social network analysts typically collect data by asking subjects to perform some more or less standardized task. Subjects are sometimes asked simply for their free recall of names of objects from some domain like fruits, musical instruments, or personal friends (Rosch et al. 1976; Romney et al. 1979; Freeman et al. 1981; Gatewood 1983, 1984; Freeman et al. 1987; Freeman and Romney 1987).

Sometimes subjects are presented with a list and asked to choose the objects from the list that fall into the domain under investigation (Freeman and Freeman 1980; Krackhardt 1984; Fagan 1992). Or they are presented with the names of objects—each on a separate card—and asked to sort the cards according to some criterion (Miller 1969; Boster 1986, 1987; Freeman et al. 1988). Or they may be required to perform a triads task (Romney and D'Andrade 1964; Burton 1972; Weller and Romney 1988; Romney et al. 1993), or to rank them (Miller 1974; Killworth and Bernard 1976; Bernard and Killworth 1979), or perhaps to assign each of them a numerical position on a scale (Andrews and Withey 1974; Thompson 1975; Bernard and Killworth 1979).

Typically, data that have been collected by any of these means are used to construct some measure of the perceived distance (or closeness) between pairs of objects. When n objects are studied, the result is a square, $n - n$, object-by-object matrix in which each cell contains an index of the distance between the object designated by the row and that designated by the column.

Matrices of this sort are usually fairly complex. Often it is difficult or impossible to grasp their overall patterning (Friendly 1979). But if their structure is simplified—if the objects and their arrangement are projected onto an ordinary two- or three-dimensional Euclidean space—we can actually see their patterning. The aim, of course, is to retain as much information about the original distances between objects as possible while we are transforming them into this simpler form.

The use of various standard methods to perform such simplification is common. These methods include MDS, principal components analysis, and correspondence analysis (Weller and Romney 1990). They differ in detail, but they all transform distance data into Euclidean form.

Using any of these transformations rests on the faith that (1) the subjects' cognitive organization of the domain in question actually is both Euclidean and of small dimension, or (2) if it is not, the small-dimensional Euclidean representation imposes no serious distortion on that cognitive organization (whatever form it really takes).¹ But, if either of these beliefs is true, we could streamline the process that leads us to collect our data in one form, scale them, and then transform them to an entirely different form. Perhaps we could begin by collecting our data directly in Euclidean form.

At least for the two-dimensional case, collecting data in Euclidean form turns out to be both simple and straightforward. Following a suggestion by David Morgan,² I programmed a micro computer to collect Euclidean data directly from subjects. The program is called MAP.³ It performs three tasks: (1) it instructs the subject, (2) it presents stimuli, and (3) it records detailed data about subjects' responses.

The program requires three files: (1) a labeling file to guarantee that each subject has a unique identification, (2) a file containing instructions for subjects and (3) a file containing the stimulus list. Each time the program is run, it updates the labeling file and it produces a uniquely labeled data file for a subject.

There is, of course, no restriction on the domains from which we might draw objects for study by this method. It could be used to see how subjects arrange objects of any sort, and it would be natural to use it to study subjects' perceptions of proximities or similarities when the objects were human beings. Thus, it could be used to study perceived group structure or the perceived structure of roles, traits, or social positions.

Suppose, for example, we want to collect data about subjects' images of a domain made up of the names of 21 animals (Henley 1969). We would write a file containing a number (say 1), another containing instructions, and a third containing the list of animal names. Then the program would be run and a subject asked to sit down at the computer and to strike the key marked <Enter>. The computer might display an instruction screen like that shown in Figure 1.

Then, when the subject strikes a key, a work area is framed on the left of the screen and the stimulus list is randomly ordered and presented on the right. The cursor is displayed prominently in the center of the work area, and reminders of the commands are listed at the bottom. A working screen is shown in Figure 2.

At this point, only the four arrow keys, the G and D keys and the <ESC> key are enabled. Pushing any other key produces a BEEP sound and no other action. Data recording begins when the subject moves the cursor so that it covers the letter associated with an animal's name and hits the G key. At that point, the label for the animal dims and a copy of that animal's letter becomes locked to the cursor. That letter is moved with each use of an arrow key. It will follow the cursor movement until it is dropped by pressing the D key.

As the task progresses, data are recorded in a file. Records include both (1) the name of each object picked up along with the time at which the pick up occurred and (2) the location at which that object was put down as well as the time of the put down.

The most important feature of the MAP method is that it requires no data transformation. Each subject produces a set of distances between each pair of objects. And these distances are explicitly both two dimensional and Euclidean. Moreover, this new method provides an additional benefit by recording more and more detailed data than other procedures. It provides a record of the order in which objects are chosen by the subject, it records information on the number of times each object is relocated, and it retains precise information on the subject's timing in picking each new object up as well as in placing that object down in its chosen location.

Thus, as compared with standard techniques of data collection, the MAP procedure produces a more direct measurement of Euclidean distance and at the same time yields much richer and more detailed data about the subjects' responses. Rather than

imposing a Euclidean space on data that are possibly non-Euclidean, MAP automatically yields a true Euclidean metric. At the same time, it generates enough detailed data to permit us to examine many of the micro aspects of the process in which objects are located in Euclidean space.

Notes

¹ This assumption is open to question. Indeed, to the degree that Eleanor Rosch (1978) and others are correct in characterizing cognitive structures as hierarchically arranged sets of categories, their structure approaches that of a mathematical form called ultrametric. But ultrametric structures—even simple ones—cannot usually be embedded in Euclidean spaces of small dimension.

² Personal communication, 1985.

³ The MAP program was written in Q-Basic and compiled to run under DOS. On request, I will send a copy of both the source deck and the executable code. Write me at the School of Social Sciences, University of California, Irvine, CA 92717. Phone: 714-856-6698. Best of all, Internet: LIN@ARIS.SS.UCI.EDU.

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Figure 1: An Instruction Screen.

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This experiment is concerned with how people think about
animals. You will be presented with a screen. The names of
several animals will be listed on the right. Each is
identified by a letter, like "F". On the left, is a blank
screen. You use the arrow keys to move the cursor around.
First move to an animal's letter, and hit G to grab it. Then
move it to some spot on the screen and hit D to drop it.
Locate all the animals anywhere you want on the screen--the
rule is simple. If a bunch of animals seem to be similar to
each other--if they resemble one another in an important way--
put them close together on the screen. Animals who seem to
have less in common should be located farther apart on the
the screen. Once you drop an animal's letter, you can always
go back and grab it again using G. Move it to where you want
then use D to drop it. When you are satisfied that all the
animals are arranged in a pattern that reflects the patterning
of similarity, just hit the key marked "ESC" to finish.

Hit any key to continue...

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Figure 2: The Working Screen.

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A BEAVER
B HORSE
C TIGER
D ANTELOPE
E SHEEP
F CHIMPANZEE
G CAT
H CAMEL
I GIRAFFE
J CHIPMUNK
K COW
L GOAT
M DEER
N ELEPHANT
O RAT
P GORILLA
Q LION
R DOG
S MONKEY
T RABBIT
U ZEBRA

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MOVE = ↑ ↓ → ←   GRAB = G   DROP = D   FINISHED = <ESC>

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