DUAL TECHNIQUES FOR THE DETERMINATION OF SPATIAL CLUSTERING OF MORTALITY IN MONTREAL, QUEBEC - 1972

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I. Introduction

The determination of spatial clustering in health related studies is crucial for optimizing the design and implementation of prophylactic measures. In addition, by noting where high morbidity and mortality rates predominate, new hypotheses may be generated which would help unravel some of the more perplexing contemporary etiological puzzles.

As an outgrowth of a larger study (Nisen, 1976), standardized mortality rates for neoplasms of all types, excluding lung and breast (ICD, eighth revision, nos. 140-163, 170-173, 180-199, 200-239), for the males in the city of Montreal, 1972, two techniques to determine spatial clustering were employed. The first involved the use of Moran's "I" statistic to statistically determine spatial autocorrelation (Moran, 1950). The second approach utilized computer assisted techniques to produce maps of the entire city for visual analysis, and of selected sets of contiguous tracts as a means of identifying homogeneous areas of high mortality rates.

II. Data Collection and Standardization

The examination of mortality due to neoplasms was

facilitated by the existence of a mortality data base maintained by the Government of Quebec and the City of Montreal. Death certificates are coded in Montreal then geocoded to the census tract of last residence and stored on the data processing system of the Quebec provincial data processing system under the auspices of the Centre Informatique.

Although the mortality information was computerized to facilitate storage and retrieval, considerable time delay was experienced between the request for information and actual examination. This delay was due to problems associated with making operational a new data base management system.

Upon receipt of the mortality tabulations, manual extraction and compilation by census tract produced a file of raw mortality data. This data was divided into deaths by sex, and then age standardized using Liddell's method (Liddell, 1965).

III. <u>Statistical Determination</u> of <u>Positive</u> <u>Spatial</u> <u>Autocorrelation</u>

Spatial autocorrelation may be visualized in the following manner. Given a finite area which is divided into n areal units, if for every pair of areal units i and j, samples yield variates, x_i and x_i , which are uncorrelated, then there is no spatial autocorrelation in the areal system on x. If the samples are not all pairwise uncorrelated, then spatial autocorrelation is said to exist (Cliff and Ord, 1973, 2).

A set of tests have been developed (Moran, 1950, Geary, 1954, Dacey, 1965, and Cliff and Ord, 1969) to detect spatial autocorrelation for nominal, ordinal, and interval data. There are three main test statistics available for use with interval scale data: Geary's "c" statistic, Moran's "I" statistic and the Cliff and Ord statistic.

Both Moran and Geary utilized a binary weighting mechanism in their spatial autocorrelation formulae. Cliff and Ord further extended upon the results of Moran and Geary to incorporate generalized weights (Cliff and Ord, 1975, 151). Since the data are recorded by discrete areal units (census tracts), no attempt is made to extend beyond binary weighting. Furthermore, the "I" statistic has been judged superior to the "c" statistic; hence, our analysis will be limited to this index (Cliff and Ord, 1975). The last remaining question to be addressed is whether the results obtained from either test for spatial autocorrelation are statistically significant.

Cliff and Ord (1975) state that for the "I" statistic it is necessary to consider two distinct forms for the null hypothesis. These are:

- Normality assumption: that the values for the set of variates x are the results of n independent drawings from a normal population (or identical populations); and
- 2. Randomization assumption: whatever the underlying distribution of the random variables, we consider the observed value of I or c within the set of all n! possible values which I or c could take on if the $[x_i]$ were randomly permuted around the areal system (Cliff and Ord, 1975, 152-153).

By reviewing the nature and form of the standardized mortality data as distributions in histogram form, it is apparent that all spatial autocorrelation analysis must be performed assuming the second form (randomization assumption) of the null hypothesis.

The calculated value for Moran's "I" statistic for neoplasms of all types in males, excluding lung and breast, was 0.19277 which is not significant at the 0.05 level.

Even if there was a significant amount of spatial autocorrelation, this type of analysis is limited by its inability to specify the location of the spatial clustering. This deficiency is remedied by use of a more traditional medical geographic approach, visual map analysis, and with recent developments in computer cartography, this technique is more powerful than ever before.

IV. <u>Visual Map Analysis</u>: <u>Problems and a Potential</u> Solution

Visual map analysis has been used extensively in

medical geographic research. The work of Snow (1885) during the cholera outbreak of 1849 in London serves as an example of the types of beneficial findings that can result from an "enlightened" inspection of a mortality or morbidity map. However, in the present study the data was collected on an areal basis, thus necessitating a choropleth display as opposed to individual occurrences which can be investigated via point pattern analysis.

Although associations in certain instances can be ascertained from choropleth maps through visual analysis, there still exist perceptual problems from the interpreter's perspective (Monmonier, 1977). The most important facet of the problem of perceiving areal associations on a mapped surface is that of map complexity. The problem of complexity has been reviewed and investigated by many individuals (Julesz, 1975; Monmonier, 1974; Muller, 1975, Olson, 1975; and Bouchard, 1976).

To evaluate the map complexity problem, two types of cartographic outputs were utilized. A standard twodimensional shaded map was produced by the POLYPS program developed at Harvard. As evidence in Figure 1, this map is confusing and difficult to interpret. Figure 2 represents the same data, but it makes the perception of high mortality rates much easier to discern. This output was produced by the PRISM program also developed at Harvard. An additional useful characteristic of the PRISM program is the ability to individually represent all the census tracts.

Due to its inherent capabilities to communicate the standardized mortality rates more effectively, the PRISM program was selected for use in an attempt to depict sets of contiguous tracts with relatively high mortality rates and which were homogeneous with respect to these rates, i.e., low variance.

The first step in determining the clusters of high mortality rates was to rank all contiguous sets of tracts by their mean value. Table 1 is an abbreviated representation of this table which illustrates not only the mean values, but the variance as well. The asterisks indicate the contiguous sets of tracts that were selected for mapping. Figure 3. is a composite cartographic representation of the six sets of tracts where high mortality rates predominate. The small cluster of tracts along Montreal's waterfront is interesting and is deserving of a closer examination.

V. Summary and Conclusions

The increase in chronic disease mortality is one of the most serious health problems faced by the developed nations. This problem becomes acute in an intra-urban context. As a means of elucidating the spatial structure of the Standard Mortality Ratios for neoplasms of all types, excluding lung and breast, (a representative chronic disease) for the male population of the City of Montreal, 1972, two types of analysis were employed. The use of a statistical determination of spatial autocorrelation, specifically Moran's "I" statistic, indicated that no significant amount of spatial clustering of high rates was present in the City of Montreal on a census tract level. Even if a significant amount of spatial autocorrelation was present, this type of analysis has a major deficiency: it cannot specify the location of the high mortality rate clusters. To attempt to solve this problem, a visual analysis of two types of cartographic output were undertaken. To minimize the problems associated with interpreting a two-dimensional shaded map, a program that produces pseudo three-dimensional representations was used. It was found that the combination of interactive computer cartographic programs coupled with rudimentary manipulative and query capabilities result in a tool that can be of tremendous value to the medical geographer and epidemiologist alike as a means of discovering health problem areas for further examination and analysis.

References

Bouchard, Diane. "Spatial Autocorrelation and the Analysis of Health Data." <u>Analysis of Non-Vectored</u> <u>Diseases in Montreal</u>, Report 3. Montreal: Department of Geography, McGill university, 1976.

Cliff, A.D. and Ord, J.K. "The Problem of Spatial Autocorrelation." London Papers in Regional Science, Vol. 1. Edited by A.J. Scott. London: Pion Ltd., 1969. Cliff, A.D. and Ord, J.K. <u>Spatial Autocorrelation</u>. London: Pion Limited, 1973.

Cliff, A.D., Haggett, P., Ord, J.K., Bassett, K. and Davies, R. <u>Elements</u> of <u>Spatial</u> <u>Structure</u>. London: Cambridge University Press, 1975.

Dacey, M.F. "A Review on Measures of Contiguity for Two and K-Colour Maps." <u>Spatial Diffusion Study</u>, Technical Report No. 2. Evanston, Ill." Department of Geography, Northwestern University, 1965.

Geary, R.C. "The Contiguity Ratio and Statistical Mapping." <u>The Incorporated Statistician</u>, Vol. 5 (1954), 115-45.

Julesz, Bela. "Experiments in the Visual Perception of Texture." <u>Scientific American</u>. Vol 232 (April, 1975), 34-43.

Liddell, F.D.K. "The Measurement of Occupational Mortality." <u>British</u> Journal of Industrial Medicine, Vol. 17 (1965), 228-233.

Monmonier, Mark S. "Measures of Pattern Complexity for Choroplethic Maps." <u>The American Cartographer</u>, Vol. 1 (1974), 159-169.

Monmonier, Mark S. "Maps, Distortion, & Meaning." <u>Commission on College Geography Resource Paper</u>, No. 75-4. Washington, D.C.: The Association of American Geographers, 1977.

Moran, P.A. "Notes on a Continuous Stochastic Phenomenon." <u>Biometrika</u>, Vol. 37 (1950), 17-23.

Muller, J.C. "Definition, Measurement, and Comparison of Map Attributes in Choroplethic Mapping. "Proceedings of the Association of American Geographers, Vol. 7 (1975), 160-164.

Nisen, W.G. "An Examination of Intra-Urban Mortality Patterns in Montreal: A Spatial Analytical Approach." Unpublished thesis, Department of Geography, McGill University, 1976.

Olsen, Judy M. "Autocorrelation and Visual Map Complexity." Annals of the Association of American Geographers, Vol. 65 (1975), 189-204.

TABLE 1

Subjective Identification of Sets of Contiguous Tracts with High Mean Mortality Values and Low Variance

TRACT SET	VARIANCE	MEAN	TRACT SET	VARIANCE	MEAN
ID 649 638 1387 1311 1311 1311 1314 1645 2033 2112 2131 2132 2112 2132 2112 2132 2112 2138 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 2112 307 307 307 307 307 307 307 307 307 307	$\begin{array}{c} 11844.513\\ 2268.8555\\ 1C40.7368\\ 2436.255\\ 457.618\\ 2045.321\\ 9333.372\\ 971.437\\ 1384.416\\ 971.437\\ 1384.416\\ 873.921\\ 1301.911\\ 227.196\\ 5853.689\\ 1301.911\\ 781.921\\ 5873.921\\ 5873.921\\ 5333.4825\\ 4770.489\\ 5333.4825\\ 4770.480\\ 213.5866\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.686\\ 441.707\\ 213.586\\ 441.685\\ 5333.859\\ 701.707\\ \end{array}$	109.225 58.5162 59.32162 59.32162 59.32162 49.4182 49.4182 49.4182 49.4182 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.4255 49.4182 49.	ID 205 10 5 142 210 210 210 217 219 215 215 215 215 215 215 215 215 215 123 215 215 215 215 215 215 215 215 215 215	155.548 303.731 680.956 627.882 179.876 4155.876 4155.876 46.484 9395.614 46.484 9395.621 46.484 9395.623 122.435 91.129 91.129 91.129 5152.435 255.605 3138.376 46.898 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 46.805 3138.376 41.801	0037, 0037
Fig. 1		👞 EXCL	S ASSOCIATED UDING BREAST MONTREAL	WITH NEOPLAS	MS

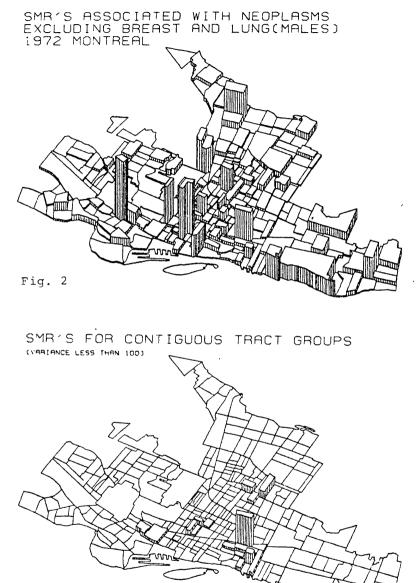


Fig. 3