

EPIDEMIOLOGICAL GIS: Understanding Emerging Critical Issues

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Abstract This paper examines the roles of GIS in epidemiological research, and identifies needs for spatial epidemiology. It argues that the objective function of spatial epidemiology is to identify patterns and correlates it to geospecific variation in health risks incidences. GIS supports disease mapping, location analysis, the characterization of populations, spatial statistics and simulation modeling. Despite these flurries of applications, they are yet to sufficiently identify disease causes and correlates. This is partly because of the failure of GIS to provide tools appropriate for epidemiological research and analysis. Hence, the need to review our understanding towards the relationship between causes of a disease and its spatio-temporal information. Thus the need for a design and implementation of a space-time information systems and process-based disease models for epidemiology

Keywords GIS, GPS, DSS, Epigis, Epidemiology, Spatial Statistics, Spatial Models, Geostatistics, Spatio-Temporal

1. Introduction

According to [1], the issue of public health concerns the knowledge, monitoring, assessment and maintenance of the health of human populations. This involves monitoring disease outcomes, through the identification, designing and implementing interventions to ameliorate health risk factors. GIS is making substantive contributions in public health at a hands-on level. Logical questions such as 'Where is the best place for the new clinic?', 'Why are disease mortality high?', 'Why are there a disease cluster?', 'Where are the source point of this disease?', are all geospecific questions that can be addressed using spatial statistical techniques and GIS [2].

As epidemiology is the study of health and disease in human populations, and, because populations are inextricably bound to 'place', it seems reasonable to expect GIS to advance Epidemiology as Geosciences [3]. Despite the many practical applications in public health, some authors believe this expectation hasn't been fully met as shown in [4]. Sources for this failure are several and include the lack of trained epidemiologists to 'think spatially'; a lack of epidemiological studies that invoke GIS applications/concepts as a Decision Support System (DSS) [5]; and the failure of commercial on-the-shelf (COTS) GIS to provide appropriate tools for spatial epidemiology [6]. As a tool for advancing epidemiology as science, GIS thus

appears to have severe limitations. To understand these limitations, this paper considers GIS both as technology and as a scientific tool.

2. GIS and Spatial Analysis Concepts in Epidemiology

The issues of Intelligent Maps (*iMaps*)

Visualization is one of the first steps in exploratory spatial data analysis [7]. GIS creates intelligent maps (*iMaps*) that are digital, interactive, queriable, and can be subjected to Boolean operations for morbidity and mortality patterns in relation to population density, putative exposures and geo-based features. It therefore constitutes a link between statistical plots, quantitative appraisals and geographic maps [5], [8], [9] forming the geostatistical (choropleth/density) maps. Geostatistics is a methodology for incorporating the spatial and temporal coordinates of observations in data processing. However, the key feature of epidemiological data is their location and display in a space-time continuum which should be incorporated in any descriptive or quantitative analysis of the data [10]. Recently, GIS has emerged as an innovative and important component of many projects in public health and epidemiology. GIS can be used to map and analyze the geographical distribution of population at risk, health outcomes and risk factors; to explore associations between risk factors and health status; and to plan public health services [11]. As it is traditional, the adaptation of mapping and cartographic techniques in search for answers begins with the formulation of spatio-epidemiologic hypothesis.

Location-Allocation Problem

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Published online at <http://journal.sapub.org/ajgis>

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The location-allocation problem addresses the need to place health facilities and services in 'the best' geographic locations as to optimize its objective functions (minimizing cost and maximizing accessibility) through the identification of the best possible location. It provides solution for initial facility problem and/or the incremental facility problem. GIS and its add-ons support the determination of optimal facility location or where health facilities are best located in relation to its users; through the creation of attribute layers for an overlay and split operations, the estimation of ambulance travel times to and from hospitals and ambulance routing path, the identification of hospital catchment areas, and the determination of the hospitals niche population. In public health, GIS has proven contributions in vector control (e.g. where to place the intervention?) to the impact (why the intervention?) to reduce malaria, and other vector-borne diseases.

Characterization of populations.

To outsiders, epidemiologists seem fixated on 'numerators' and 'denominators', and with good reason: In a disease rate, the numerator is the number of the disease cases while the denominator is the population at risk of contracting the disease. In spatial epidemiology, the characterization of geographic populations is an important task directly supported by GIS. This is because GIS is interested in giving answer to where, what is where and why. This entails exposure assessments involving the identification of high-risk populations, the interplay of factors among the attributes and why risks incidences are high.

Spatial statistics.

Spatial statistics quantify geographic variation in geographic variables. They can identify violations of assumptions of independence required by many epidemiological statistics; and measure how populations, their characteristics, covariates and risk factors vary in geographic space [12], [13], and [5]. Methodological researches on disease cluster techniques for spatial epidemiology are indispensable for advancing spatial epidemiological science.

Spatial models.

Recent developments of spatial models in public health according to [14] include Bayesian smoothing of disease rates, geostatistical models, and Haggastrand Mean Information Field (MIF) for the Monte Carlo simulation estimation techniques. These models address important issues of the stabilization of disease rates, interpolation, map presentation, information flow, contagion rates and the estimation of variables at 'not gauged' locations.

While these models are formidable, they support descriptive epidemiology and do not directly increase our ability to link human health outcomes to putative exposures and environmental risk factors. At best they can identify a local excess of disease that maybe suggestive using geographic proximity as a surrogate for a possible cause. This seems poor payoff for what is often a costly and time consuming geographic investigation.

Spatial Data

Spatial data is information about the location and shapes of geographic features, in the form of either vector or raster. These features are created as graphic maps in three geometric forms as points, lines, and polygons (areas). Associated with these graphic data are data records that provide the identifying and descriptive data attributes. An outcome from this enterprise reinforces the *where* answer to epidemiological data.

3. Needs of Spatial Epidemiology

This lack of scientific payoff is rooted in technological determinism whereby GIS tools determine the hypotheses that can be addressed. The questions one can ask of the data depend implicitly on the data models, spatial data structures, and Boolean queries employed by the GIS. The adaptation of health data unto geographic virtual models explains the process-pattern link governing spatial epidemiological research.

Process-pattern link.

We need an increased understanding of the relationships between disease processes and the resulting disease patterns. Spatial disease patterns are the outcomes of space-time processes, and several alternative processes may give rise to similar spatial patterns [15] making it very difficult to interpret spatial disease patterns in other than broadly descriptive terms. The trajectory of this link requires a clear understanding of the models, the data and the data's dynamic process while the estimation of such a model's parameters requires space-time information systems.

Models of process.

Models of process are expressed in terms of the physical and biological mechanisms underlying the system under study. This contrasts with models of data that are expressed in terms of the data's statistical properties. At present most if not all spatial models in public health are models of data, rather than models of process. While useful for prediction, the parameters of models of data have limited epidemiological utility. In contrast, the parameters of process models are by definition directly interpretable in terms of underlying diseases. Compartmental analysis according to [16] is a powerful approach for constructing models of space-time processes that holds great promise in epidemiology.

Space-time information systems (STIS)

Estimation of the parameters of such process models will require information systems capable of dealing with GIS in epidemiology using spatial and temporal referencing. While several authors recognize the need for 'time GIS' in order to represent space-time data in general (e.g. [17], [18]) and human disease data in particular [19], its essential role in the estimation of parameters of space-time process models has received little recognition. Because STIS will make possible estimation of the parameters of space-time models, they are expected to be indispensable for advancing spatial epidemiologic science.

According to [20] and [5], in designing statistics for spatial epidemiology certain data attributes must be considered. These are:

- The spatial coordinates of the data noting their proximity measurements (determining their nearest neighbor index, and accessibility index),
- An understanding of their point – pattern display (determining their Variance-Mean-Ratio), and
- Identifying certain data nominal attributes such as relevant descriptors of place, including its spatial location, exposure and covariates.

It must be noted that for diseases with multiple casual factors an integrating approach is essential. GIS techniques can assess a multiplicity of external factors, such as physical variables, technological hazards and social/demographic factors [21]. More so, it is also desirable to integrate biological indicators into exposure assessment, with respect to the timing of exposures, induction and latency.

4. Discussion and Conclusions

The above presentation has been to stimulate thoughts on observed problems in spatial epidemiological modeling and on the inventing technologies required to solve them. In that, a problem solution stimulates the development of critical technology; and secondly, a technological advancement that results to problems that were not the target of the innovation, [4]. The GIS adaptation in spatial epidemiology fell within the second group. This problem is more exposed in the non-availability of reliable exposure data. In most cases, the use of calculated or measured data is, at best, only an approximation of the actual exposure. Data on past exposure and location has to be detailed and accurate, concerning mobility and other outcomes that are dependent on time variation. The results from analyses on regional data will of course depend on the spatial scale used. This sophistication display possibilities that can bring a false sense of accuracy. Noting that results from analyses can never be more accurate than the quality of the data captured [22], [23] and [24].

Within the last 3 years, Nigeria has experienced some catastrophic events that clearly provide evidence of the importance of a spatially-induced health information system. The September, 1998 C130 plane crash; May 2002 EAS airline crash; October, 2005 Bellview plane crash; December, 2005 Sosoliso crash; 2006 ADC airline and June 2012 DANA crash and the associated health hazard has exposed the inappropriate management of post-plane crash era. Thus provides an enormous evidence of the need for a spatially-induced digital platform as a decision support in the post emergency management efforts. Lack of a functional EpiGIS platform in Nigeria has militated against efficient and sustainable post disaster management techniques. With constant outbreak of meningitis in the Northern part of Nigeria and severe water pollution in the south causing disastrous outbreak of typhoid presents strategic failures in

epidemiological post-crisis management.

It is the view of this paper that there is an important need to carefully design appropriate GIS technology for epidemiological research and management; and adaptations that will lead to the evolution of Spatial Decision Support System (SDSS) in public health which will eventually drive the database engine for the health system. This system will provide centralized access to national health and disease database and a repository of GIS epidemiological tools. These will result to:

- An increased understanding of the relationships between health status, disease processes and patterns;
- Evolution of process-based disease and health models;
- Understanding space-time information systems for epidemiology; and
- Development of statistical models for spatio-temporal epidemiological indices.

This paper should be seen as a contribution to the flurry of papers on spatial epidemiology, and concludes that sustainable research on epidemiology must recognize the spatio-temporal information on diseases. The model arising from this situation must be able to calibrate space-time data on health outcomes into epidemiologically meaningful measures. The geo-specificity of the health database constitutes the foundation of Epidemiological Geographic Information Systems (EpiGIS).

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