Effects of study area size on geographic characterizations of health events: Prostate cancer incidence in Southern New England, USA, 1994–1998
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Abstract

Background: We consider how representations of geographic variation in prostate cancer incidence across Southern New England, USA may be affected by selection of study area and/or properties of the statistical analysis.

Method: A spatial scan statistic was used to monitor geographic variation among 35,167 incident prostate cancer cases diagnosed in Massachusetts, Connecticut and Rhode Island from 1994 to 1998, in relation to the 1990 populations of men 20+ years of age living in that region. Results from the combined-states analysis were compared to those from single-states. Impact of scanning procedures set to examine up to 50% or no more than 10% of at-risk populations also was evaluated.

Results: With scanning set to 50%, 5 locations in the combined-states analysis were identified with markedly distinct incidence rates. Fewer than expected cases were estimated for nearly all Connecticut, Rhode Island and West Central Massachusetts, whereas census tracts on and around Cape Cod, and areas of Southwestern Connecticut and adjacent to greater Boston were estimated to have yielded more than expected incidence. Results of single-state analyses exhibited several discrepancies from the combined-states analysis. More conservative scanning found many more locations with varying incidence, but discrepancies between the combined- and single-state analysis were fewer.

Conclusion: It is important to acknowledge the conditional nature of spatial analyses and carefully consider whether a true cluster of events is identified or artifact stemming from selection of study area size and/or scanning properties.

Background

Spatial analyses of health events can provide helpful information that informs our understanding of the determinants and control of diseases within populations. Geographic studies have discerned the non-random nature of health hazards [1], at-risk populations [2], disease occurrence [3,4], progression [5], screening [6], treatment [7,8], and end results [9,10].
By their nature, spatial analyses of health events are zero sum problems wherein overall rates (cases per 100,000 persons), proportions (deaths among diagnosed cases) or case counts (birds carrying West Nile virus) applicable to a defined study area (nation, region, state) are disaggregated to measure relative differences across smaller analytic units (counties, census tracts, exact coordinates). Changing event counts and/or underlying populations by expanding or contracting study area size and/or modifying properties of the spatial statistic can affect estimates of spatial variation therein. Similarly, the sensitivity of the spatial statistic to identify event clusters specific to any given location is known to vary by changing the proximity of that location to a study area’s boundaries [11]. As such, findings for any given spatial analysis are best considered as conditional and modifiable as a consequence of study area size and/or properties of statistical procedures.

Nationwide efforts to foster regional health information networks/organizations (RHINs/RHIOs) that span traditional geo-political boundaries demand greater understanding of how aggregating health and population data may affect analysis and interpretation of disease patterns. To date, there are few opportunities to evaluate consequences of study area size selection. Sharing health related data across states or regions is uncommon, if not restricted, in the interest of protecting individual privacy and confidentiality of information. Moreover, inconsistencies across states regarding their use of geocoding references, statistical and mapping software further limit possibilities to pool/combine data for multi-state studies. Hence, researchers have limited guidance as to how (and to what extent) findings may change by modification of the study area.

We previously described provisional qualities of spatial analyses related to geocoding [12], choice of statistical methods [13,14]; surveillance period [15], covariate adjustment [16], and spatial units chosen for analysis [17]. Here, we consider the conditional effect of study area size on the representation of health events. The situation is illustrated using data on prostate cancer incidence in Southern New England, USA. The data reported here are unusual regarding both their level of detail (census tract for cases and population) and coverage (3 contiguous jurisdictions – states). Discerning the potentials and pitfalls of spatial analytic methods will facilitate the dissemination of methods among researchers and practitioners focused on health surveillance and health system evaluation.

### Results
Among the 35,167 geocoded records across Southern New England, an average annual age-adjusted incidence rate of 194.6 incident cases per 100,000 at-risk men was observed (Table 1). Within Connecticut, the statewide average annual age-adjusted incidence rate of 201.7 prostate cancers per 100,000 at-risk men (5.2% above the composite rate among men outside Connecticut) was noted. Comparable incidence rates were 182.8 for Rhode Island (93% of the rate outside Rhode Island) and 192.6 for Massachusetts (97% of the rate outside Massachusetts).

#### Combined-states study area
Our first comparison of combined and single-state analyses employs the spatial scan statistic with properties set to evaluate rates within locations containing as much as 50% of a study area’s at-risk population (1.8 million men in the combined-states study area, 580 K within Connecticut, 174 K within Rhode Island and 1.05 million within Massachusetts). Rate variation was observed across Southern New England regardless of the size of the study area (Figure 1 and Table 2).

When data regarding Connecticut, Rhode Island and Massachusetts were examined simultaneously, 5 locations were identified as having incidence rates likely to differ significantly (p < 0.05) from elsewhere across the combined-states study area. For most of Rhode Island, Connecticut and West Central Massachusetts (Area 1) the age-adjusted average annual incidence rate was estimated to be 91% of expectation relative to the rate among men living elsewhere within the Southern New England region. For a considerably more circumscribed location north of Greater Boston (Area 4), the incidence rate was estimated to be only 49% of the rate observed elsewhere around the combined-states study area.

### Table 1: Prostate cancer incidence (1994–1998) and population (1990) for Southern New England and single states, USA.

|                          | Reported Cases | Coded Cases (%) | Population at-risk | Avg. Annual incidence | Census tracts |
|--------------------------|----------------|-----------------|--------------------|------------------------|---------------|
| CT-MA-RI                 | 38,956         | 35,167 (90)     | 3,606,842          | 194.6                  | 2,400         |
| Massachusetts (MA)       | 22,525         | 20,243 (90)     | 2,097,746          | 192.6                  | 1,331         |
| Connecticut (CT)         | 12,501         | 11,735 (94)     | 1,160,925          | 201.7                  | 834           |
| Rhode Island (RI)        | 3,930          | 3,189 (81)      | 348,171            | 182.8                  | 235           |

1 1990 U.S. Census of males, 20+ years old. 2 Per 100,000 at-risk persons.
By comparison, census tracts on and around Cape Cod (Area 2) were estimated to have had an incidence rate 27% greater than other locations within the combined-states study area. Tracts in Southwestern Connecticut (Area 3) and those to the immediate southwest of Greater Boston (Area 5) were observed to have rates estimated to be 33% and 26% higher, respectively, than rates found elsewhere within the study area.

**Single-state study areas**

While much consistency between the combined and single-state analyses was evident, several important differences were evident. When considering Massachusetts by itself, for example, 6 locations were identified with rates that differed markedly from elsewhere around the State. Consistent with earlier results, census tracts in the western half of the State (Area 6) yielded an age-adjusted average annual incidence rate 81% of that observed outside that location. Additional locations with markedly low incidence again were found along the Massachusetts borders with Rhode Island (Area 9) and New Hampshire (Area 10).

As in the combined-states analysis, the most likely location of elevated cancer incidence specific to Massachusetts was found among census tracts on and around Cape Cod (Area 7) where the incidence rate was estimated to be 1.26-times greater than expectation. Census tracts around greater Boston (Area 8) revealed a significantly high rate of disease (1.15-times greater than expectation) that spatially encompassed considerably more area, cases and persons at-risk than previously detected within Area 5.

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*Figure 1*

Places of significant variation in invasive prostate cancer incidence rates (using the default scanning window of up to 50%) for Southern New England and its constituent states, USA, 1994–1998.
Equally noteworthy, the state-specific analysis yielded a location north east of the city (Area 11), with a significantly elevated incidence rate (RR = 1.20) that previously was not identified by the combined-states analysis. Whether this location merits specific attention for disease control efforts depends on which study area is selected for analysis.

Rates specific to Connecticut includes 4 locations that differed from the statewide pattern and finding based on the combined-states study area. Census tracts of Southwestern Connecticut (Area 12) were found, as in Area 3, to have greater than expected incidence (RR = 1.28 in relation, this time, to the rate elsewhere around Connecticut). Whereas the combined-states analysis found the bulk of census tracts across the state to have a lower than expected incidence rate (Area 1), the single-state analysis identified much of the state as having rates that were not remarkably different from the statewide experience. Here, only among at-risk men living within census tracts along the State's northern border (Area 17) was it estimated that prostate cancer occurred at a rate below that (87%) of what occurred elsewhere around the Rhode Island. In the single-state analysis, men living in South Central census tracts situated around Narragansett Bay (Area 16) were found to have experienced an average annual age-adjusted incidence rate 1.39-times greater than expectation.

Use of a more restrictive scanning setting

The problem of edge effects and selection of study area can be overcome, to a degree, by modifying properties of the spatial statistic. By rejecting the default SaTScan settings and limiting the spatial scan procedure to include a smaller portion of a study area's at-risk population it is possible to reduce the likelihood that identified clusters will reach or span single-state boundaries. To illustrate, we compared results of combined and single-state analyses when the size of the spatial scan was limited to include no more than 10% of a study area's at-risk population (361 K for the combined-states analysis, 116 K within Connecticut, 35 K within Rhode Island and 210 K within Massachusetts).

Predictably, limit on scanning properties yielded more (but smaller) places of likely rate variation. Using the

| Study Area | Location* | Population at-risk | Size (sq km) | Cases | O/E** | p-value |
|------------|------------|--------------------|-------------|-------|-------|---------|
| CT, MA & RI | 1 | 1,779,267 | 28,174.1 | 16,055 | 0.91 | 0.0001 |
| | 2 | 200,603 | 13,809.5 | 2,729 | 1.27 | 0.0001 |
| | 3 | 115,636 | 973.1 | 1,626 | 1.33 | 0.0001 |
| | 4 | 31,229 | 36.1 | 127 | 0.49 | 0.0001 |
| | 5 | 63,951 | 382.9 | 830 | 1.26 | 0.0001 |
| Massachusetts (MA) | 6 | 408,741 | 37,868.3 | 3,465 | 0.81 | 0.0001 |
| | 7 | 200,603 | 13,809.5 | 2,729 | 1.26 | 0.0001 |
| | 8 | 518,190 | 2,102.5 | 5,076 | 1.15 | 0.0001 |
| | 9 | 29,381 | 497.2 | 114 | 0.45 | 0.0001 |
| | 10 | 31,229 | 36.1 | 127 | 0.48 | 0.0001 |
| | 11 | 177,096 | 865.7 | 2,194 | 1.20 | 0.0001 |
| Connecticut (CT) | 12 | 140,890 | 1,313.8 | 1,897 | 1.28 | 0.0001 |
| | 13 | 353,654 | 2,249.7 | 3,144 | 0.86 | 0.0001 |
| | 14 | 64,714 | 3,261.4 | 433 | 0.69 | 0.0001 |
| | 15 | 116,216 | 713.5 | 1,293 | 1.22 | 0.0001 |
| Rhode Island (RI) | 16 | 66,477 | 1,794.5 | 721 | 1.39 | 0.0001 |
| | 17 | 168,340 | 1,699.7 | 1,366 | 0.87 | 0.0001 |

* Illustrated in Figure 1. ** Risk of prostate cancer for residents of a given location, relative to the risk elsewhere around the designated study area.
default setting produced 5 significant locations for the combined-states analysis, along with 6, 4 and 2 significant locations for the respective single-state studies, whereas the more restrictive scanning setting yielded 18, 13, 5 and 3 significant locations, respectively (See Figure 2 and Table 3). High rates based on the combined-states study are, for example, now found on and around Cape Cod (Area 19), communities proximate to Boston (Areas 22, 25, 29 and 35), Greater Hartford (Area 28) and Southwest Connecticut (Area 21). Lower than expected incidence characterized census tracts of Western and Central Massachusetts (Areas 20 and 30), various portions of Connecticut (Areas 26, 32 and 33), and most of Rhode Island.

Patterns observed in the combined-states analysis generally hold for analyses specific to Massachusetts and Connecticut, but as before, the story regarding Rhode Island changes more notable depending on the study area examined. The combine-state analysis yielded evidence of a localized area along the eastern shore of Narragansett Bay with greater than expected incidence, while the northern portion of the state revealed a lower than expected rate of disease. Analysis specific to Rhode Island, however, identified a larger area presumed to have elevated disease rates, while the remainder of the state (with exception of 2 small pockets of low incidence) exhibited rates consistent with the statewide pattern.

**Discussion**

This paper examined prostate cancer incidence in Southern New England, 1994–1998, in order to describe whether and how selection of study areas and/or properties of the statistical method could affect estimates of geographic variation among health events. We found combined- and single-state analyses to share much in...
common, but several discrepancies were noted between approaches. Our analysis also discerned that the extent of discrepancies between combined- and single-state analyses could be reduced by modifying properties of the statistical analysis; limiting the capacity to scan at-risk population for potential disease clusters, reduced the likelihood that identified clusters would span area boundaries.

In essence, 'artifact' resulting from study area size and selection of scanning properties is inevitable in spatial analysis of health events. Understanding the origins of such 'error' is an important step in effectively utilizing available technologies for better disease control. Data analysts must balance a desire for specificity of location of possible event clusters (using a restrictive scanning window) with practical considerations of needing to draw valid generalizations about patterns across large populations.

Table 3: Geographic variation of prostate cancer incidence using of a restrictive (10%) scanning window according to selected study areas across Southern New England, USA, 1994–1998.

| Study Area       | Location* | Population at-risk | Size (sq km) | Cases | O/E** | p_value |
|------------------|-----------|--------------------|--------------|-------|-------|---------|
| CT, MA & RI      | 18        | 340,925            | 3,306.1      | 2763  | 0.79  | 0.0001  |
|                  | 19        | 200,603            | 13,809.5     | 2729  | 1.27  | 0.0001  |
|                  | 20        | 239,448            | 17,572.6     | 1917  | 0.77  | 0.0001  |
|                  | 21        | 115,636            | 973.1        | 1626  | 1.33  | 0.0001  |
|                  | 22        | 376,304            | 1,935.3      | 3925  | 1.16  | 0.0001  |
|                  | 23        | 5,741              | 5.3          | 0     | 0.00  | 0.0001  |
|                  | 24        | 31,229             | 36.1         | 127   | 0.49  | 0.0001  |
|                  | 25        | 177,096            | 865.7        | 2194  | 1.21  | 0.0001  |
|                  | 26        | 117,550            | 1,179.9      | 966   | 0.77  | 0.0001  |
|                  | 27        | 6,027              | 0.4          | 0     | 0.00  | 0.0001  |
|                  | 28        | 116,216            | 713.5        | 1293  | 1.24  | 0.0001  |
|                  | 29        | 20,338             | 12.8         | 220   | 1.66  | 0.0001  |
|                  | 30        | 91,701             | 1,626.0      | 774   | 0.82  | 0.0003  |
|                  | 31        | 17,970             | 86.9         | 211   | 1.50  | 0.0010  |
|                  | 32        | 28,139             | 42.5         | 228   | 0.71  | 0.0016  |
|                  | 33        | 32,013             | 779.3        | 225   | 0.73  | 0.0118  |
|                  | 34        | 19,704             | 28.8         | 159   | 0.69  | 0.0220  |
|                  | 35        | 3,526              | 1.3          | 42    | 2.33  | 0.0479  |
| Massachusetts (MA) | 36    | 164,636            | 13,224.3     | 1325  | 0.73  | 0.0001  |
|                  | 37        | 185,321            | 13,167.3     | 2546  | 1.27  | 0.0001  |
|                  | 38        | 29,381             | 497.2        | 114   | 0.45  | 0.0001  |
|                  | 39        | 5,741              | 5.3          | 0     | 0.00  | 0.0001  |
|                  | 40        | 31,229             | 36.1         | 127   | 0.48  | 0.0001  |
|                  | 41        | 177,096            | 865.7        | 2194  | 1.20  | 0.0001  |
|                  | 42        | 6,027              | 0.4          | 0     | 0.00  | 0.0001  |
|                  | 43        | 204,367            | 1,542.7      | 2174  | 1.19  | 0.0001  |
|                  | 44        | 144,083            | 3,295.9      | 1193  | 0.82  | 0.0001  |
|                  | 45        | 39,483             | 169.3        | 546   | 1.37  | 0.0001  |
|                  | 46        | 20,338             | 12.8         | 220   | 1.64  | 0.0001  |
|                  | 47        | 14,521             | 7.9          | 126   | 1.70  | 0.0010  |
|                  | 48        | 19,704             | 28.8         | 159   | 0.69  | 0.0098  |
| Connecticut (CT) | 49        | 105,999            | 920.8        | 1487  | 1.32  | 0.0001  |
|                  | 50        | 105,495            | 1,488.9      | 860   | 0.75  | 0.0001  |
|                  | 51        | 64,714             | 3,261.4      | 433   | 0.69  | 0.0001  |
|                  | 52        | 116,216            | 713.5        | 1293  | 1.22  | 0.0001  |
|                  | 53        | 28,139             | 42.5         | 228   | 0.69  | 0.0002  |
| Rhode Island (RI)| 54        | 36,996             | 912.2        | 424   | 1.51  | 0.0001  |
|                  | 55        | 31,645             | 31.8         | 215   | 0.71  | 0.0002  |
|                  | 56        | 28,593             | 28.3         | 198   | 0.75  | 0.0245  |

* Illustrated in Figure 2. ** Risk of prostate cancer for residents of a given location, relative to the risk elsewhere around the designated study area.
tions or study areas (using a less restrictive scanning window). Studies suspecting focused clusters and those involving limited geographic area may be suitable for more restrictive scanning windows, whereas exploratory analyses and those involving large geographies (regions, nations) may find such restriction impractical (given the likelihood of identifying a large number of clusters) or ill-advised (given the greater potential for Type II error). The volume and severity of edge effects produced necessarily will vary by such decisions.

It could be that when the reference rate for pooled data exceeded the rate of an individual state, the higher expected rate for the expanded study area reduced the likelihood that particular places would exhibit rates that were significantly higher than that new expectation, whereas the likelihood observing places where rates were significantly below that expectation would have been somewhat greater. In most but not all findings reported in Tables 2 and 3, the changes in estimated ratios of observed-to-expected incidence for specific locales were consistent with the modification of the baseline rates from the combined to the single-state study areas. Most illustrative of were changes related to Rhode Island data.

Here, we considered geographic variation in cancer incidence across three jurisdictions (states) that are distinguishable regarding their respective social, political, economic, health care and environmental systems and for which data furnished from three independent registries that may have differed regarding definitions, techniques and standards for reporting cancer incidence. Hopefully, the analysis, unusual for its detail (census tract data) and coverage (3 states), will foster greater appreciation of the opportunities and challenges of pooling health data from contiguous jurisdictions.

Conclusion

Fully distinguishing 'real' variation due to the geographic distribution of risk, rather than artifact attributable to study area, statistical procedures and/or data systems may not be possible. How particular findings might differ by adding/deleting adjacent areal units to a study area is not typically considered by investigators. Consequently, spatial analysis results may be rightfully considered conditional upon the particular geography selected for study. Unlike epidemiology studies of disease etiology or clinical effect where sampling assures cases are representative of an underlying population, geographic studies of health events do not similarly sample locations within a 'population' of possible places for study, but rather, rely on contiguous areas often aggregated by administrative/political reasons. Short of analyzing entire geographies, there are no a priori ways to distinguish the appropriate size or location of study area. Decisions typically rest on suspicion/

For the future, states and similar jurisdictions must pursue strategies that maximize potential for data to be pooled and analyzed across conventional geopolitical boundaries. Investment in geocoding, reference street files, data systems and GIS software should commit to principles of data sharing at the same time that procedures are implemented to maintain privacy of personal and group information.

Methods

Southern New England states of Connecticut, Massachusetts and Rhode Island consist of 17,644 square miles spatially organized within 3 States, 27 counties, 559 towns and 2,400 census tracts. It is home to approximately 3.6 million men 20 years of age and older. The geography of cancer incidence during this period was examined in relation to the populations-at risk within census tracts as enumerated by the 1990 U.S. Decennial Census of the Population, broken down according to ten-year age categories (i.e., 20–29 years, 30–39, 40–49, 50–59, 60–69, 70–79, 80+) [18]. Between 1994 and 1998, a total of 38,956 incident invasive prostate cancers (ICD-9-CM code # 185) were recorded by statewide tumor registries in Massachusetts, Connecticut and Rhode Island. For 35,167 records (90%), the census tract of residence at the time of diagnosis was known and successfully assigned geographic coordinates for analysis; 3,789 records lacked sufficient information to assign a census tract and therefore were excluded from further analysis. The proportion of records assigned census tract locations was substantially greater for Connecticut (94%) and Massachusetts (90%) than Rhode Island (81%). Excluded records typically contained no, incomplete or ambiguous street addresses or addresses that cited P.O. Boxes in place of street addresses. Reason for differences across states was not readily discerned. Previous work suggests that failure to geocode cancer events somewhat under-represented cases among urban dwellers [12].

Based on records available for study, variation in average annual age-adjusted incidence rates across census tracts was evaluated using a spatial scan statistic [19]. The procedure utilizes a large number of scanning circles (>100 K) of varying size and location to search for places (1 or more census tracts independent of conventional geo-political boundaries) where the number of observed cases deviated from a null hypothesis that incidence was proportional to population density (random). Age-adjusted case counts
and disease rates within and outside particular circles were determined by SatScan 5.1 software [20].

The spatial scan statistic is well suited for disease surveillance, as it does not require a priori assumptions about the number, place or size of locations or direction of effect that may be identified. It takes into account the uneven geographic distribution of the population at risk and, as required, accounts for any number of possible confounding variables. The significance of identified clusters is evaluated using Monte Carlo procedures, with adjusted p-values for multiple testing, to designate locations (clusters of census tracts) where incidence varied from the null. Results of the spatial scan statistic are considered to be conservative estimates of the likelihood of observing events within given locations, relative to places elsewhere around the study area [21].

Spatial analyses of prostate cancer incidence were completed for the 3 state study area of Southern New England, along with analyses specific to Connecticut, Rhode Island or Massachusetts. Findings of significantly high or low concentrations of incident cases are reported in Tables 2 and 3 and illustrated, using Maptitude® software [22], in Figures 1 and 2.

Competing interests
The authors have not received reimbursements, fees, funding or salary from an organization that may in any way gain or lose financially from the publication of this manuscript, nor do they hold stocks or shares in an organization or other competing financial interests that may in any way gain or lose financially from the publication of this manuscript.

Authors’ contributions
D Gregorio conceived of the study and supervised all aspects of its implementation. H Samociuk and L DeChello assisted with the study and completed the analyses. H Swede assisted in interpretation of findings and manuscript preparation.

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