

# Spatial Analysis of Health Risk

Gerard Rushton  
Department of Geography and College of Public Health  
The University of Iowa

Specialist Meeting on Spatial Analysis of Health Risk Perception  
Center for Spatially Integrated Social Science  
University of California, Santa Barbara  
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# Common Approaches

- A variety of methods for searching for statistically significant spatial clusters after adjusting rates for known risk factors.
- Data are generally “area-based” but sometimes are “point-based.”
- Methods are generally different for the two types of data.

# Common problems in these approaches: Measurement Errors

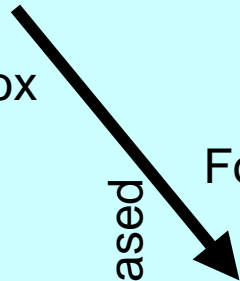
- exposures of individuals or groups
- health outcomes (population based registries?)
- locations of people or exposures—geocoding errors
- spatial relationships between people and exposures
- co-variates
- in protecting the privacy of individuals and institutions

Common problems: accounting for co-variates that condition effects of health risks

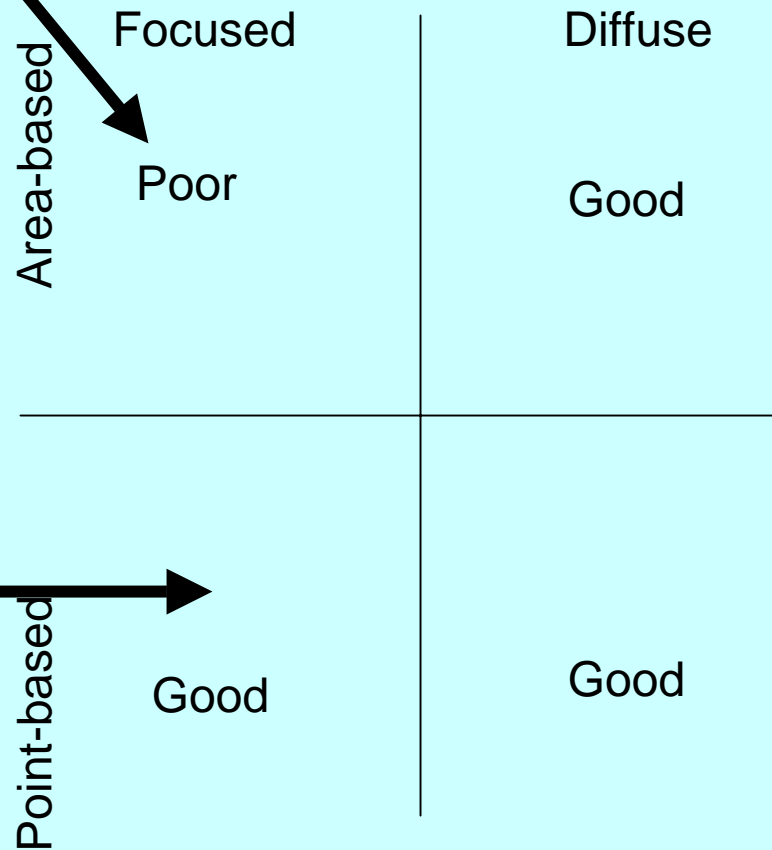
- adjusting for age and sex effects
- Adjusting for social or economic effects
- Adjusting for differential screening effects

# Ability to Measure Health Risk by Spatial Analysis

Most spatial analyses of health risk occur in this box



Spatial Patterns of Risks



Need to move the practice of spatial analyses of health risk to this box



# Ability to Measure Health Risk by Spatial Analysis

e.g. human environmental exposures in relation to superfund sites

## Spatial Patterns of Risks

Area-based

Focused	Diffuse
Poor	Good

e.g. Relationship between childhood asthma and ozone levels

But problems of ecological fallacy

e.g. human environmental exposures to CAFO emissions

(confined animal feedlots)

Health Data

Point-based

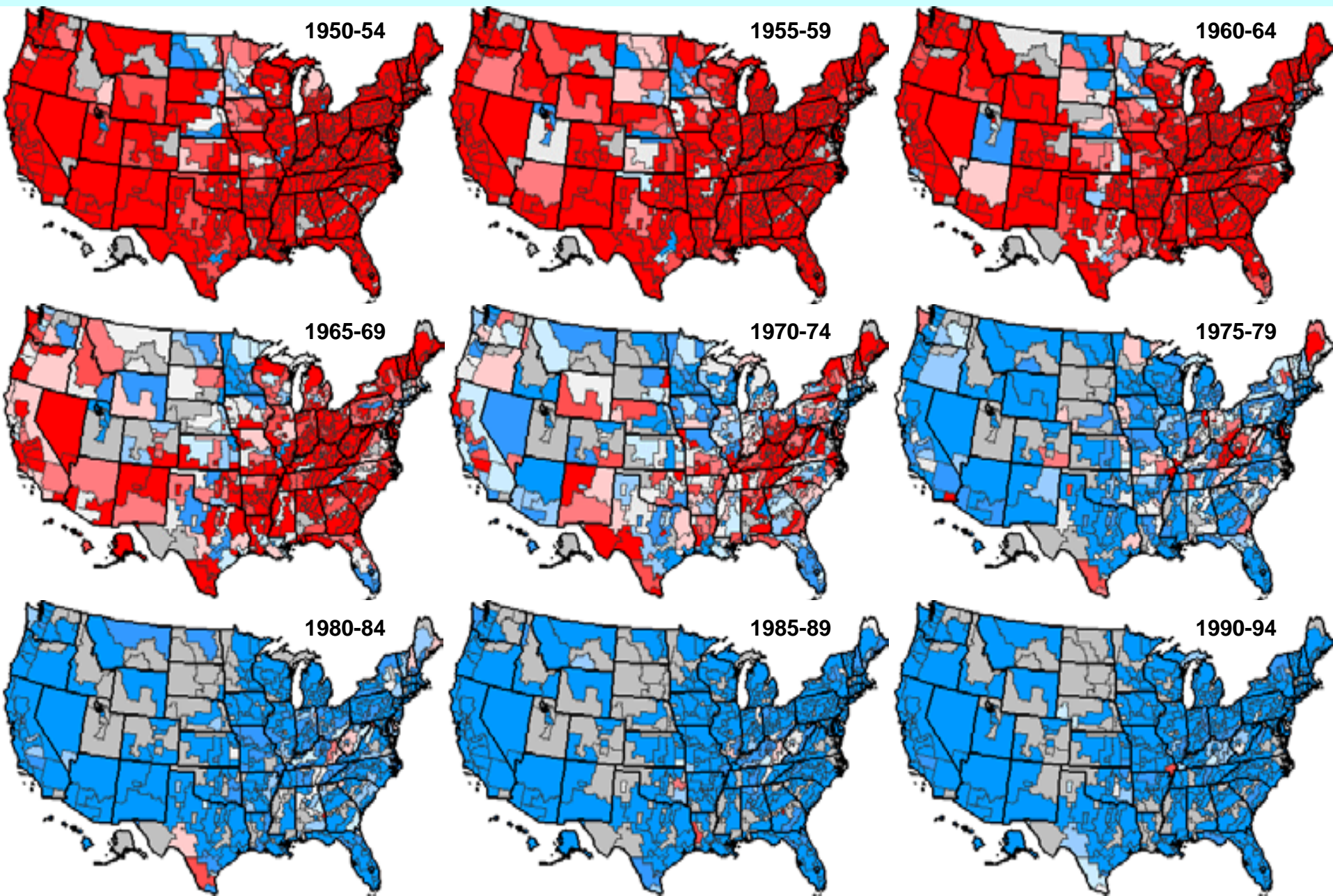
Good	Good
------	------

Come to this sector and avoid ecological fallacy problems

# Cervix uteri cancer, white females, 1950-94, state economic area

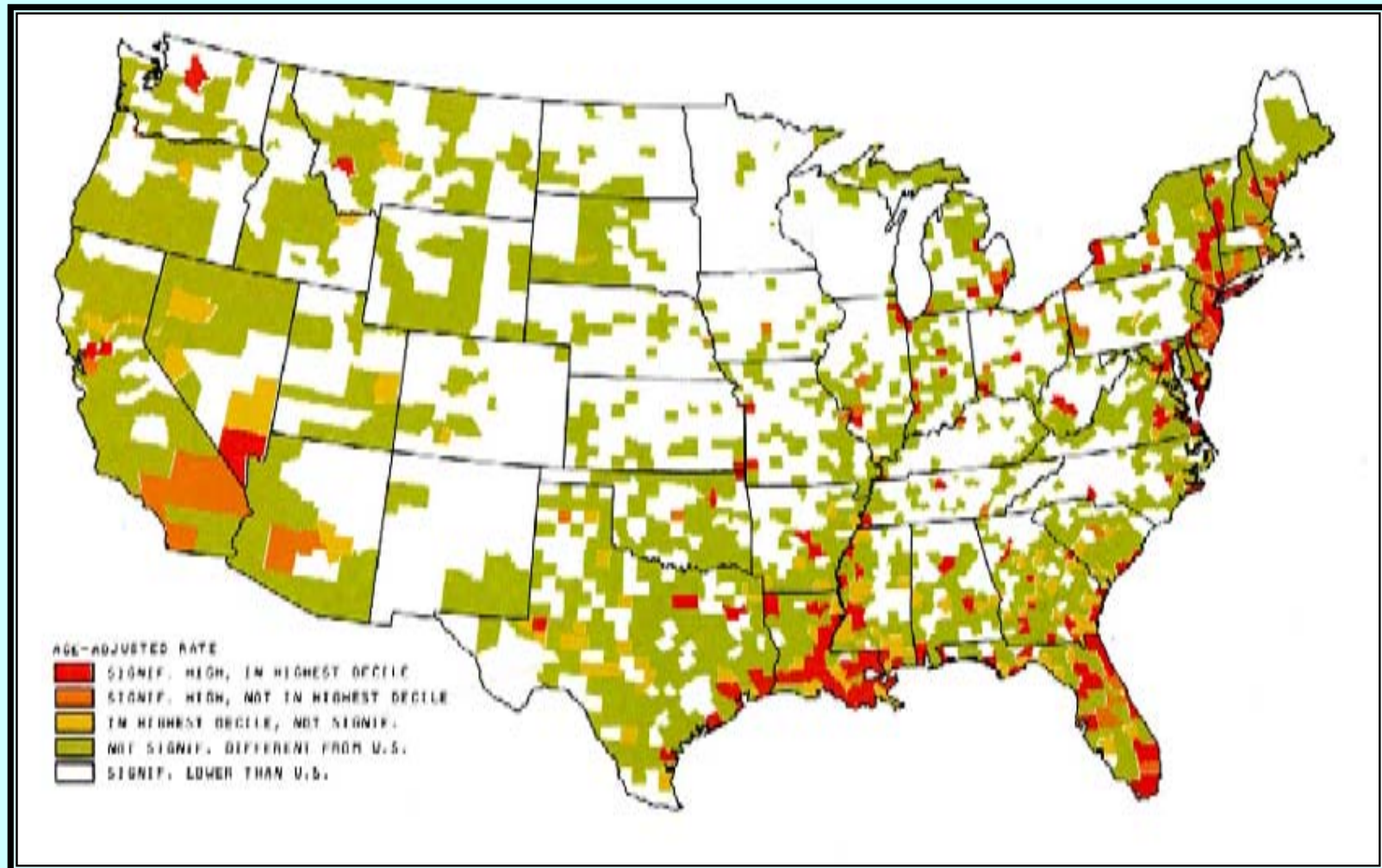
## Time trends

Source: [www.nci.nih.gov/atlasplus](http://www.nci.nih.gov/atlasplus)





# Lung Cancer Mortality Rates, 1950-69, White Males

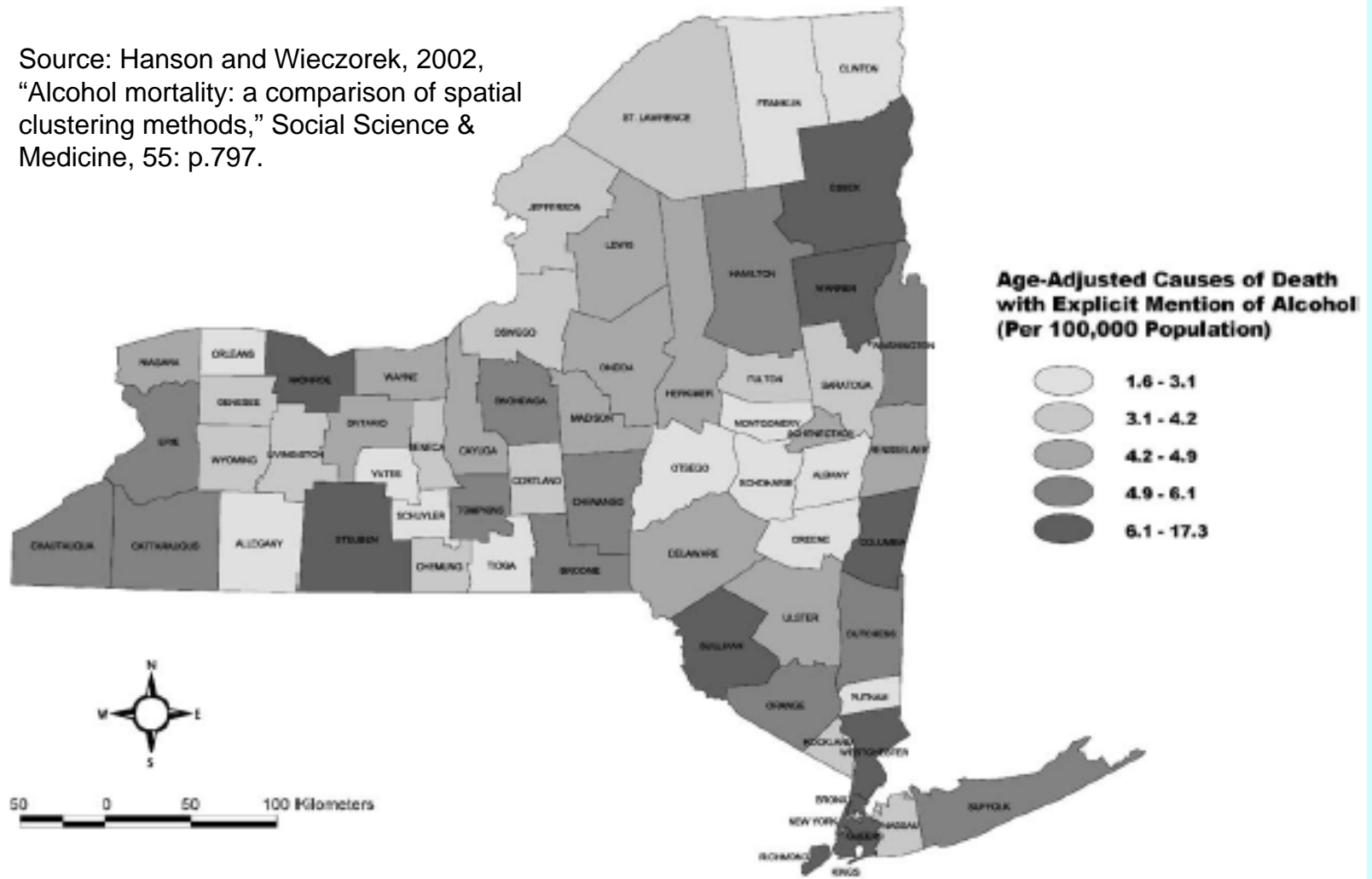


Source: Mason et al., *Atlas of Cancer Mortality for U.S. Counties, 1950-1969*. NIH, 1975.

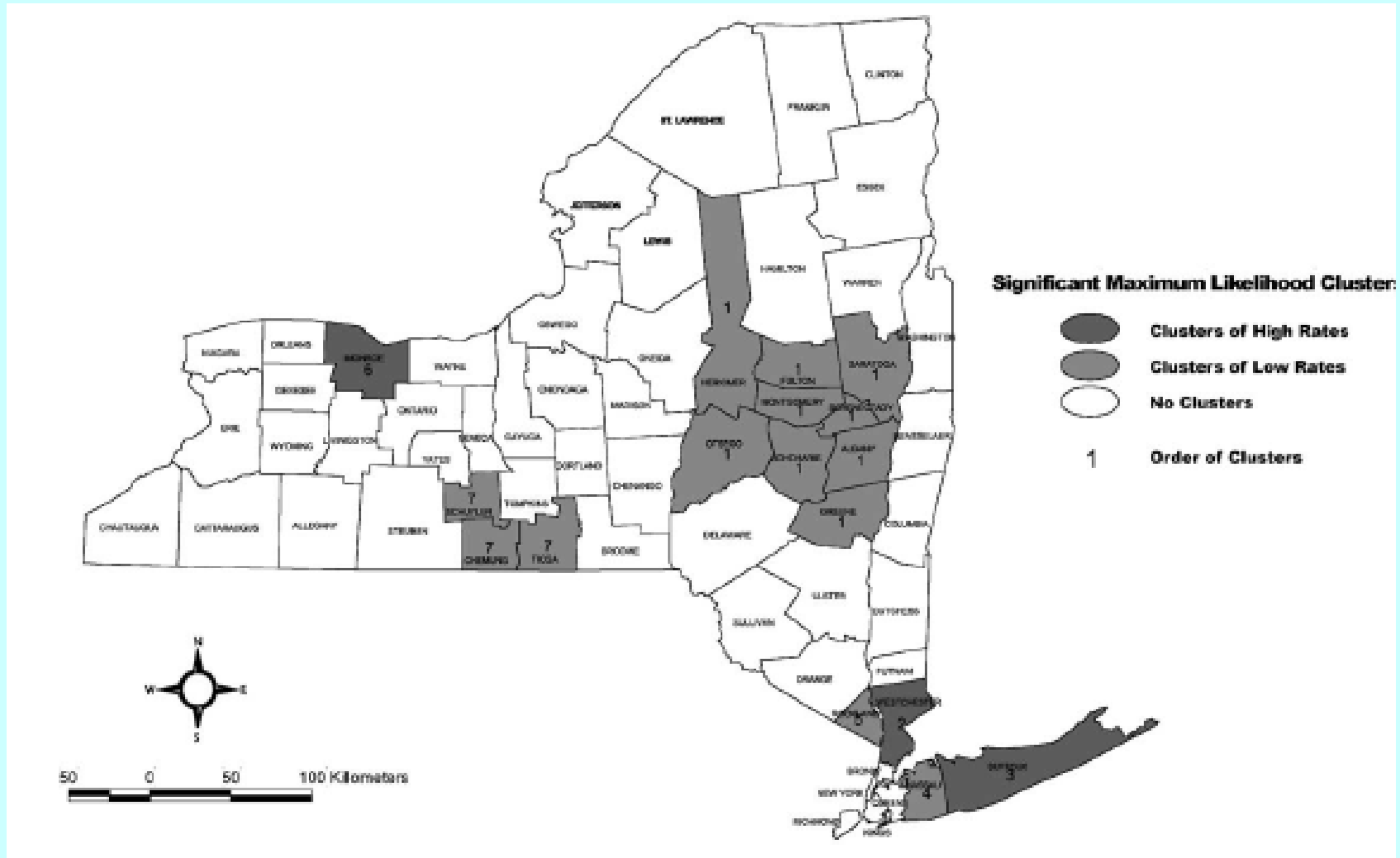


# Map of direct age-adjusted cause of death with explicit mention of alcohol (per 100,000 population).

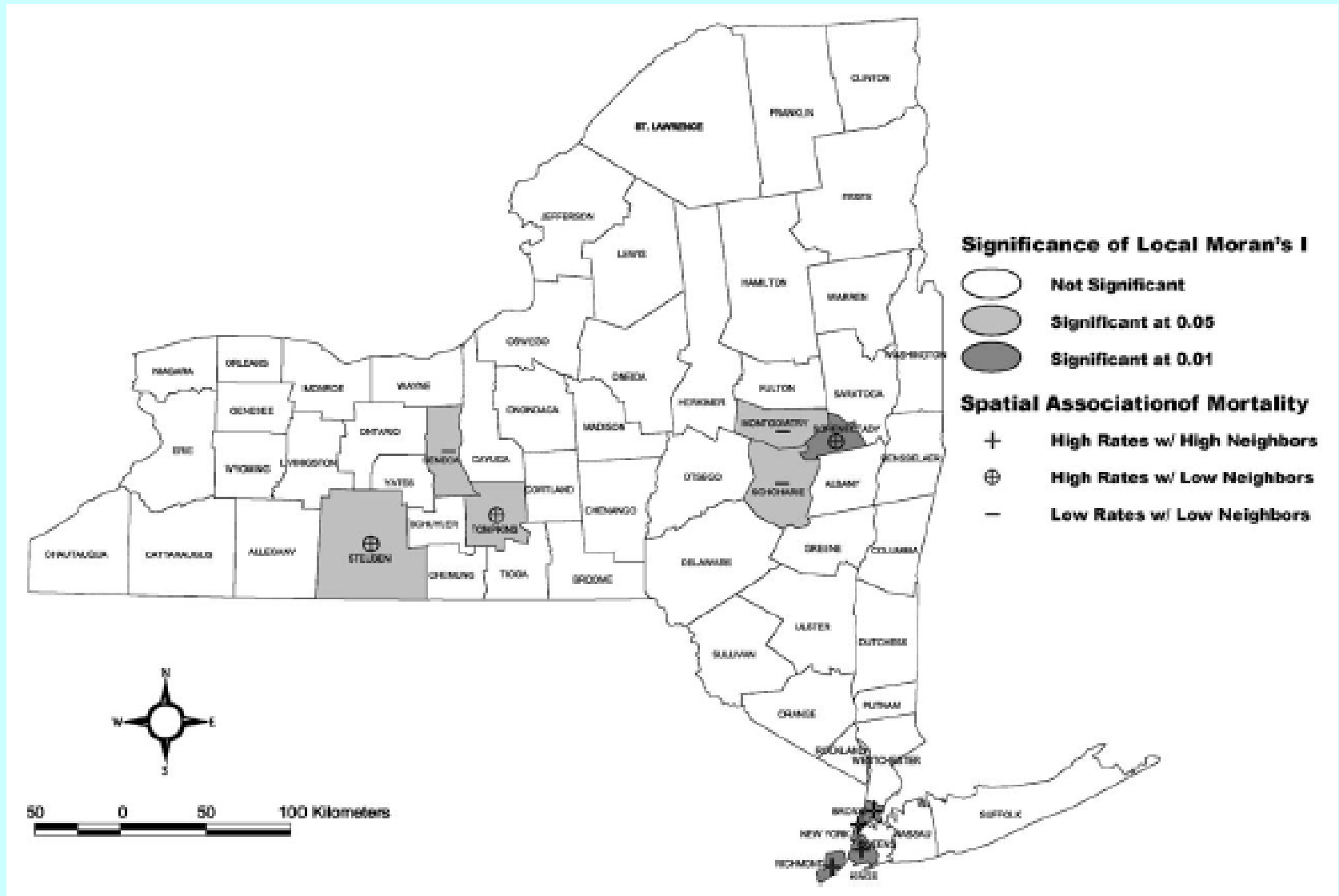
Source: Hanson and Wieczorek, 2002, "Alcohol mortality: a comparison of spatial clustering methods," *Social Science & Medicine*, 55: p.797.



SATScan (Kulldorff, 1997) Significant spatial scan clusters of alcohol-explicit mortality. Source: Hanson and Wieczorek, 2002, SS&M, p.799.



LISA (Anselin 1995, G.A.) Areas of significance of spatial association of alcohol-explicit mortality. Hanson and Wieczorek, 2002, SS&M, p.798



# Comparison of Three Methods of Identifying Areas of Risk

Source: B. Sue Bell. 2002. Spatial Analysis of Disease – Applications. Ch. 8 in C. Beam, ed. Biostatistical Applications in Cancer Research. Boston: Kluwer Academic Publishers, 2002, pp. 151-182.

“The research question was “Does the birth outcome of intrauterine growth retardation (IUGR) co-vary ecologically with characteristics at the neighborhood level?” IUGR (Usher and McLean 1969; Frisbie et al. 1997) is an outcome measure to categorize births that are small for gestational age. The project team analyzed geocoded birth records for Harris County (Houston), Texas for 1991.” p. 160.

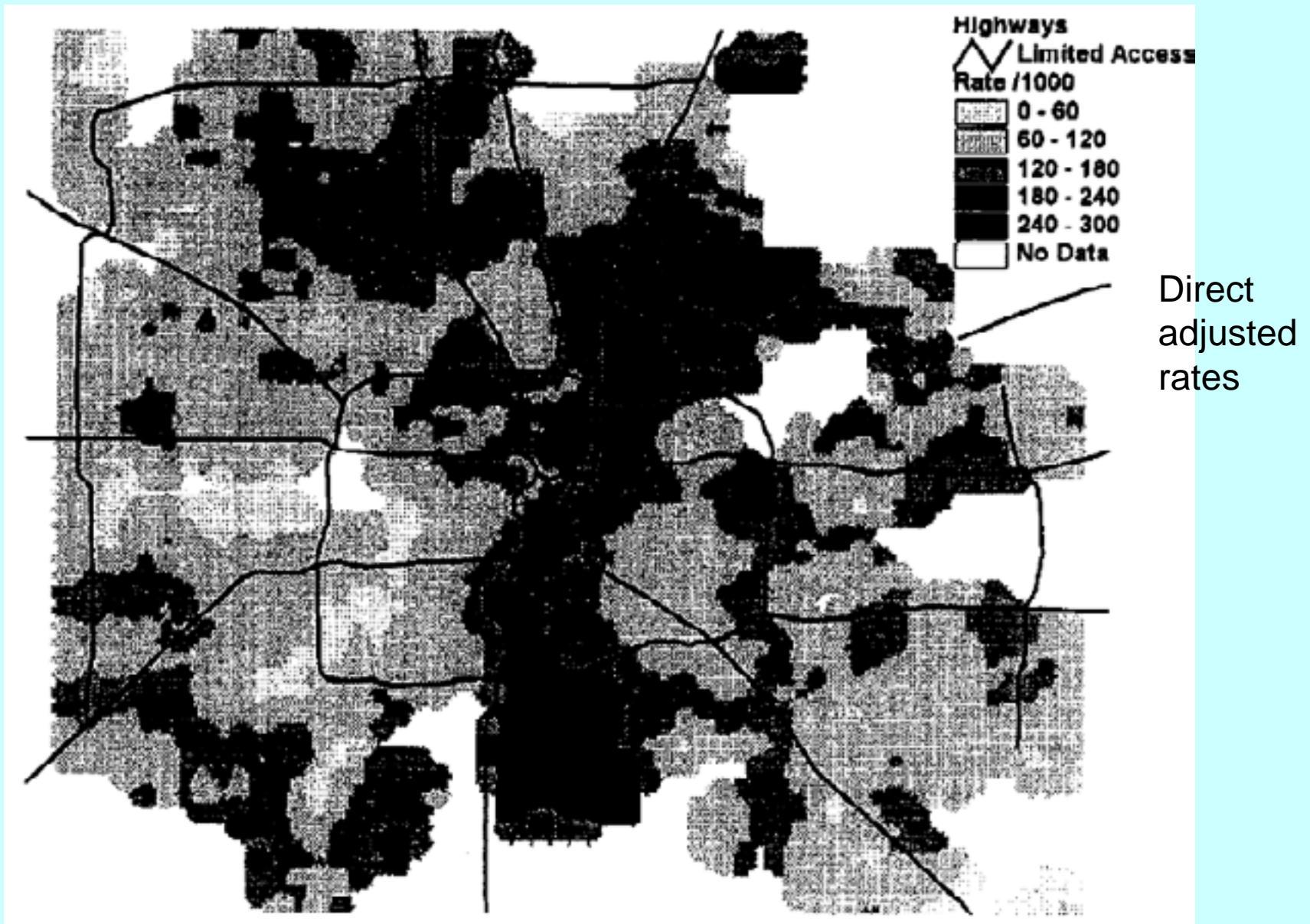
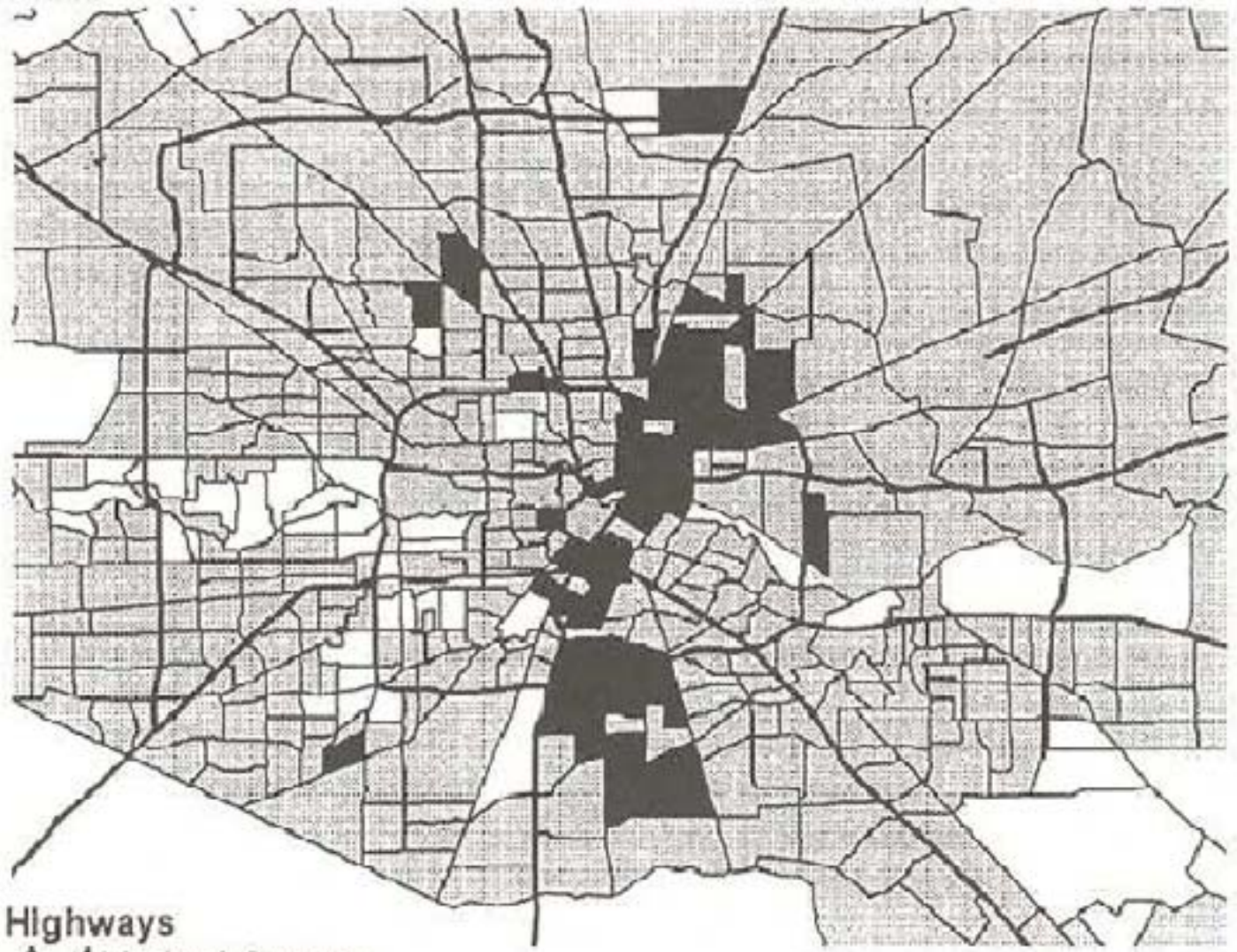


Fig. 1. Map based on spatial filtering using a half-mile grid and a half-mile filter





Highways  
Limited Access  
Concentrated Disadvantage Index  
10% least disadvantaged  
80% mid range  
10% most disadvantaged

Fig. 2. Map highlighting economically disadvantaged census tracts



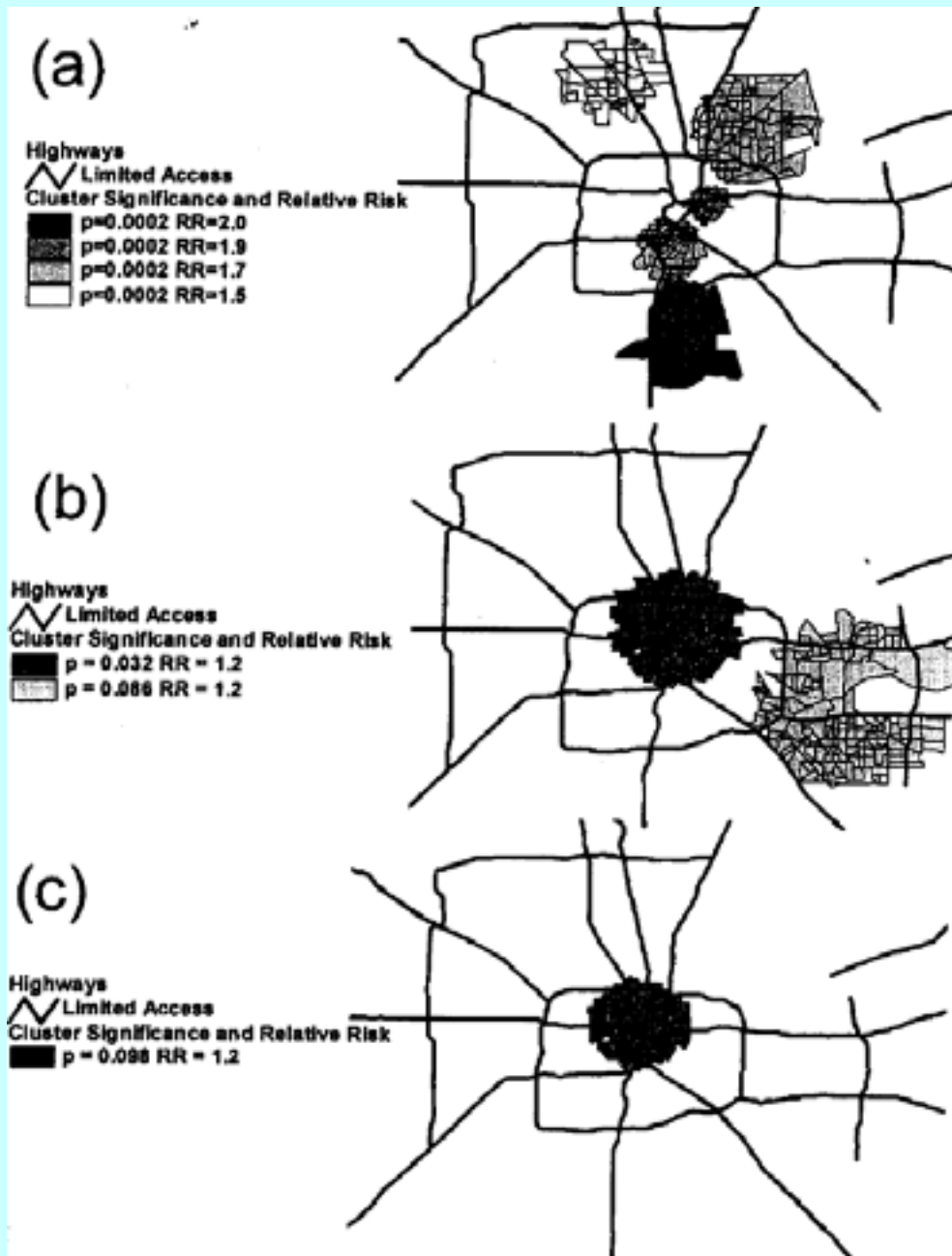
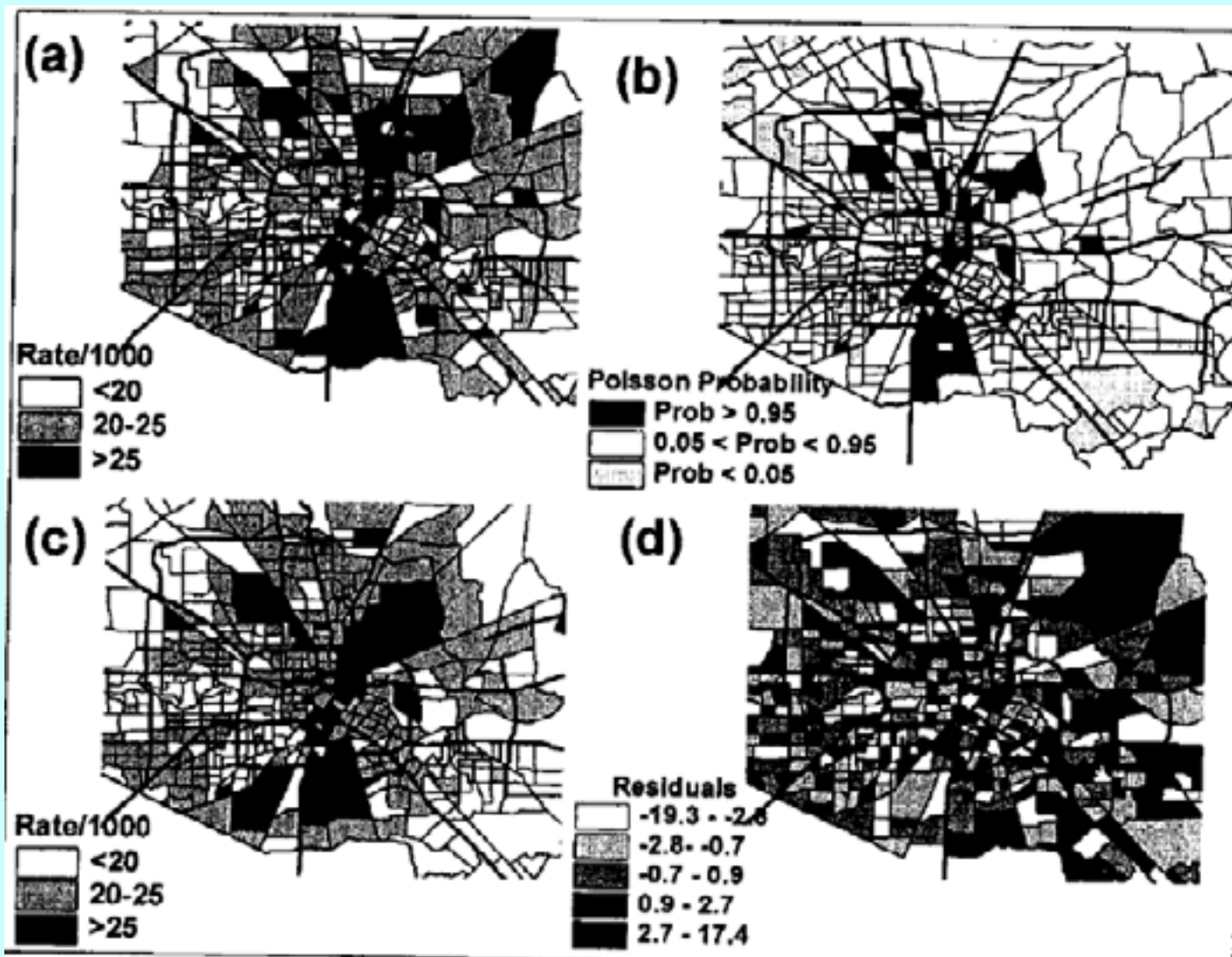


Fig. 3. Comparison of clustering results from SaTScan with

- (a) no adjustment for confounding variables,
- (b) adjustment for race/ethnicity of the mother, and
- (c) adjustment for race/ethnicity and age of mother. (Note: first cluster in each group with the darkest shading is the primary cluster).

Source: Bell, 2002, p. 168



(c). The CAR model is a conditional autoregression model. .."It incorporates the concept of spatial neighbors where the definition of neighbor is left to the analyst." (Bell 2002, p.173).

Fig. 5. Maps of rates, probabilities and residuals (a) rate before spatial modeling (b) map of statistically high rates and statistically low rates assuming rates have a Poisson distribution with constant mean, © map of predicted rates fit by a CAR spatial model, (d) map of residuals after fit by a CAR spatial model. (Bell, 2002, p.177).



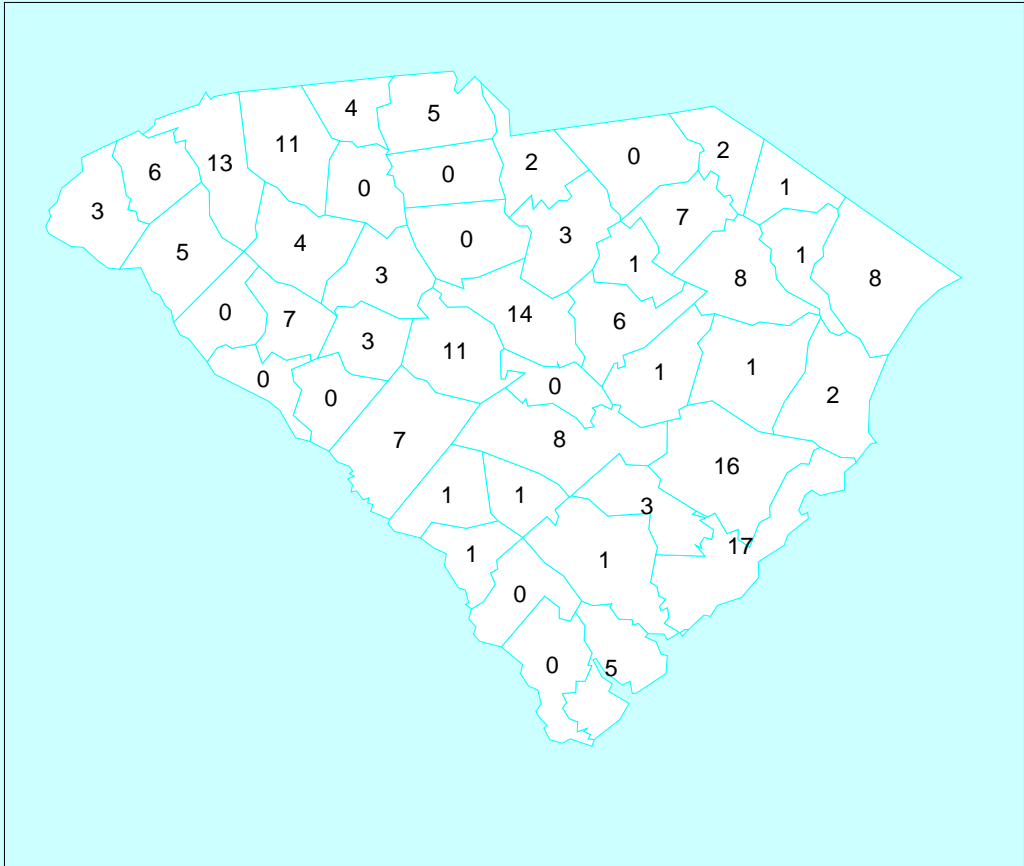
# Introduction to Bayesian Mapping Methods

Andrew B. Lawson

Arnold School of Public Health

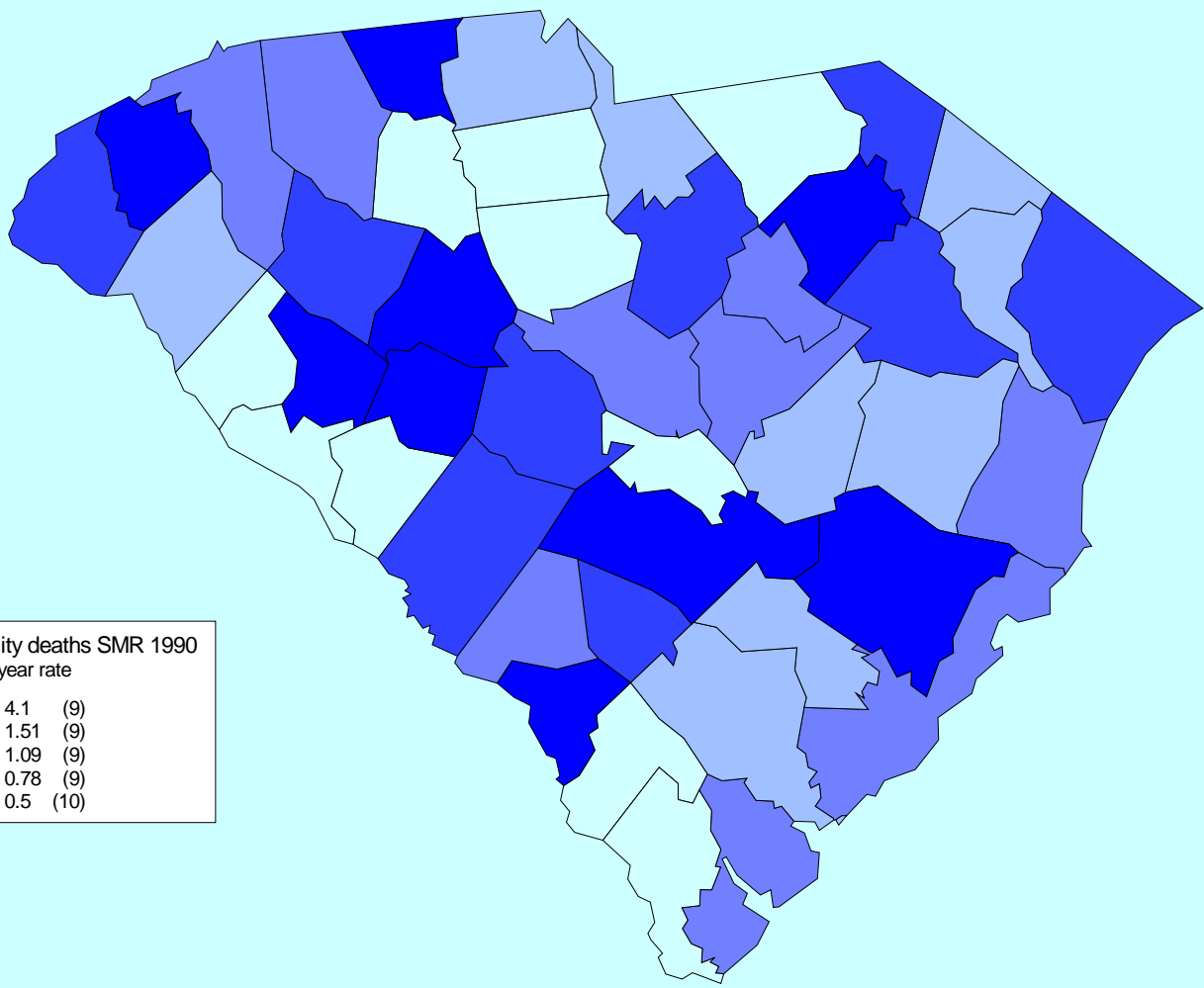
University of South Carolina

- South Carolina congenital abnormality deaths 1990



Congenital abnormality deaths SMR 1990  
using 8 year rate

- 1.51 to 4.1 (9)
- 1.09 to 1.51 (9)
- 0.78 to 1.09 (9)
- 0.5 to 0.78 (9)
- 0 to 0.5 (10)



# Some notation

- For each region on the map:
  - $y_i$  is the count of disease in the  $i^{\text{th}}$  region
  - $e_i$  is the expected count in the  $i^{\text{th}}$  region
  - $\theta_i$  is the relative risk in the  $i^{\text{th}}$  region
- 
- The SMR is just  $smr_i = y_i / e_i$
  - This is just an estimate of  $\theta_i$



# SMR problems

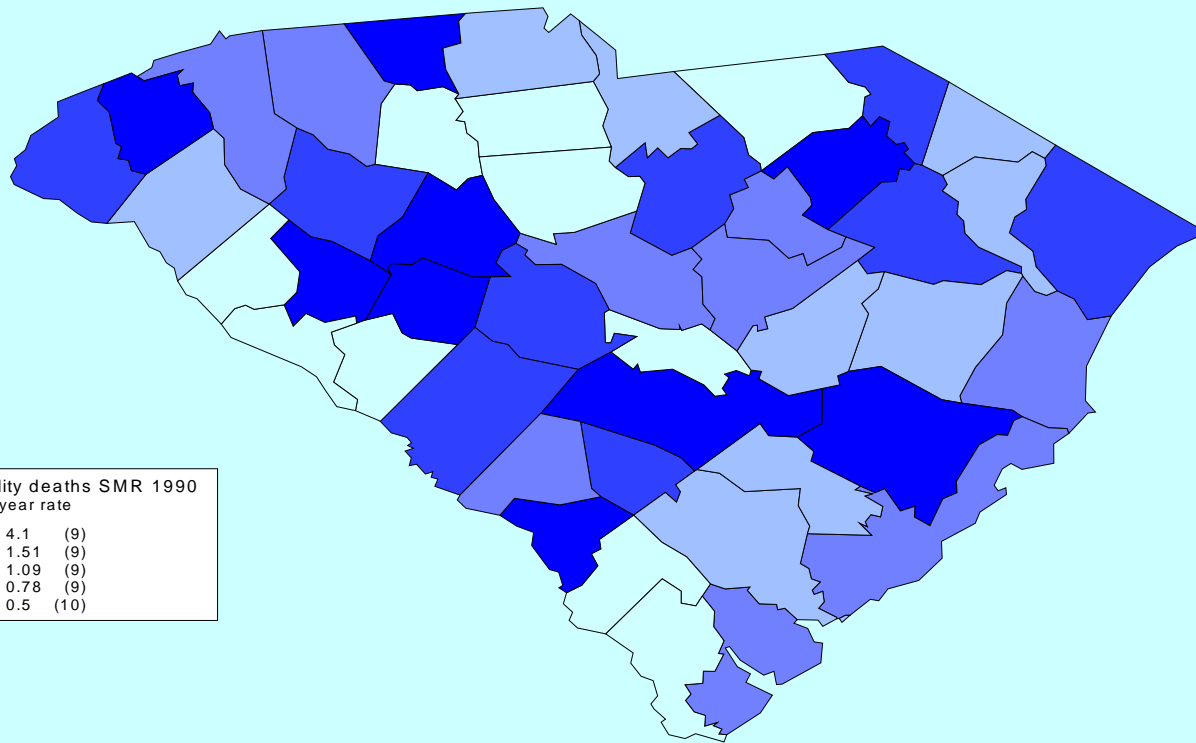
- Notoriously unstable
- Small expected count can lead to large SMRs
- Zero counts aren't differentiated
- The SMR is *just the data!*

# Bayesian Modeling

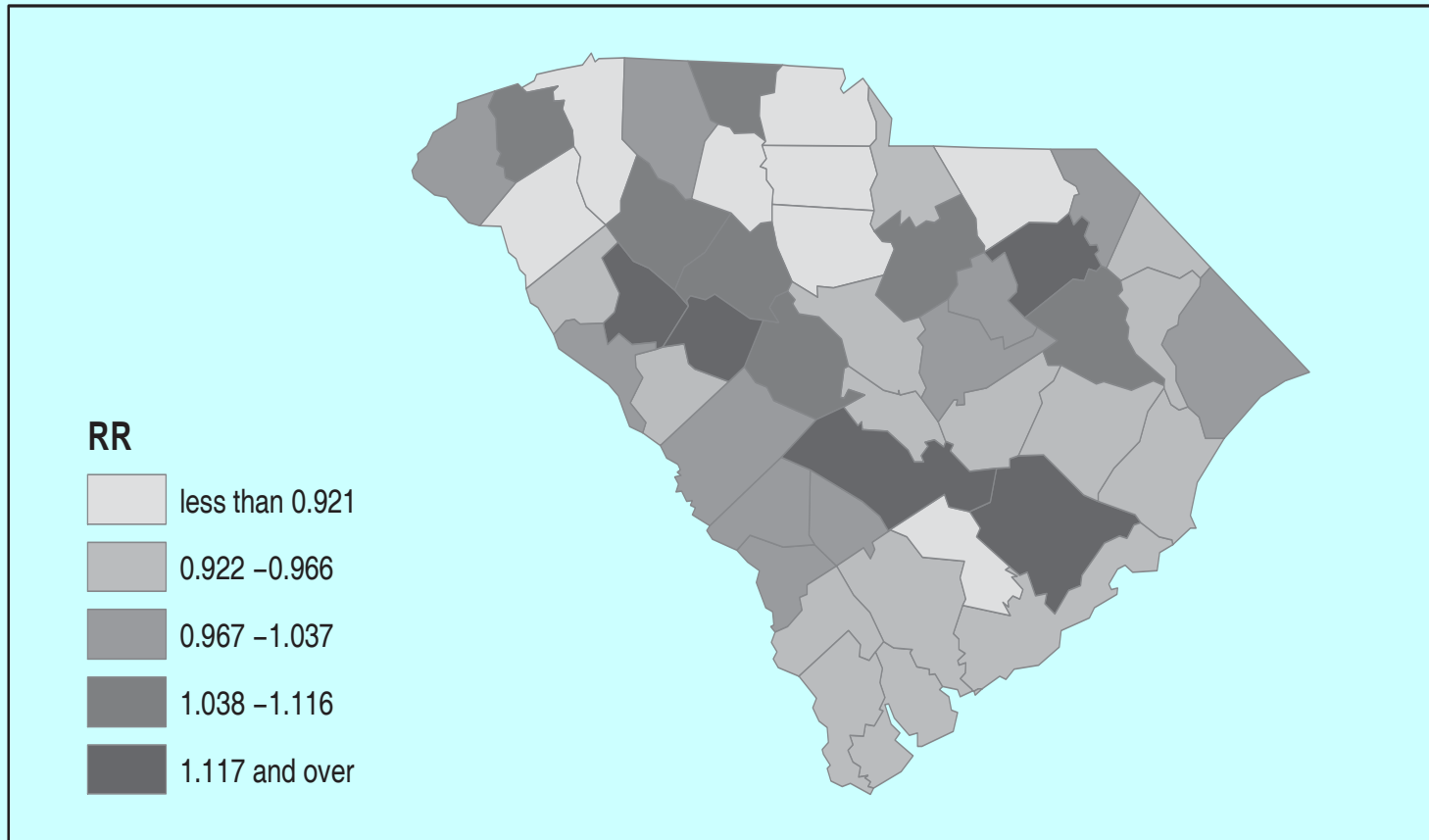
Some statistical ideas:

- Likelihood.....we usually assume that counts of disease have a Poisson distribution so that  $y_i$  has a Poisson distribution with expected value  $e_i \theta_i$
- We usually write this as  $y_i \sim \text{Pois}(e_i \theta_i)$  for short
- The counts have a Poisson likelihood

# SMR for congenital anomalies



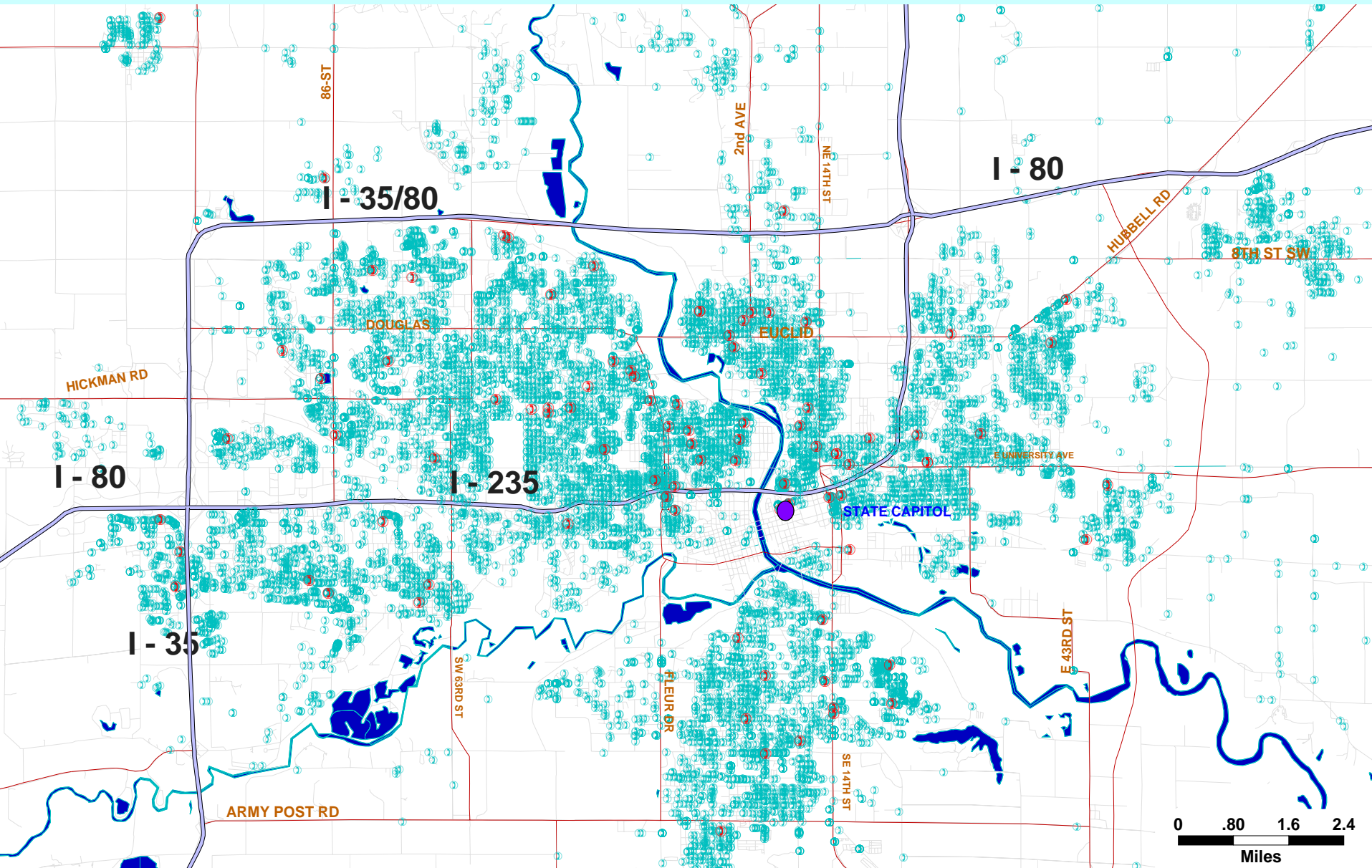
# Gamma Poisson model: WinBUGS



# Conclusions (Lawson 2003)

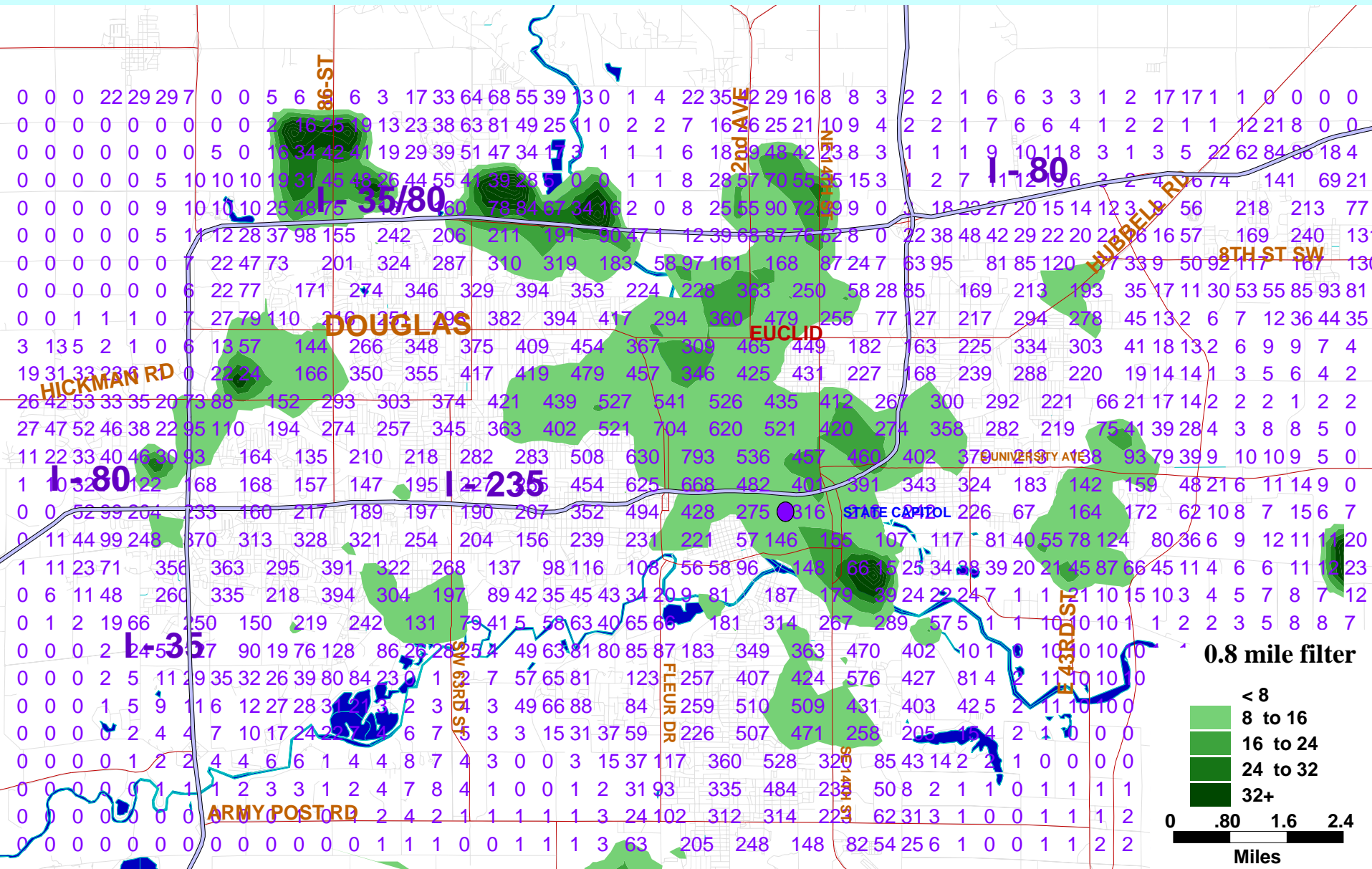
- WinBUGS provides a *free* and relatively easy-to-use tool for disease mapping with small area count data
- Allows state-of-the-art approach to relative risk and ecological regression
- Available from:  
*[www.mrc-bsu.cam.ac.uk/bugs](http://www.mrc-bsu.cam.ac.uk/bugs)*

# Location of Infant Births and Deaths: 1996-98

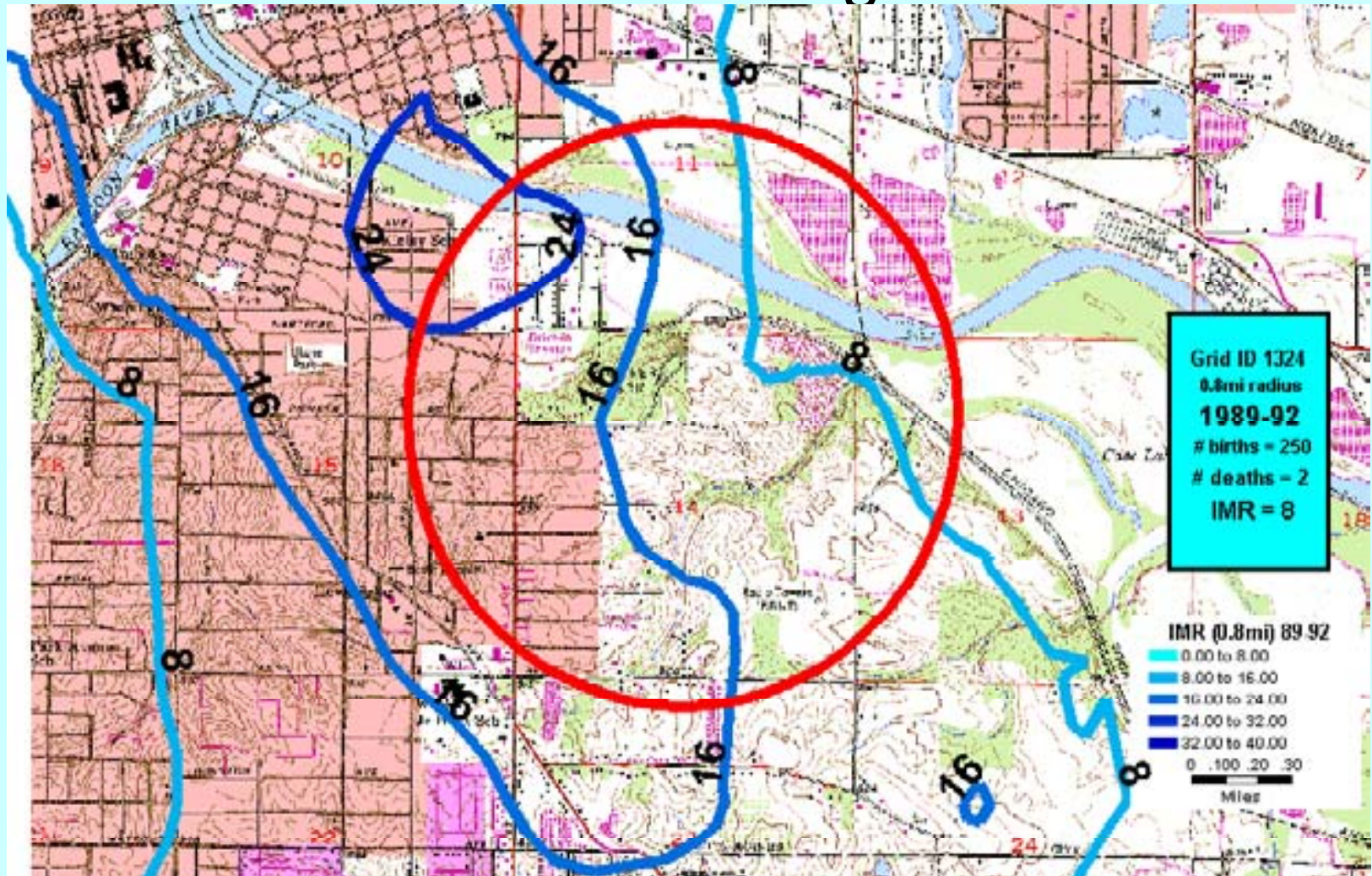




# Des Moines, Iowa: Births and Infant Mortality Rates: All Races, 1996-98



# Part of Des Moines, Iowa: infant mortality rates 1989-92 draped on USGS 1:24,000 TIF Image





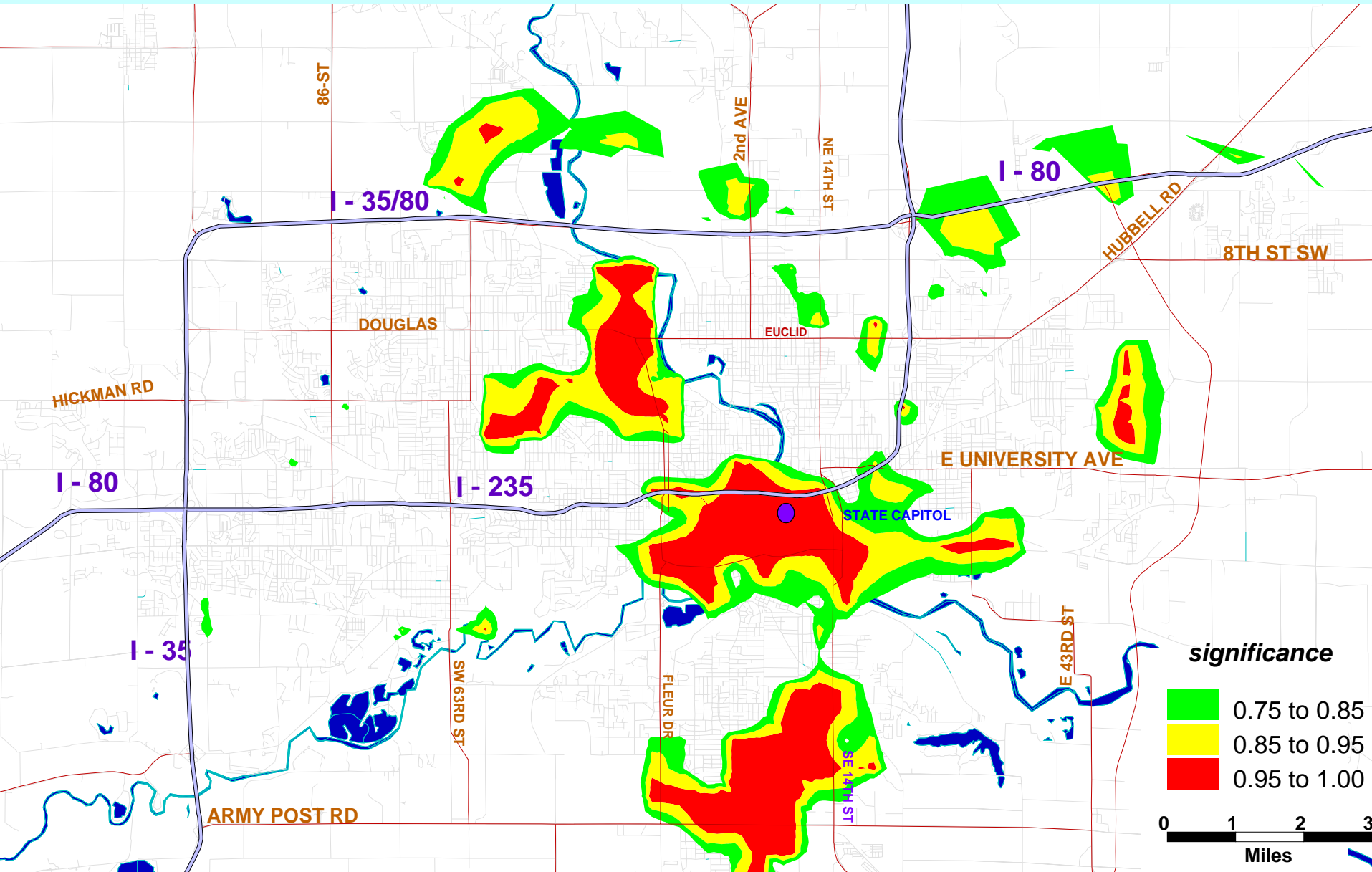
# *Not all areas of the map have equally reliable rates*

- Some local rates are based on a large amount of information; other local rates are based on little information.
- Need to compute the local reliability of the observed rate
- Generate a map of the expected disease pattern according to the null hypothesis that each person at risk has an equal and identical chance of having the disease

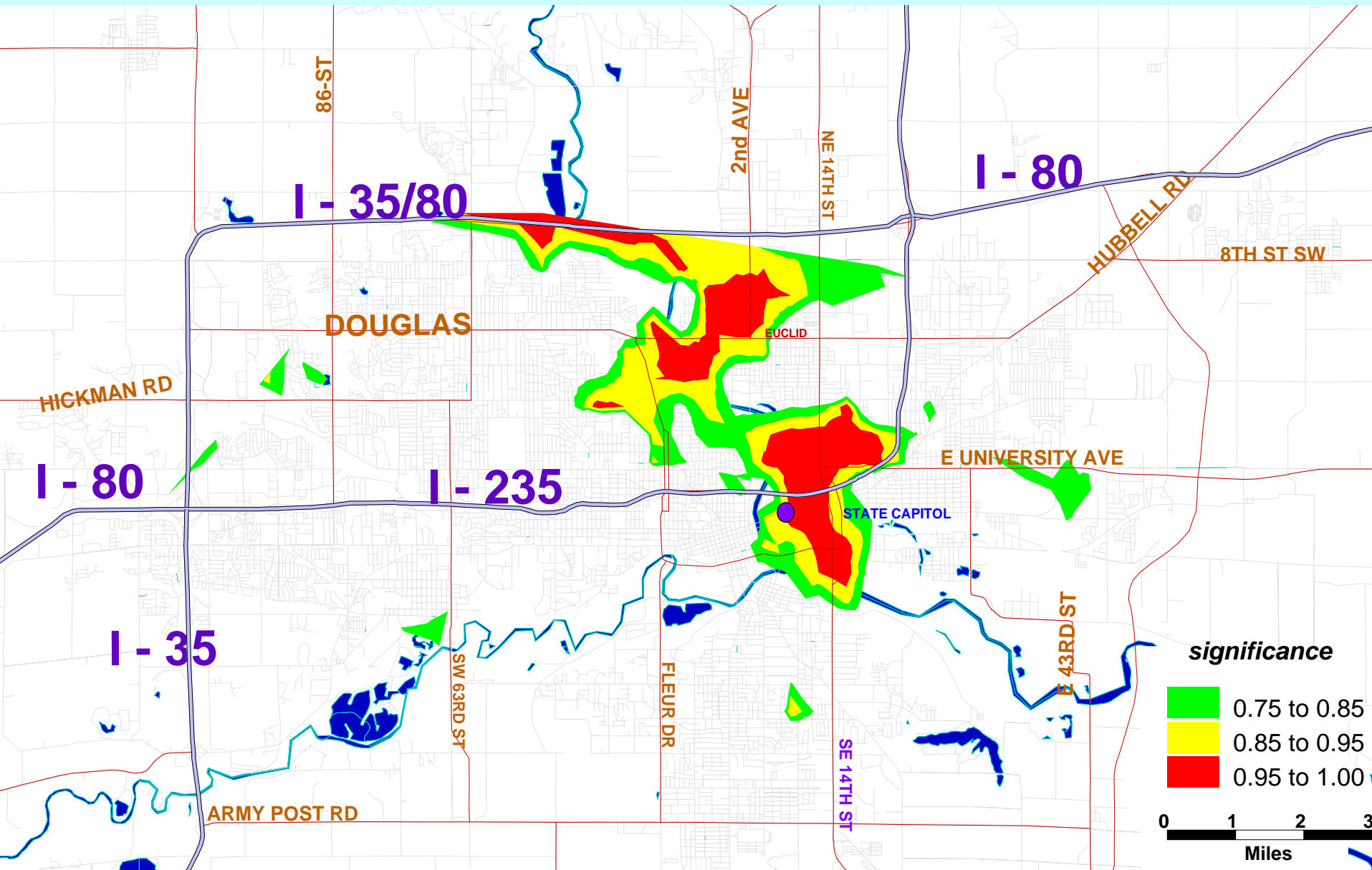
# Significance of Rates

- let each birth in 1989-91 have an identical probability equal to the county-wide infant mortality rate of 9.5 of becoming an infant death
- generate 1,000 geographical distributions of simulated infant deaths
- for each of these maps compute the infant mortality rate at each grid location
- compute the proportion of the simulated rates at each grid location that are smaller than the observed infant mortality rate
- define regions with significantly high infant mortality rates where a high proportion of the simulated rates are smaller than the observed rate

# Significance of Infant Mortality Rates: 1993-95



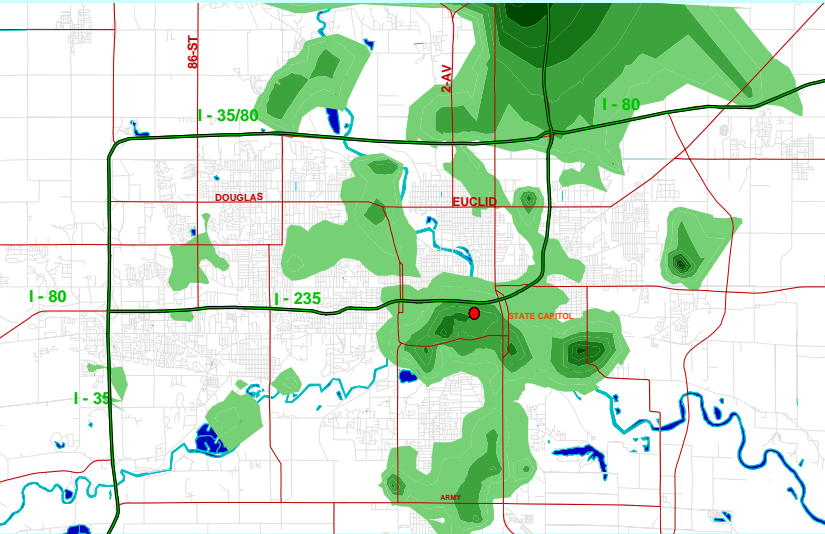
# Significance of Infant Mortality Rates: 1996-98



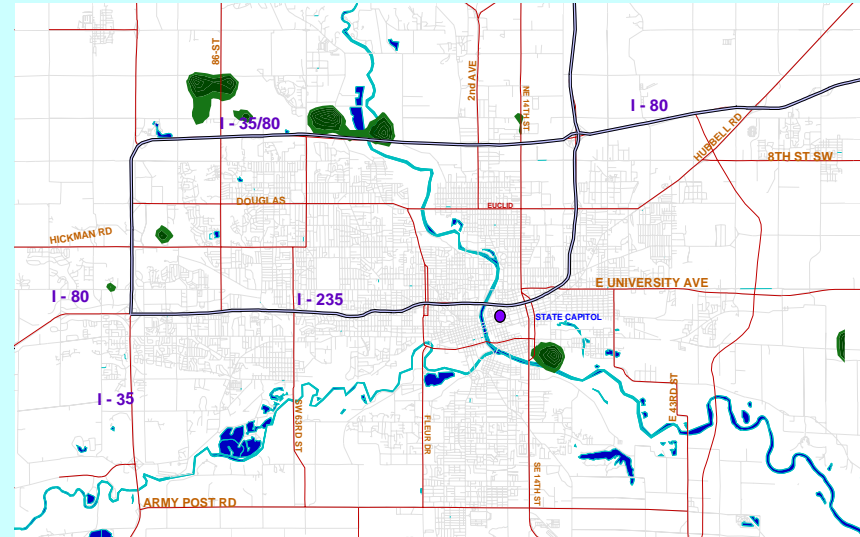


# Change in Infant Mortality 1993-1998

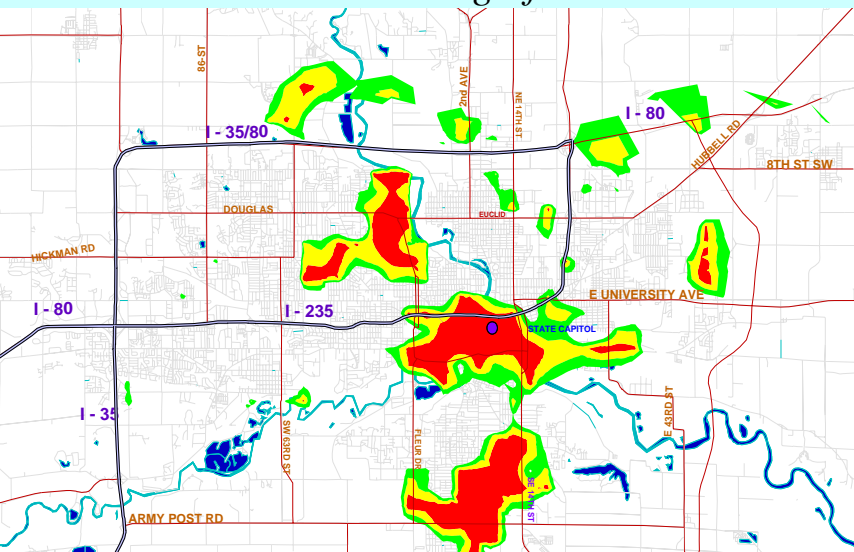
1993-95 IMR



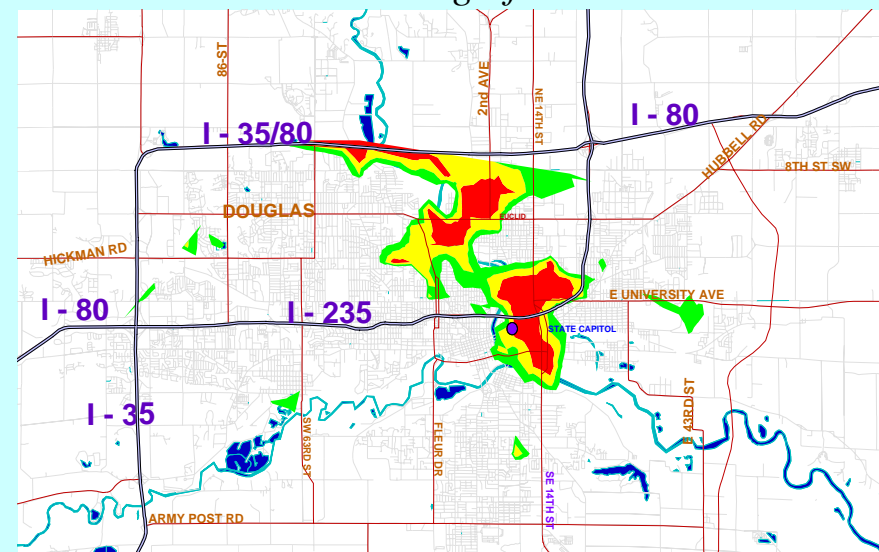
1996-98 IMR



1993-95 Significance



1996-98 Significance



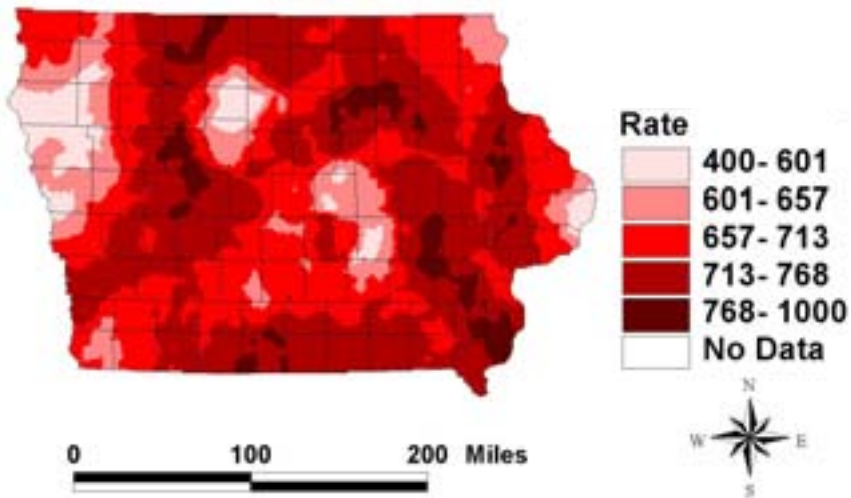
# Examples

- Most searches for spatial clusters of cancer use geographic units such as health service areas, state economic areas or counties.
- But suppose areas of high risk for cancer are much more localized?

# Spatially Adaptive Filters

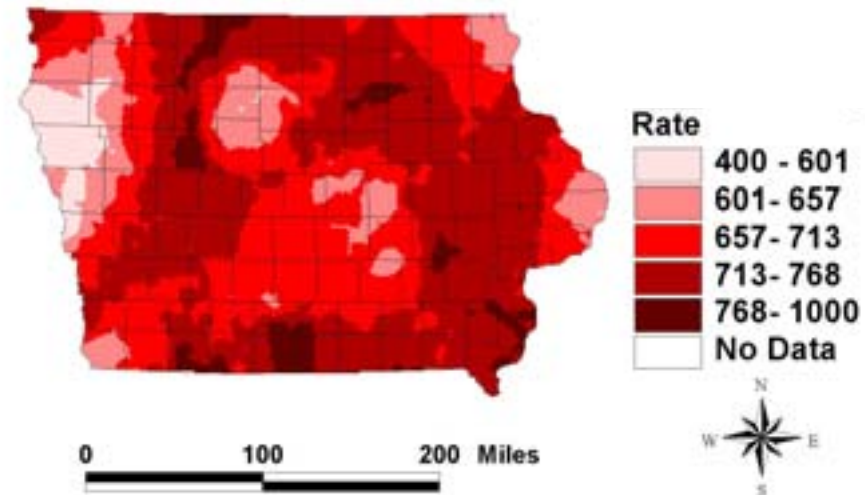
Rates of Late-Stage Colorectal Cancer at Time of First Diagnosis

**Spatially Adaptive Filter  
Threshold > 100**



**Using Spatially Adaptive Filters**

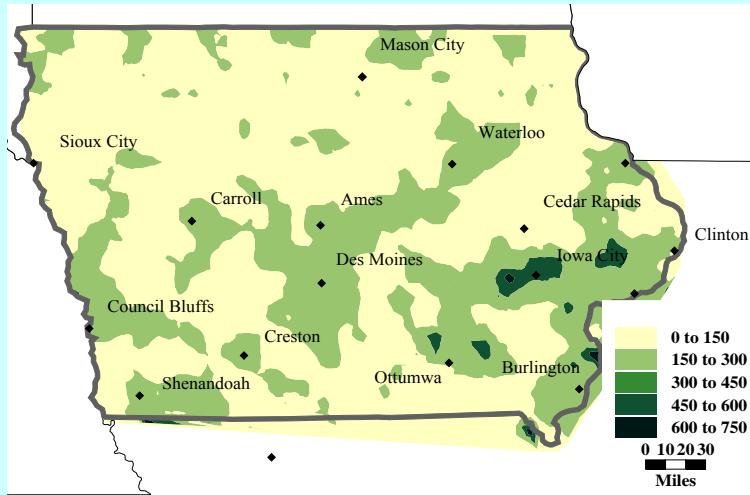
**24 Mile Filter**



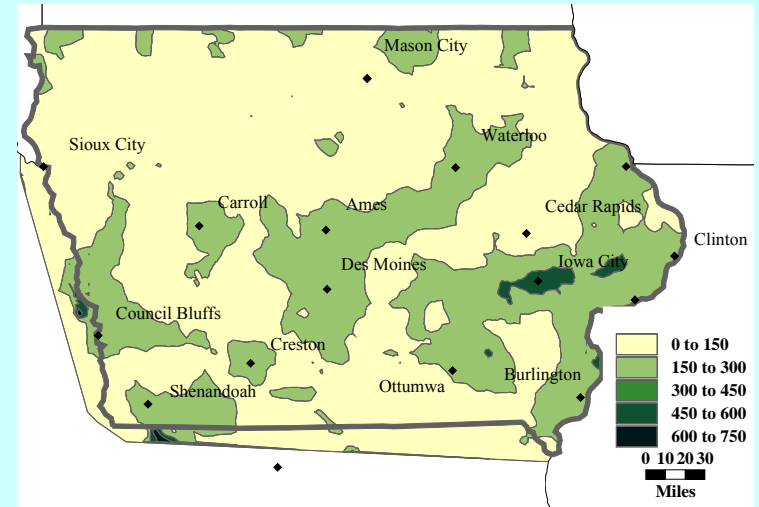
**Using Fixed Filter Size of 24 Miles**

# Lumpectomy With Radiation Rates

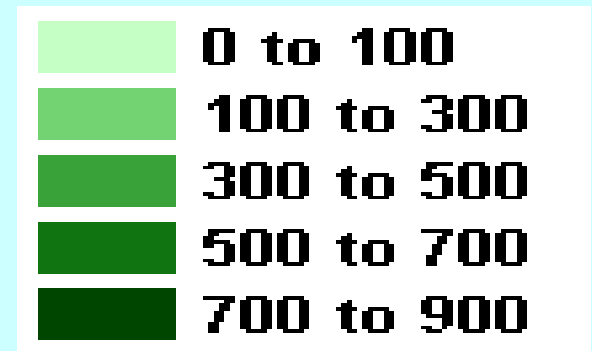
## 10 Mile Filter



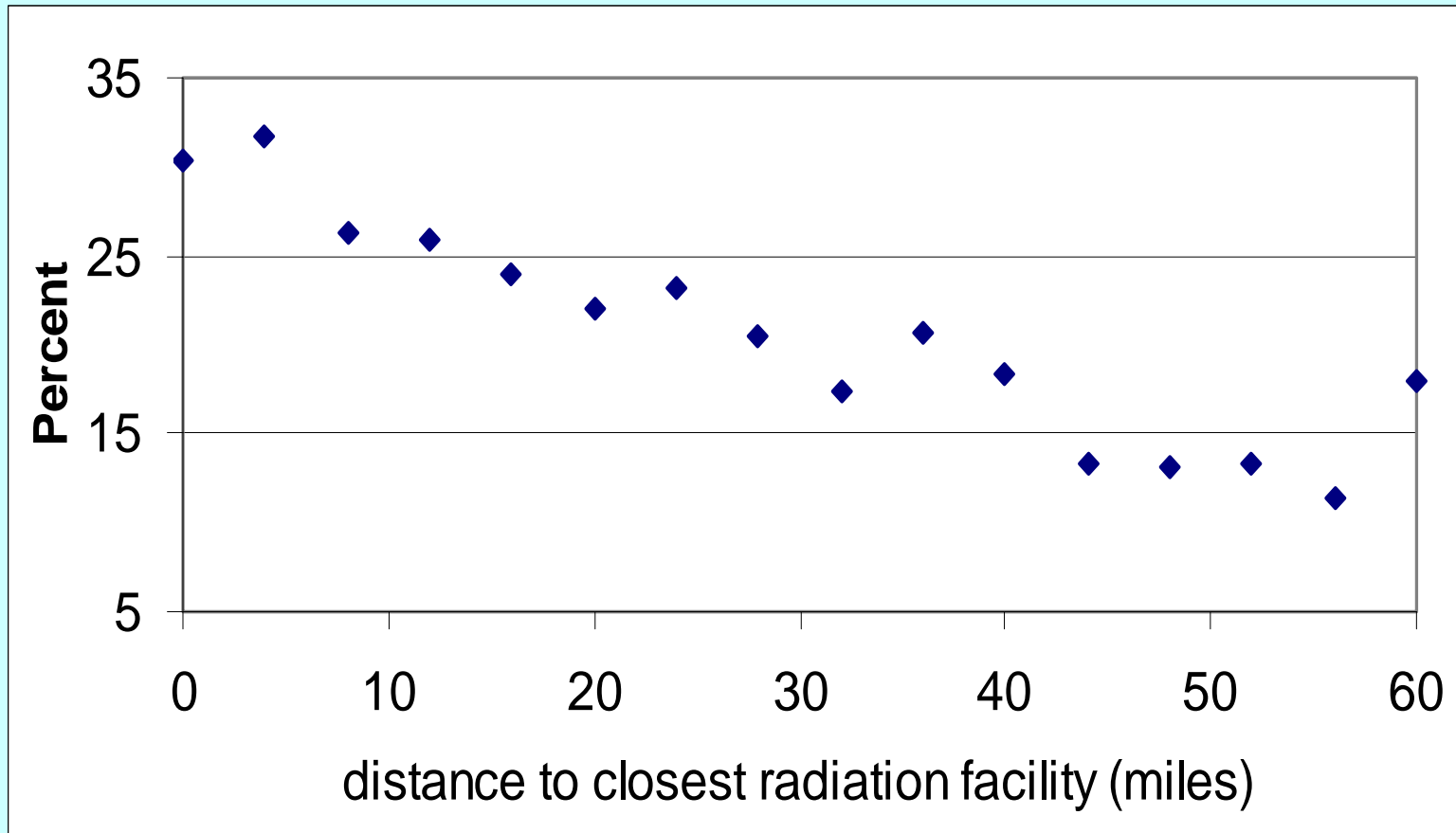
## 12.5 Mile Filter



## 15 Mile Filter



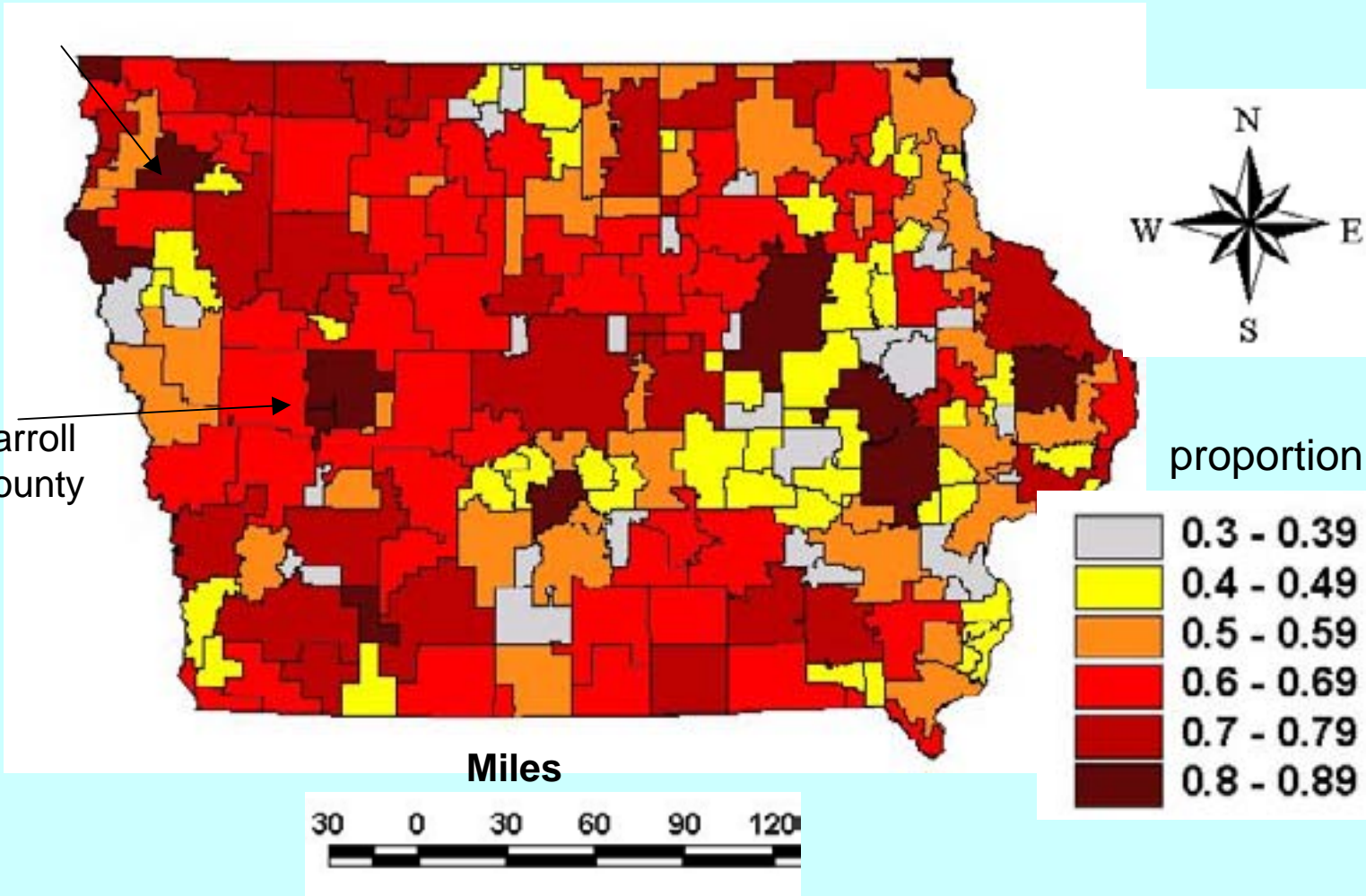
# Percent Selecting Lumpectomy/Radiation Treatment



# Problem with geocoding practices

- Geocoding is not taken too seriously in most environment exposure and health applications
- The principle of 'fitness for use' must be understood and used.

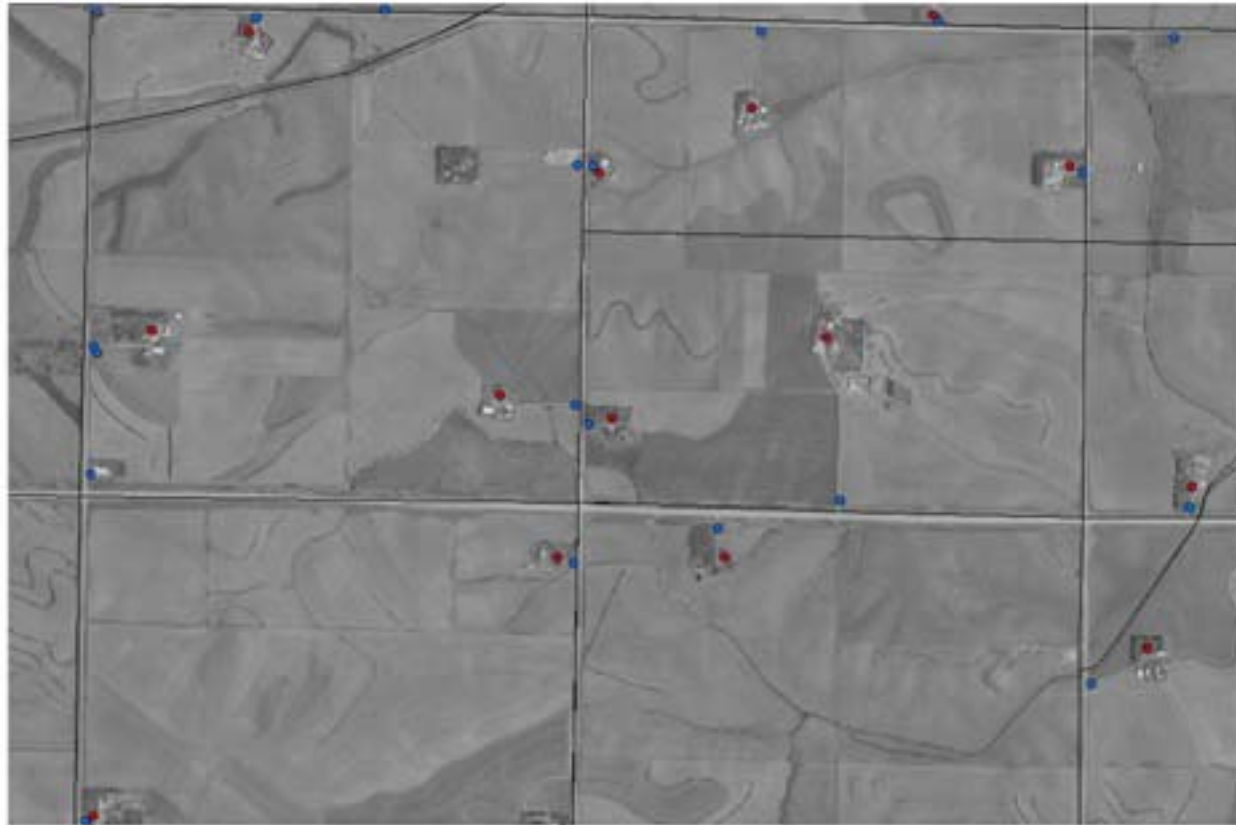
# Iowa: Primary Care Service Areas—proportion of primary care patient visits made inside the local service area



Source: data files from [pcs.a.hrsa.gov](http://pcs.a.hrsa.gov) See Goodman et al. 2003



In this map, there are 14 houses and 20 addresses. Each house has a corresponding address point located where the entrance road to the house leaves the public road. There are two possible reasons for the other six addresses that do not have a corresponding house. The first is that the address is not a residential address, the second is that the house of the resident address can not be identified by this interpretation of the orthophoto map.



Legend

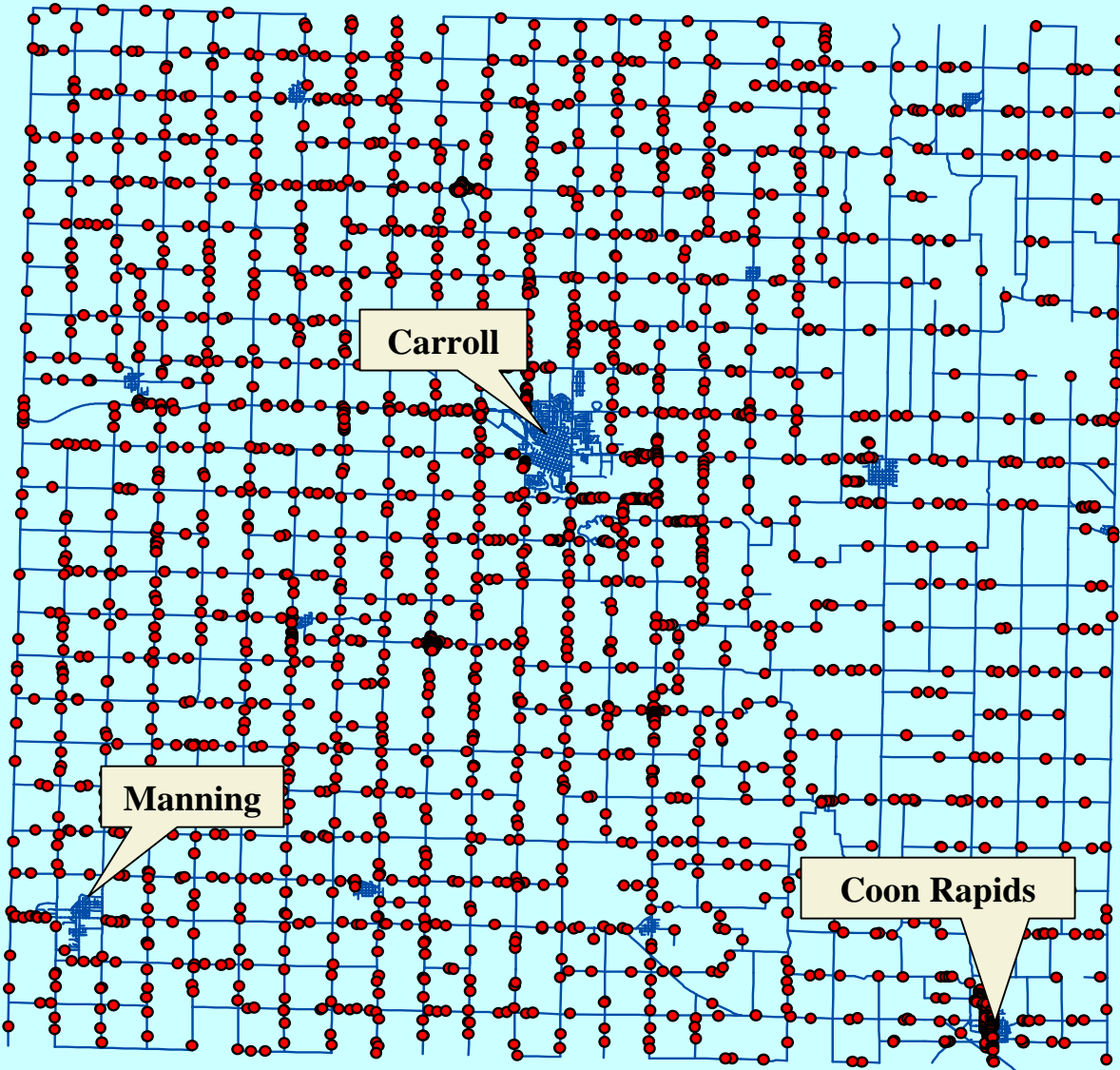
- addresses
- houses
- Block boundaries

1,000 0 1,000 2,000 Feet





# Address point map containing the locations of separately addressable units in Carroll County



This map(07/21/03) shows the current extent of the address point file. This file contains the locations of 100% of the addresses in the E-W direction and 2/3rds of the addresses in the N-S direction.

## Legend

- Address point
- Roads

# *Long Island Breast Cancer Study and the GIS-H (Health)*

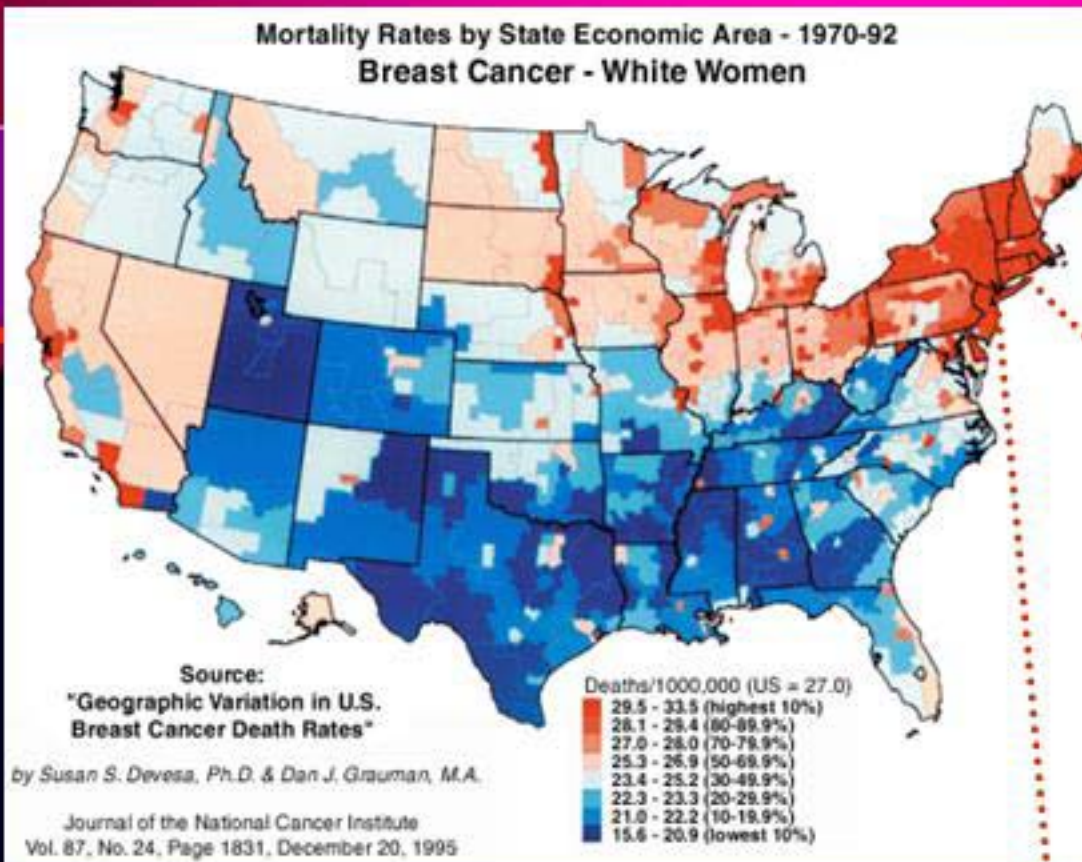


Edward J. Trapido, Sc.D.  
Associate Director

Epidemiology and Genetics Research Program, DCCPS/NCI

COMPREHENSIVE APPROACHES TO CANCER CONTROL  
September, 2003 Atlanta, GA



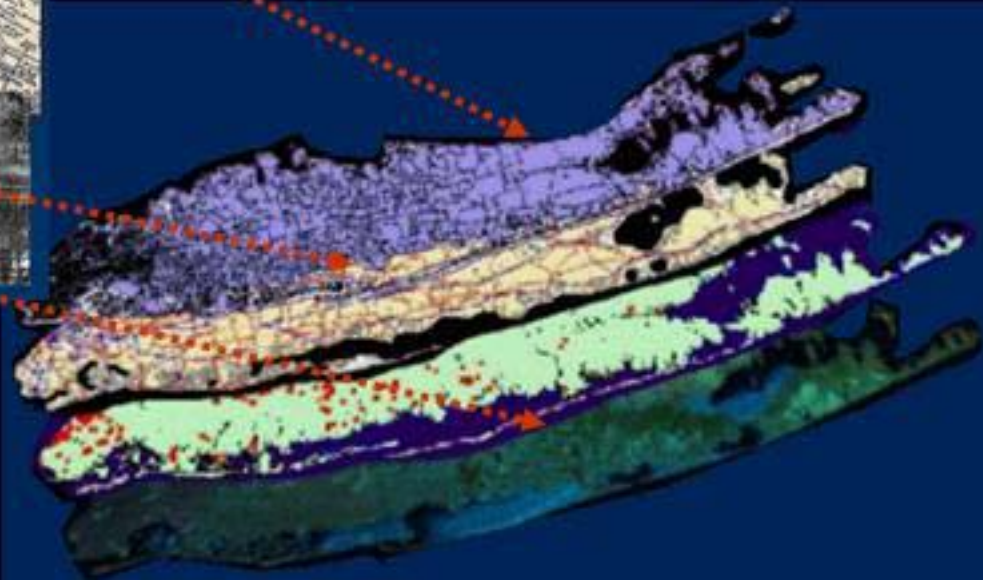
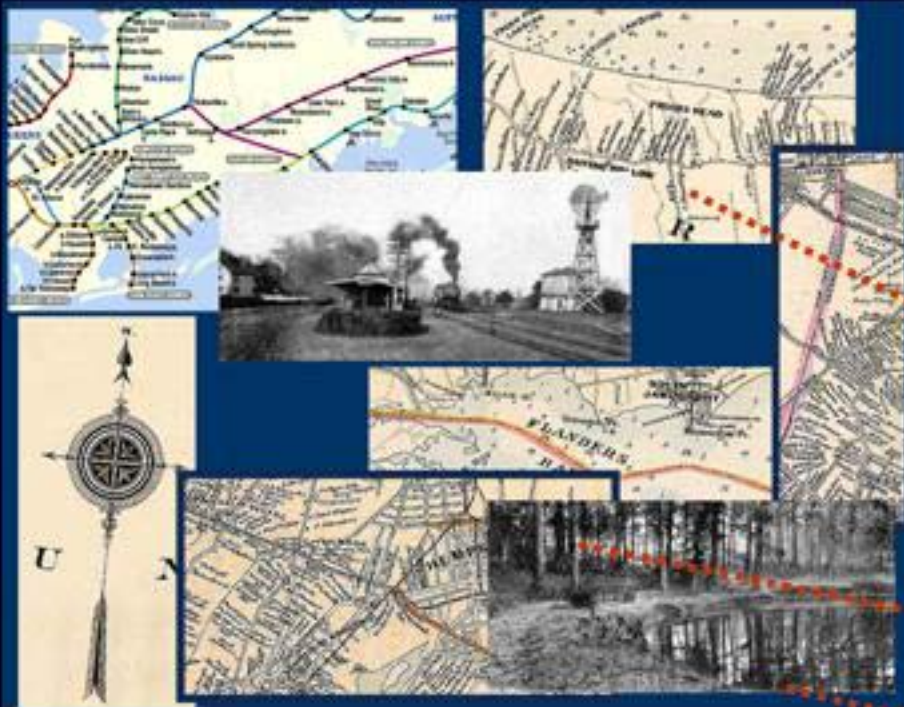


The northeastern United States has had high rates of breast cancer.

The Long Island Breast Cancer Study Project (LIBCSP) focuses on Long Island (Nassau and Suffolk counties) in New York.



# A Tool for Studying Environment & Breast Cancer







# Geographical Information System for Health (GIS-H)

Part of the Long Island Breast Cancer Study Project (LIBCSP)

*A new tool for research on breast cancer on Long Island*



- Overview
- FAQ
- Reports & Activities
- Advisory Groups
- Links
- NCI
- LIBCSP
- GIS-H Project

## GIS-H and the Public

- Go — Find out what is in the GIS-H Data Warehouse using the Metadata Browser
- Go — View the geographic extent of the GIS-H.
- Use Interactive Maps to look at environmental and demographic information about Long Island

## GIS-H and Researchers

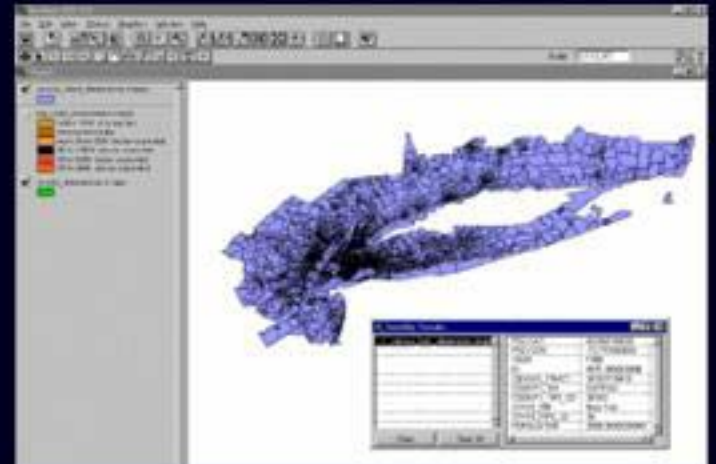
- Go — What does the GIS-H offer Researchers?
  - A data warehouse providing data from over 60 different sources

[www.healthgis-li.com](http://www.healthgis-li.com)



# *Data Included in the GIS-H*

- ▶ Geospatial
- ▶ Demographic and Behavioral
- ▶ Health
- ▶ Environmental







Home > Metadata Browser > FGDC Reports

# FGDC Reports

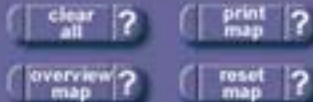
## FGDC Reports Directory

- Geographic
- 2000 TIGER/Line Files
- 2000 Census County Boundaries
- 2000 Census Minor Civil Divisions
- 2000 Census Place Boundaries
- 2000 Census Tract Boundaries
- 2000 Census Block Groups
- 2000 Census Block Boundaries
- 1998 TIGER/Line Files
- 1992 TIGER/Line Files
- 1990 Census County Boundaries
- 1990 Census Minor Civil Divisions
- 1990 Census Place Boundaries
- 1990 Census Tract Boundaries
- 1990 Census Block Groups
- 1990 Census Block Boundaries
- 1980 Census Tract Boundaries
- NYS Railroad Lines and
- ESRI Zip Code Boundaries

The GIS-H Data Warehouse comprises geographic data from a variety of sources in addition to geographic data that have been developed specifically for the GIS-H. The metadata for geographic information in the GIS-H Data Warehouse follow the [Federal Geographic Data Committee \(FGDC\)](#) guidelines. In cases where data were supplied by commercial or government organizations, the metadata were taken from the metadata supplied by the distributor. Where the metadata were supplied in FGDC format, the GIS-H staff transposed the information to the GIS-H metadata repository. Where FGDC data were not supplied, the GIS-H staff conformed the metadata that were supplied to the FGDC guidelines as much as possible. For data that were created by GIS-H staff, the metadata conform to the FGDC guidelines. The metadata are presented here for the benefit of GIS-H users.

It is critical for the users of the GIS-H to understand the limitations of the geographic data in the GIS-H. The user should review the metadata carefully, paying special attention to accuracy, consistency, quality, and use constraints as well as pertinent geospatial specifications. The GIS-H staff transforms the geographic data into a format appropriate for loading into the GIS-H Data Warehouse. The data are stored in Oracle in ESRI's proprietary format and are accessed through ESRI's SDE facility. Any

Expand All Collapse All Help Home Frames Contact Us Floating Menu



# Long Island Income Demographics

zoom in zoom out zoom last pan identify measure ?

intro search buffer legend layers help ?

Click a point on the map or hold the mouse to draw a rectangle to zoom in



**Legend - Microsoft Internet Explorer**

**1990 Income Demographics**

- 0 - \$24,999
- \$25,000 - 34,999
- \$35,000 - 49,999
- \$50,000 - 99,999
- \$100,000 and Up

**Breast Cancer Incidence by Zip Code**

- More than 100% above expected
- 50 to 100% above expected
- 15 to 49% above expected
- Within 15% of expected
- 15 to 50% below expected
- More than 50% below expected
- Very sparse data
- Rates not available

**County Boundaries**

## Layers

[Layer Information](#)

Redraw Map

### Display

- County Boundary
- 1970 Income Demographics
- 1990 Income Demographics
- Breast Cancer Incidence by Zip Code

0 15 miles

Display Units: Miles

Active 1970 Income Demographics

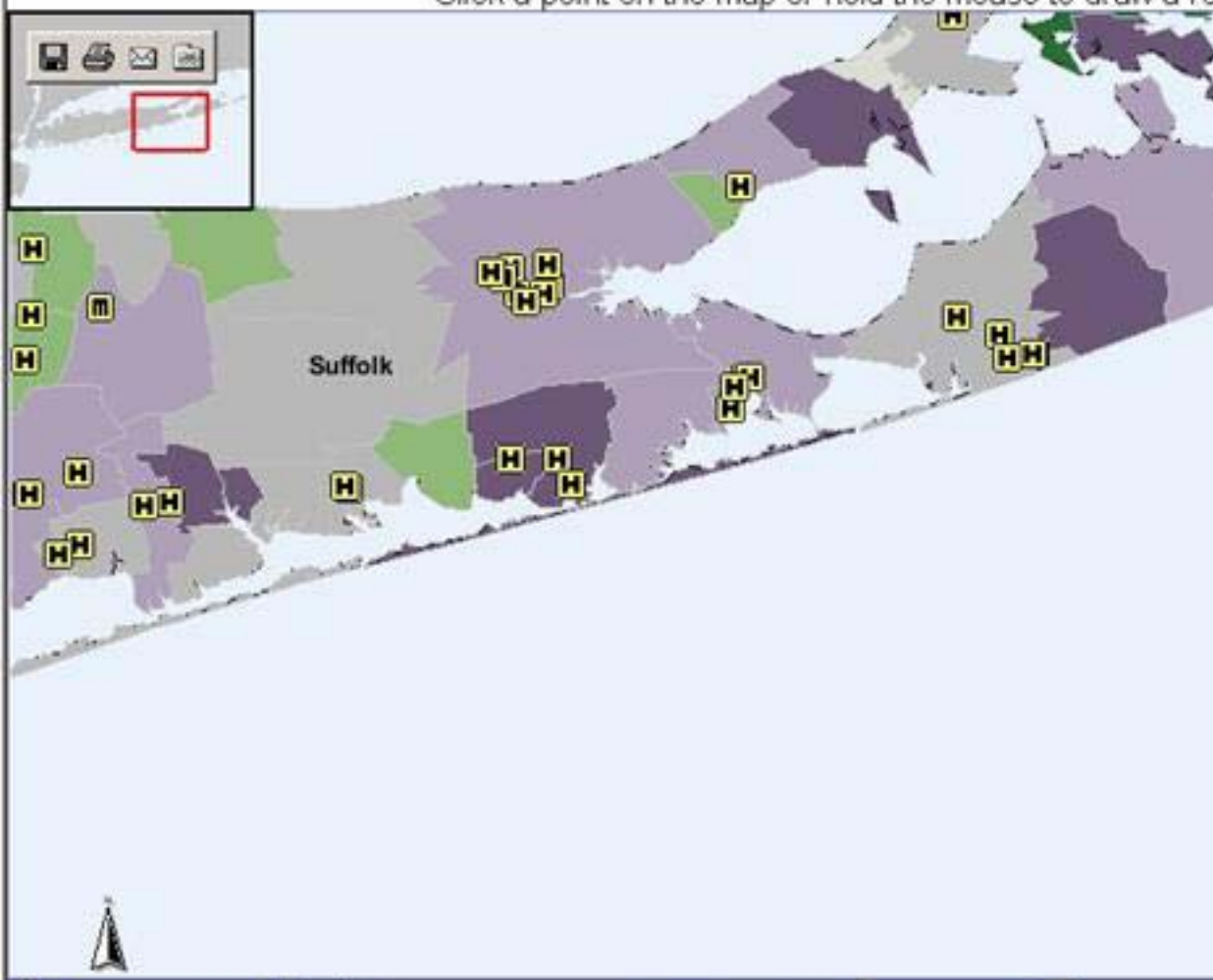
Demographic and 1970 Income Demographics you will only see the 1970 information because 1970 Income





# Long Island Health Facilities

Click a point on the map or hold the mouse to draw a rectangle to zoom in



## Layers

[Layer Information](#)

### Display

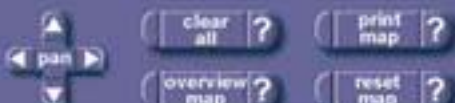
- Health Facilities
- Mammography Clinics
- Zip Code Boundaries
- County Boundary
- Breast Cancer Incidence by Zip Code

0

Display Units:

Active

Other ...

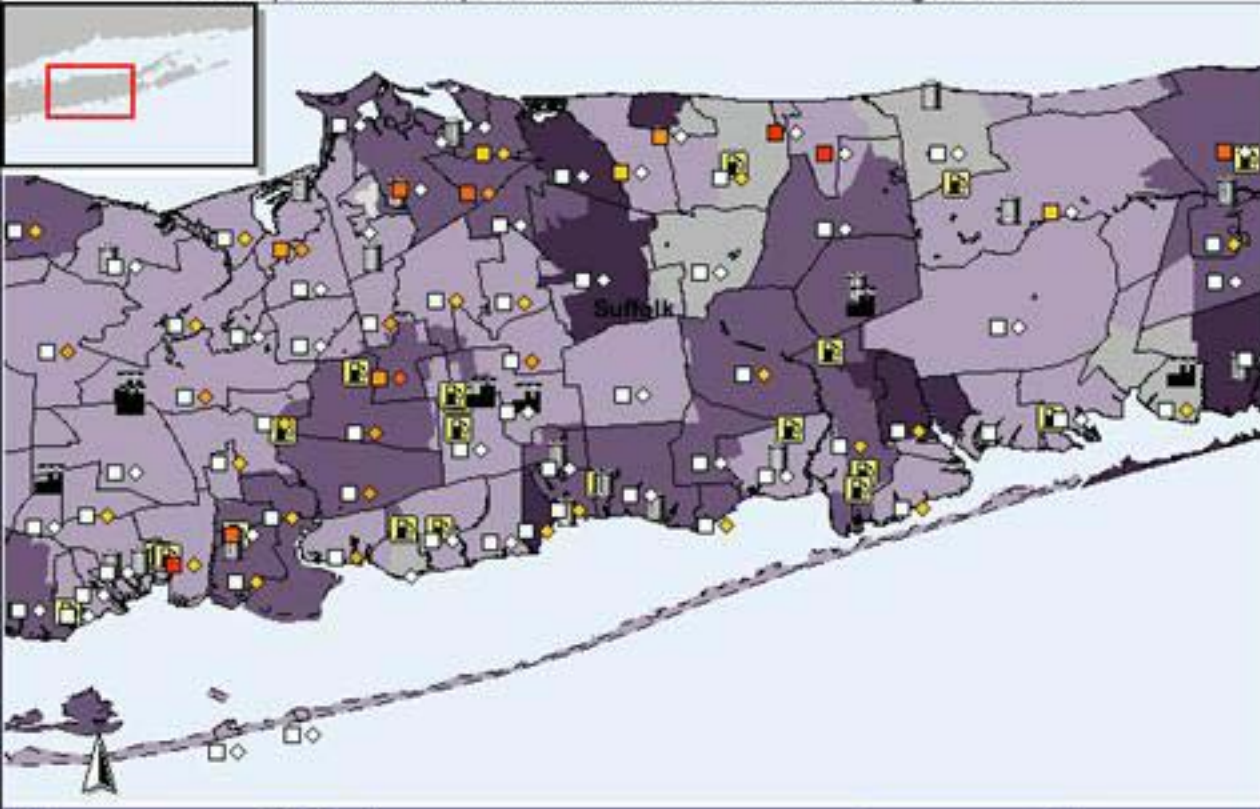


## VOC Detections in Suffolk County Wells

zoom in zoom out zoom last pan identify measure ?

intro search buffer legend layers help ?

Click a point on the map or hold the mouse to draw a rectangle to zoom in



### Layers

[Layer Information](#)

Redraw Map

#### Display

- Cis-12-Dichloroethene Detection by Place
- MTBE Detection by Place
- Tetrachloroethene Detection by Place
- Tetrachloroterephthalic Acid Detection by Place
- 12-Dichloroethane Detection by Place
- TRI Water Release
- Petroleum Tanks
- Spills
- Gas Stations
- Dry Cleaners
- Suffolk Well VOC Detection

0 | 6.75 miles

Display Units: Miles

Active Cis-12-Dichloroethene



# Metadata Browser

## FGDC Reports Directory

- [-] Geographic
  - [+] 1998 TIGER/Line Files
  - [+] 1992 TIGER/Line Files
  - [+] 1990 Census County Boundaries
  - [+] 1990 Census Minor Civil Division Boundaries
  - [+] 1990 Census Place Boundaries
  - [+] 1990 Census Tract Boundaries
  - [+] 1990 Census Block Group Boundaries
  - [+] 1990 Census Block Boundaries
  - [+] 1980 Census Tract Boundaries
  - [+] NYS Railroad Lines and Stations
  - [+] ESRI Zip Code Boundaries
  - [+] ESRI Zip Code Centroids
  - [+] ESRI Golf Courses
  - [+] ESRI Cemeteries
  - [+] US Geologic Survey (USGS) 1:100K Digital Line Graphs(DLG)
  - [+] USGS 1:24K Digital Line Graph (DLG)
  - [+] USGS Digital Elevation Models (DEM)
  - [+] USGS Land Use/Land Cover
  - [+] C-CAP Land Cover
  - [+] Suffolk County Water Authority (SCWA) Pipelines
  - [+] SCWA Pressure Zones
  - [+] NCDC Daily Weather Summaries
  - [+] Nassau County Base Maps
  - [+] Suffolk County Base Maps
  - [+] Long Island Traffic Counts
- [+] Imagery

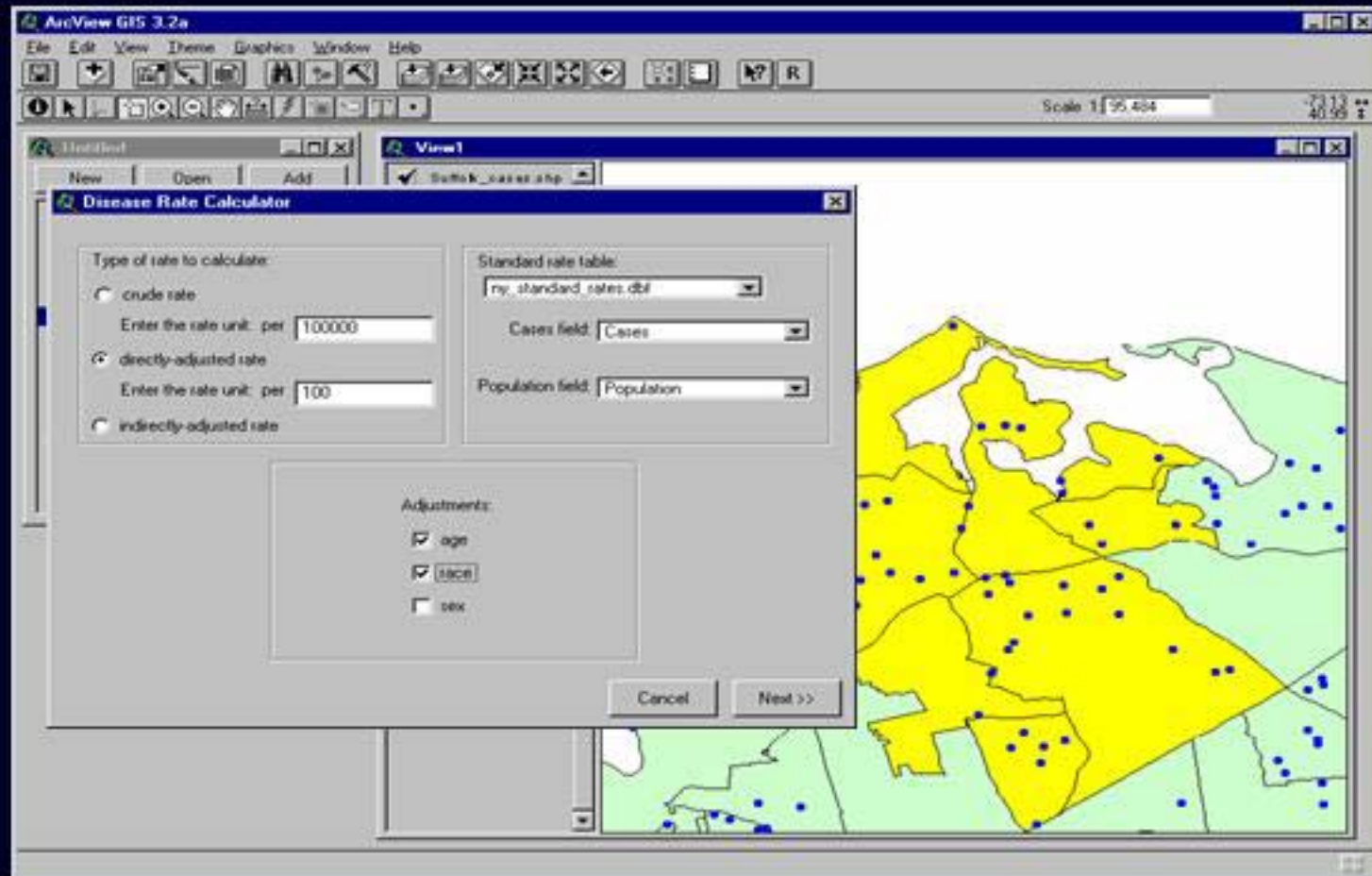
## Source Datasets Directory

- [+] Demographic
- [+] Medical Facility
- [+] Behavioral Surveys
- [+] Air Quality
- [-] Water Quality and Water Use
  - [+] EPA Permit Compliance System (PCS)
  - [+] USGS Relation of Ground-water Quality to Land Use
  - [+] USGS Pesticides in Surface Waters
  - [+] USGS Pesticides in Wells of Suffolk County
  - [+] USGS State Water Use
  - [+] Suffolk County Sewage Treatment Facilities
  - [+] Suffolk County Water Quality for Community Supply
  - [+] Suffolk County Water Quality for Non-community Supply
  - [+] Suffolk County Distribution Drinking Water
  - [+] Suffolk County Carbamate Pesticide Study
  - [+] Suffolk County Private Well Database
- [+] Industrial Sites and Hazardous Materials
- [+] Radioactive Sites or Materials
- [+] Miscellaneous



# Disease Rate Calculator

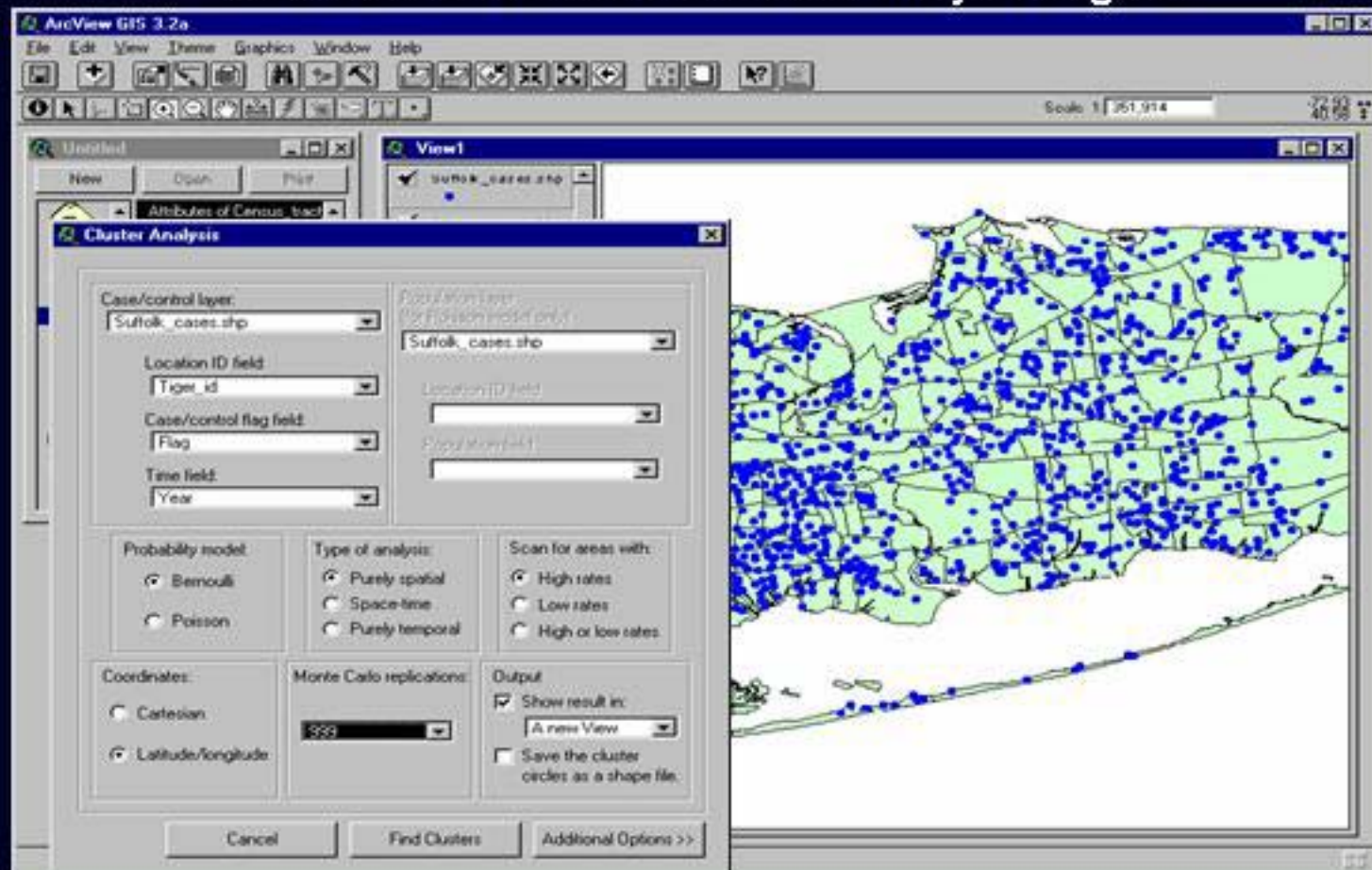
Calculating directly-adjusted rate for selected census tracts.





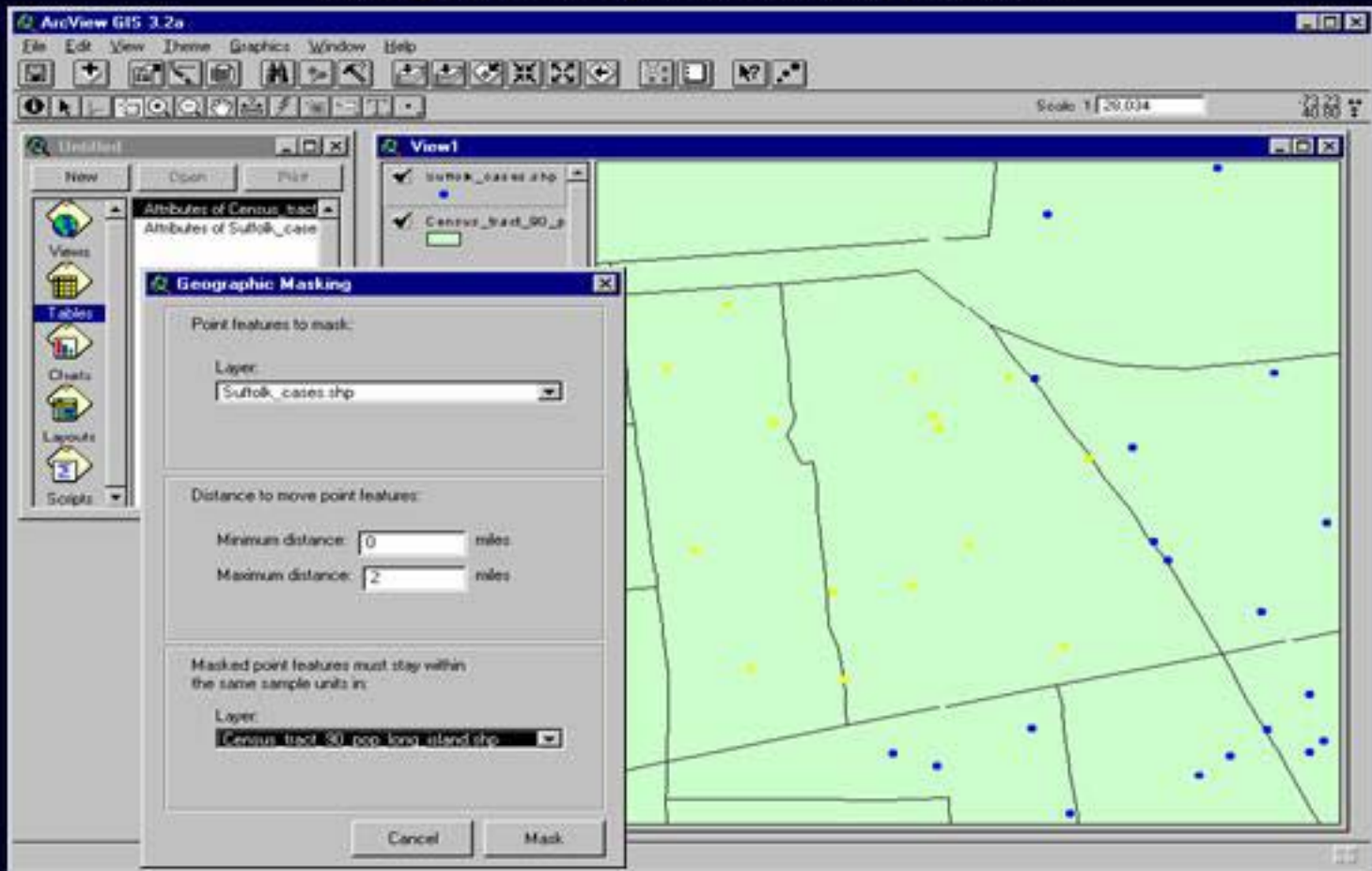
# Cluster Analysis

Checking for clusters of sample cases  
uses SatScan software as cluster analysis engine.

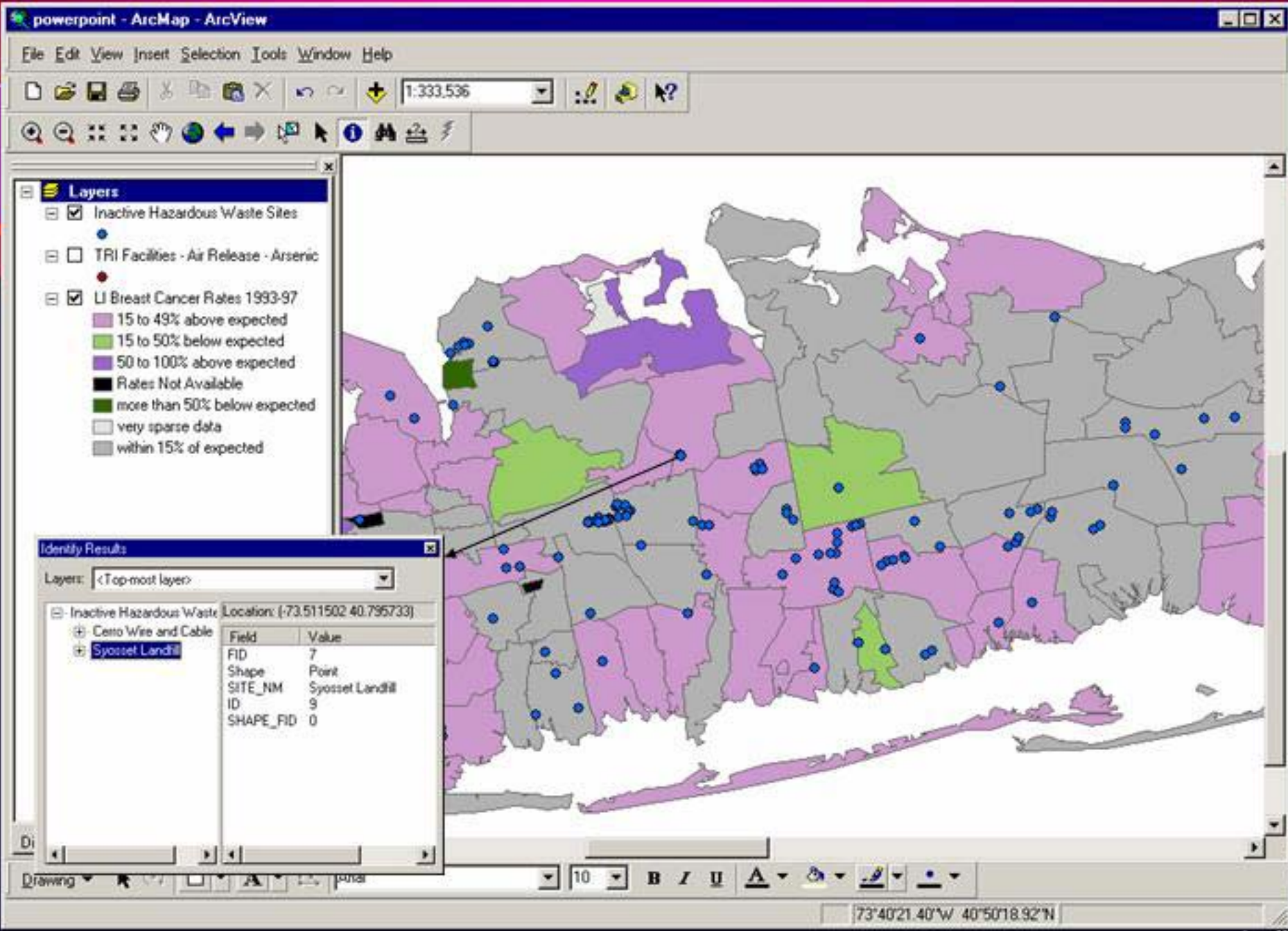


# Geographic Masking

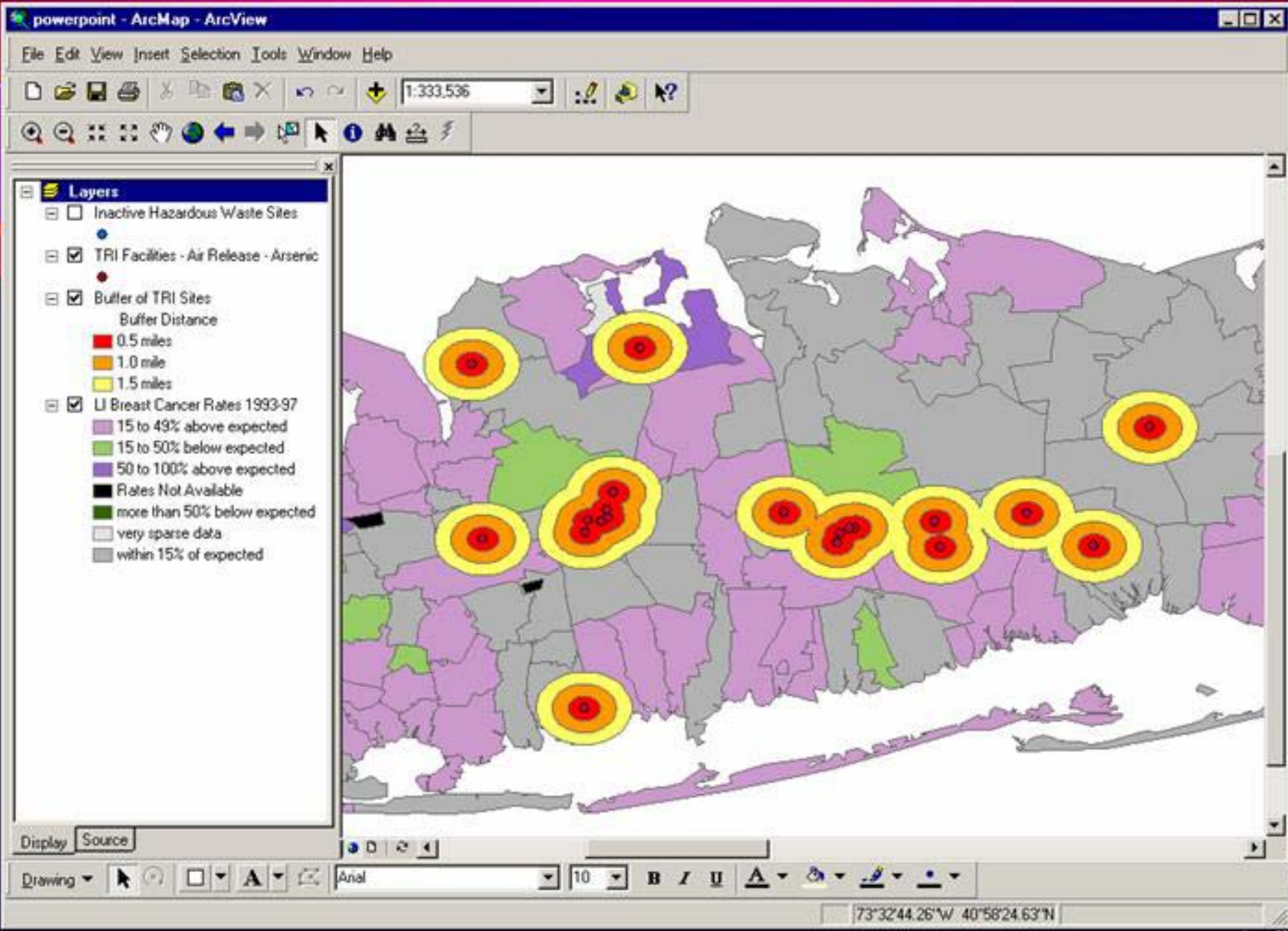
Masking selected (in yellow) sample cases using random perturbation method.











## *What is Availability of GIS-H?*

- ▶ Available now to researchers with approved projects
- ▶ Public mapping features available soon



## *In Summary, the GIS-H is ...*

- ▶ **Comprehensive, integrated data warehouse (> 80 datasets)**
- ▶ **Flexible and expandable**
- ▶ **Can integrate external datasets**
- ▶ **Sophisticated researcher's toolbox**
- ▶ **Community input and access**
- ▶ **Systematic attempt to include high quality data, comprehensive metadata**
- ▶ **A prototype and resource for future studies**



*Apply!*



- ▶ **Access to researcher site is limited to investigators with approved protocols**
- ▶ **For additional information, visit GIS-H website**

[www.healthgis-li.com](http://www.healthgis-li.com)

*Inquiries may be directed to:*

- ▶ **Burdette (Bud) Erickson, M.Sc.**
- ▶ **301.435.4913**
- ▶ **[berikso@mail.nih.gov](mailto:berikso@mail.nih.gov)**





# Women and Heart Disease

An Atlas of Racial and Ethnic  
Disparities in Mortality  
Second Edition

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Michele L. Casper

Elizabeth Barnett

Joel A. Halverson

Gregory A. Elmes

Valerie E. Braham

Zainal A. Majeed

Amy S. Bloom

Shaun Stanley

**CDC**  
Centers for Disease Control and Prevention

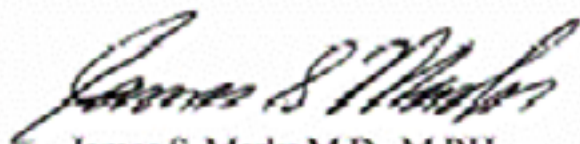


**West Virginia University**  
ROBERT C. BEALL HEALTH SCIENCE CENTER

Office for Social Environment and Health Research • West Virginia University  
National Center for Chronic Disease Prevention and Health Promotion • Centers for Disease Control and Prevention

**“Men and Heart Disease indicates where those programs are most needed and can have the greatest benefit.**

**It is my hope that Men and Heart Disease: An Atlas of Racial and Ethnic Disparities in Mortality will be used to guide the distribution of funds and resources to those communities of men experiencing excess mortality from heart disease and will promote the development of culturally sensitive prevention strategies.”**



James S. Marks M.D., M.P.H.

Director, National Center for Chronic Disease Prevention and Health Promotion  
Centers for Disease Control and Prevention



## In *Direct* Adjustment:

- Locally measured disease rates are applied to a standard population
- “The formula calls for computing the weighted average of the age-specific death rates in a given area, using as weights the age distribution of the standard population.” M&M Demography p. 419
- This answers the question: given the age-sex specific disease rates of this area, how many more (or less) incidences would occur in this area if it had the demographic characteristics of the standard population.

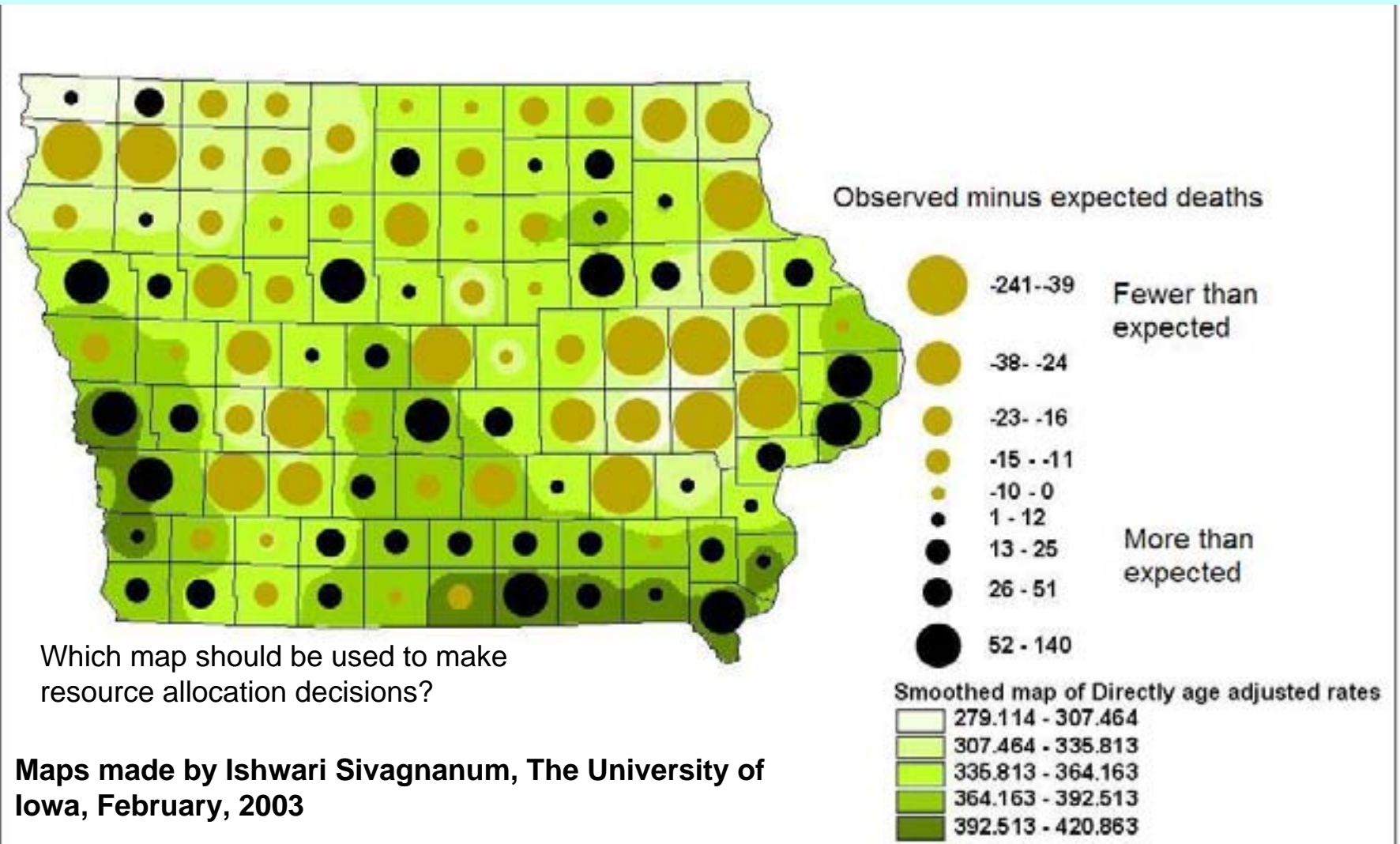
## **In *indirect* adjustment**

- **In indirect adjustment, national or regional rates are applied to the demographics of each small area.**
- **The ratio of actual/expected incidences is the standardized disease incidence rate of the area.**
- **This answers the question, “given the demographics of this area, how many incidences of the disease would we expect to find?”**

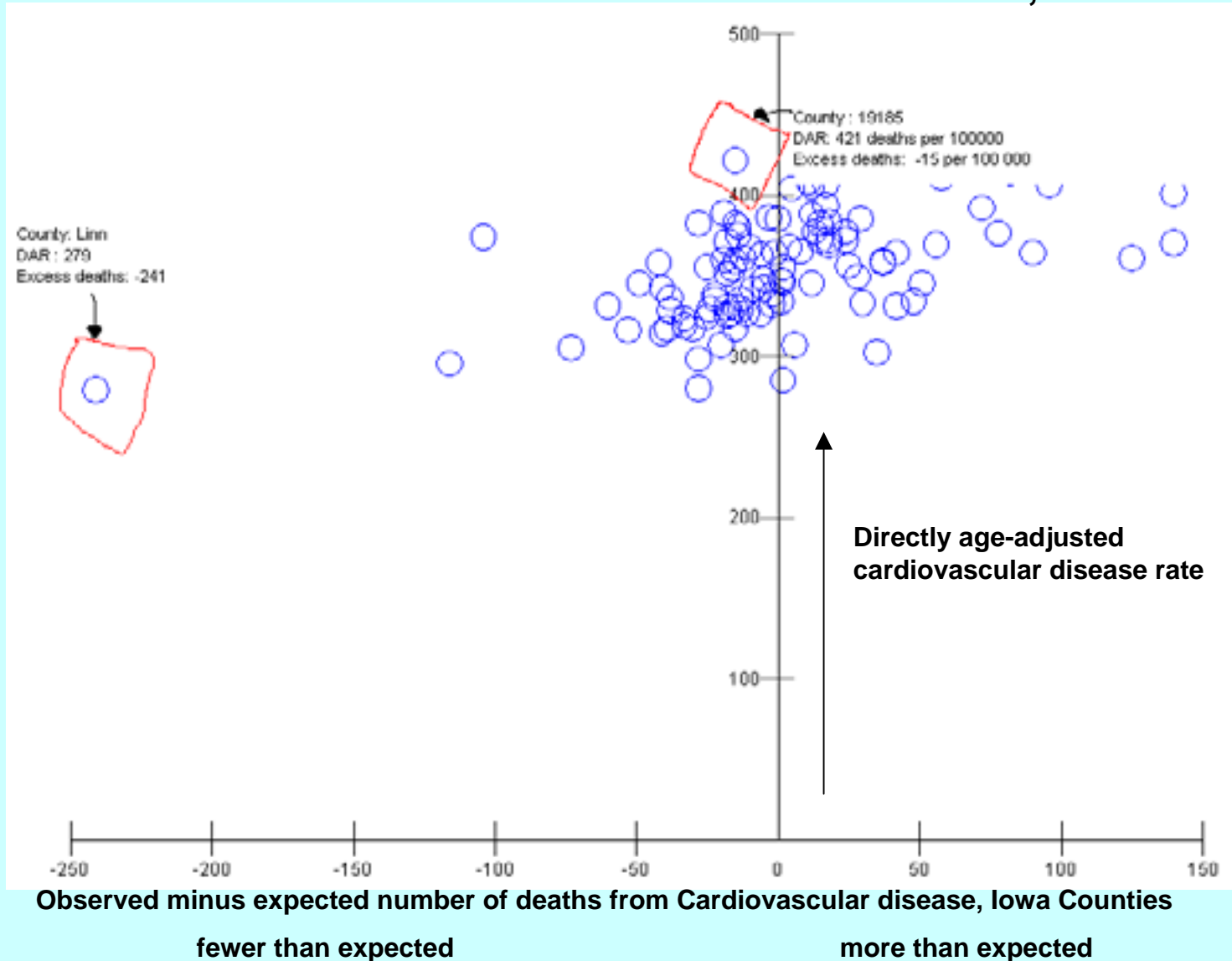
The Green Map is the Cardiovascular Death Rate in Iowa, 1995 - 2000.

(Directly Adjusted Rates--smoothed)

Circles show the difference between the observed and the expected number of deaths (from Indirectly Age Adjusted Rates). Black is excess deaths—greater than expected; Brown is fewer than expected number of deaths.



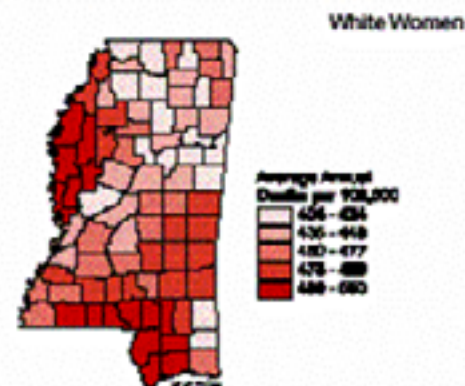
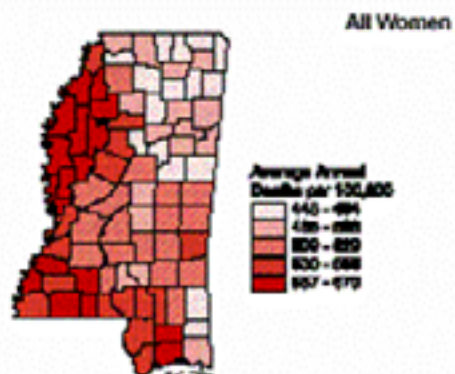
# Relationship Between Direct Age-Adjusted Cardiovascular Disease Rates and the difference between observed and expected numbers of deaths from cardiovascular disease. Iowa counties, 1995 - 2000





## Mississippi

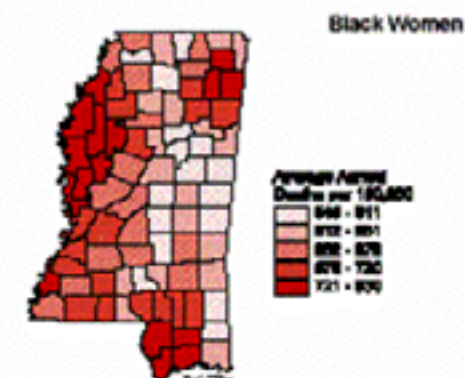
### Smoothed County Heart Disease Death Rates, 1991-1995



### State Profile — Mississippi

Race or Ethnicity	State Population 1995	State Heart Disease Death Rate, 1991-1995*
All Women	690,726	531
American Indian and Alaska Native Women	1,720	337
Asian and Pacific Islander Women	1,434	Insufficient Data
Black Women	206,726	696
Hispanic Women	3,600	Insufficient Data
White Women	480,836	472

\* Average annual age-adjusted rate (deaths per 100,000) for women ages 35 years and older. Data for Hispanics was also included within each of the four categories of race.



## Indirectly adjusted rates can be mapped for smaller spatial units than directly adjusted rates

- the indirect rate involves multiplying national or statewide rates by local age-sex demographics;
- direct rates involve multiplying local age-sex rates by national or statewide demographics.
- the indirect rate involves multiplying stable disease rates by stable population totals;
- the direct rate involves multiplying unstable disease rates by stable population totals.



# Alternative Data Models for GIS and Health

It is often necessary to work simultaneously with all three data models

