

Quantifying the Health Impacts of Air Pollution

Day 3: Mapping Risks

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OUTLINE

Spatial Epidemiology

Disease Mapping

Statistical Methods for Smoothing Risks

Spatial Epidemiology

WHAT IS SPATIAL EPIDEMIOLOGY?

- ▶ Epidemiology is the study of the distribution of diseases in populations.
- ▶ Disease risk depends on the person (genetics/behaviour), place and time.
- ▶ Spatial epidemiology focuses on the second of these.
- ▶ Place is a surrogate for exposures present at that location
 - ▶ environmental exposures in water/air/soil
 - ▶ lifestyle characteristics of those living in particular areas.

GROWING INTEREST IN SPATIAL EPIDEMIOLOGY

- ▶ Public interest in effects of environmental hazards/pollution.
- ▶ Epidemiological interest in differences in disease rates across different areas.
- ▶ Data availability: collection of health data at different geographical scales.
- ▶ Increase in computing power and methods
 - ▶ Geographical Informations Systems (GIS).
- ▶ Development of statistical/epidemiological methods for investigating disease 'clusters'.

THE NEED FOR SPATIAL METHODS

- ▶ Many epidemiological studies are spatial
 - ▶ many are spatio-temporal!
- ▶ When do we need to 'worry'?
 - ▶ are we explicitly interested in the spatial pattern of disease incidence?
 - ▶ disease mapping
 - ▶ cluster detection.
 - ▶ is the clustering a nuisance quantity that we wish to acknowledge, but are not explicitly interested in?
 - ▶ spatial regression.

TYPES OF SPATIAL DATA

- ▶ **Point data**
 - ▶ 'exact' residential locations exist, e.g. for cases and controls.
- ▶ **Count data**
 - ▶ aggregation
 - ▶ typically over administrative units.

Disease Mapping

OVERVIEW OF DISEASE MAPPING

- ▶ The estimation and presentation of summary measures of health outcomes.
- ▶ The aims of disease mapping include
 - ▶ simple description
 - ▶ hypothesis generation
 - ▶ allocation of health care resources, assessment of inequalities
 - ▶ estimation of background variability in underlying risk in order to place epidemiological studies in context.
- ▶ Unfortunately there are well-documented difficulties with the mapping of raw estimates since, for small areas and rare diseases in particular, these estimates will be dominated by sampling variability.

EXAMPLE: COPD IN ENGLAND

- ▶ Incidence rates of hospital admissions for Chronic Pulmonary Obstructive Disease (COPD) in 324 local authorities of England, between in 2001–2010.
- ▶ Data:
 - ▶ observed and ‘expected’ number of cases (based on the local authority age-sex profiles)
 - ▶ this allows the calculation of the standardised morbidity ratio (ratio of the observed to the expected cases).

EXPECTED NUMBERS

- ▶ The expected number of deaths/disease are calculated using *indirect* standardisation.
- ▶ Rates from a reference population are applied to the population of interest.
- ▶ The expected number

$$E = \sum_k N_k r_k$$

where r_k is the rate in the reference population and N_k is the population in the study. Commonly, k would denote age–sex categories.

STANDARDISED MORTALITY/MORBIDITY RATIOS

- ▶ The observed number of deaths can be compared to the expected number using the standardised mortality/morbidity ratio (SMR).
- ▶ Let O be the observed number of deaths/disease cases in the population of interest, and E be the expected number.

$$\text{SMR} = \frac{O}{E}$$

- ▶ An SMR of 1 means that the rates of death/disease in the population of interest are the same as in reference population.
- ▶ If it is greater than 1, we have more deaths/disease cases than expected; if it is less than 1 we have less.

EXAMPLE: COPD IN ENGLAND

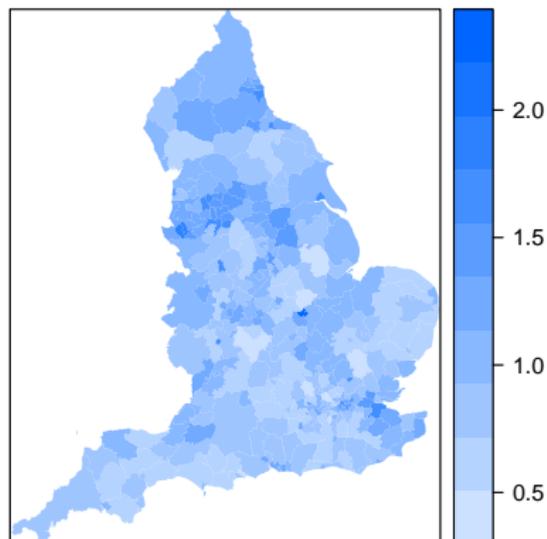


Figure: SMRs for hospital admissions of COPD in 324 local authorities in England.

NON-STATISTICAL ISSUES

- ▶ There is a trade-off when a geographical scale is chosen
 - ▶ larger geographical areas provide more stable rates
 - ▶ risks may be distorted due to the large aggregation of individuals.
- ▶ The size of the areas chosen also determines the sort of questions that can be posed
 - ▶ smaller areas often mean greater contrasts in relative risks and exposures
 - ▶ localized effects can only be detected with data at a smaller level of aggregation
 - ▶ high (and low) relative risks will be diluted under aggregation
 - ▶ if the relative risk shows marked variation within a particular area this information may be lost under aggregation.

STATISTICAL ISSUES

- ▶ Unfortunately there are well-documented difficulties with the mapping of raw estimates of risk.
- ▶ For small areas and rare diseases in particular, estimates will be dominated by sampling variability.
- ▶ The SMR is given by

$$\text{SMR}_i = \frac{O_i}{E_i}$$

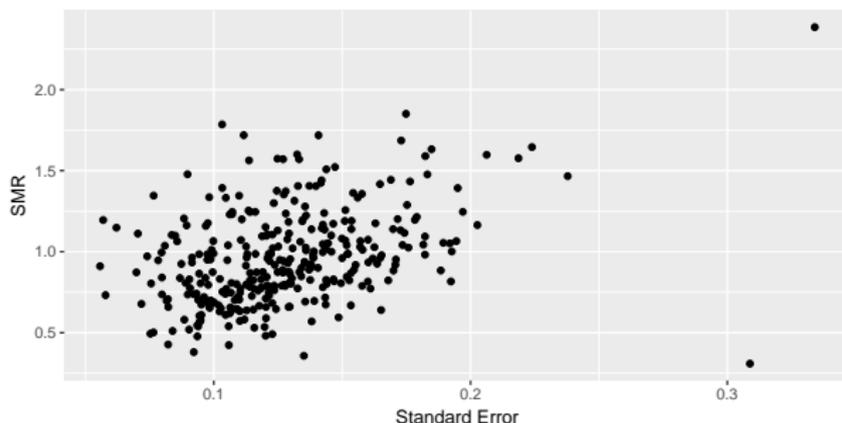
and it has variance

$$\text{var}(\text{SMR}_i) = \frac{\text{SMR}_i}{E_i} = \frac{O_i}{E_i^2}$$

therefore areas with smaller expected numbers, E_i have higher variance.

EXAMPLE: SCOTTISH LIP CANCER

- ▶ For COPD in England, the expected numbers are highly variable, with range 3.2–368.9.
- ▶ Are the extreme SMRs based on small expected numbers?
 - ▶ many of the large, sparsely populated, rural areas in the north have high SMRs.



SMOOTHING MODELS

- ▶ The above considerations have led to the development of models to *smooth* SMRs.
- ▶ These approaches use data from surrounding areas to improve the stability of estimates.

EXAMPLE: COPD IN ENGLAND

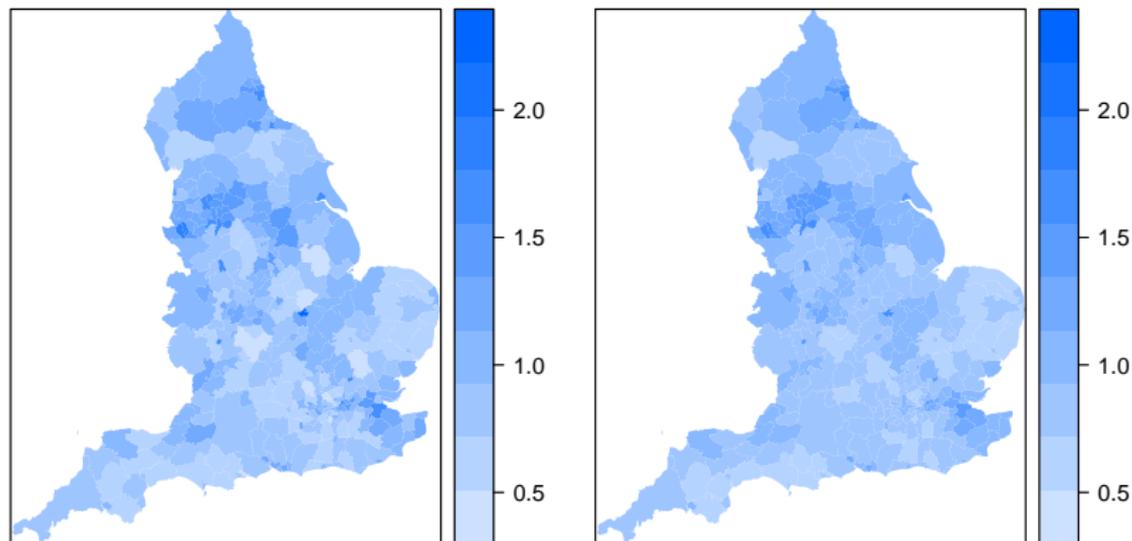


Figure: (Left) Unsmoothed SMRs and (Right) Smoothed SMRs for hospital admissions of COPD in England.

Statistical Methods for Smoothing Risks

SMOOTHING

- ▶ Defined in terms of 'neighbouring areas'.
- ▶ Often defined as when areas share a *common boundary*.
- ▶ May use an intrinsic conditional autoregressive (ICAR) distribution.

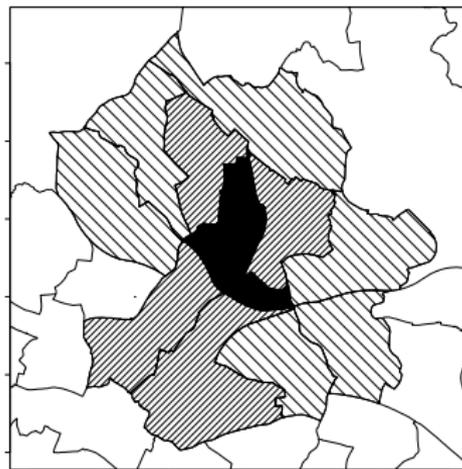


Figure: Close-up of a region within Birmingham, UK.

BAYESIAN ANALYSIS

- ▶ Spatial smoothing models are not easily fit using traditional techniques, and so Bayesian methods are commonly used.
- ▶ Unfortunately, Bayesian models are not available in most standard software packages.
- ▶ Markov chain Monte Carlo (MCMC)
 - ▶ Win/OpenBUGS – A package that allows very general Bayesian modelling
 - ▶ GeoBUGS – Module of WinBUGS that contains spatial models and mapping facilities
 - ▶ R – Packages for example CARBayes.
- ▶ For fast computation on big datasets
 - ▶ R-INLA.

EXAMPLE: COPD IN ENGLAND

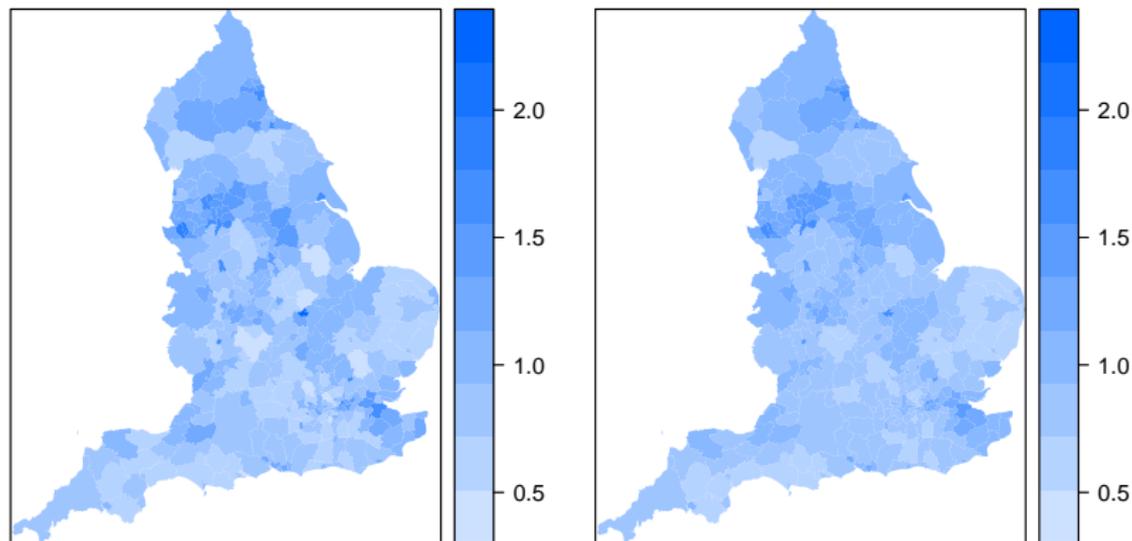


Figure: (Left) Unsmoothed SMRs and (Right) Smoothed SMRs for hospital admissions of COPD in England.

SUMMARY OF SMOOTHING IN DISEASE MAPPING

- ▶ The aim is to reduce the inherent instability in SMRs that are based on small expected numbers (from small populations).
- ▶ This is achieved by fitting a smoothing model.
- ▶ Comparing raw and smoothed SMRs is important.
- ▶ If there are big changes, are they appropriate?
 - ▶ were the expected numbers small?
 - ▶ can inform whether the statistical model was sensible or not.

