

# Small Area Analyses

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# Aim

- To provide an oversight of the outcomes of a programme of integrated projects focussed on the derivation of small area indicators of health need

# Small Area Indicators

- People live in areas
- Agencies have area-based responsibilities
- BUT
- Home and work
- Changing boundaries
- Non-bounded processes
- No routine data available

# Method: Synthetic Estimation

- Method to estimate unknown area-level measures based on other known variables
- Unknown measures: health themes
- Known variables: census data, marketing data, sample data, administrative data...typically large national surveys

# Method: Synthetic Estimation

- Why needed?
  - No alternative
  - Survey sample sizes too small for direct estimates
  - Survey samples clustered so no data for large % of areas

# Method: Synthetic Estimation

- Crosstabulation approach
  - Step 1: derive crosstabulation from national survey data
  - Step 2: apply cell percents from step 1 to small area crosstabulations of (usually) census data
- Limitations
  - limited number of limited cross-tabulations available
  - Erases geography
- Advantages:
  - computationally simple
  - transparent

# Method: Synthetic Estimation

- Modelling approach
  - Step 1: model relationship between outcome and predictor variables in a survey
  - Step 2: apply coefficients from step 1 to known small area counts of predictors from (usually) census data
- Individual predictors
  - computationally equivalent to the crosstabulation approach
  - same problems
- Area predictors
  - aggregate survey results to lower scale
  - same problems

# Method: Synthetic Estimation

- Multilevel approach
  - Both individual and area level predictors
  - Area predictors in a multilevel framework
  - Apply coefficients to known small area counts of predictors from (usually) census data
- Advantages:
  - Both individual and area-level factors affect health outcomes
  - Hierarchical nature of survey data lends itself to a multilevel model
- Disadvantages:
  - complexity

# Method: Synthetic Estimation

- Or... just use a proxy indicator of poor-health, such as a deprivation index
- Choice of model:  $f(\text{complexity, data, objectives})$
- Note: estimates: ‘risk’ measures, ‘probable prevalence given...’; synthetic: educated best guess

# Two examples

- Environmental health outcome
  - Exposure to ETS
- Environmental health predictor
  - Air pollution and chronic respiratory disease

# Exposure to ETS

- NHS Health Scotland
- Scottish Health Survey
- Sample: 16,915
- Non-smokers aware of being exposed to environmental tobacco smoke
  - Covers range of settings
- Objective: small area estimates

# Exposure to ETS

```
nssetsawareijk ~ Binomial(denomijk, πijk) }  
nssetsawareijk = πijk + ε0ijkbconsijk  
logit(πijk) = βijkcons + 0.23172(0.04452)maleijk + -0.23673(0.04601)singleijk + 0.11505(0.17051)alijk + -0.14914(0.04810)a3ijk +  
-0.28453(0.05057)a4ijk + -0.37483(0.05118)a5ijk + -0.63531(0.06766)a6ijk + -0.01644(0.06536)smijk +  
0.35047(0.17697)salijk + -0.00481(0.00406)rm6crijk + 0.00156(0.00511)s.rm6crijk + -0.01048(0.00486)m.rm6crijk +  
0.01591(0.00729)sm.rm6crijk + -0.01766(0.00542)a6.rm6crijk + -0.00490(0.00591)tenprcrijk +  
-0.04803(0.00956)al.tenprcrijk + -0.00416(0.00218)tenlahacrijk + -0.00707(0.00303)a1.tenlahacrijk +  
-0.00418(0.00254)sc3m45crijk + 0.00741(0.00328)a5.sc3m45crijk + -0.04615(0.01030)scnotcrijk +  
0.04333(0.01304)s.scnotcrijk + 0.03958(0.01307)m.scnotcrijk + -0.04399(0.01901)sm.scnotcrijk +  
0.02948(0.00911)a3.scnotcrijk + 0.02258(0.00887)munempcrijk + -0.03360(0.01034)s.munempcrijk +  
-0.03872(0.01062)m.munempcrijk + 0.04339(0.01509)sm.munempcrijk + -0.01201(0.00528)a34.munempcrijk +  
0.01071(0.00447)kidscrijk + -0.01851(0.00488)s.kidscrijk  
βijk = -0.17209(0.05059) + νijk + uijk
```

Prevalence decreases with age (the youngest married males are an exception). Since prevalence of Current Smoking does *not* increase overall with age, this decrease is largely a measure awareness or exposure, not of a decreasing pool of non-smokers.

# Exposure to ETS

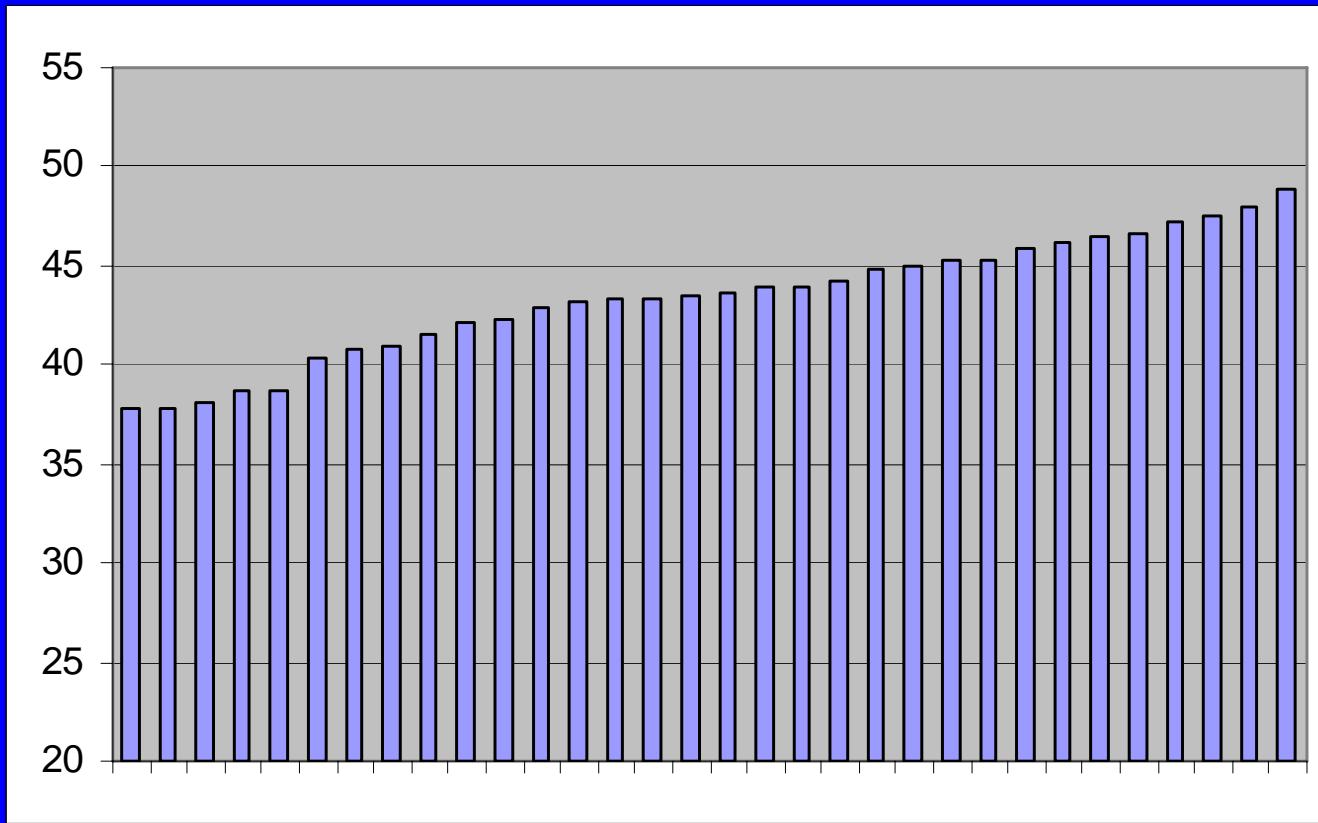
$$[v_{jk}] \sim N(0, \Omega_v) : \Omega_v = [0.01416(0.00711)]$$

$$[u_{jk}] \sim N(0, \Omega_u) : \Omega_u = [0.05390(0.01162)]$$

$$\text{bcons}_{jk}^* = \text{bcons}_{jk} [\pi_{jk}(1 - \pi_{jk})/\text{denom}_{jk}]^{0.5}$$

$$[e_{ijk}] \sim (0, \Omega_e) : \Omega_e = [1.00000(0.00000)]$$

# Exposure to ETS



LA Scale

# Air Pollution and CRD

- Multilevel Synthetic Estimation
- Consolidated 1998-2004 database
- 95,000 respondents aged 20+
  - HSfE, SHS, NIH&SCS
  - census
  - CRD as a function of (1) age, sex, ethnicity; (2) socioeconomic, PM10

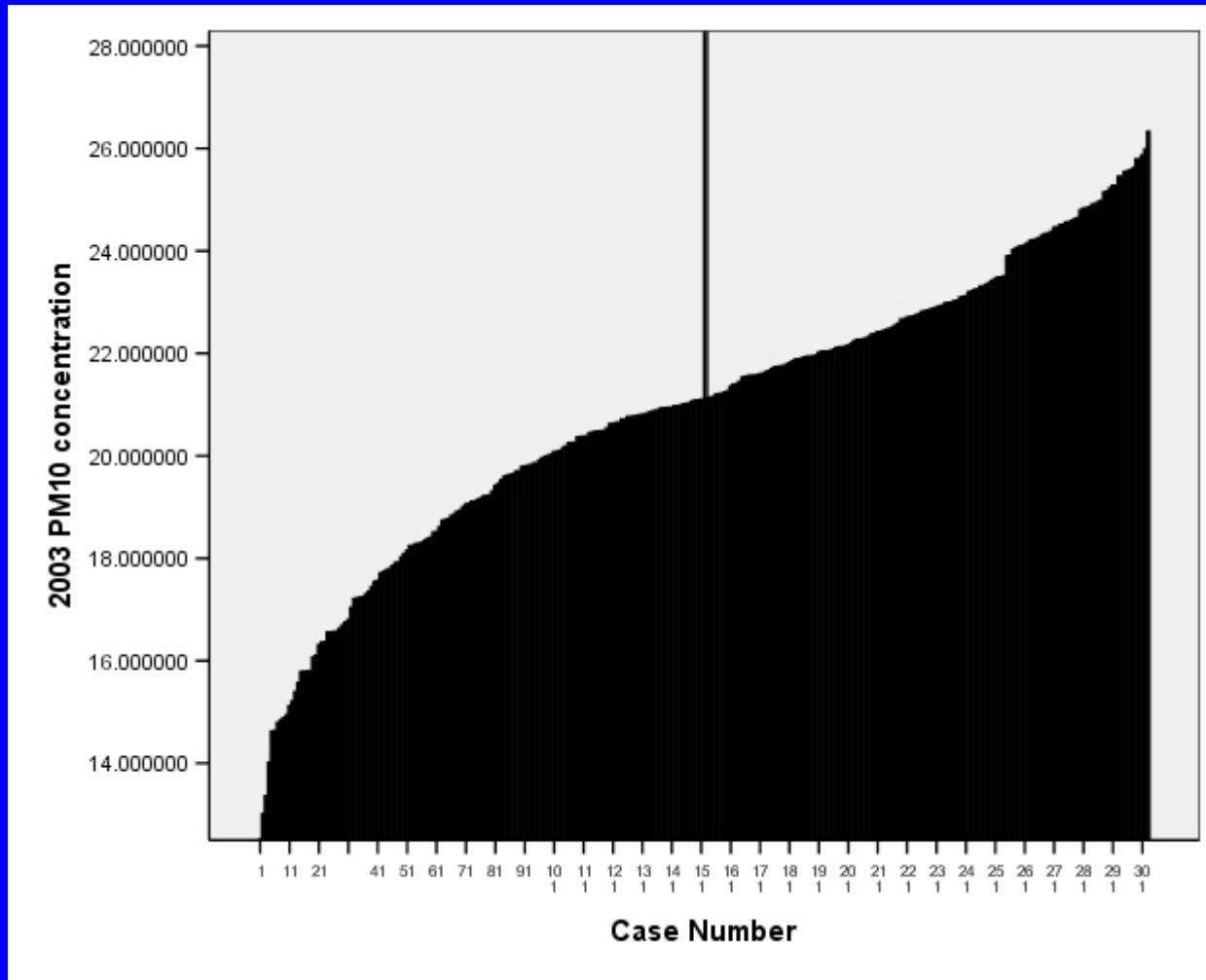
# Air Pollution and CRD

- PM10 data
  - modelled an overall PM10 profile across scattered empirical observation sites.
  - Estimated PM10 concentrations for 1km grid derived from modelled profile
  - These were aggregated to Lower Special Output Areas (in England)

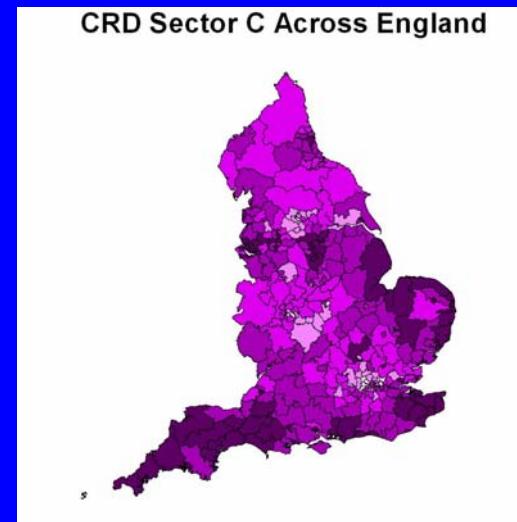
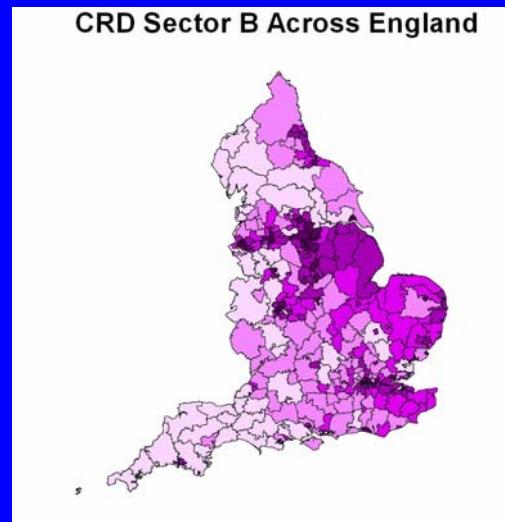
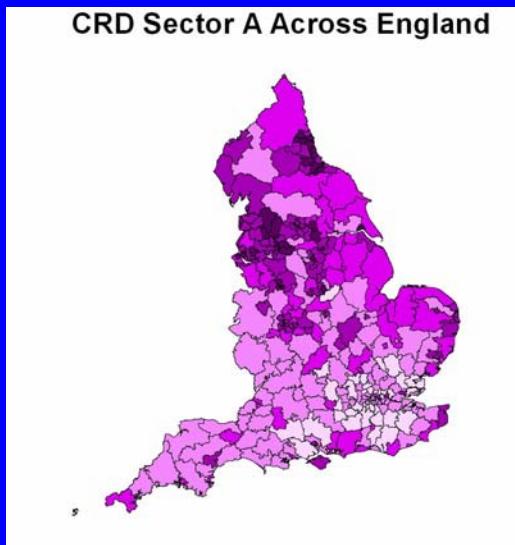
# Air Pollution and CRD

- PM10 data
  - We can only work with PCO data
  - Aggregate LSOA to PCO
  - 107 LSOAs/PCO in England
  - Population versus area weighting
  - Chose population weighting

# Air Pollution and CRD



# Chronic Respiratory Disease



- A: diagnosed emphysema, asthma or bronchitis
- B: Symptoms: undiagnosed but smokes and wheezes or smokes and poor lung function
- C: Signs: Undiagnosed but smokes

# Conclusions

- Major utility in
  - health profiling
  - care planning
- Clear advantages in absence of other data