

A Socio-Economic Mortality Index for Scotland

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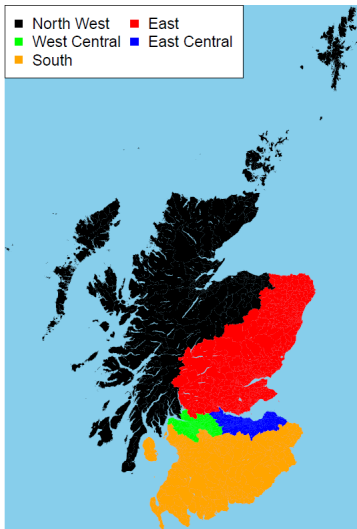


- Motivation and objectives
- Inequalities in mortality rates
- Data
- Methodology: Local linear regression
- Results
- Local sensitivity
- Summary

Objectives and motivation

- To develop a customised mortality index for Scotland
- Mortality heterogeneity: mortality varies between subgroups.
Main reason: prevalence of causal risk factors varies between groups (e.g. smoking)
- What are the most significant **socio-economic factors** that are associated with mortality rates?
Note: relative significance varies with age
- How much **inequality** is there in mortality rates at different ages?
- Prior work: Longevity Index for England (LIFE)
Objective: to extend this to the rest of the UK
Focus here on Scottish mortality

Background: Relative mortality by region



Actual/Expected Deaths
Variation by Region
Ages 70-89, Years 2011-19

Region	Regional Relative To National Mortality (%)	
	Males	Females
Scotland	100	100
North West	90.5	88.9
East	95.0	95.0
West Central	113.6	109.6
East Central	96.2	96.7
South	98.9	102.6

Values show **actual deaths** (ages 70-89) by region as a **percentage of expected deaths** using national age-specific death rates

Background: Relative mortality by deprivation quintile (SIMD)

Actual/Expected Deaths Variation by Region Ages 70-89, Years 2011-19		
Region	Males	Females
North West	90.5	88.9
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South	98.9	102.6

Actual/Expected Deaths Variation by Deprivation (SIMD) Quintile Ages 70-89, Years 2011-19		
Quintile	Males	Females
1	134.0	128.9
2	113.1	109.1
3	98.2	98.4
4	87.8	89.6
5	75.3	76.3

Regional variation < variation by Scottish Index of Multiple Deprivation (SIMD)

- West Central has higher levels of deprivation than other regions
⇒ 'naturally' higher mortality rates
- But is there an additional regional effect in addition to deprivation?
Or does deprivation fully explain regional differences?

Does deprivation explain all variation in mortality by region

- West Central has higher levels of deprivation
⇒ 'naturally' higher mortality rates
- But is there an additional regional effect in addition to deprivation?
Or does deprivation fully explain regional differences?
- **Four potential hypotheses:**
 - H_0 : year and age explains everything at the **neighbourhood** level
⇒ $m(t, x)$ national death rates (REJECT!)
 - H_1 : year + age + deprivation (SIMD)
 - H_2 : year + age + **more general socio-economic data**
 - H_3 : none of the above!

Data: Neighbourhoods: “LSOA”s

- Small neighbourhoods:
“Data Zones”: DZ
- $L = 6976$ DZs
- Socio-economically homogeneous
- Average size ≈ 780 persons
- DZs $i = 1, \dots, L$,
single years ($t = 2011-2019$),
single or 5-year age groups, x :
 - Deaths: $D(i, t, x)$ (5-yr age groups) (by normal place of residence)
 - Exposures: $E(i, t, x)$ (single ages)
- Many potential predictive variables, $\Rightarrow$ vector $X(i)$

Methodology

- $D(i, t, x)$, $E(i, t, x)$ deaths and exposures by DZ
- National death rates (all t and x)

$$m(t, x) = \frac{D(t, x)}{E(t, x)} \quad (\text{single ages})$$

- Neighbourhood death rates ($i = 1, \dots, L$)
 - local death rates: $m(\mathbf{i}, t, x)$
 - how to model $m(i, t, x)$?
- Predictive variables $X(i)$ plus $m(t, x) \longrightarrow m(i, t, x)$
- Method: Local Linear Regression (LLR); Random Forest; XGboost; ...

Methodology: General Approach

- Over a limited age range (e.g. 70-89); and
- Over a limited range of years (e.g. 2011-2019):

neighbourhood i ,
$$m(i, t, x) = m(t, x)F_1(i)$$

- $F_1(i)$ = neighbourhood-specific *relative risk* due to socio-economic and other non-spatial characteristics, $X(i)$
Estimate $F_1(i)$ using **Local Linear Regression**

Local Linear Regression (LLR): in brief

- i = neighbourhood index
- $D(i)$ = actual deaths
- $\hat{D}_0(i)$ = expected deaths based on national mortality, $m(t, x)$
- LLR: Locally regress
 - $Y_0(i) = D(i)/\hat{D}_0(i)$ response variable against
 - $\tilde{X}(i) = (X(i, 1), \dots, X(i, M))^T$ predictive variables
 - Local weights $w(i, j)$ depend on proximity of $\tilde{X}(j)$ to $\tilde{X}(i)$
- $F_1(i) = E[Y_0(i)|\tilde{X}(i)]$
- $\hat{F}_1(i) = \hat{a}(i) + \hat{b}(i)^T \tilde{X}(i)$

Feature Selection/Predictive Variables: what to include in $\tilde{X}(i)$

Sources of data available at the neighbourhood level:

- Scottish Index of Multiple Deprivation (SIMD)
 - Various domains (e.g. employment)
- Census 2022
 - Non-sensitive data publicly available
 - Based on census questions

Advantage of LLR in feature selection process:

- During feature selection process: set weight $w(i, i) = 0$
 $\Rightarrow$ each $D(i, t, x)$ is “out of sample”
- In theory: maximise Poisson log-likelihood
- In practice: different age groups; males and females
Want to choose the same set of predictive variables for each group

Choice of Predictive Variables: what to include in $\tilde{X}(i)$

- **SIMD:**

- Employment deprivation
- Education deprivation: Proportion of the population with no qualifications
- Crime rank (low rank $\Rightarrow$ high crime rate)

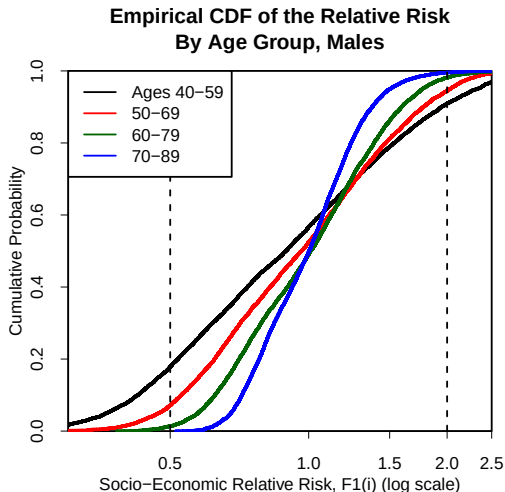
- **Census data:**

- Communal establishments: proportion of the population aged 65+ in a communal establishment (mostly care homes in the 65+ group)
- Proportion married: ages 35+
- Proportion widowed or divorced: ages 35+
- Proportion born in Scotland: all ages

- **2011-2019 population data:**

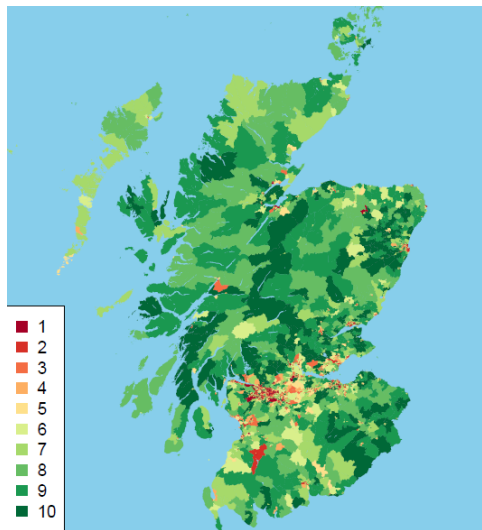
- Old-age dependency ratio: $\#$ 65+ / $\#$ 20-64; males and females separately
- Female to male ratio: ages 15+

Socio-Economic Relative Risk, $F_1(i)$: Significant Inequality



- High levels of inequality
[Higher levels of inequality than England.]
- Inequality diminishes with age
- Age 40–59 main predictors:
unemployment, married, education
- Age 70–89 main predictors:
care homes, education, dependency ratio
- Males and females:
Females slightly less spread out
High correlation between male and female relative risks

Geographic Spread of Inequality: Males 60-79



- Males 60-79
- Deciles by estimated $F_1(i)$
- Decile 1 (dark red): high mortality
- Decile 10 (dark green): low mortality
- Decile 1
High mortality neighbourhoods
Mostly concentrated in dense urban areas
- Rural areas:
More green
Lighter mortality

Regional Deaths Adjusting for Socio-Economic Variables

Actual / Expected Deaths

$$m(i, t, x) = m(t, x)$$

Variation by Region

Ages 70-89, Years 2011-19

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North West	90.5	88.9
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Actual / Adjusted Expected Deaths

$$m(i, t, x) = m(t, x)F_1(i)$$

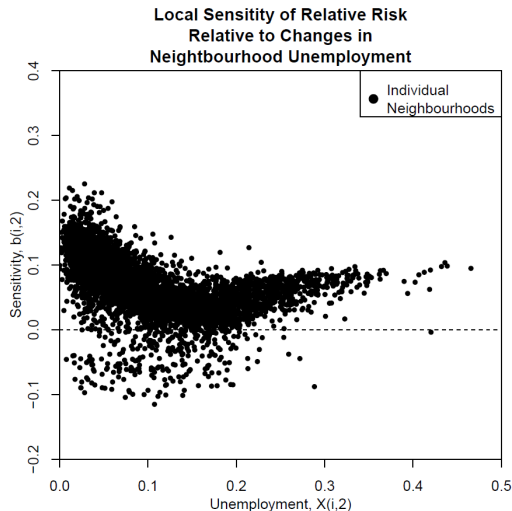
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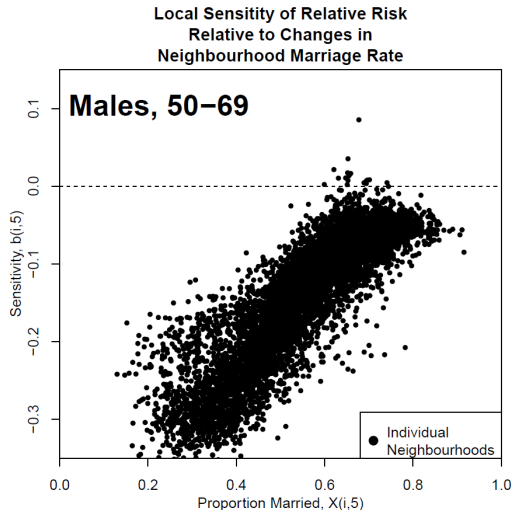
- Socio-economic modelling reduces significantly the variation between regions but it's not perfect

Local Sensitivity of Results to Specific Predictive Variables



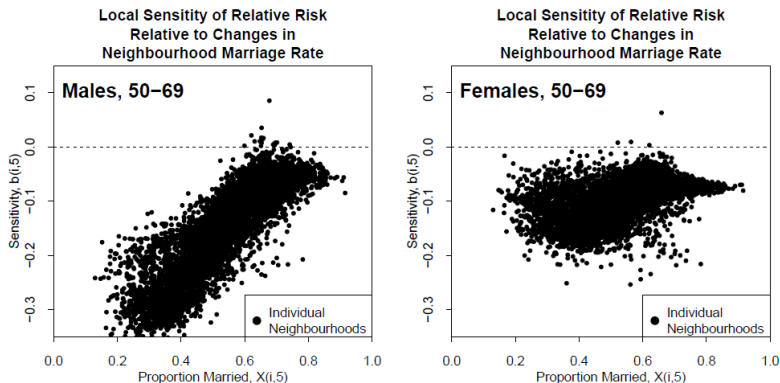
- $\hat{F}_1(i) = \hat{a}(i) + \sum_j b(i,j)X(i,j)$
- $b(i,j)$ measures the local sensitivity of $\hat{F}_1(i)$ to changes in $X(i,j)$
- Here: $j = 2$, unemployment
- Males, 60-79
- Consistent, mostly positive level
⇒ relative risk rises with unemployment
- But also noisy
 - Sampling variation in death counts
 - Genuine variation across the range of predictive variables (unemployment and other variables)

Sensitivity to Proportion Married



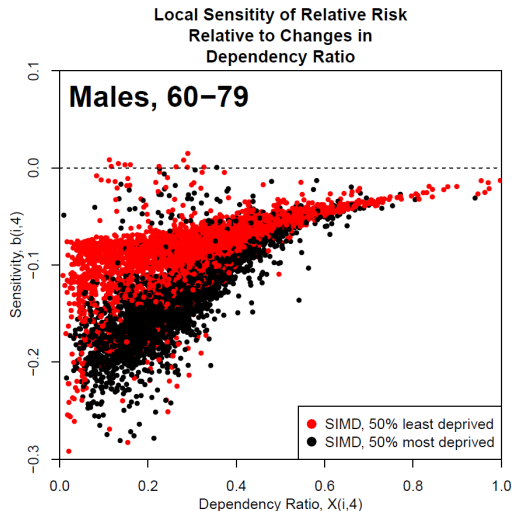
- Sensitivity to the proportion married (or in a civil partnership)
- Males 50-69
- Strong negative sensitivity
⇒ strong correlation between marriage and lower mortality
- Stronger influence in neighbourhoods with lower marriage rates
- Potential inference:
marriage is a possible predictor of a healthier lifestyle

Sensitivity to Proportion Married: Males vs Females



- Relative risk for females much less sensitive than males
- Inference:
 - Female mortality much less dependent on marital status
 - Male lifestyles more heavily influenced by marital status

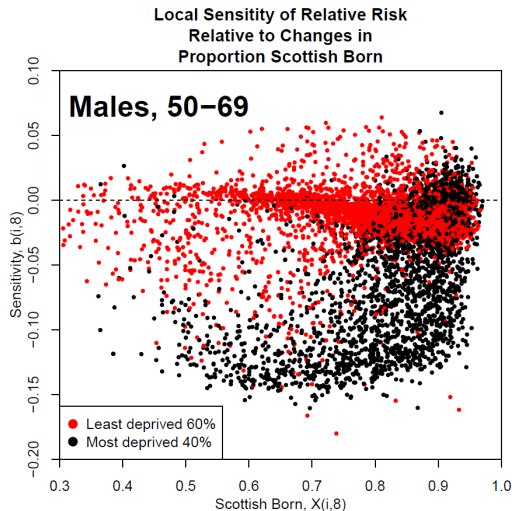
Sensitivity to Dependency Ratio, by Deprivation



$$\text{Dependency Ratio, DR} = \frac{\text{population 65+}}{\text{population 20-64}}$$

- Dependency ratio depends on:
 - high/low death rates
 - migration into/out of the neighbourhood as individuals get older
- Strong negative sensitivity
⇒ migration in is a strong predictor of lower mortality
- SIMD deciles (black/red):
The strength of the DR effect is very dependent on the level of general deprivation

Sensitivity to Proportion Born in Scotland



- **Least deprived 60%:**
Limited sensitivity to Scottish birth
- **Most deprived 40%:**
Significant effect
Higher proportion $\Rightarrow$ lower mortality
- This is the opposite of a healthy immigrant effect!
- But for females at higher ages sensitivity is positive

Discussion: Neighbourhood Effects and Identifiability

- Dependency ratio effect $\Rightarrow$ two explanations
 - Healthy people are more likely to move to “nice” or “healthy” areas on retirement $\Rightarrow$ bf individual “migrant” status
 - Age-related immigration and a larger population of 65+ creates a positive **environment** for retirees
- Similarly: proportion married
 - Is an **individual's marital status** a strong predictor?
 - Or is the **environment** in which they live a strong predictor?
- With neighbourhood data only:
we cannot differentiate between the two explanations
 $\Rightarrow$ we have an identifiability problem
- Reality (?):
Both individual status and environment are likely to be influential.

Summary

- Customised mortality index is much better at explaining mortality inequalities than the generic Scottish Index of Multiple Deprivation (SIMD)
 - Inequality between different socio-economic groups is very significant: greatest at younger ages; and narrows significantly with age
 - Key predictive variables depend on age group
 - Local Linear Regression $\Rightarrow$ complex dependencies between variables that need careful analysis and interpretation
- Sensitivities vary significantly within the predictive variable space
- Annuities: implications for pricing and reserving?

Thank You!

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Coming soon: **T**He **I**ndex of **S**cottish **L**ife **E**xpectancy – **THISLE**

