



SOCIETY OF ACTUARIES

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Modelling mortality by cause of death and socio-economic stratification: an analysis of mortality differentials in England



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Agenda

- ▶ Motivation
- ▶ Modelling mortality by cause of death (CoD)
- ▶ Modelling mortality by CoD and socio-economic stratification
- ▶ Case study: Mortality by deprivation in England
- ▶ Conclusions

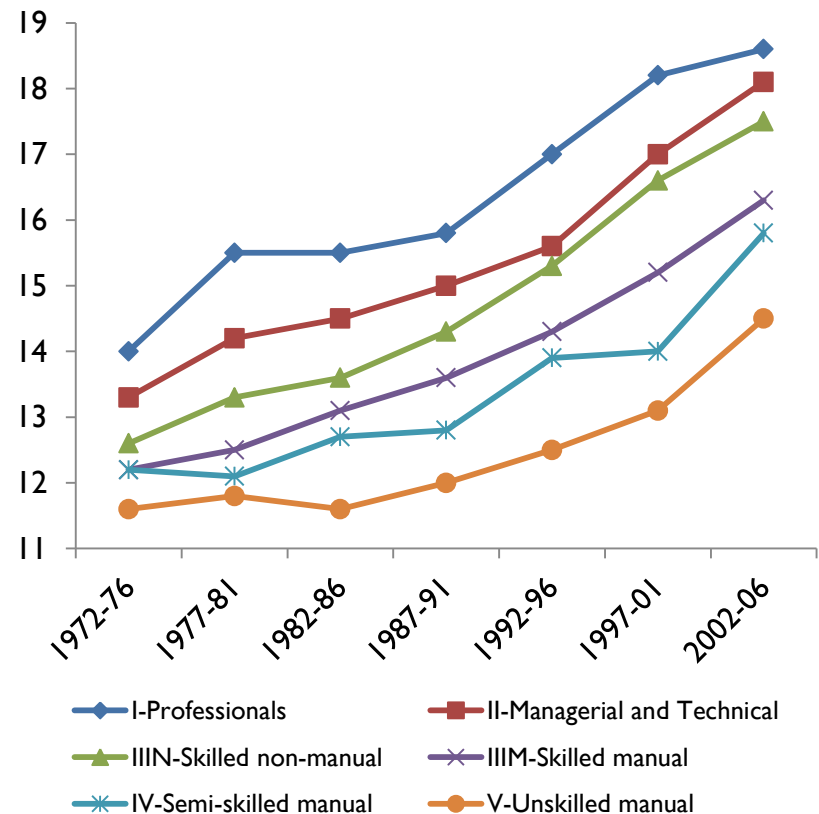


Motivation

Socio-economic differences in mortality

- ▶ Well-documented relationship between mortality and socioeconomic variables
 - ▶ Education
 - ▶ Income
 - ▶ Occupation
- ▶ Important implications on social and financial planning
 - ▶ Public policy for tackling inequalities
 - ▶ Social security design
 - ▶ Annuity reserving and pricing
 - ▶ Longevity risk management

Male life expectancy at age 65 by social class -England and Wales



Source: ONS Longitudinal Study

Motivation

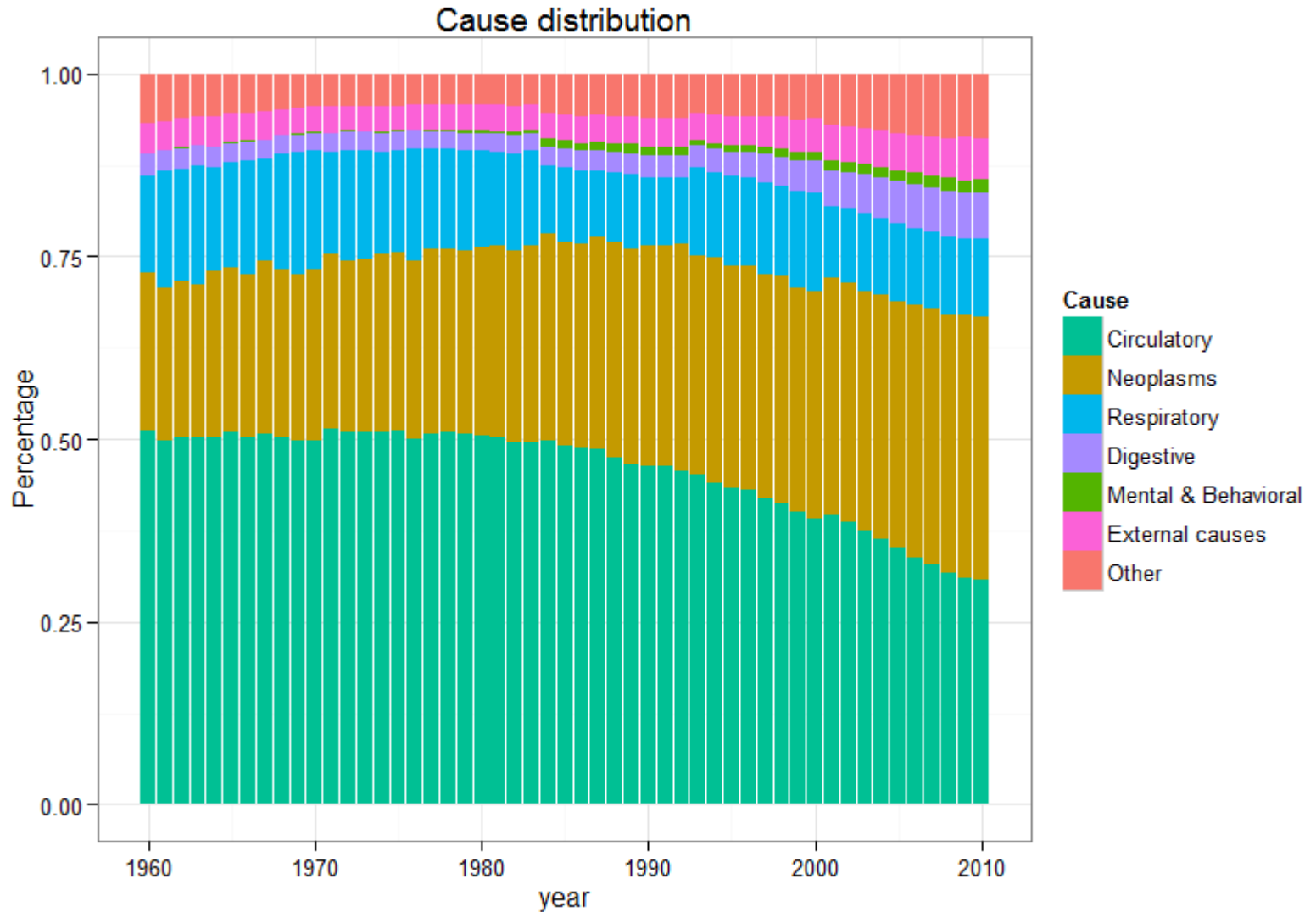
Cause-specific mortality

- ▶ Forecasts of cause-specific mortality required for many purposes
 - ▶ E.g Estimation of health care costs
- ▶ Inform the assumptions underlying overall mortality projections
- ▶ Shed light on the drivers of
 - ▶ Mortality change
 - ▶ Mortality differentials



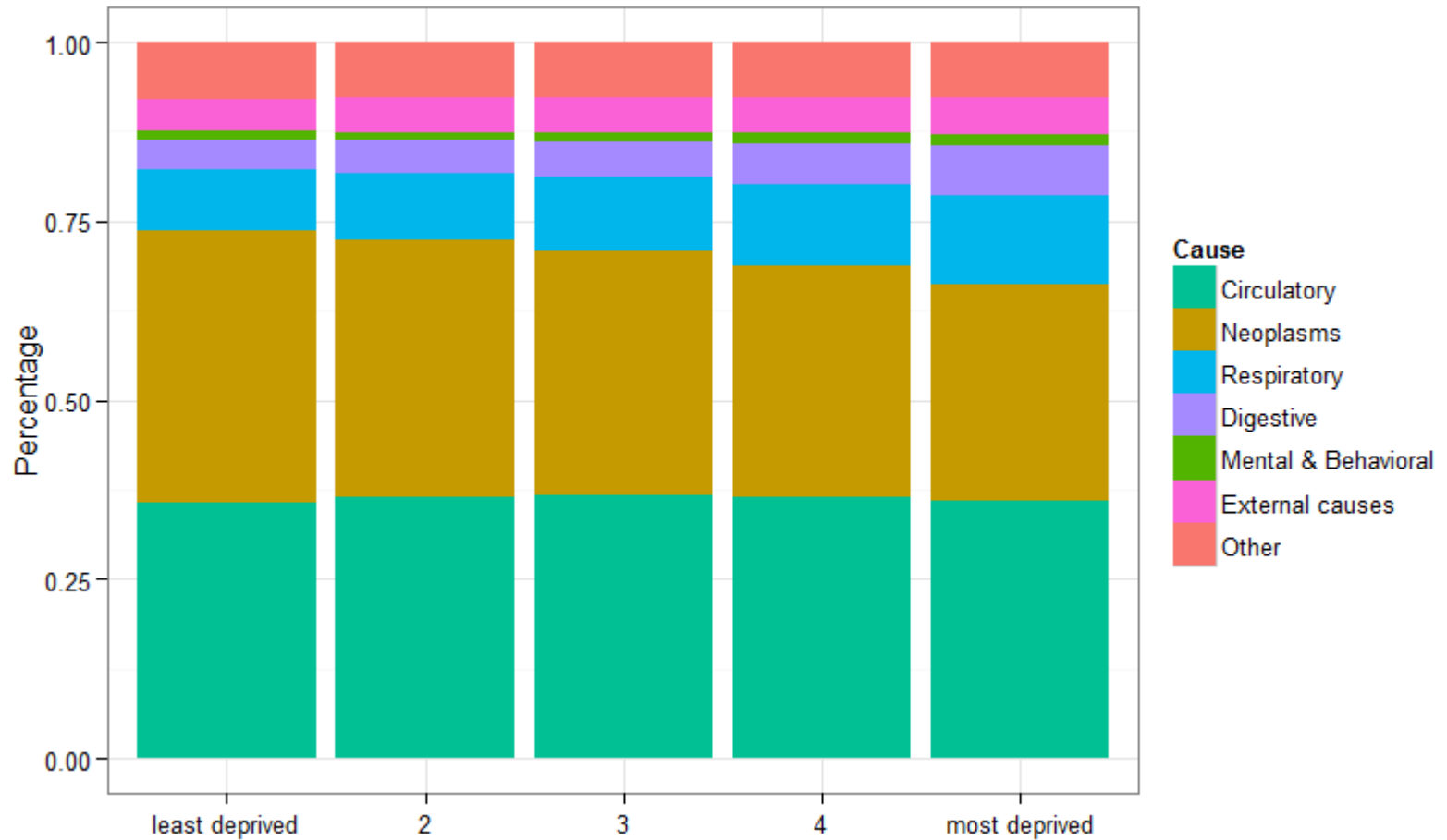
Causes of mortality in England and Wales

Causes distribution in time (ASDR males age 25-84)



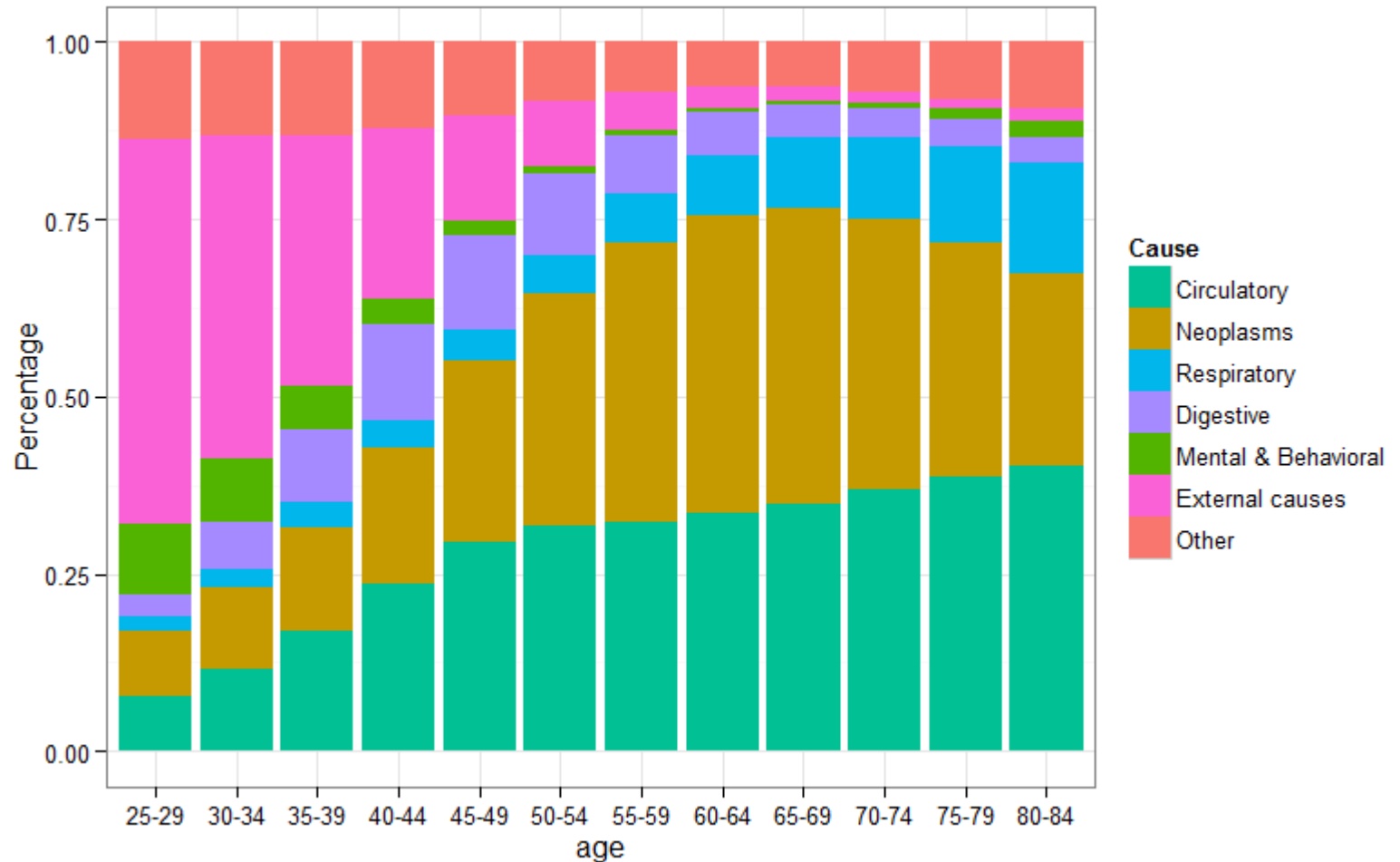
Causes of mortality in England and Wales

Causes distribution by deprivation quintile (males 25-84 2001-2007)



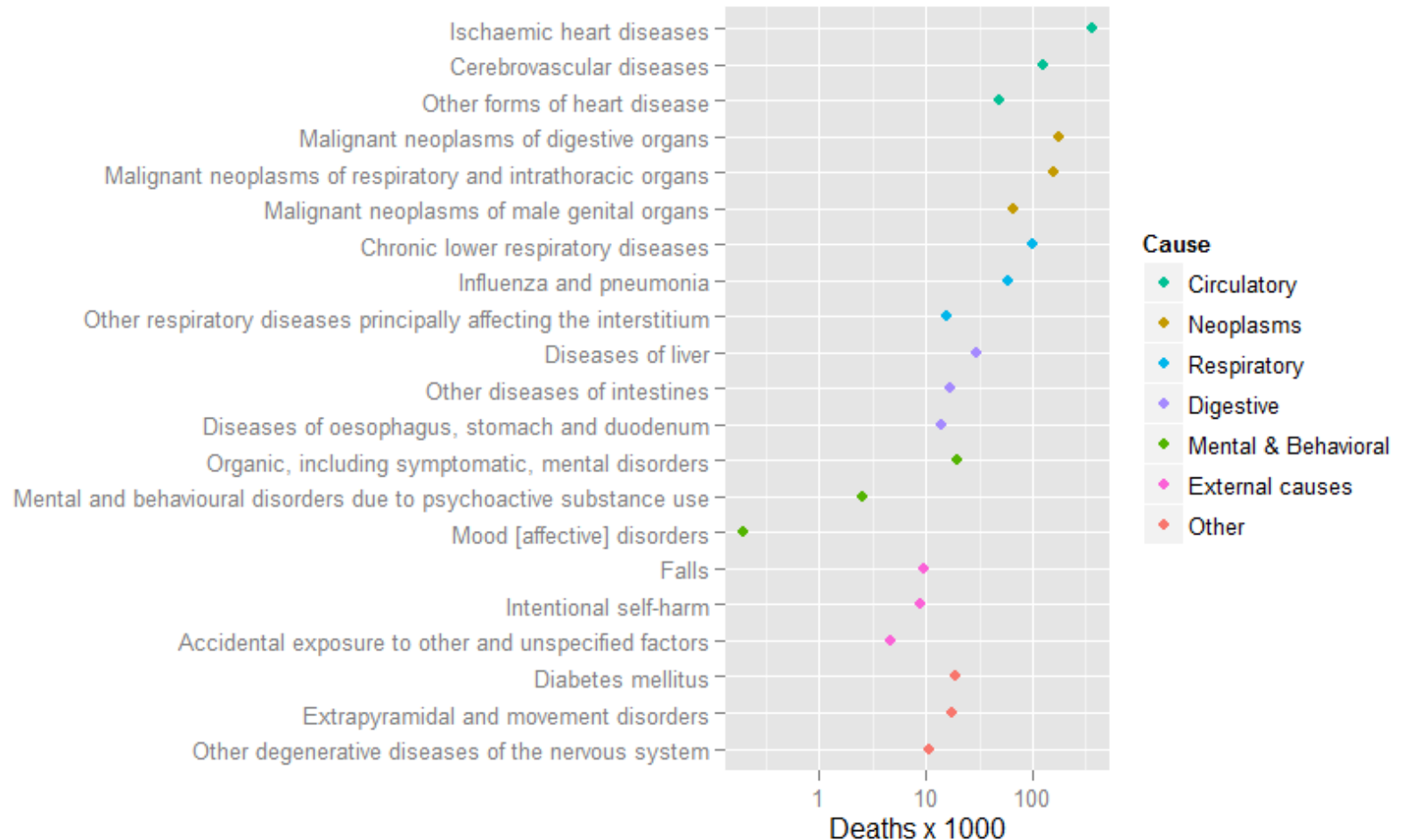
Causes of mortality in England and Wales

Causes distribution by age (males 2001-2010)

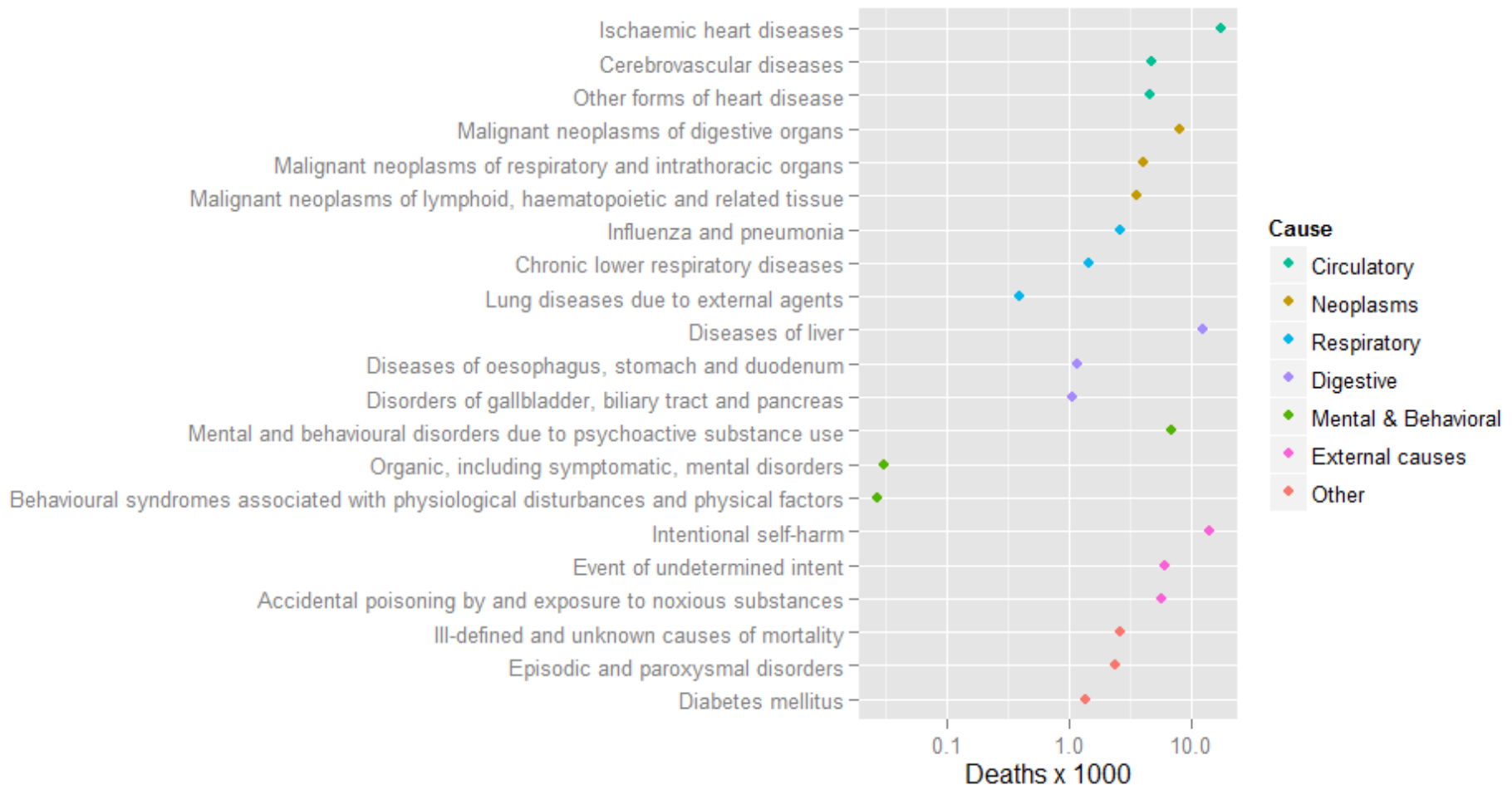


Causes of mortality in England and Wales

Main causes for males aged 50-84 (2001-2010)



Causes of mortality in England and Wales Main causes for males aged 25-49 (2001-2010)



Modelling mortality by cause of death

Challenges

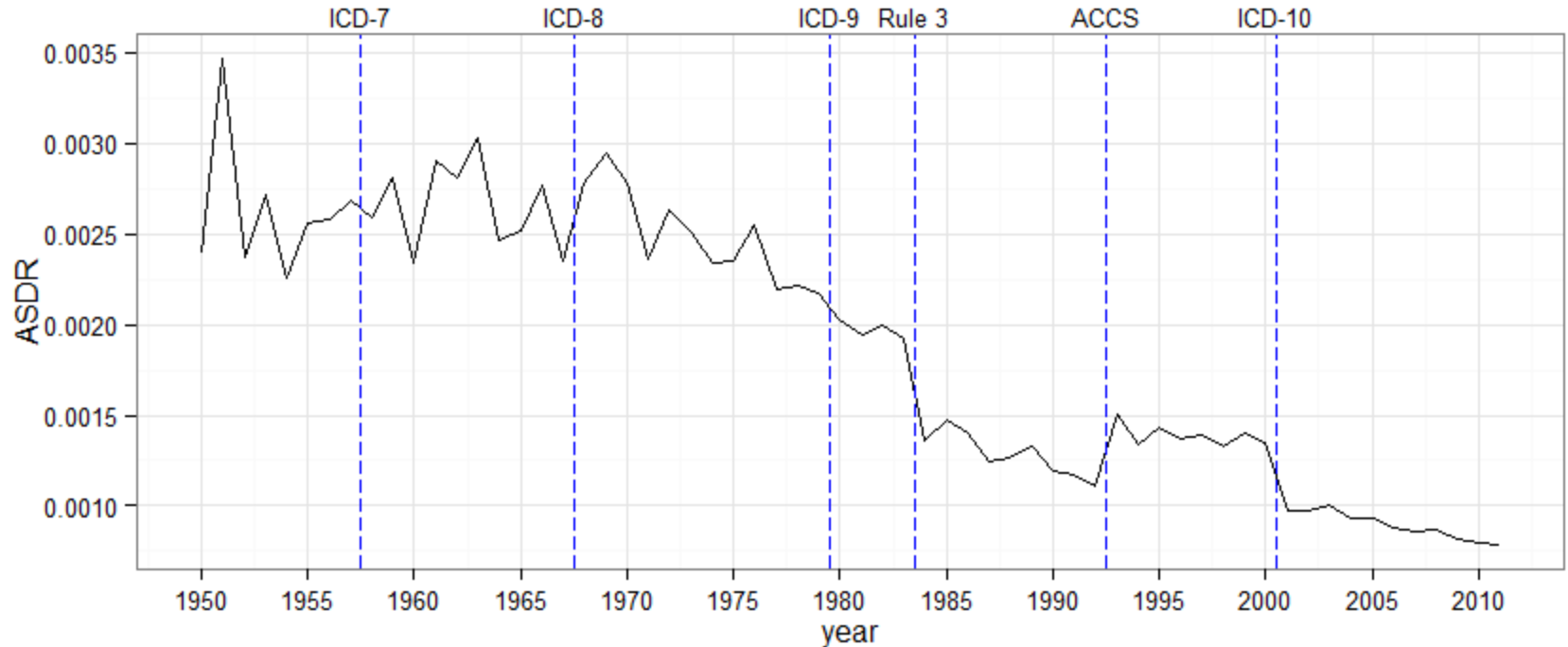
- ▶ **Correlation between causes**
 - ▶ Same risk factor can affect several causes (e.g. smoking and some cancers and heart diseases)
 - ▶ Reduction in the relative importance of one cause can lead to further improvements on other causes
- ▶ **Increase in dimensionality induced by the disaggregation**
 - ▶ The same modelling methods might not be appropriate for all causes
 - ▶ Major empirical exercise
- ▶ **Changes in classification of causes of death difficult the analysis of trends**



Modelling mortality by cause of death

Cause of death coding changes

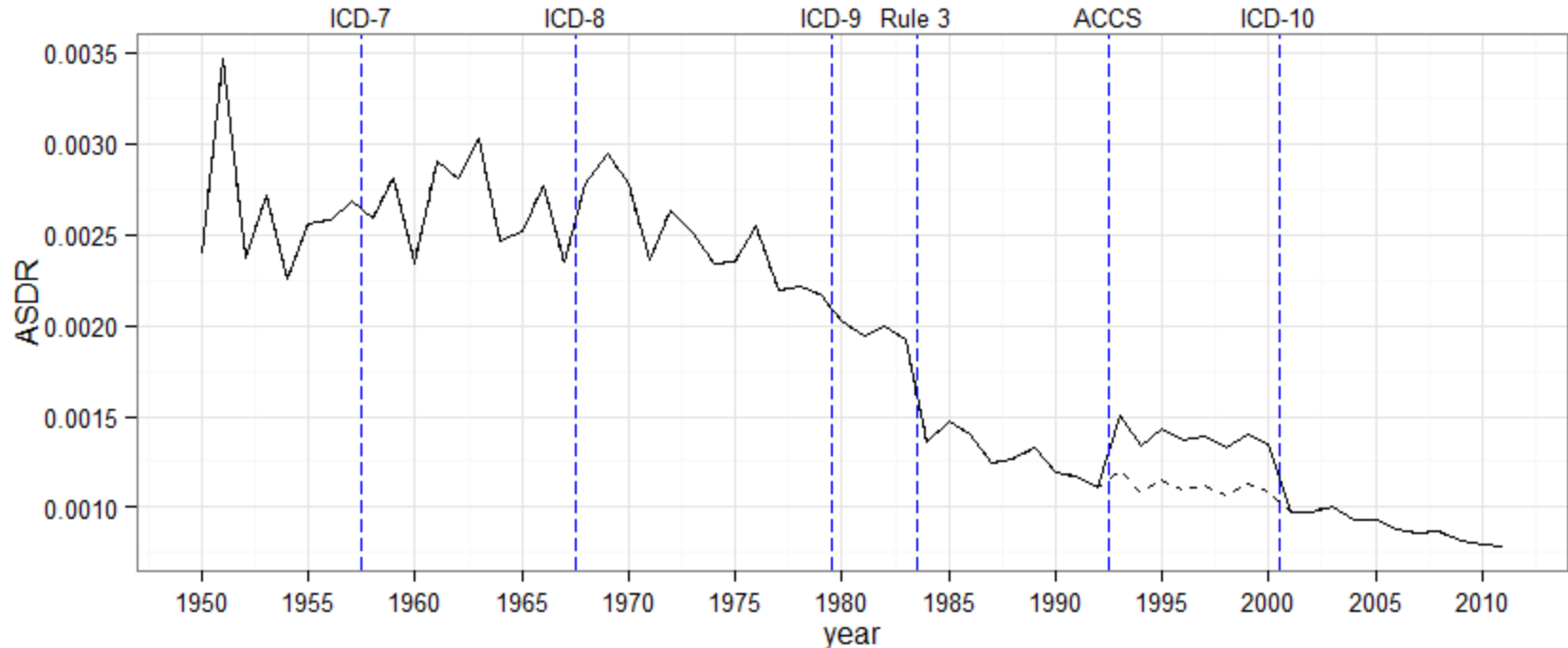
Age-standardised mortality rate for respiratory diseases
(Male age 25-84 – England and Wales)



Modelling mortality by cause of death

Cause of death coding changes

Age-standardised mortality rate for respiratory diseases
(Male age 25-84 – England and Wales)



▶ Adjustment methods

- ▶ Bridge coding and comparability ratios (e.g. ONS for ICD-9 to ICD10)
- ▶ Statistical correction methods (e.g. Rey et al (2009), Park et al (2006))



Modelling mortality by cause of death

Lee-Carter model with coding changes

$$\log \mu_{xt} = \alpha_x + \beta_x \kappa_t$$



Modelling mortality by cause of death

Lee-Carter model with coding changes

$$\log \mu_{xt} = \alpha_x + \beta_x \kappa_t$$

Age-specific mortality pattern

Age-modulating parameters

Overall time trend of mortality



Modelling mortality by cause of death

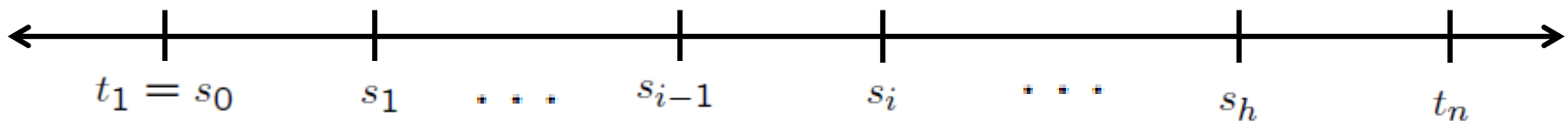
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Modelling mortality by cause of death

Lee-Carter model with coding changes

$$\log \mu_{xt} = \alpha_x + \beta_x \kappa_t + \sum_{i=1}^h \delta_x^{(i)} f^{(i)}(t)$$

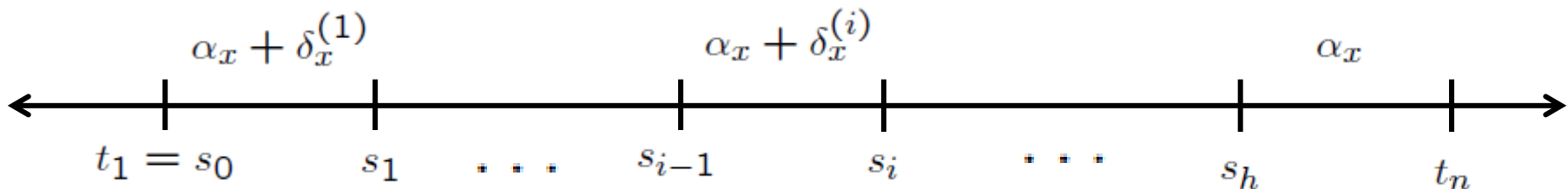
Age-specific mortality pattern

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Overall time trend of mortality

Adjustment for coding changes

$$f^{(i)}(t) = \mathcal{I}_{\{s_{i-1} \leq t < s_i\}}$$



Modelling mortality by cause of death

Lee-Carter model with coding changes – Invariant transformations

$$\log \mu_{xt} = \alpha_x + \beta_x \kappa_t + \sum_{i=1}^h \delta_x^{(i)} f^{(i)}(t)$$

This specification is invariant to the following parameter transformations

Standard Lee-Carter transformations

$$\{\tilde{\alpha}_x, \tilde{\kappa}_t\} = \{\alpha_x + b_1 \beta_x, \kappa_t - b_1\}$$

$$\{\tilde{\beta}_x, \tilde{\kappa}_t\} = \left\{ \frac{1}{b_2} \beta_x, b_2 \kappa_t \right\}$$



Modelling mortality by cause of death

Lee-Carter model with coding changes – Invariant transformations

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New transformations

$$\left\{ \tilde{\delta}_x^{(i)}, \tilde{\kappa}_t \right\} = \left\{ \delta_x^{(i)} + a_i \beta_x, \kappa_t - a_i f^{(i)}(t) \right\} \quad i = 1, \dots, h$$



Modelling mortality by cause of death

Lee-Carter model with coding changes – Identifiability constraints

Standard Lee-Carter

Make the last year in the data the reference

$$\{\tilde{\alpha}_x, \tilde{\kappa}_t\} = \{\alpha_x + b_1\beta_x, \kappa_t - b_1\} \longrightarrow \kappa_{t_n} = 0$$

Normalise the age gradient

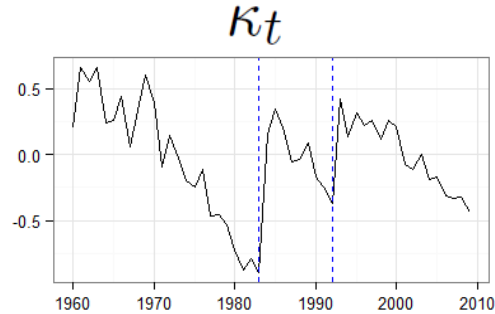
$$\{\tilde{\beta}_x, \tilde{\kappa}_t\} = \left\{ \frac{1}{b_2}\beta_x, b_2\kappa_t \right\} \longrightarrow \sum_x \beta_x = 1$$



Modelling mortality by cause of death

Lee-Carter model with coding changes – Identifiability constraints

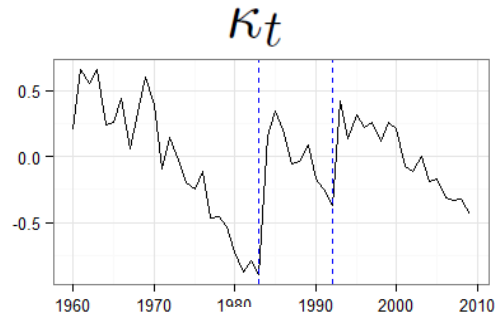
$$\left\{ \tilde{\delta}_x^{(i)}, \tilde{\kappa}_t \right\} = \left\{ \delta_x^{(i)} + a_i \beta_{x, \kappa_t} - a_i f^{(i)}(t) \right\}$$



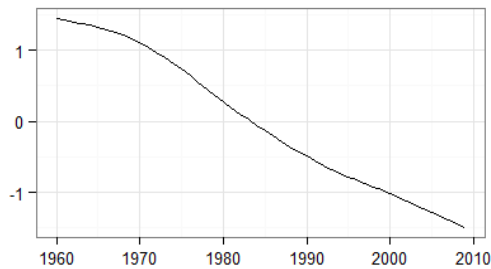
Modelling mortality by cause of death

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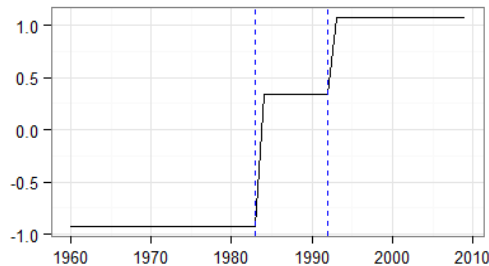


$g(t)$



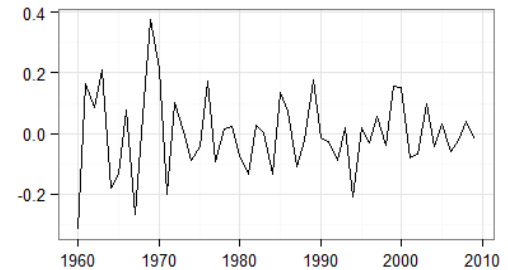
+

$$\sum_{i=1}^h a_i f^{(i)}(t)$$



+

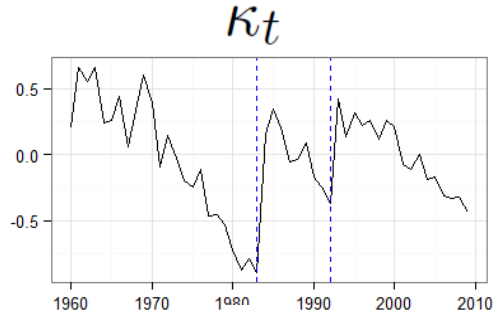
ϵ_t



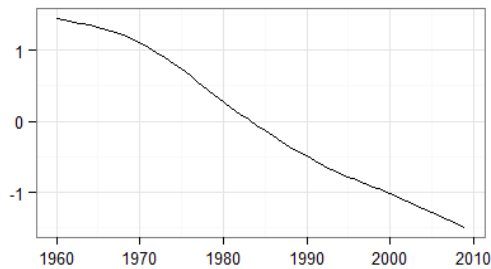
Modelling mortality by cause of death

Lee-Carter model with coding changes – Identifiability constraints

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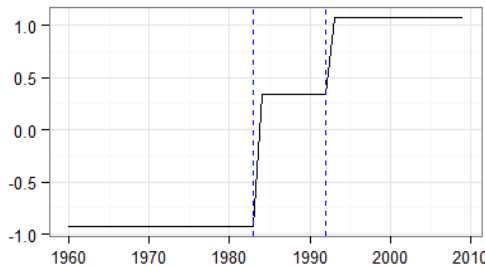


$g(t)$



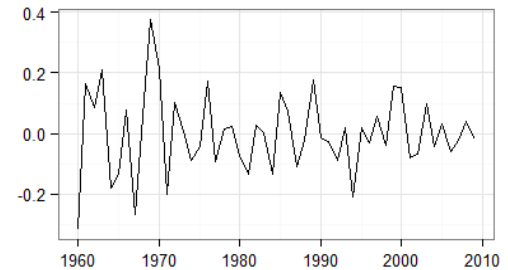
+

$$\sum_{i=1}^h a_i f^{(i)}(t)$$

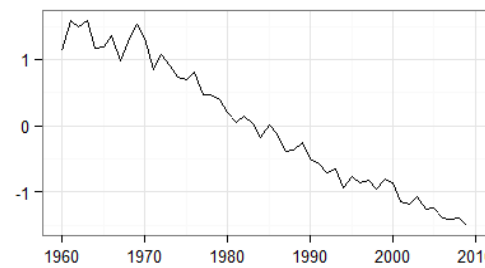


+

ϵ_t



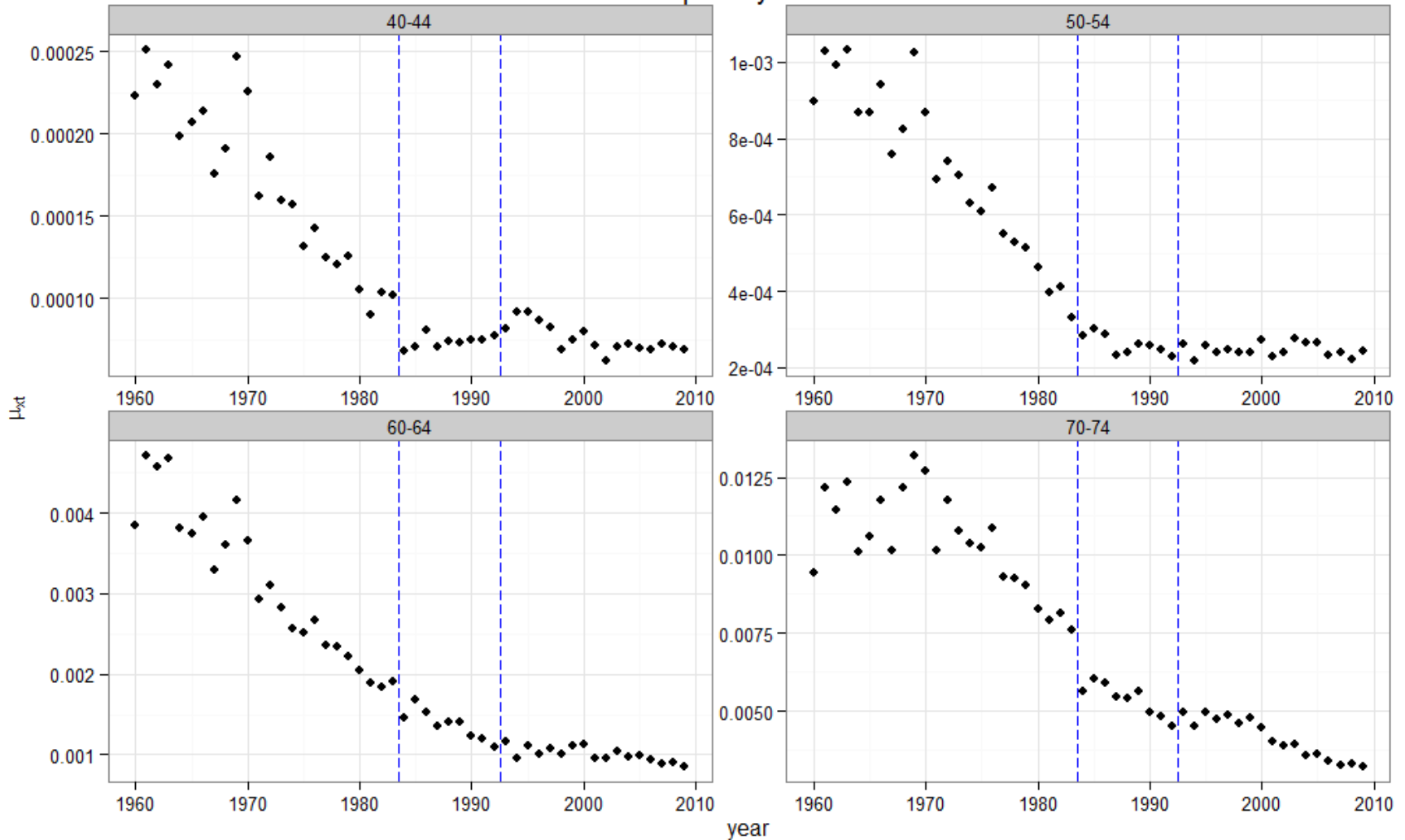
$$g(t) + \epsilon_t = \tilde{\kappa}_t = \kappa_t - \sum_{i=1}^h a_i f^{(i)}(t)$$



Modelling mortality by cause of death

Lee-Carter model with coding changes – Example

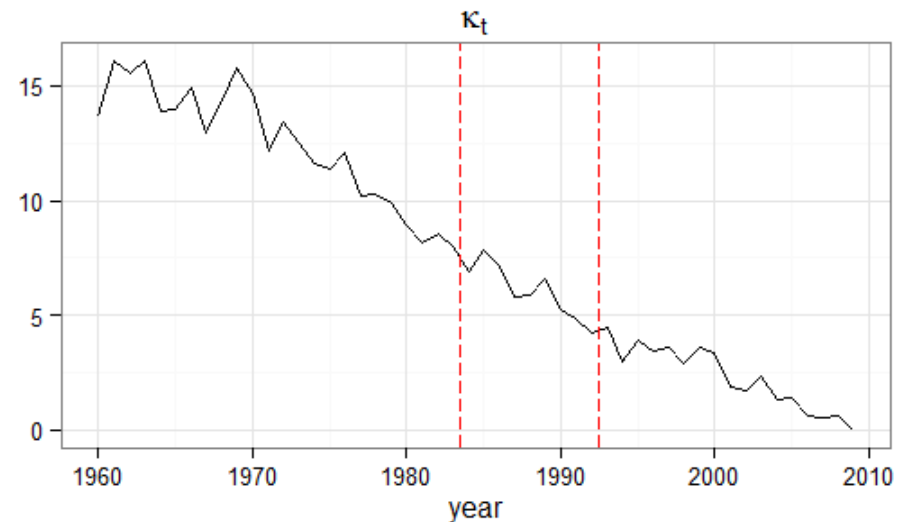
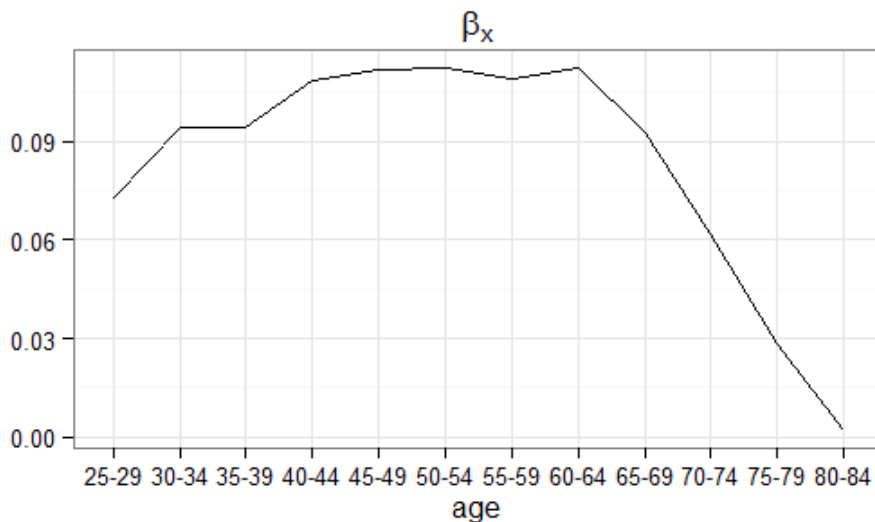
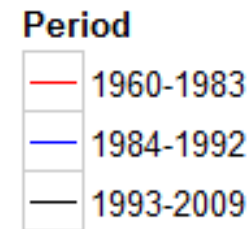
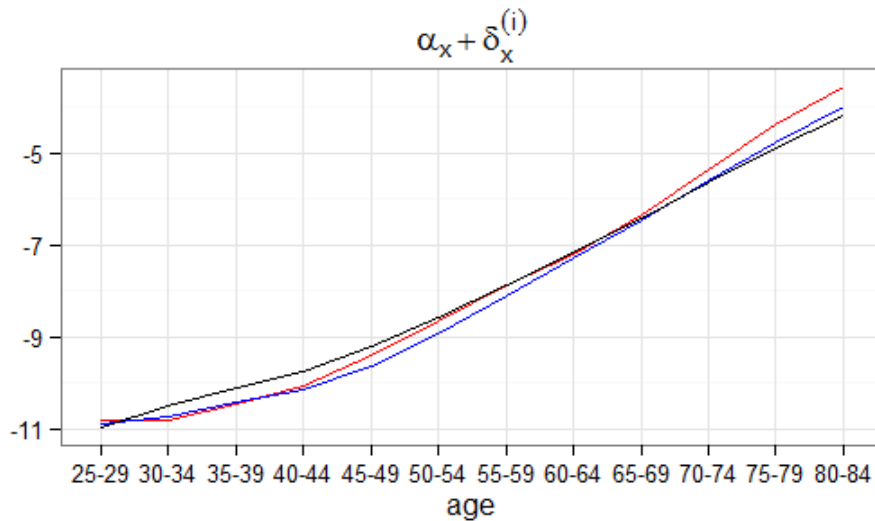
Respiratory - Males



Modelling mortality by cause of death

Lee-Carter model with coding changes – Example

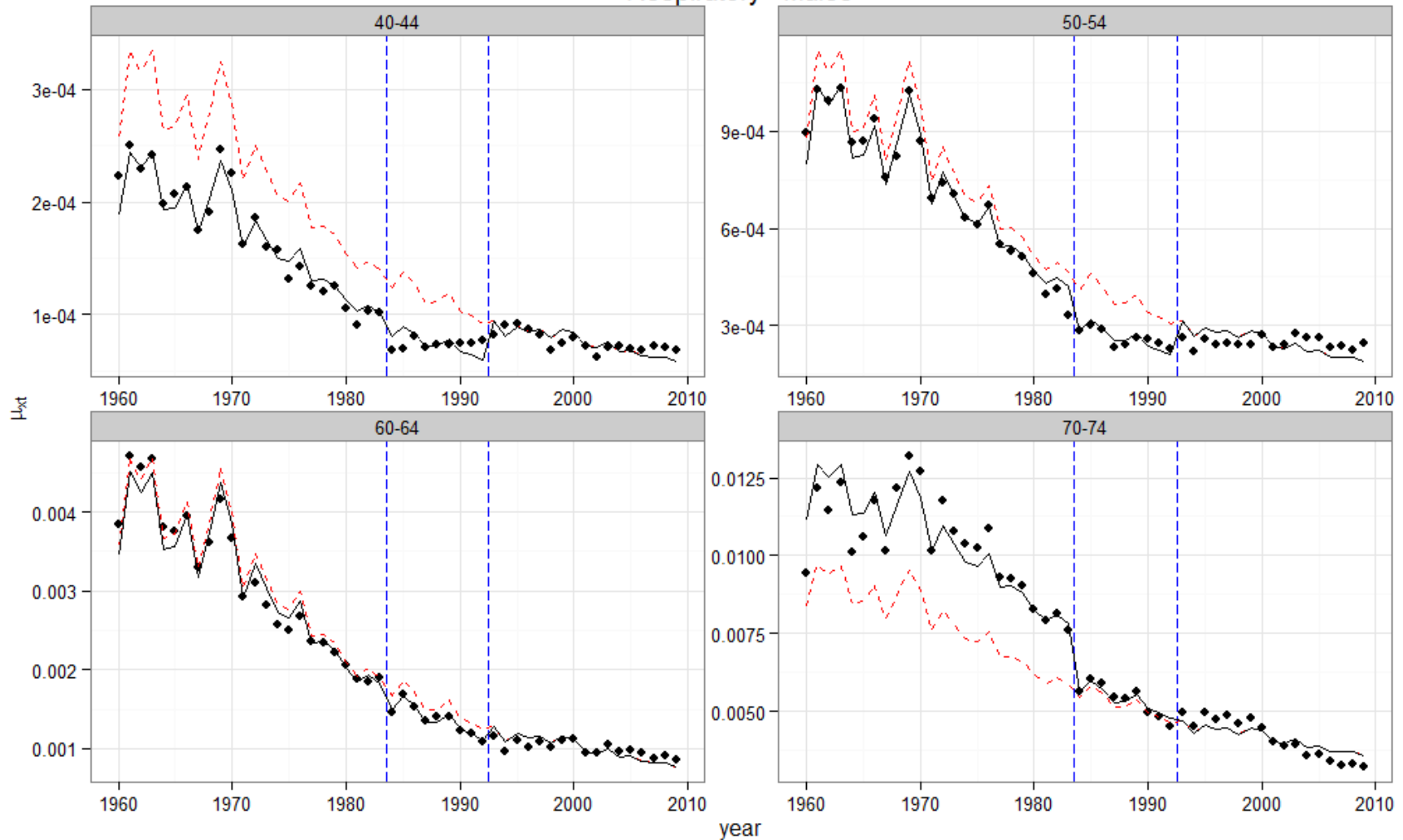
Respiratory-Males



Modelling mortality by cause of death

Lee-Carter model with coding changes – Example

Respiratory - Males



Modelling by CoD and socio-economic stratification

Three-way Lee-Carter model (Russolillo et al, 2011)

$$\log \mu_{xtg}^c = \alpha_x^c + \beta_x^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Modelling by CoD and socio-economic stratification

Three-way Lee-Carter model (Russolillo et al, 2011)

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$

Level differentials



Modelling by CoD and socio-economic stratification

Three-way Lee-Carter model (Russolillo et al, 2011)

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$

Level differentials

Improvement differentials



Modelling by CoD and socio-economic stratification

Three-way Lee-Carter model

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$

- ▶ Estimate the model parameters using a two stage estimation procedure with a reference population
 - ▶ National population data available for longer periods of time than socio-economic disaggregated data
 - ▶ More precise estimation of the long-run mortality trend
 - ▶ Coherency with the national mortality trend
- ▶ Stage I:
 - ▶ Estimate $\alpha_x^c, \beta_x^c, \kappa_t^c, \delta_x^{c,(i)}$ using the reference population data
- ▶ Stage II:
 - ▶ Estimate $\alpha_{xg}^c, \lambda_g^c$ conditional on $\alpha_x^c, \beta_x^c, \kappa_t^c, \delta_x^{c,(i)}$



Case study: Mortality by deprivation in England

Application data

Subpopulation data

- ▶ England population disaggregated by deprivation quintile using the 2007 version of the English Index of Multiple Deprivation (IMD 2007)
- ▶ Ages: 25-29,30-34,...,80-84
- ▶ Period: 1981-2007

Reference population data

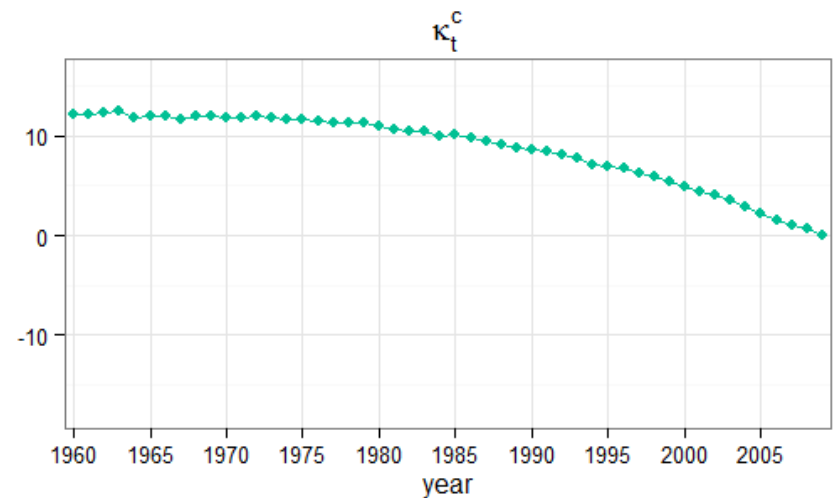
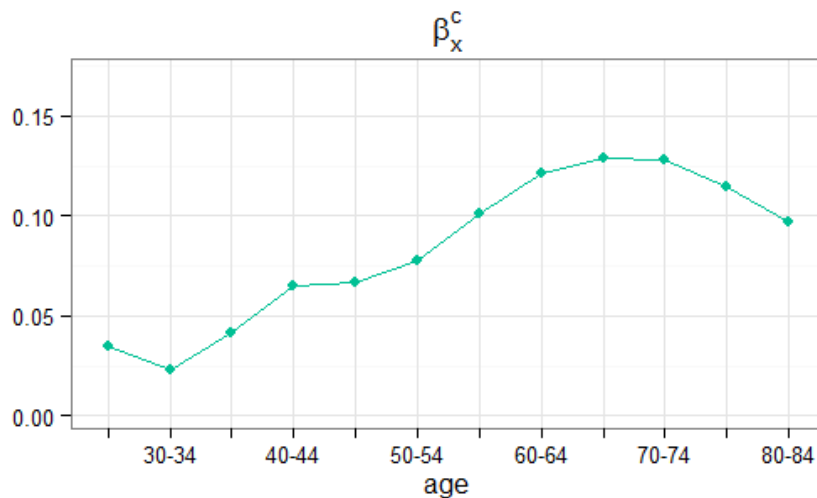
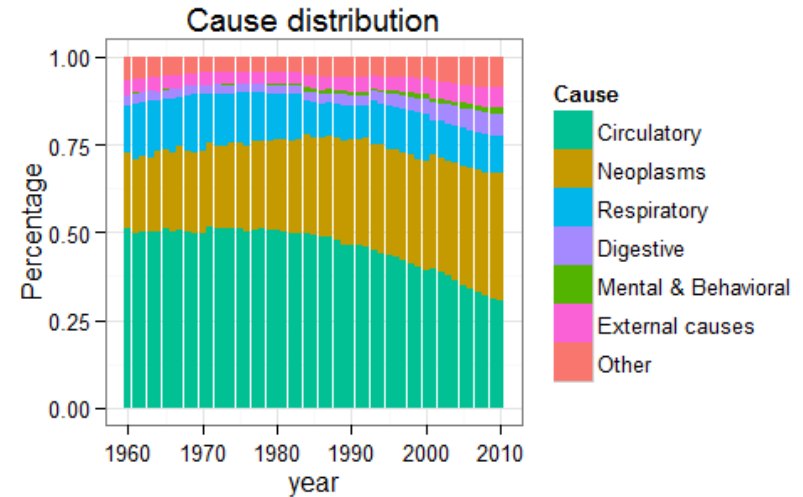
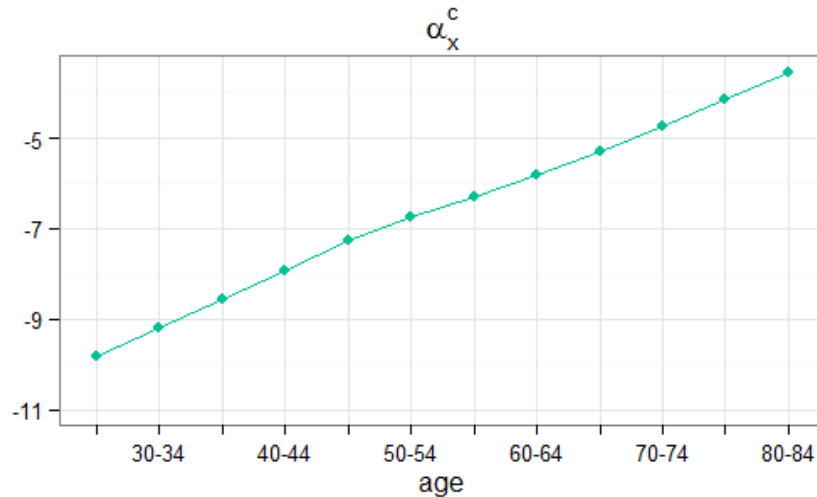
- ▶ England and Wales population
- ▶ Ages: 25-29,30-34,...,80-84
- ▶ Period: 1960-2009



Case study: Mortality by deprivation in England

England and Wales Male population parameters

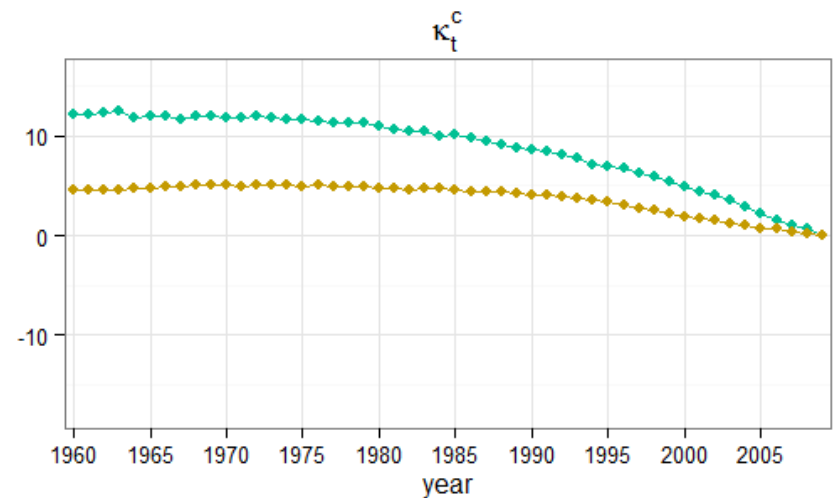
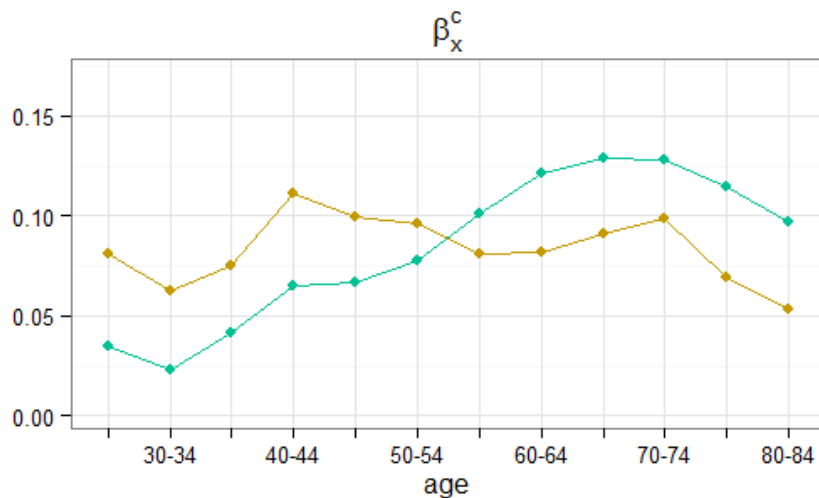
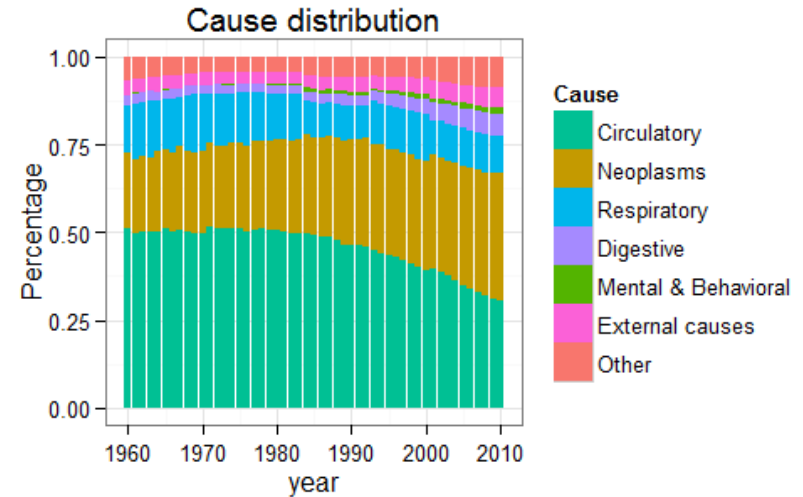
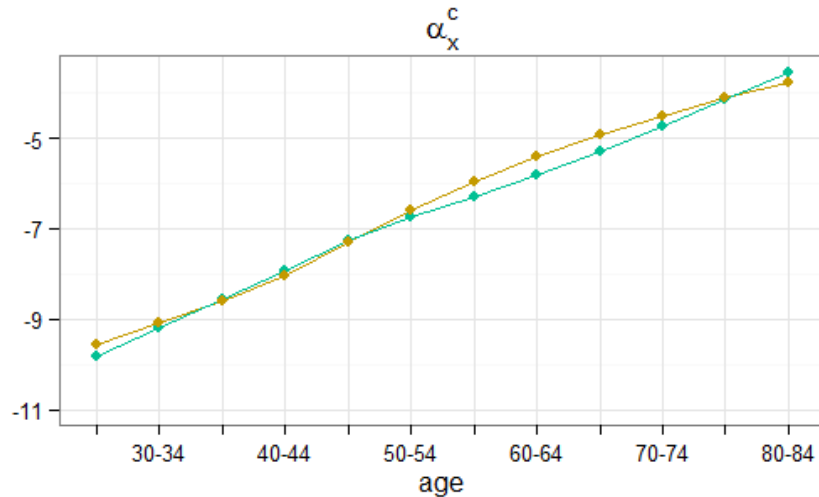
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Case study: Mortality by deprivation in England

England and Wales Male population parameters

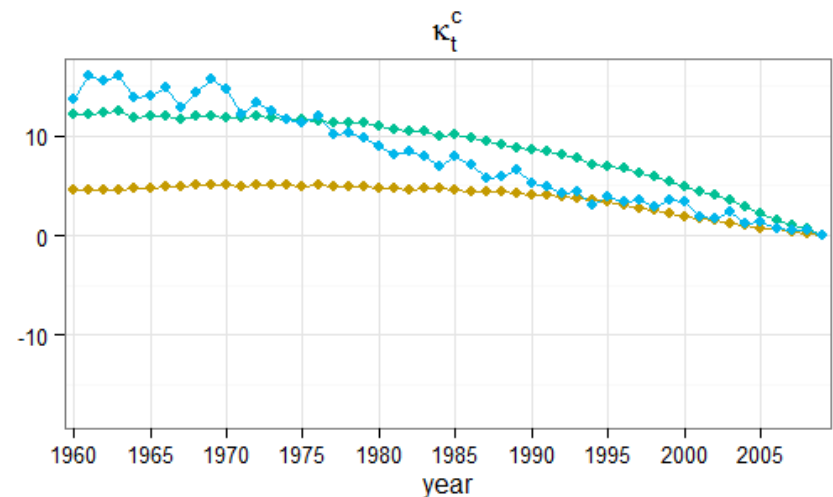
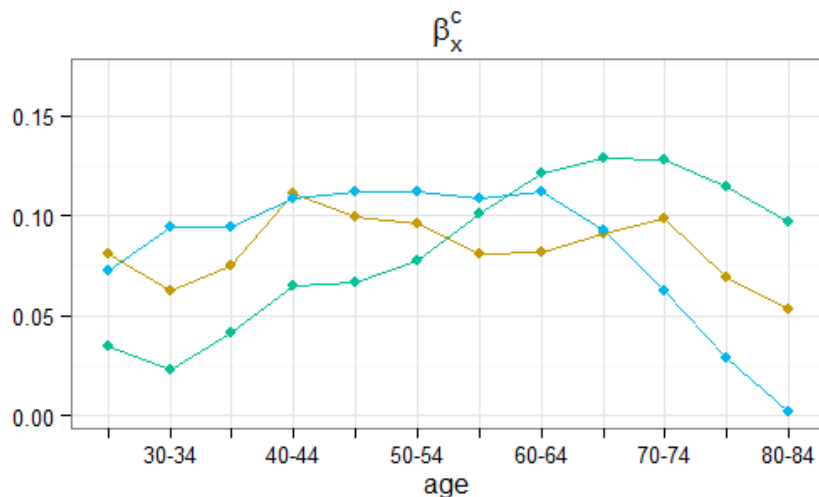
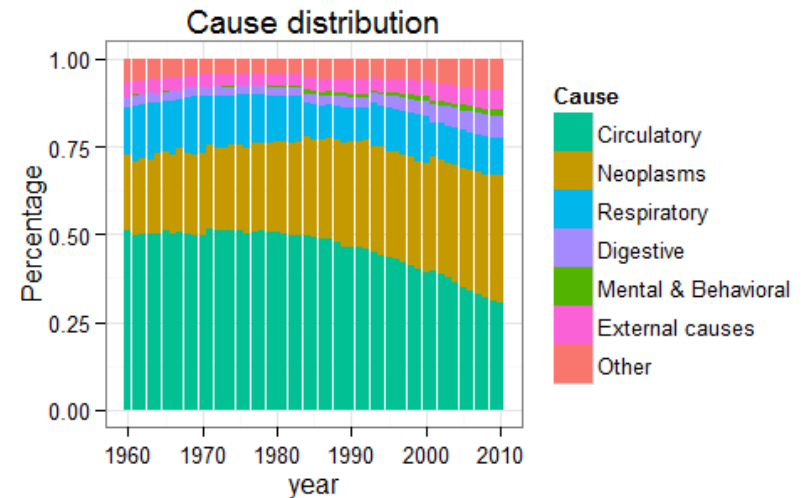
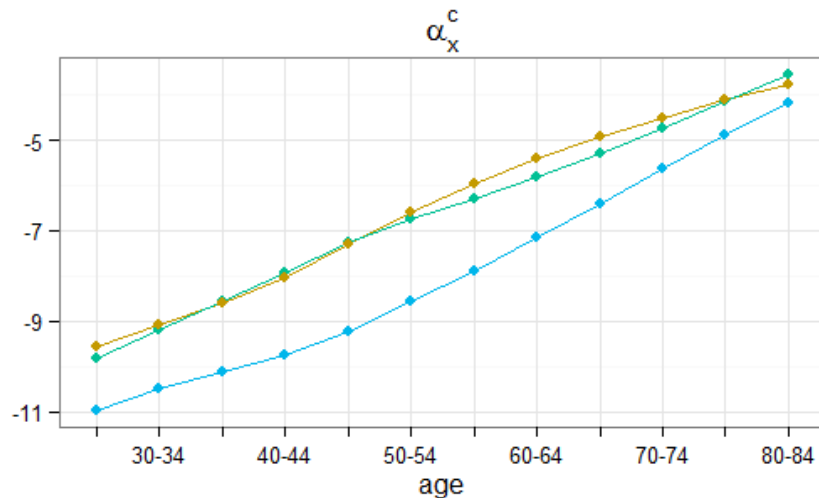
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Case study: Mortality by deprivation in England

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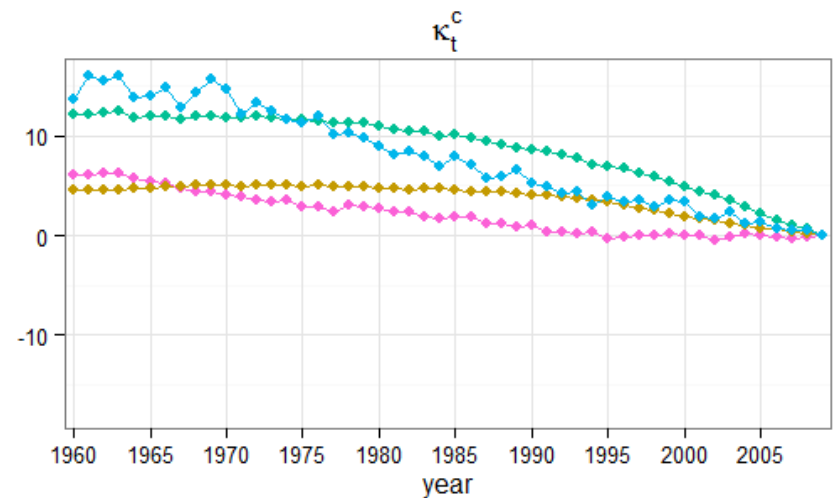
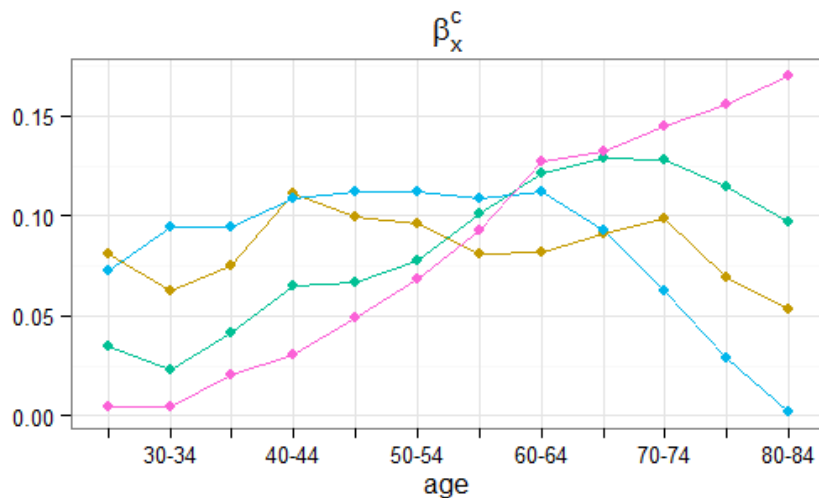
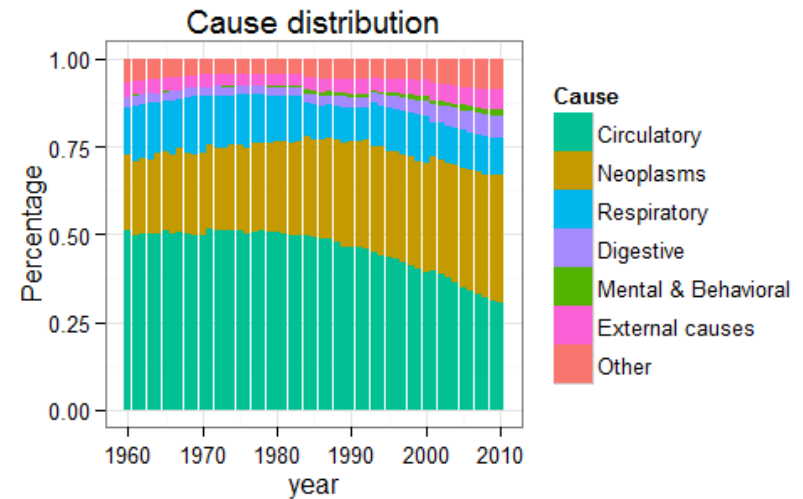
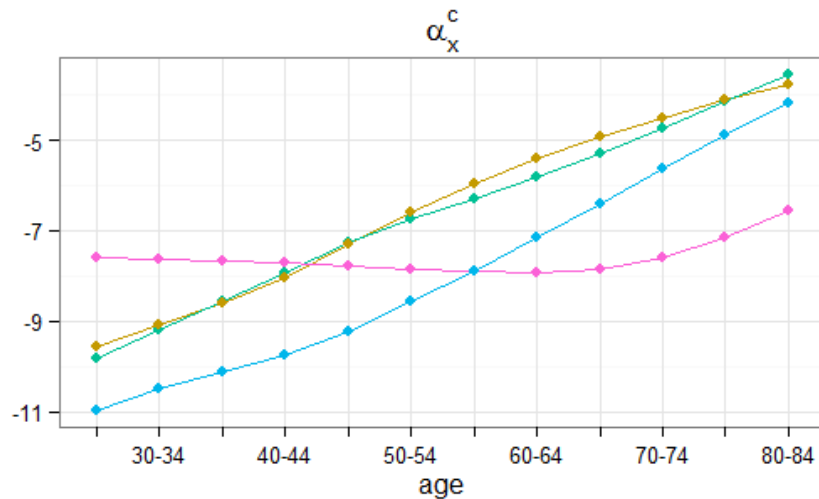
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Case study: Mortality by deprivation in England

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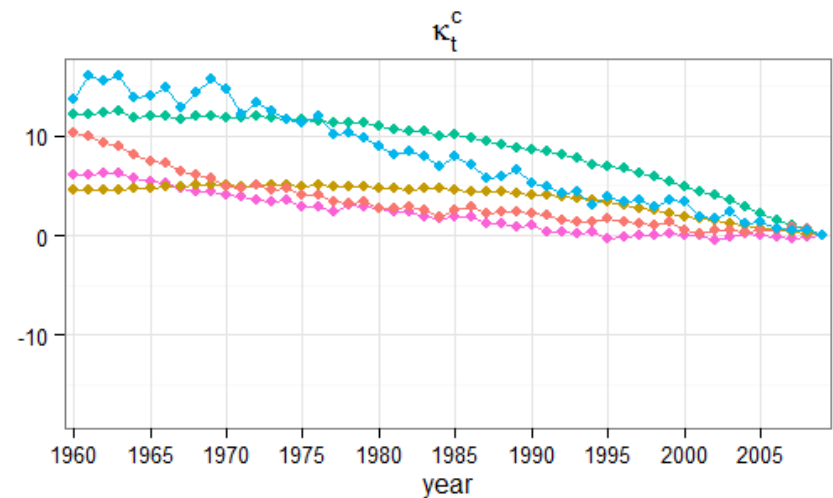
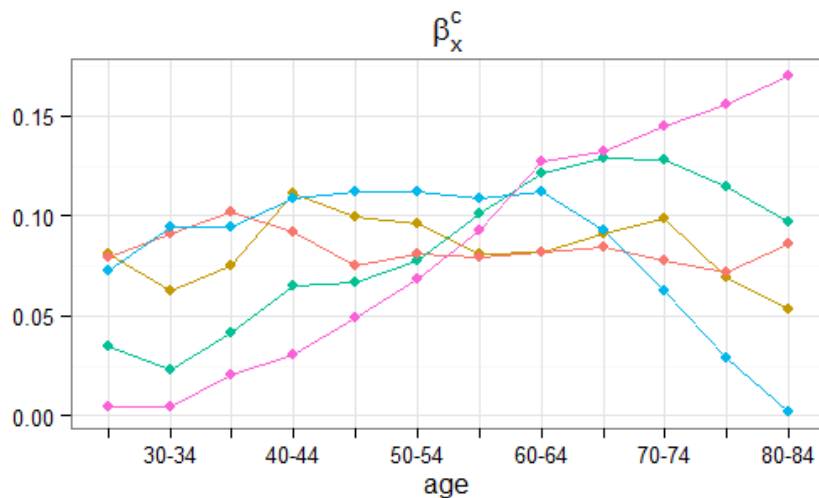
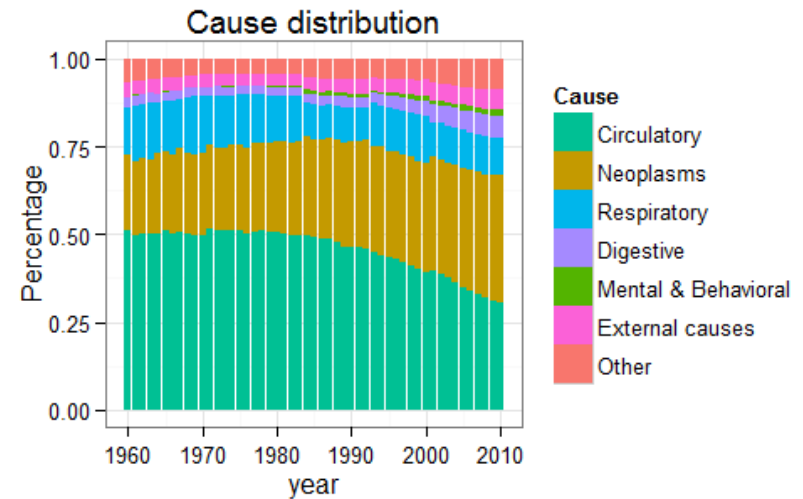
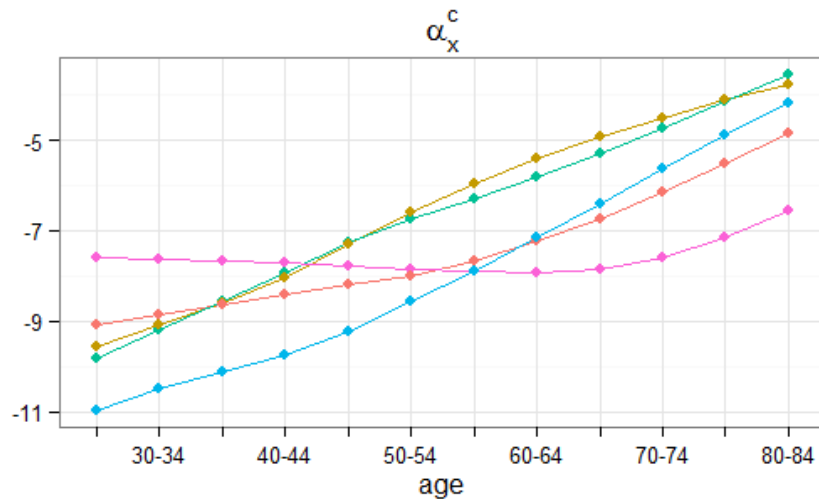
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Case study: Mortality by deprivation in England

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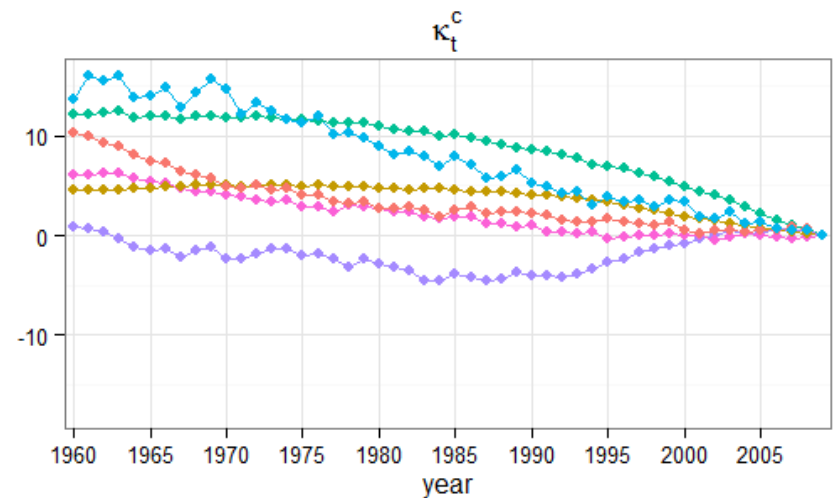
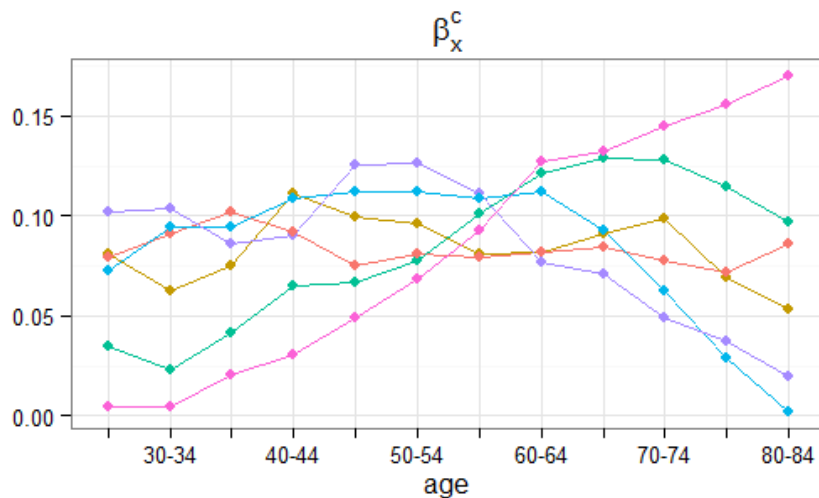
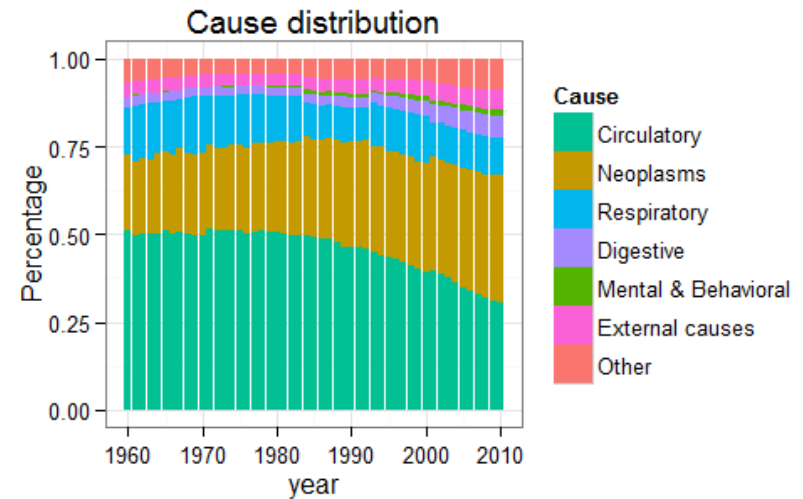
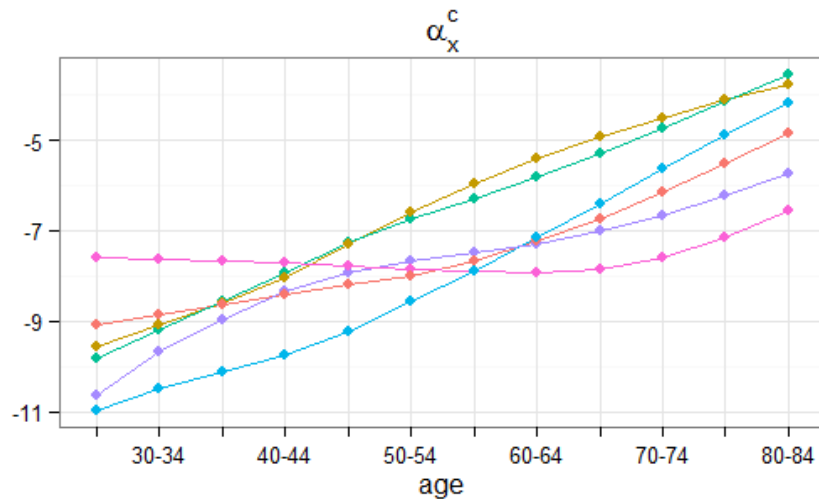
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Case study: Mortality by deprivation in England

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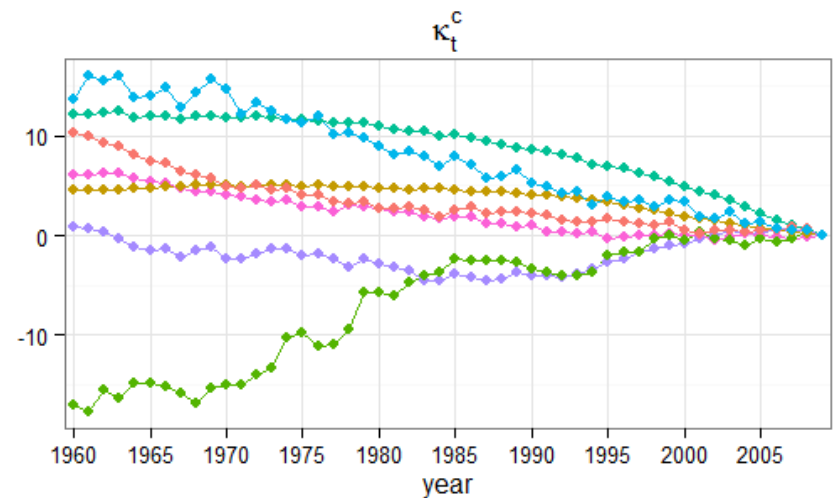
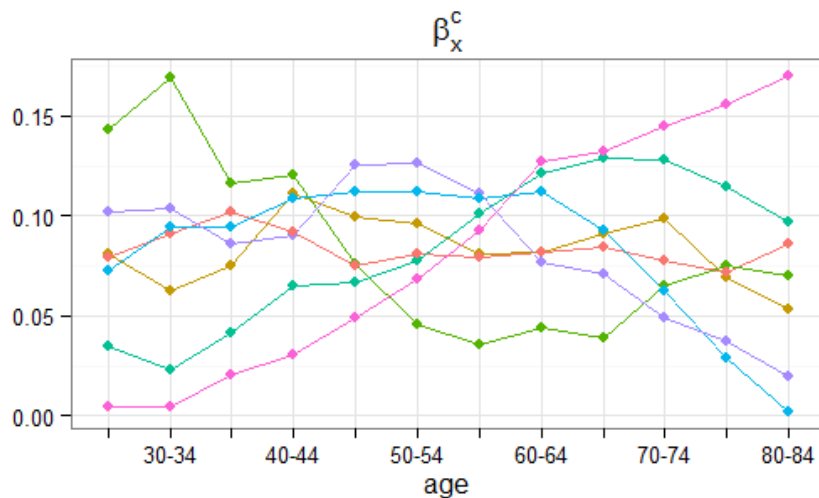
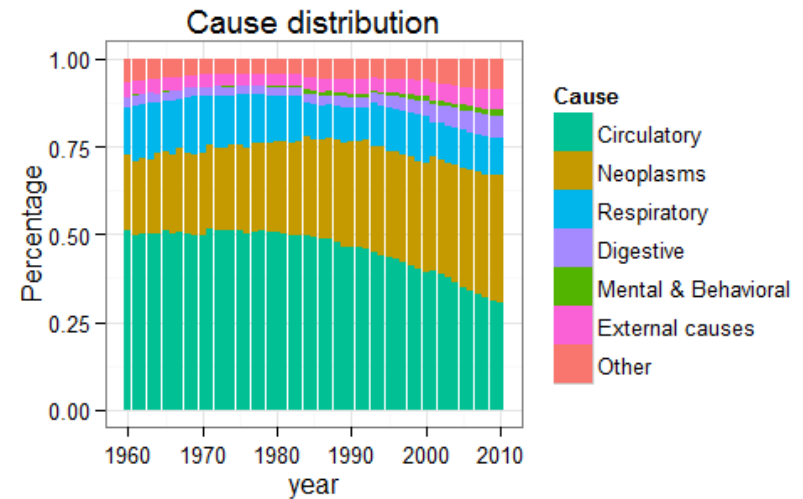
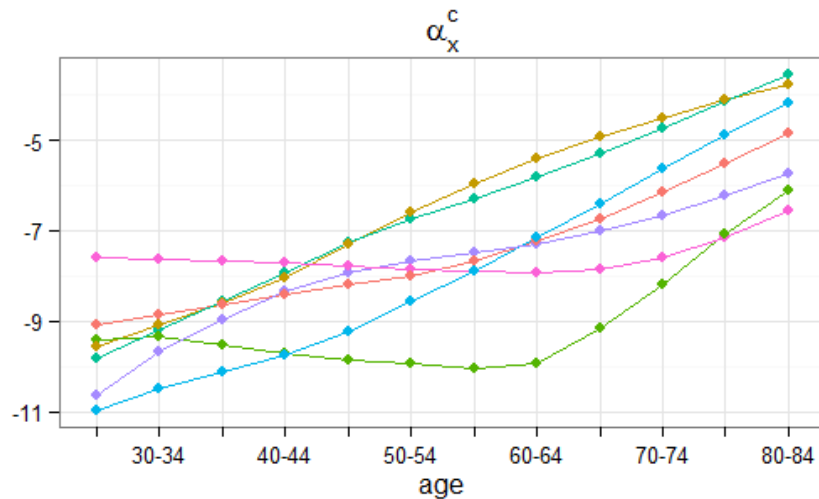
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Case study: Mortality by deprivation in England

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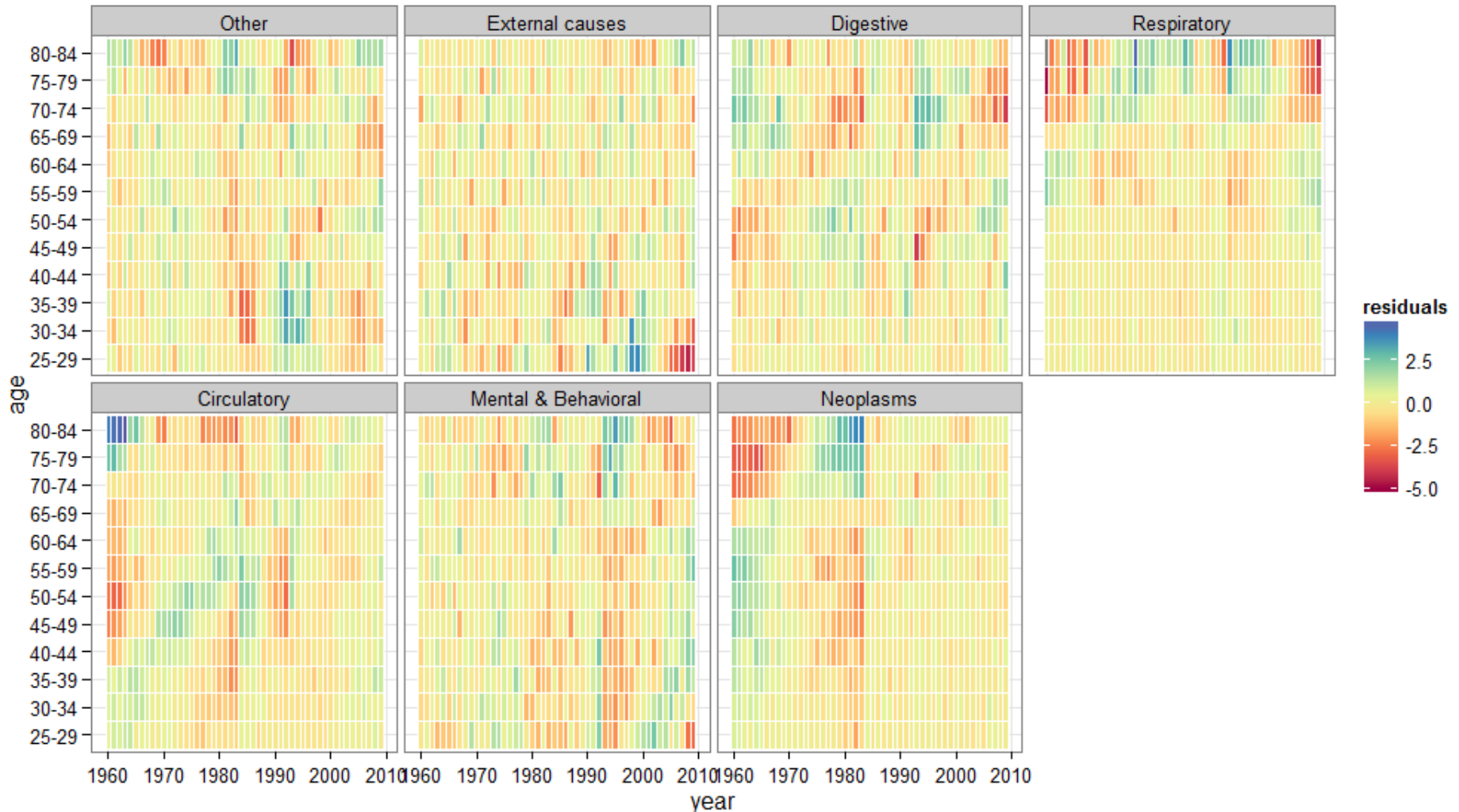
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Case study: Mortality by deprivation in England

England and Wales Male population - Residuals

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$

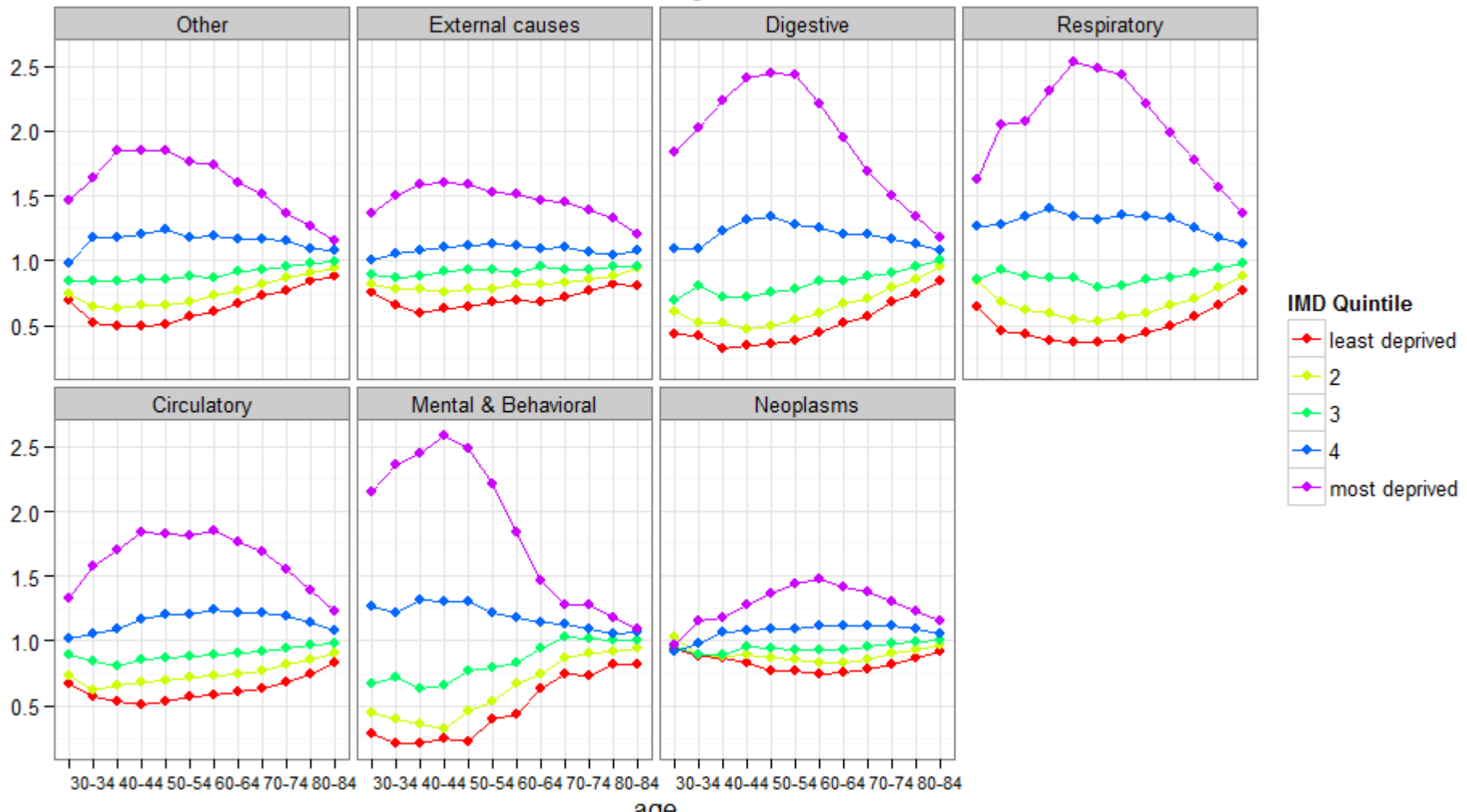


Case study: Mortality by deprivation in England

Level differences by deprivation quintile

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$

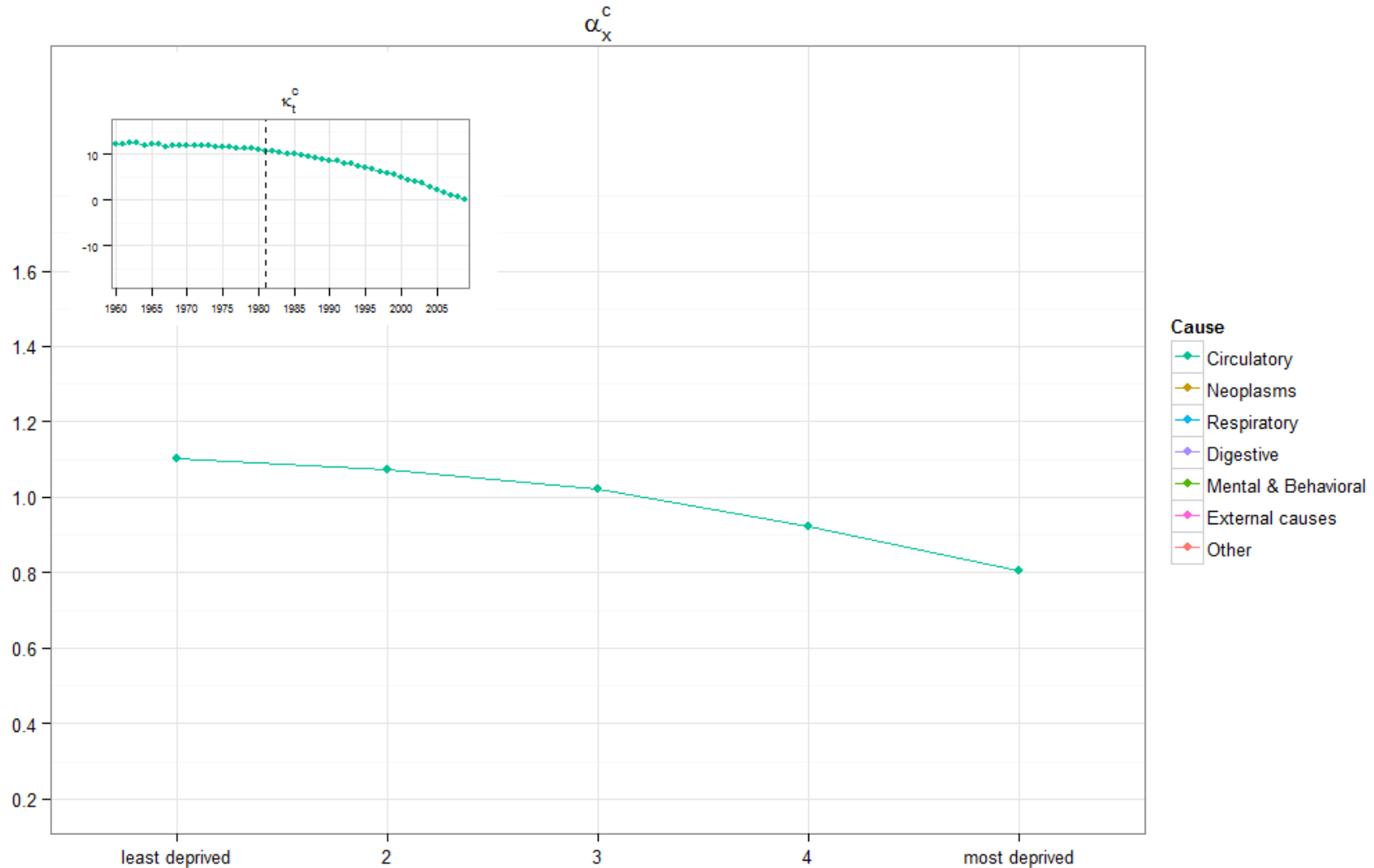
$\exp(\alpha_{xg}^c)$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

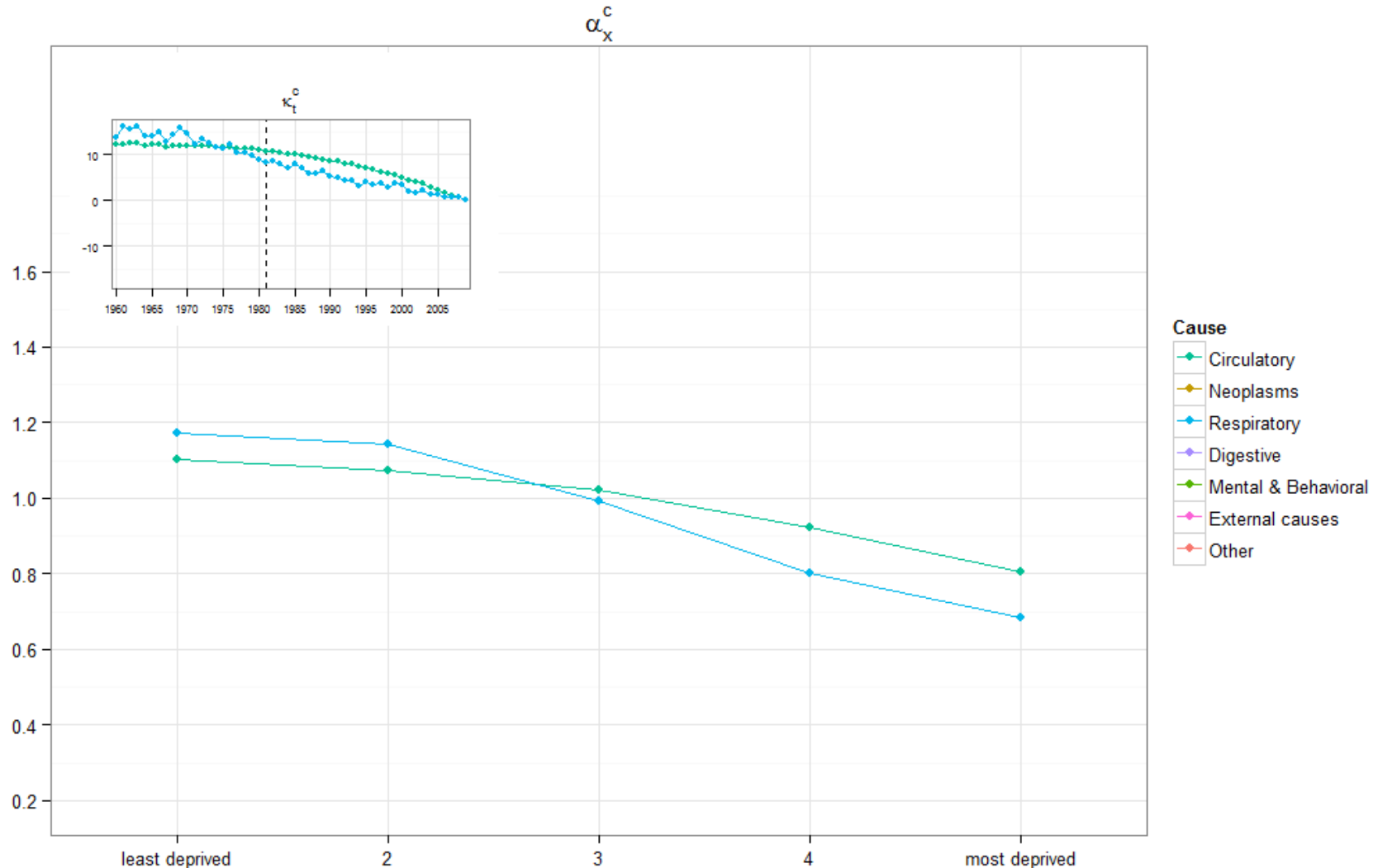
$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

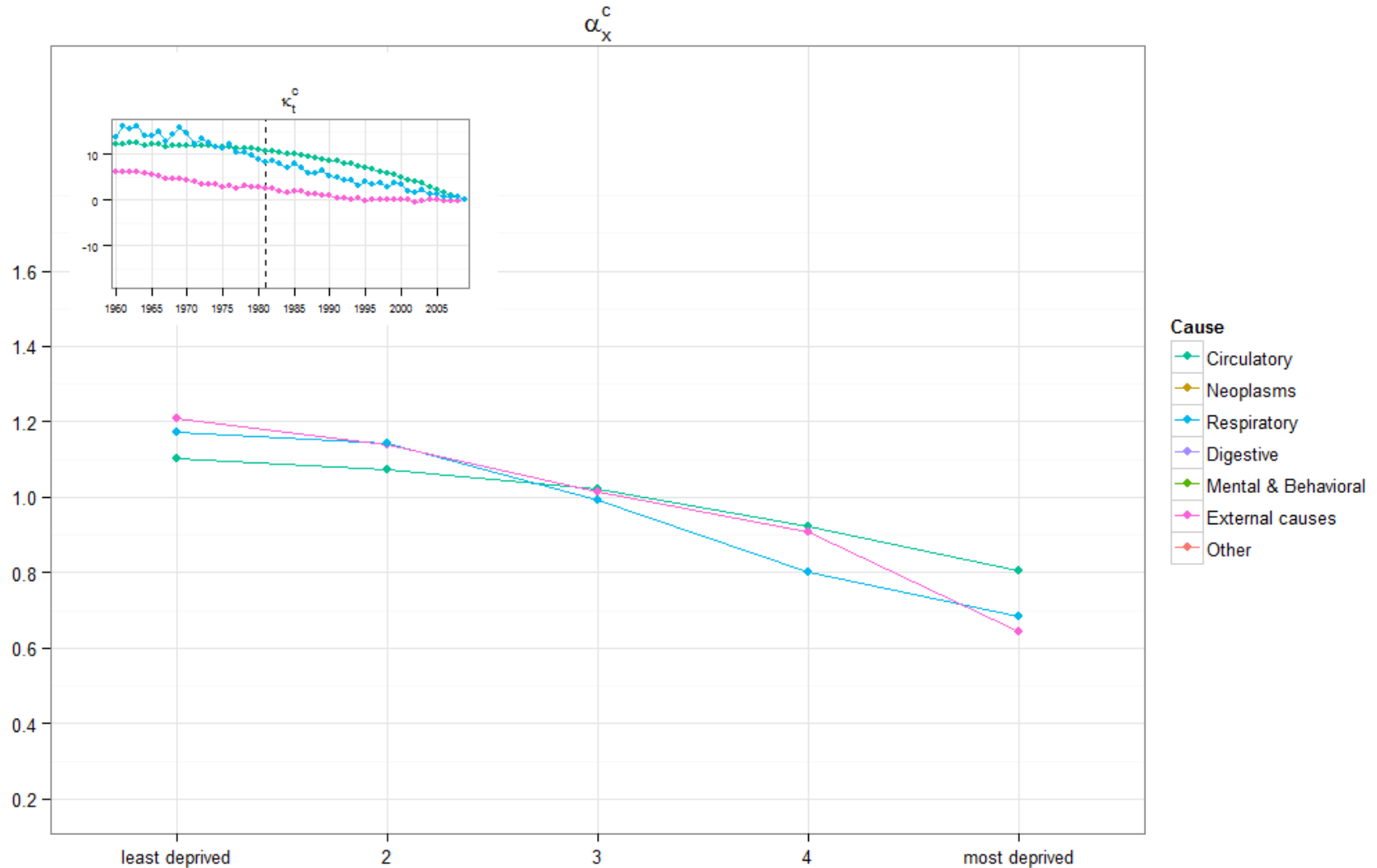
$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

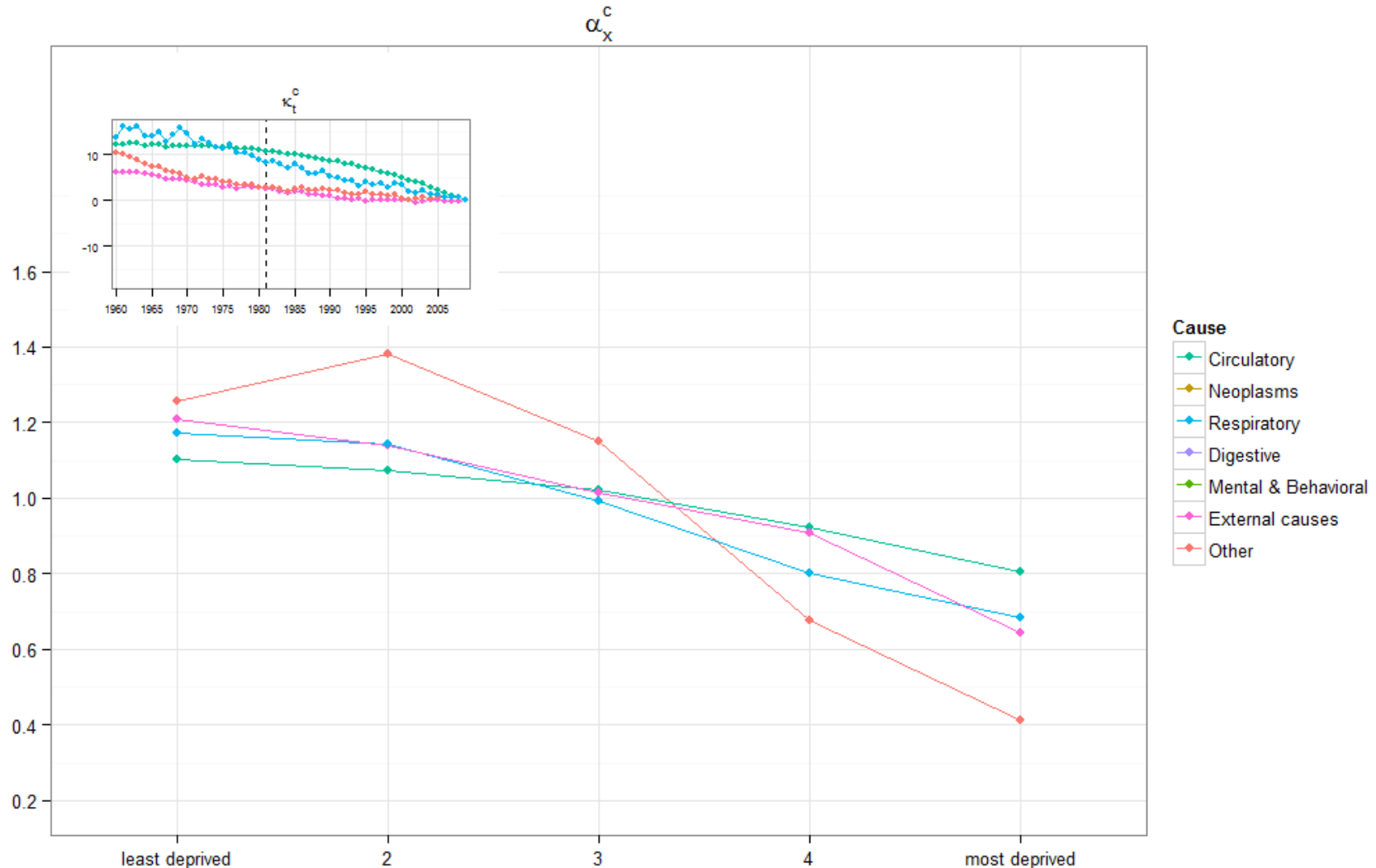
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Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

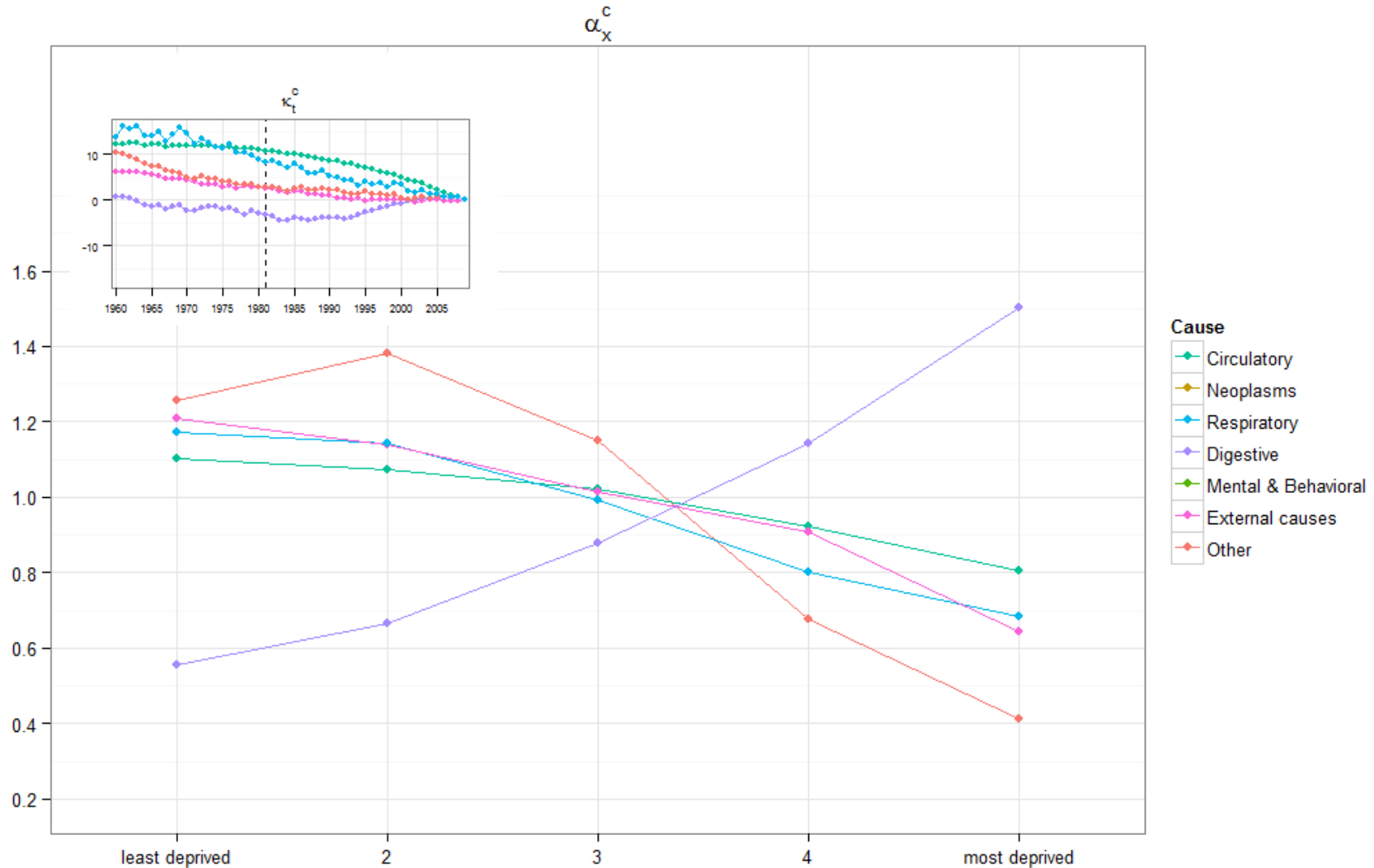
$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

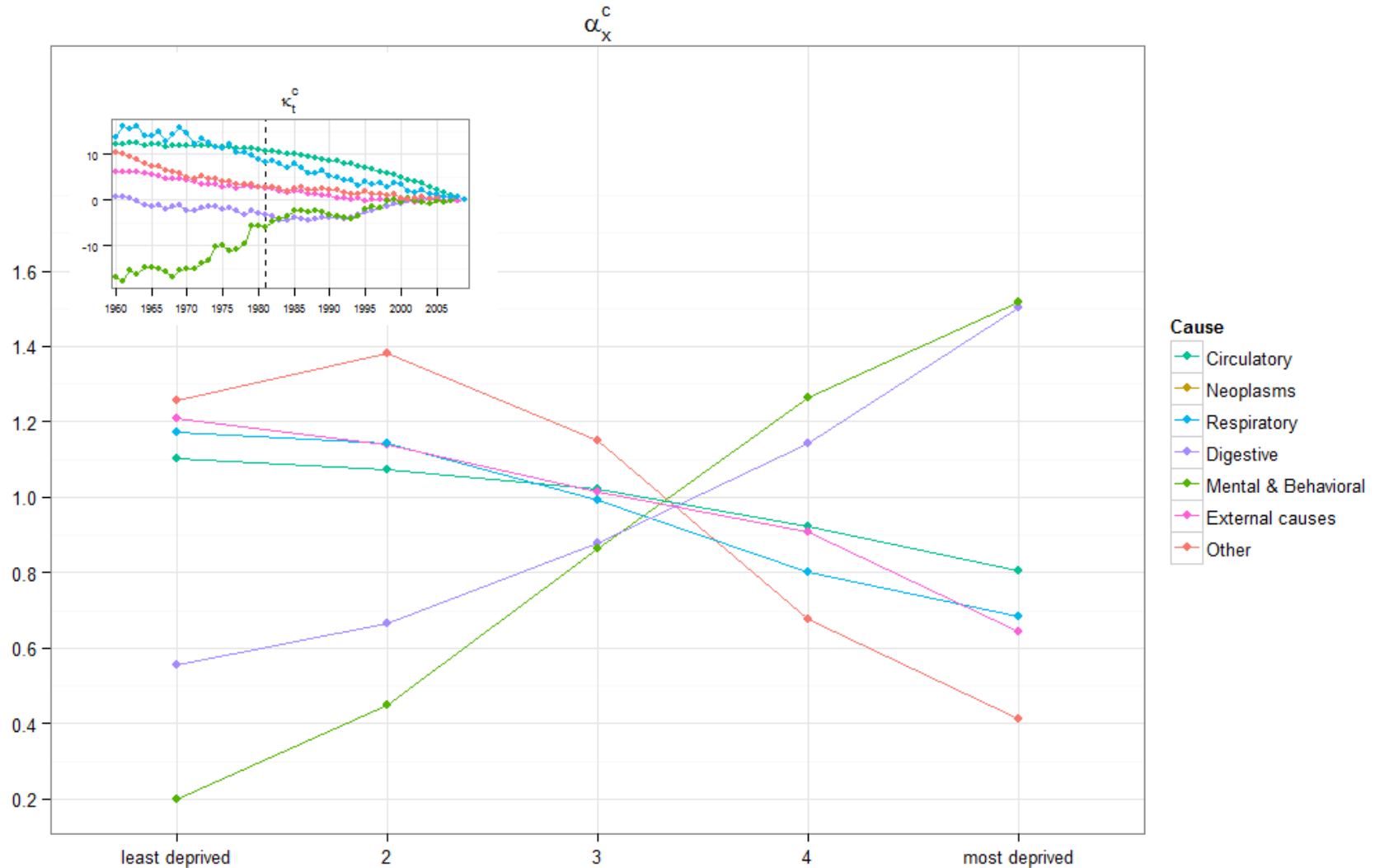
$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

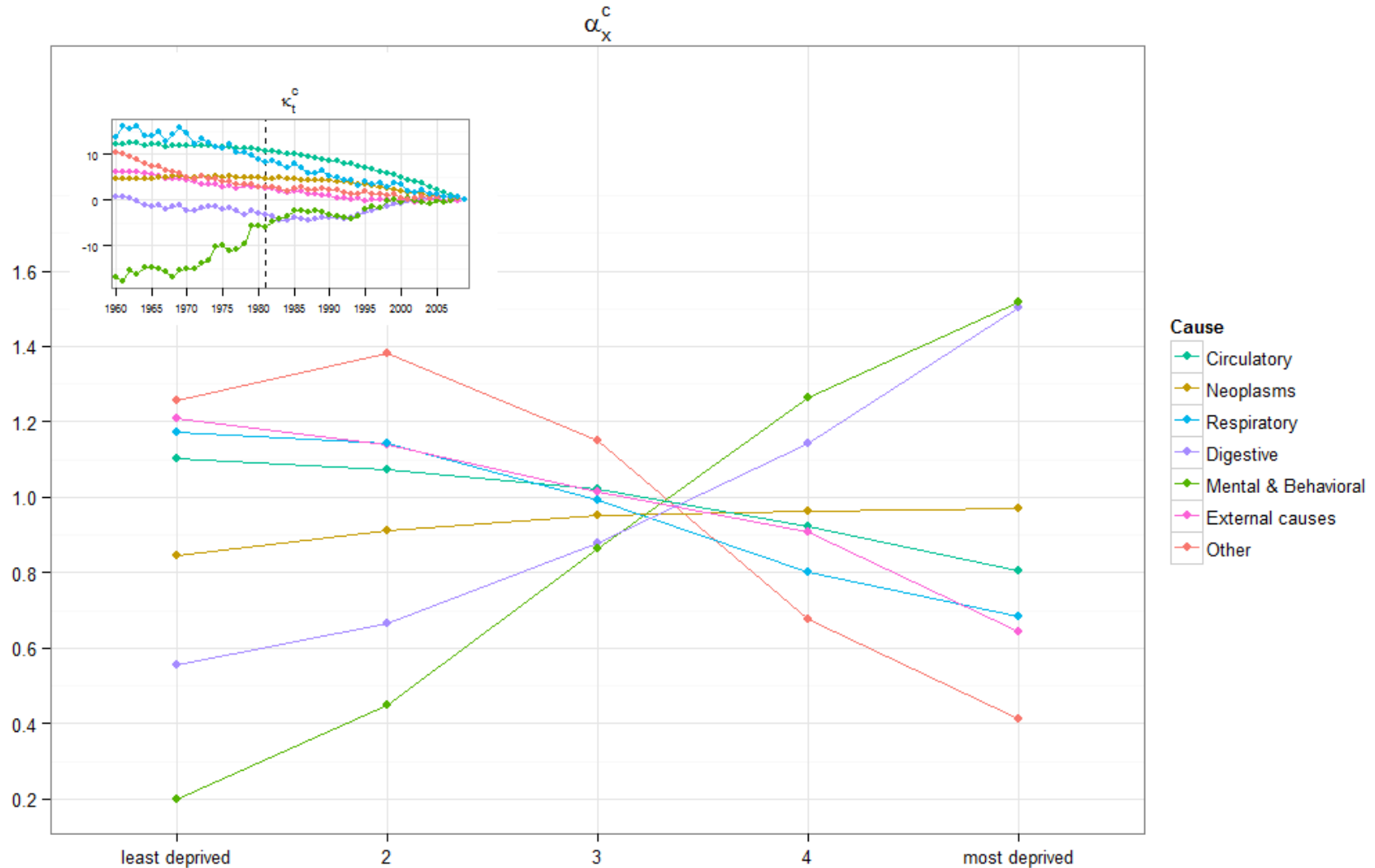
$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c + \sum_{i=1}^h \delta_x^{c,(i)} f^{(i)}(t)$$



Conclusions

- ▶ Introduce an extension of the Lee-Carter model to deal with production changes in cause-specific mortality
- ▶ Embed this model in a multipopulation framework to assess socio-economic differences in cause of death
- ▶ Application in the analysis of the extent of mortality differentials across deprivation subgroups in England for the period 1981- 2007
 - ▶ Clear inverse relationship between area deprivation and mortality for all causes
 - ▶ Reduction of differentials in cancer mortality
 - ▶ Offset of this reduction by marked differentials in digestive, respiratory and mental and behavioural diseases



Thank you!

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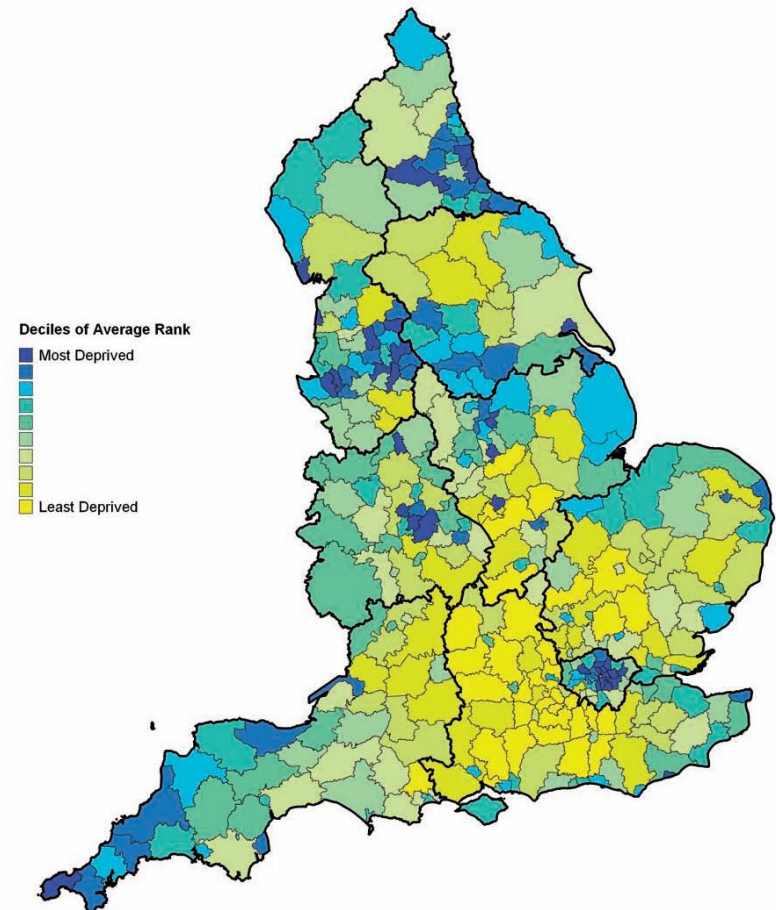
Reserve Slides

Case study: Mortality by deprivation in England

Application data - IMD 2007

- ▶ Socio-economic classification of the population obtained using the Index of Multiple Deprivation 2007 (IMD 2007)
- ▶ IMD 2007 combines indicators across 7 deprivation domains into a single deprivation score for each geographically defined Lower Layer Super Output Area (LSOA)
 - ▶ Income, employment, health, education, housing and services, crime, and living environment
- ▶ 32,482 LSOA in England with approximately 1,500 people each
- ▶ LSOAs ranked from 1 to 32,482 by their IMD 2007 score and grouped into quintiles
 - ▶ Q1: Least deprived quintile
 - ▶ Q5: Most deprived quintile

England - Average Rank District Level
Summary of the IMD 2007



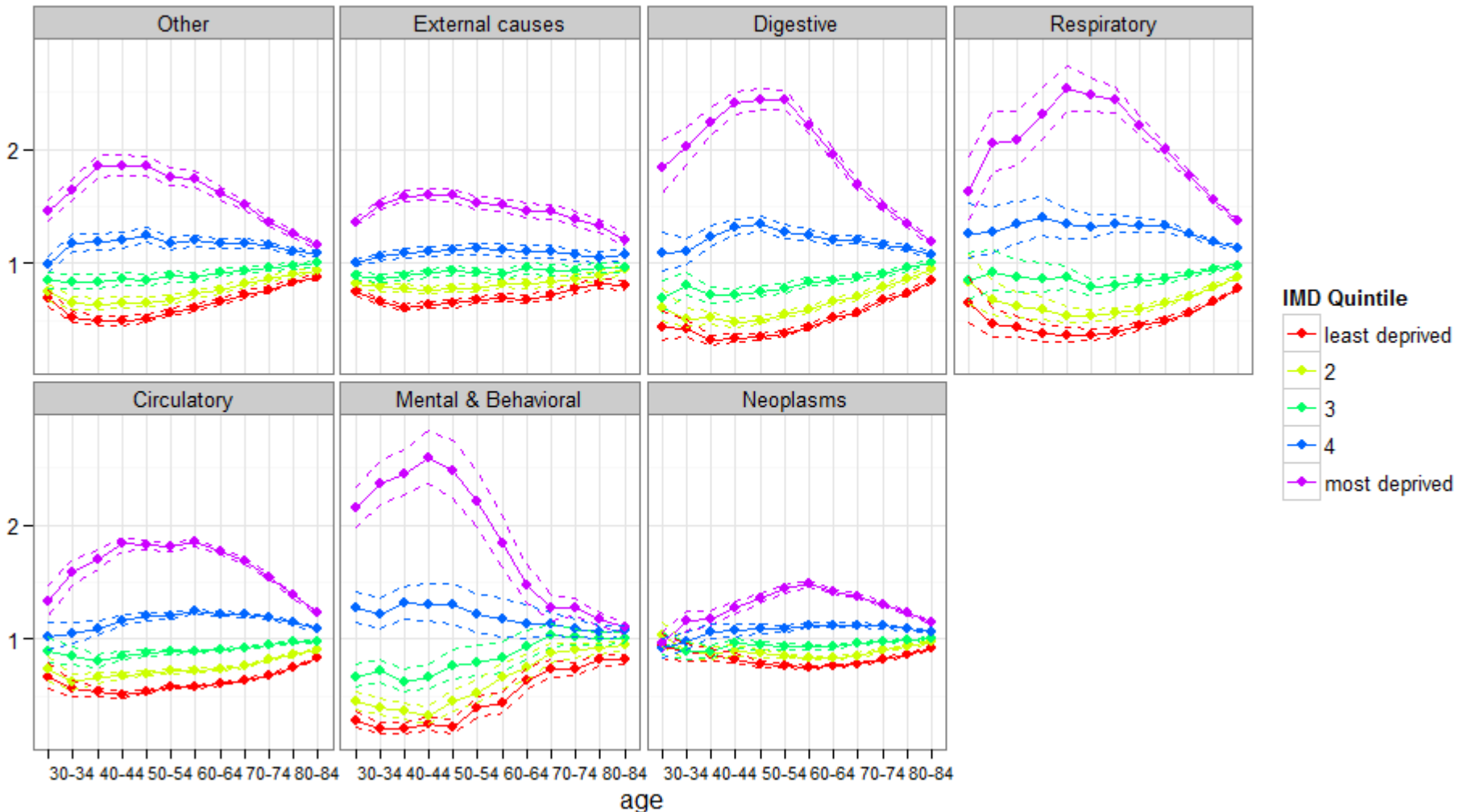
Source: Noble et al (2007)

Case study: Mortality by deprivation in England

Level differences by deprivation quintile

$$\log \mu_{xtg}^c = \alpha_x^c + \alpha_{xg}^c + \beta_x^c \lambda_g^c \kappa_t^c$$

$\exp(\alpha_{xg}^c)$



Case study: Mortality by deprivation in England

Trend differences by deprivation quintile

