

Inference and forecasting in the age-period-cohort model with unknown exposure with an application to Mesothelioma mortality

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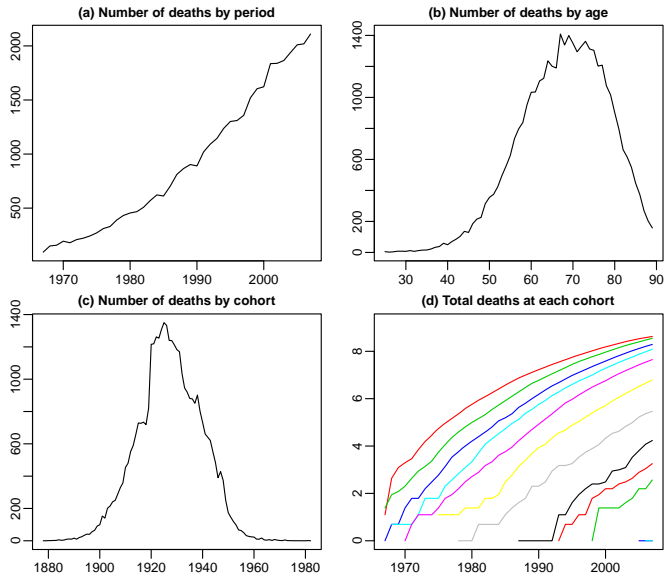
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Aim: forecast Mesothelioma mortality.

Methodological issues:

- Data: counts of deaths, but no exposure.
- Model number of deaths not mortality ratios
- Age-period-cohort model has identification problem
- Apply time series methods for forecast

Data



Data: counts of deaths in GB by age for 1967–2007.

Model: Distribution

Dose-Response Model

Y_{ij} counts of deaths, Z_{ij} exposure, by age i and period j .

Apply Poisson model with

$$\mathbf{E}(Y_{ij}|Z_{ij}) = \exp(\nu_{ij})Z_{ij} = \exp(\nu_{ij} + \log Z_{ij})$$

Modelling without exposure

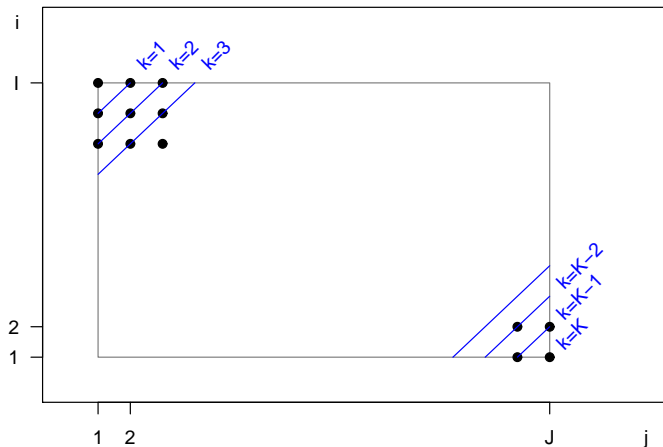
Y_{ij} counts of deaths, by age i, j .

Apply Poisson model with

$$\mathbf{E}(Y_{ij}) = \exp(\mu_{ij})$$

In Generalized Linear Model literature μ_{ij} is called *predictor*.

Model: Indices



$i = 1, \dots, I$ is age

$j = 1, \dots, J$ is period

$k = 1, \dots, K$ is cohort

where $k = I - i + J$ and $K = I + J - 1$.

Model: Age-Period-Cohort predictor

$$\mu_{ij} = \alpha_i + \beta_j + \gamma_k + \delta$$

Identification problem:

$$\mu_{ij} = (\alpha_i + a + di) + (\beta_j + b - dj) + (\gamma_k + c + dk) + (\delta - a - b - c - dI)$$

Traditional solution:

ad hoc identify by allocating **2** linear trends to **3** time effects.

Parsimonenous parametrisation:

$$\begin{aligned} \mu_{ij} &= \mu_{I1} + (i - I)(\mu_{I1} - \mu_{I-1,1}) + (j - 1)(\mu_{I2} - \mu_{I1}) \\ &\quad + \sum_{t=i}^{I-2} \sum_{s=t}^{I-2} \Delta^2 \alpha_{s+2} + \sum_{t=3}^j \sum_{s=3}^t \Delta^2 \beta_s + \sum_{t=3}^k \sum_{s=3}^t \Delta^2 \gamma_s \\ &= X'_{ij} \xi \quad \text{so } X'X \text{ invertible and } \xi \text{ freely varying} \end{aligned}$$

Analysis of model

With $\mu_{ij} = X'_{ij}\xi$ the likelihood is

$$\ell(\xi; Y) = \sum_{i,j} Y_{ij} X'_{ij} \xi - \sum_{i,j} \exp(X'_{ij} \xi).$$

ξ varies freely so model is *regular exponential family*.

Multinomial sampling

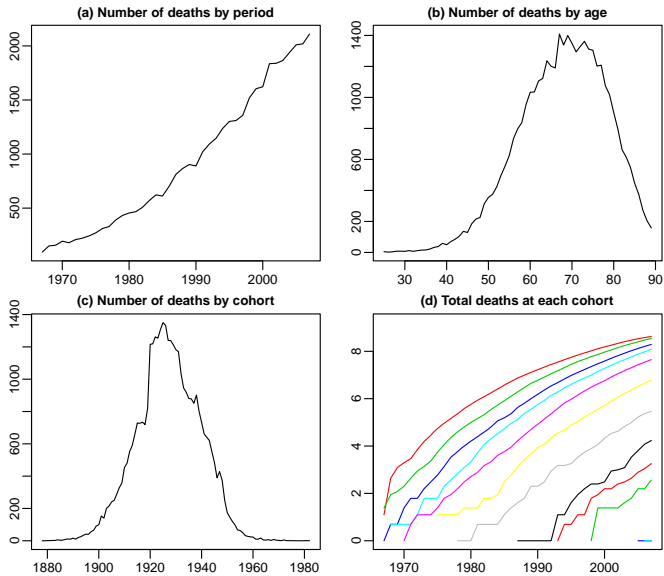
Let $\tau = \sum_{i,j} \exp(X'_{ij}\xi)$, $Y_{\bullet} = \sum_{i,j} Y_{ij}$, $\pi_{ij} = \exp(X'_{ij}\xi)/\tau$.

Then likelihood satisfies

$$\ell(\xi; Y) = \underbrace{Y_{\bullet} \log \tau - \tau}_{\text{Poisson}} + \underbrace{\sum_{i,j} Y_{ij} \log \pi_{ij}}_{\text{Multinomial}}.$$

Conduct inference about π_{ij} in conditional multinomial distribution of Y_{ij} given Y_{\bullet} .

Data

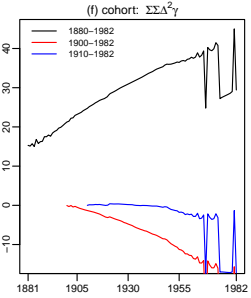
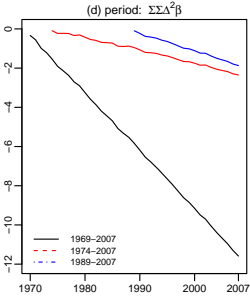
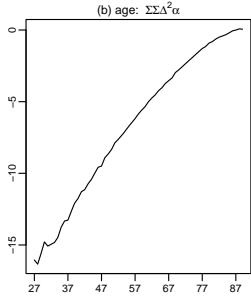
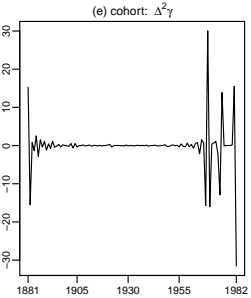
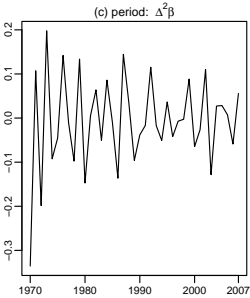
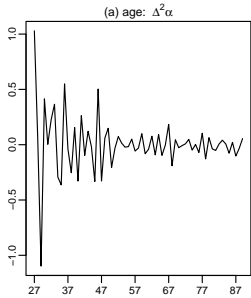


Data: counts of deaths in GB by age for 1967–2007.

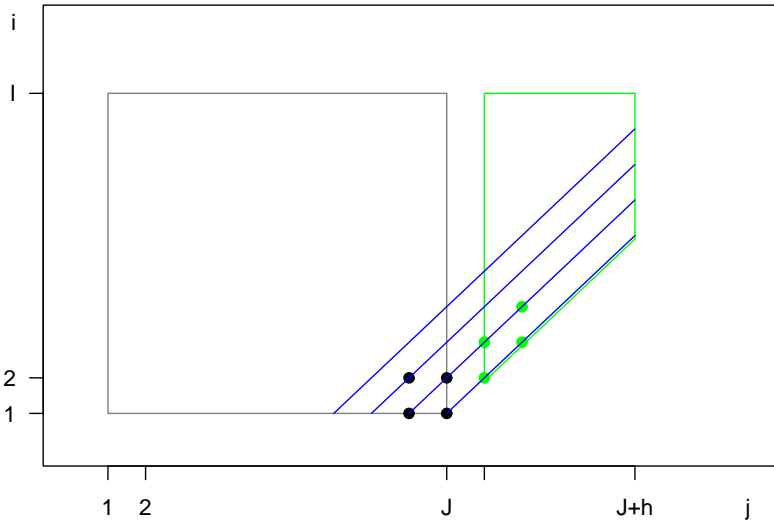
Deviances: APC or AC?

Model	Deviance	df	p
APC	2384.9	2457	0.852
AC vs. APC	56.8	39	0.033

Plots of Estimates



Forecast region



With $\mu_{ij} = \alpha_i + \beta_j + \gamma_k + \delta$
need to extrapolate β_j .

Extrapolation

With ad hoc identified parameters

forecasts should not depend on ad hoc identification.

This holds *iff* extrapolation is linear trend preserving.

With parsimonious parameter this is not constraint.

Extrapolation method: Let

$$x_j = \sum_{t=3}^j \sum_{s=3}^t \Delta^2 \beta_s \quad \text{for } j = 3, \dots, J.$$

Apply time series model $x_j = \nu_c + \nu_\ell j + \varepsilon_j$

Extrapolation

$$\tilde{x}_{J+h} = \hat{\nu}_c + \hat{\nu}_\ell(J+h) + \underbrace{\hat{\varepsilon}_J}_{\substack{\text{intercept} \\ \text{correction}}} = x_J + \hat{\nu}_\ell h.$$

Forecast

Point forecast

$$\begin{aligned}\tilde{\mu}_{i,J+h} &= \hat{\mu}_{I1} + (i - I)(\hat{\mu}_{I1} - \hat{\mu}_{I-1,1}) + (J + h - 1)(\hat{\mu}_{I2} - \hat{\mu}_{I1}) \\ &+ \sum_{t=i}^{I-2} \sum_{s=t}^{I-2} \Delta^2 \hat{\alpha}_{s+2} + \sum_{t=3}^J \sum_{s=3}^t \Delta^2 \hat{\beta}_s + \sum_{t=3}^k \sum_{s=3}^t \Delta^2 \hat{\gamma}_s + \hat{\nu}_\ell h\end{aligned}$$

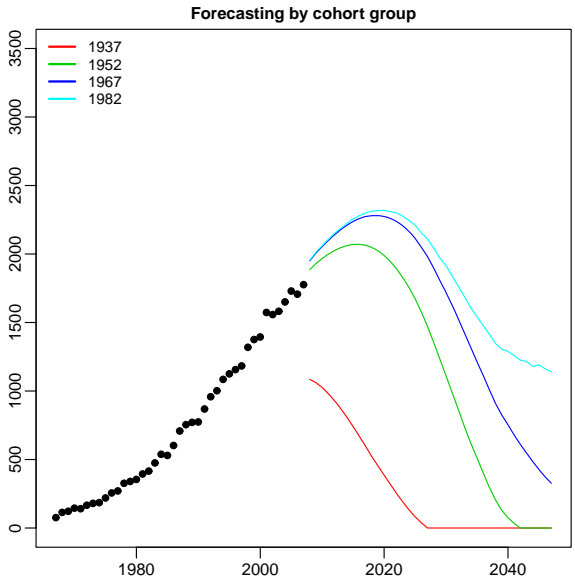
Distribution forecast

$$\tilde{Y}_{i,J+h}^{distribution} = \exp(\tilde{\mu}_{i,J+h}) + \mathbf{N}(0, s_{i,J+h}^2)$$

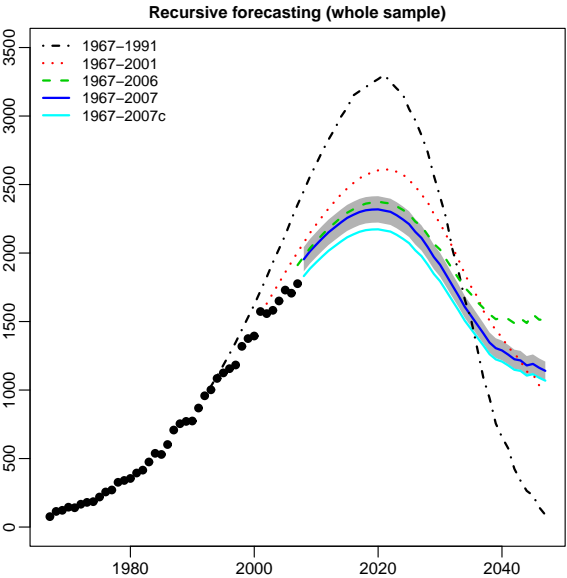
where

$$s_{i,J+h}^2 = s_{extrapolation\ error}^2 + \underbrace{s_{estimation\ error}^2}_{\text{negligible for near future}} \cdot$$

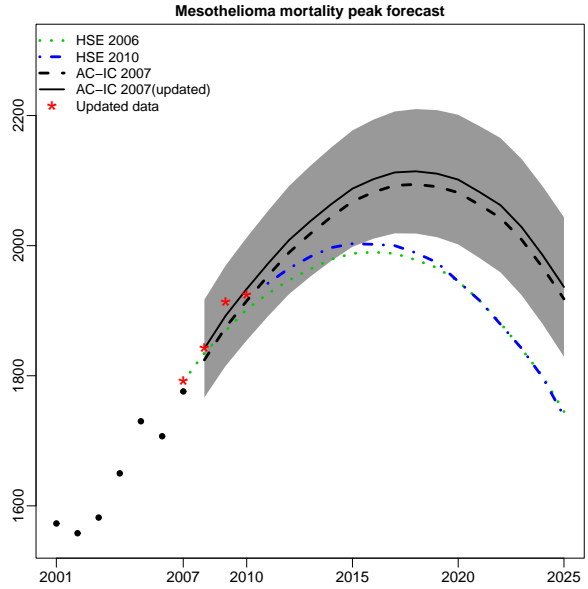
Forecast: Cohort components



Forecast: Recursive



Forecast: Recent data



References

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