

A Comprehensive Framework for Mortality Forecasting

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Models for central mortality rates m_{xt} over age x and time t include:

- Generalised bilinear (e.g. Lee Carter with cohort)

$$\log m_{xt} = \alpha_x + \beta_x \kappa_t + \gamma_{t-x}$$

- Generalised linear (e.g. APC with age-period interaction)

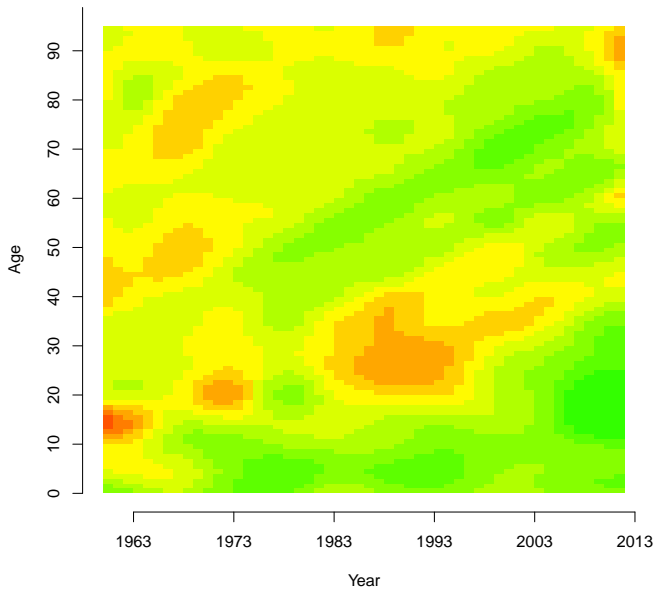
$$\log m_{xt} = \alpha_x + t\beta_x + \kappa_t + \gamma_{t-x}$$

- semi-parametric

$$\log m_{xt} = s(x, t)$$

- generalised additive (GAM)

$$\log m_{xt} = s_\alpha(x) + t s_\beta(x) + \kappa_t + s_\gamma(t - x)$$



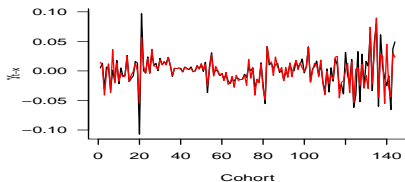
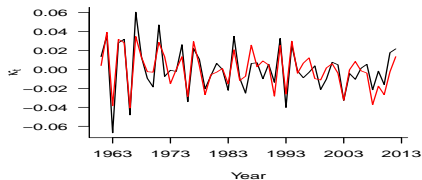
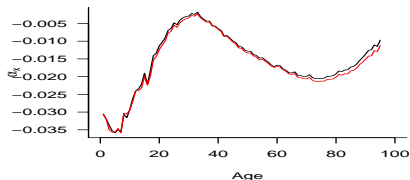
Age-period-cohort (APC) model for mortality improvements

$$\log \frac{m_{xt}}{m_{x,t-1}} = \alpha_x + \kappa_t + \gamma_{t-x} \quad (1)$$

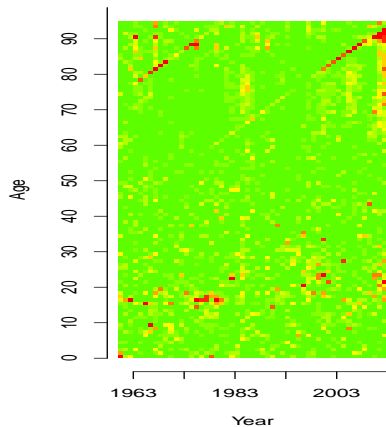
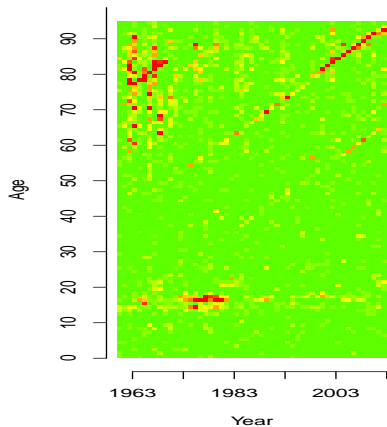
or equivalently APC model for mortality rates, with age-period interaction

$$\log m_{xt} = m_{x0} + \alpha_x t + \kappa_t + \gamma_{t-x} \quad (2)$$

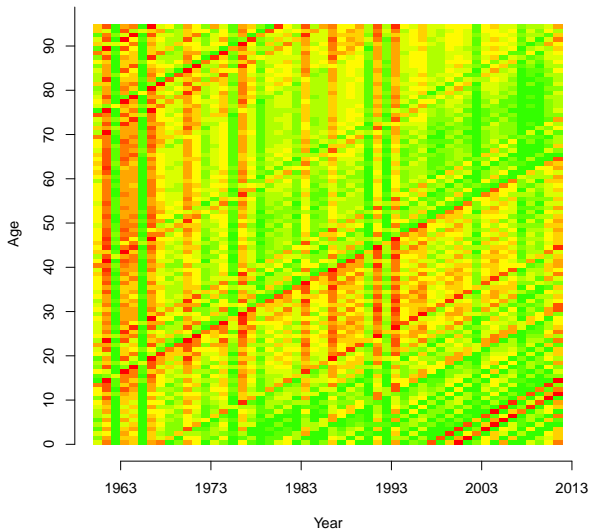
Estimation of the model parameters can be based on the negative binomial log-likelihood: which accounts for over dispersion (relative to Poisson).



Maximum likelihood estimates of the parameters of model (1) under the Poisson model (black line) and the negative binomial model (red line), data for males 1961-2013.



The P-spline approach allowing for overdispersion (left panel) and model (1) under the negative binomial distribution (right panel).

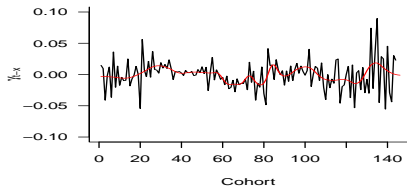
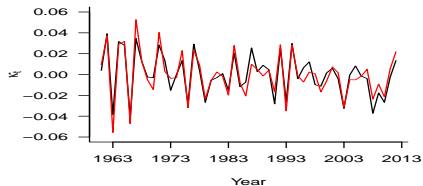
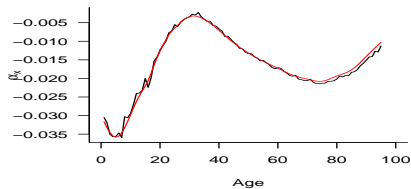
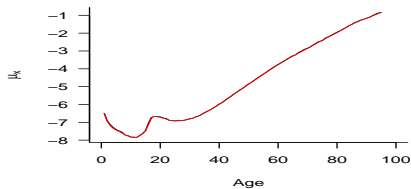


To obtain smoother estimates modify (2) to a generalised additive model (GAM):

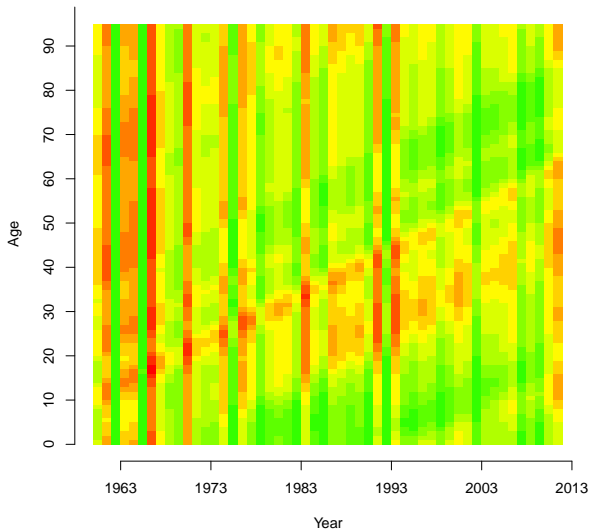
$$\log m_{xt} = s_{\mu}(x) + s_{\alpha}(x)t + \kappa_t + s_{\gamma}(t - x). \quad (3)$$

where s_{μ} , s_{α} and s_{γ} are arbitrary smooth functions.

This is a *generalised additive model* which can be robustly fitted in standard statistical software (Wood, 2006).



Estimates of the parameters of model (3), data for males 1961-2013, (red lines) superimposed over the corresponding estimates for model (2) (black lines).



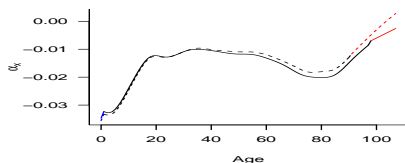
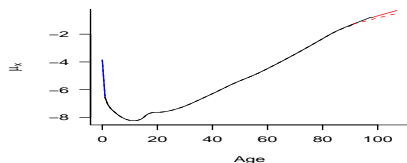
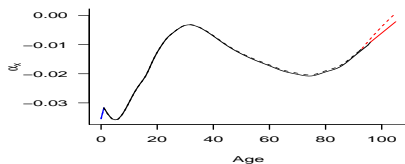
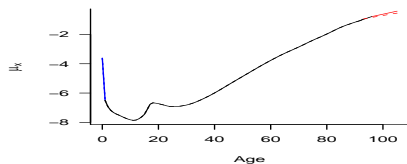
For the highest ages x , use parametric models:

$$\log m_{xt} = \mu_0 + \mu_1 x + (\alpha_0 + \alpha_1 x)t + \kappa_t + s_\gamma(t - x) \quad x > x_0 \quad (4)$$

or

$$\log \frac{m_{xt}}{\beta - m_{xt}} = \mu_0 + \mu_1 x + (\alpha_0 + \alpha_1 x)t + \kappa_t + s_\gamma(t - x) \quad x > x_0 \quad (5)$$

where κ_t , $s_\gamma(t - x)$ are estimates obtained from fitting (3) to the main body of the data ($0 < x \leq x_0$).

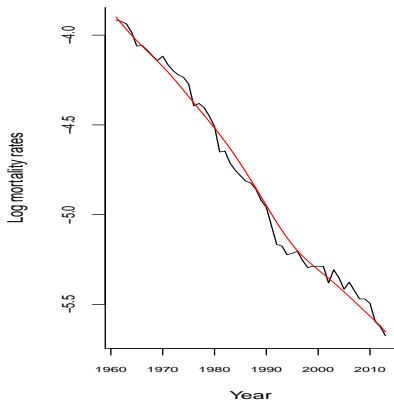
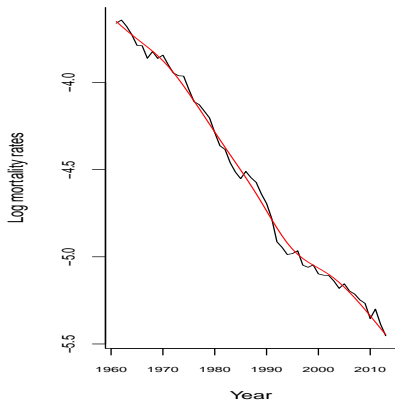


Estimates of the parameters of models (3), (4) and (5), 1961-2013, for males (upper panels; $x_0=95$ for log-linear model (solid line) and $x_0=92$ for logistic model (dashed line)) and females (lower panels; $x_0=97$ for log-linear model (solid line) and $x_0=90$ for logistic model (dashed line)).

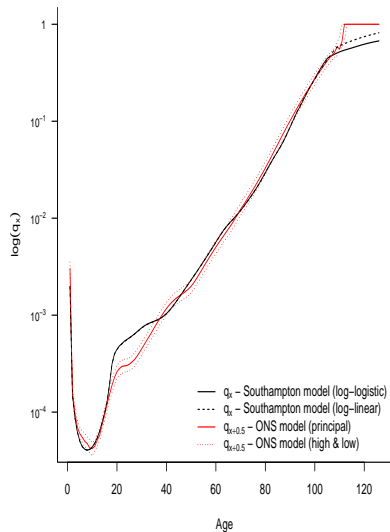
For infants (age 0) we use:

$$\log \mu_{0t} = \mu_0 + \alpha_0 t + s_\gamma(t - x) \quad (6)$$

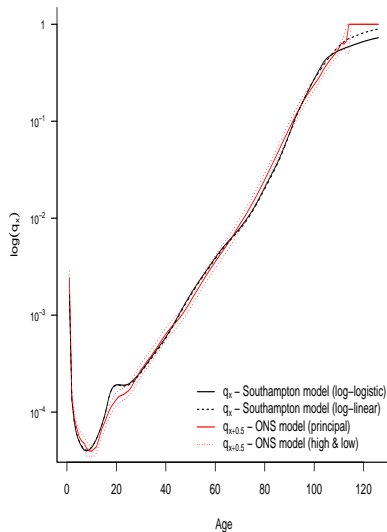
where $s_\gamma(t - x)$ are estimates obtained from fitting model (3) to the main body of the data ($0 < x \leq x_0$).



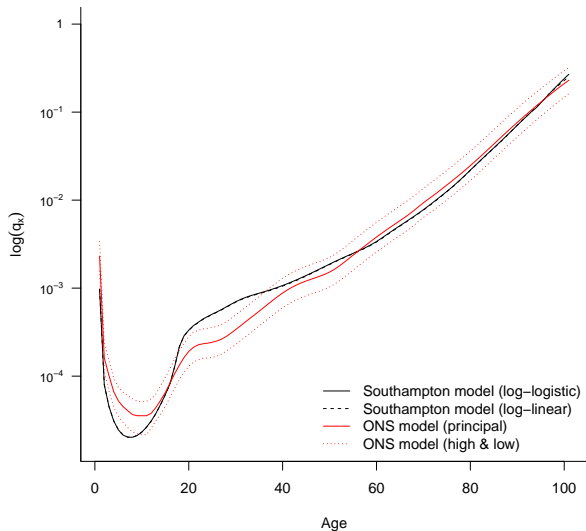
Estimates of infant mortality rates, 1961-2013, for males (left panel) and females (right panel) using model (6; red lines), compared with observed rates (black lines).



(a) Males



(b) Females



- Completely integrated estimation
- Bayesian approach with expert opinion and full uncertainty quantification
- Time-varying old-age threshold and/or mortality asymptote
- Models by causes of death

Thank you for listening