

# A matter of life and death: mortality estimation and prediction

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# Why mortality matters

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Mortality estimates forecasts are vitally important

- Tax and Expenditure
- Healthcare
- Pensions and Insurance
  
- **Goal:** Forecast future mortality with a realistic quantification of uncertainty.

# The life table

A static summary of the distribution of age at death for a population:

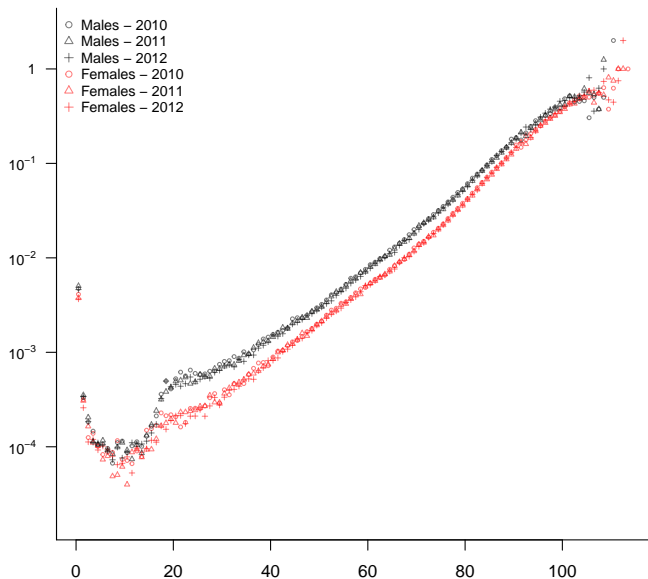
English Life Tables No 17

Period expectation of life

Based on data for England and Wales for the years 2010-2012

Age	Males								
	$x$	$m_x$	$q_x$	$l_x$	$d_x$	$L_x$	$T_x$	$\mu_x$	$e_x$
	0	0.004757	0.004746	100000	475	99576.3	7896837		78.97
	1	0.000306	0.000306	99525	30	99510.2	7797072	0.000369	78.34
	2	0.000207	0.000207	99495	21	99484.6	7697562	0.000246	77.37
	3	0.000147	0.000147	99474	14	99467.0	7598078	0.000172	76.38
	4	0.000115	0.000115	99460	12	99453.9	7498612	0.000128	75.39
	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
	109	0.676172	0.491440	8	4	5.5	11	0.661588	1.43
	110	0.701065	0.503943	4	2	2.8	5	0.685990	1.39
	111	0.725677	0.516003	2	1	1.4	3	0.709972	1.34
	112	0.750015	0.528125	1	1	0.7	1	0.733841	1.30

# Crude mortality rates 2010-2012



## A basic smoothing model

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We estimate the *Central Mortality Rate*

$$m_x \equiv \frac{\text{Expected number of deaths aged } x \text{ in 1 year}}{\text{Population aged } x \text{ at risk}}$$

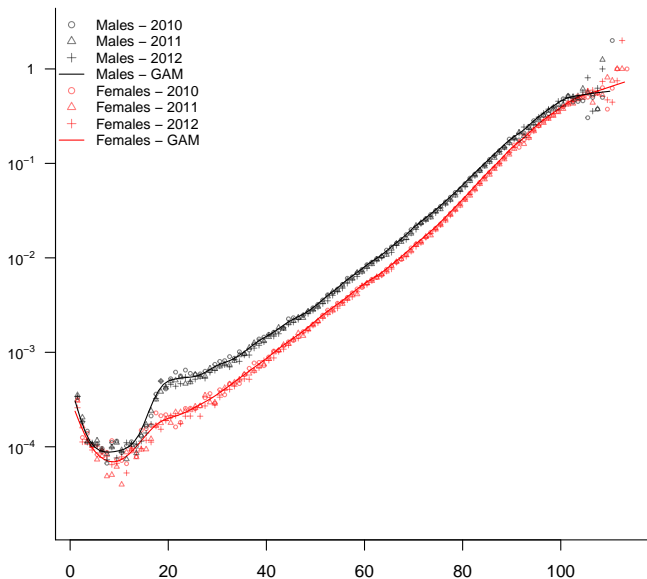
In a generalised additive (smooth) statistical model, we estimate the  $m_x$  by

$$\hat{m}_x = \exp s(x)$$

where  $s$  is a function which 'trades off' smoothness as a function of age with how closely the  $\hat{m}_x$  are to the corresponding observed rates

$$\frac{\text{Observed number of deaths aged } x \text{ in 1 year}}{\text{Population aged } x \text{ at risk}}$$

# Smooth mortality rates 2010-2012



## Models for older ages and extrapolation

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To obtain a more robust fit at older ages, and to extrapolate the mortality function  $m_x$  beyond the range of the observed data, one might use the log-linear Gompertz model

$$\log m_x = \beta_0 + \beta_1 x, \quad x \geq x_0$$

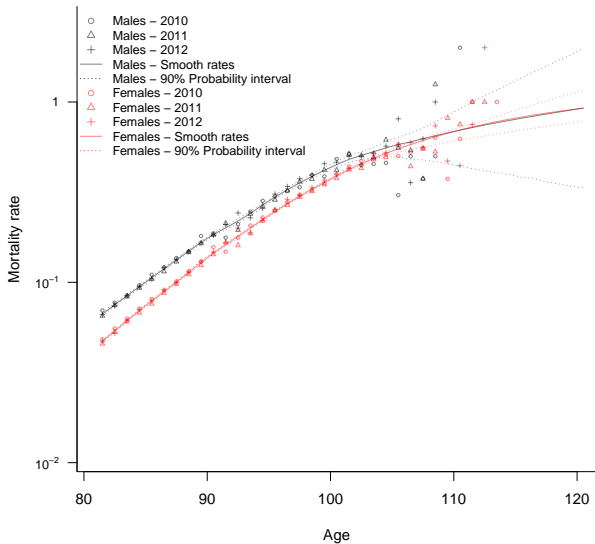
where  $x_0$  is a suitable threshold

or

$$m_x = \frac{\beta_2 \exp(\beta_0 + \beta_1 x)}{1 + \exp(\beta_0 + \beta_1 x)}, \quad x \geq x_0$$

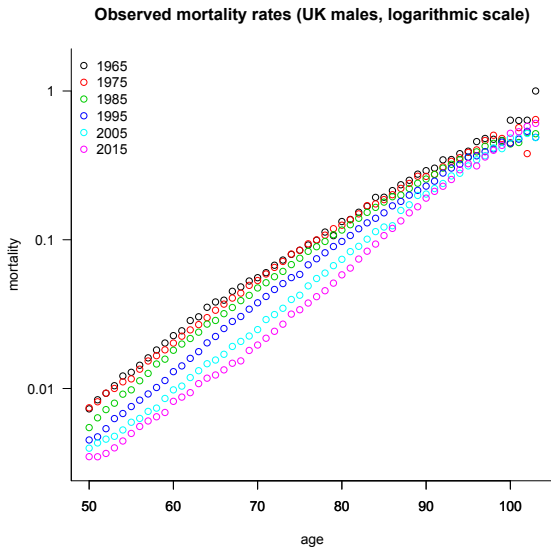
where mortality rates flatten off, converging to the limit  $\beta_2$  as  $x \rightarrow \infty$ .

# ELT17 modelling at high ages





# Mortality improvement



## Dynamic models and projection

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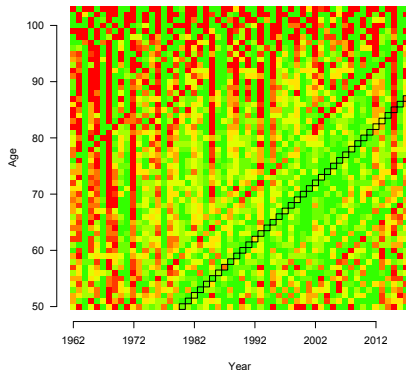
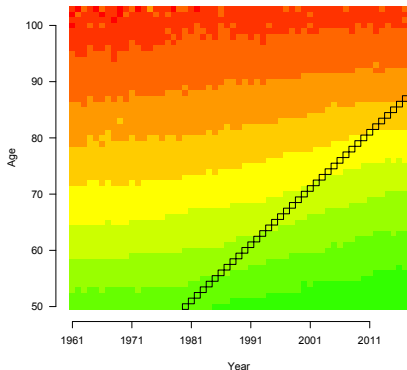
Now mortality varies not just by age, but over time as well.

We denote by  $m_{xt}$  the central mortality rate aged  $x$ , in year  $t$ , in population of interest, for  $t = 1, \dots, T = \text{present}$ .

Statistical mortality models provide a framework for projecting  $m_{xt}$  etc for  $t = T + 1, T + 2, \dots$

Hence providing us with the information we need for planning.

# The three time dimensions



A cohort is a subpopulation sharing a common birth-year. (1930 birth cohort identified above)

# The model

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

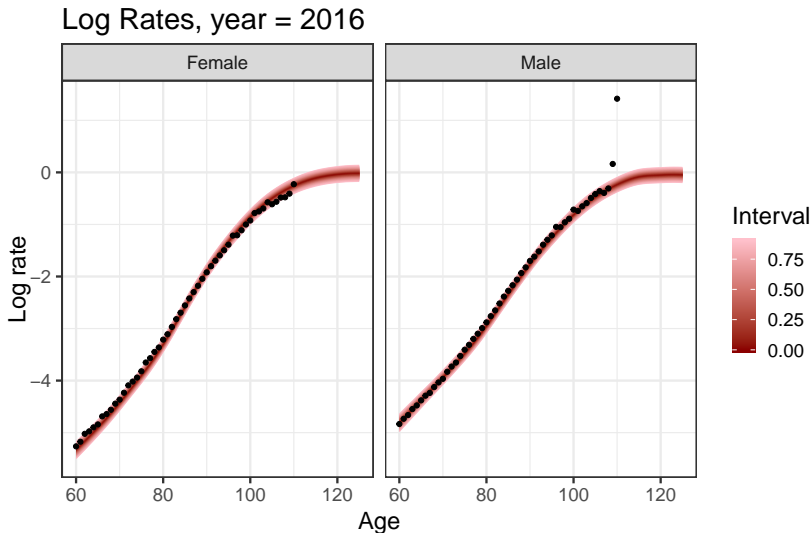
$$\log \frac{m_{xt}}{m_{x,t-1}} = s_{\alpha}(x) + \kappa_t + s_{\gamma}(t - x)$$

For the highest ages  $x > x_0$ , use the structured model

$$m_{xt} = \frac{\beta \exp(\mu_0 + \mu_1 x + (\alpha_0 + \alpha_1 x)t)}{1 + \exp(\mu_0 + \mu_1 x + (\alpha_0 + \alpha_1 x)t)} \exp(\kappa_t + s_{\gamma}(t - x))$$

# Forecast (with uncertainty)

Fit on data up to 2006, 10 year projection.



# Forecast life expectancy (with uncertainty)

Fit on data up to 2006.

