

Modelling the Age Patterns of Demographic Events

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- 1 Motivation for using model schedules and examples.
- 2 Mathematical Models.
- 3 Relational Models.
- 4 Model Life Tables.
- 5 Schedules over time.

- Intensity of demographic events varies sharply with age.
- Large sets of data can be cumbersome or unavailable.
- Demographers have searched for more parsimonious representations (a.k.a. model schedules) of age variation.
- Model schedules allow:
 - ① Identification of irregularities in the data.
 - ② Simplify assumptions for population projections.
 - ③ Estimation of incomplete data.
 - ④ Explanation of behavioral patterns of populations.

Model Schedules

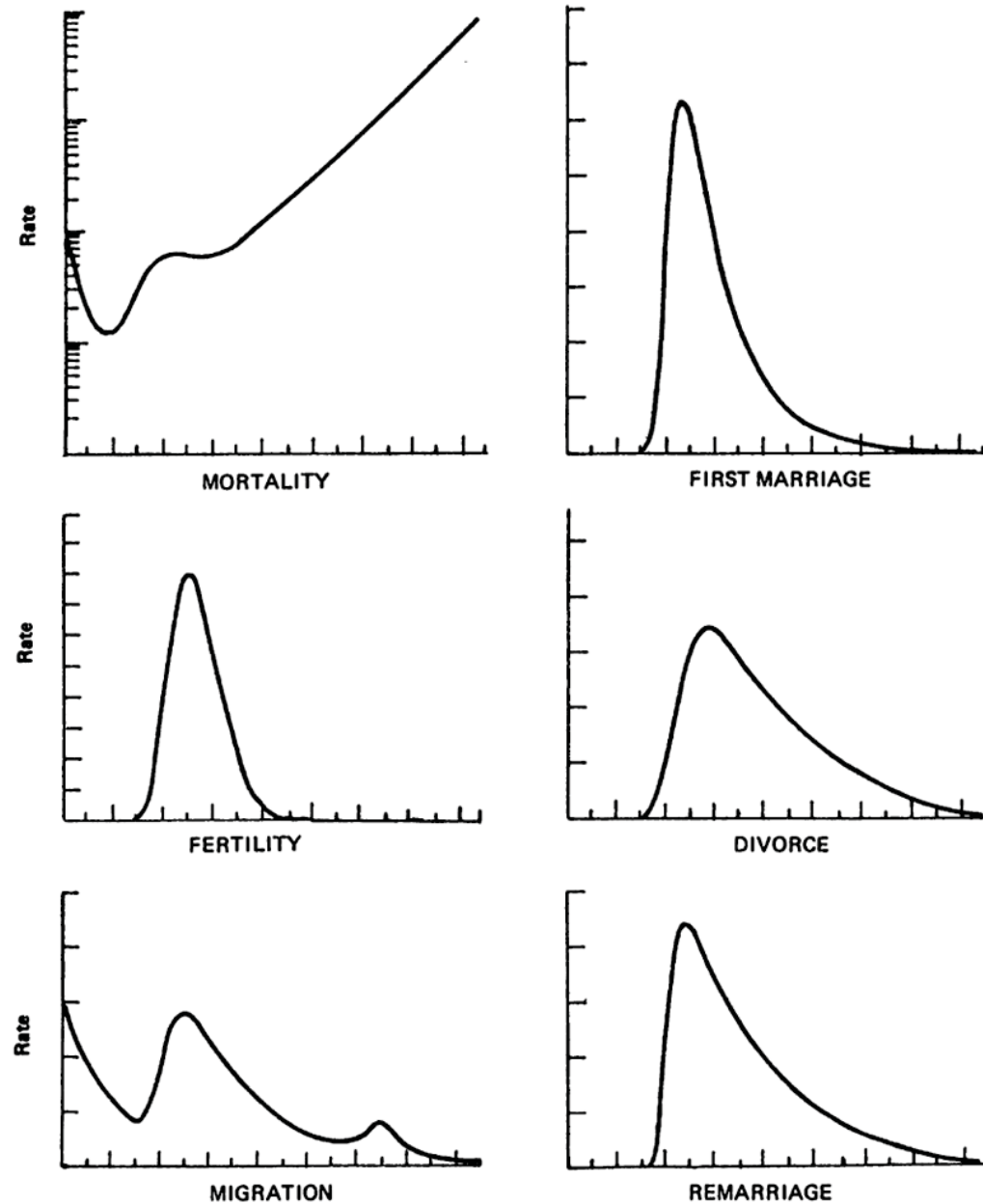
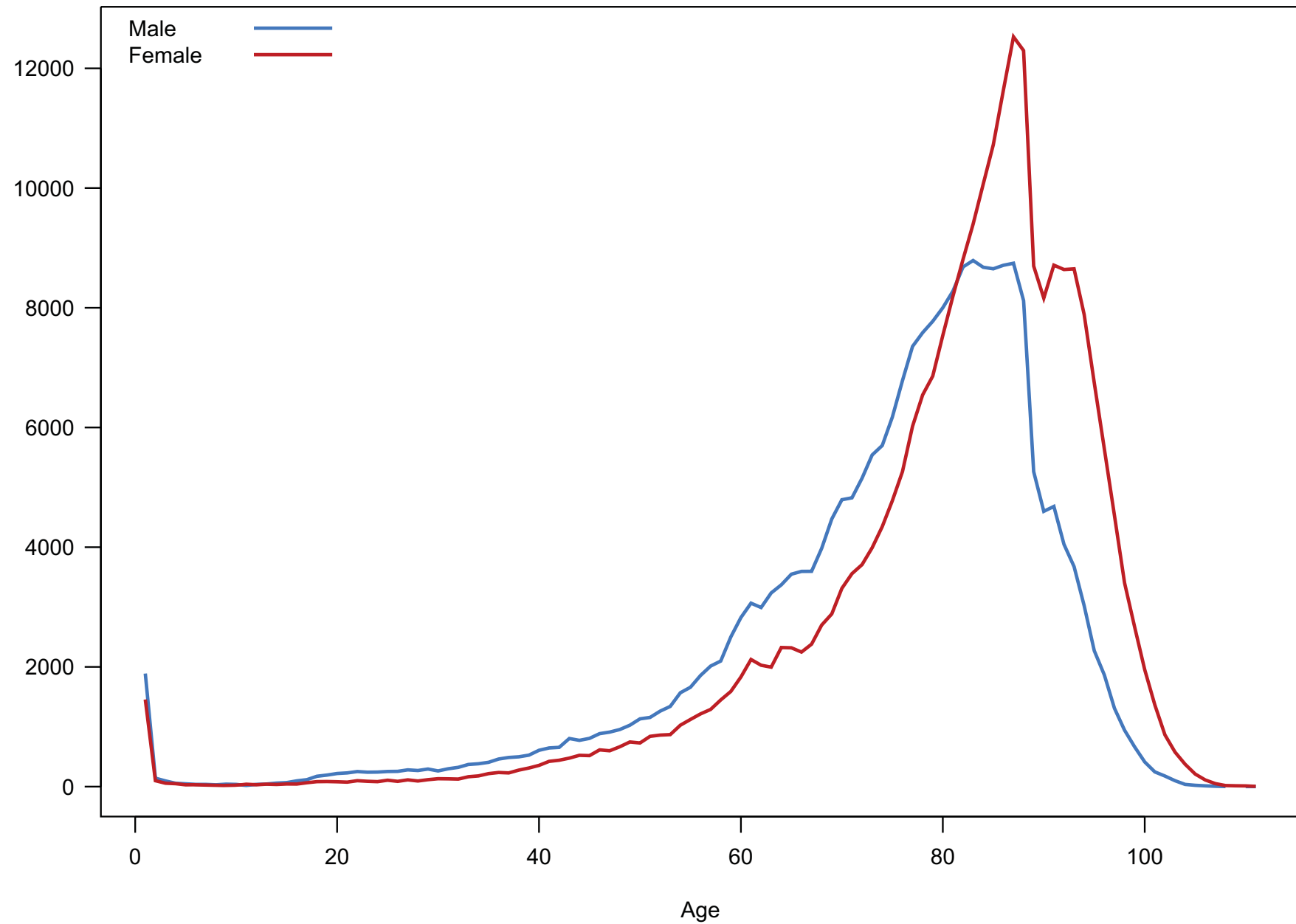
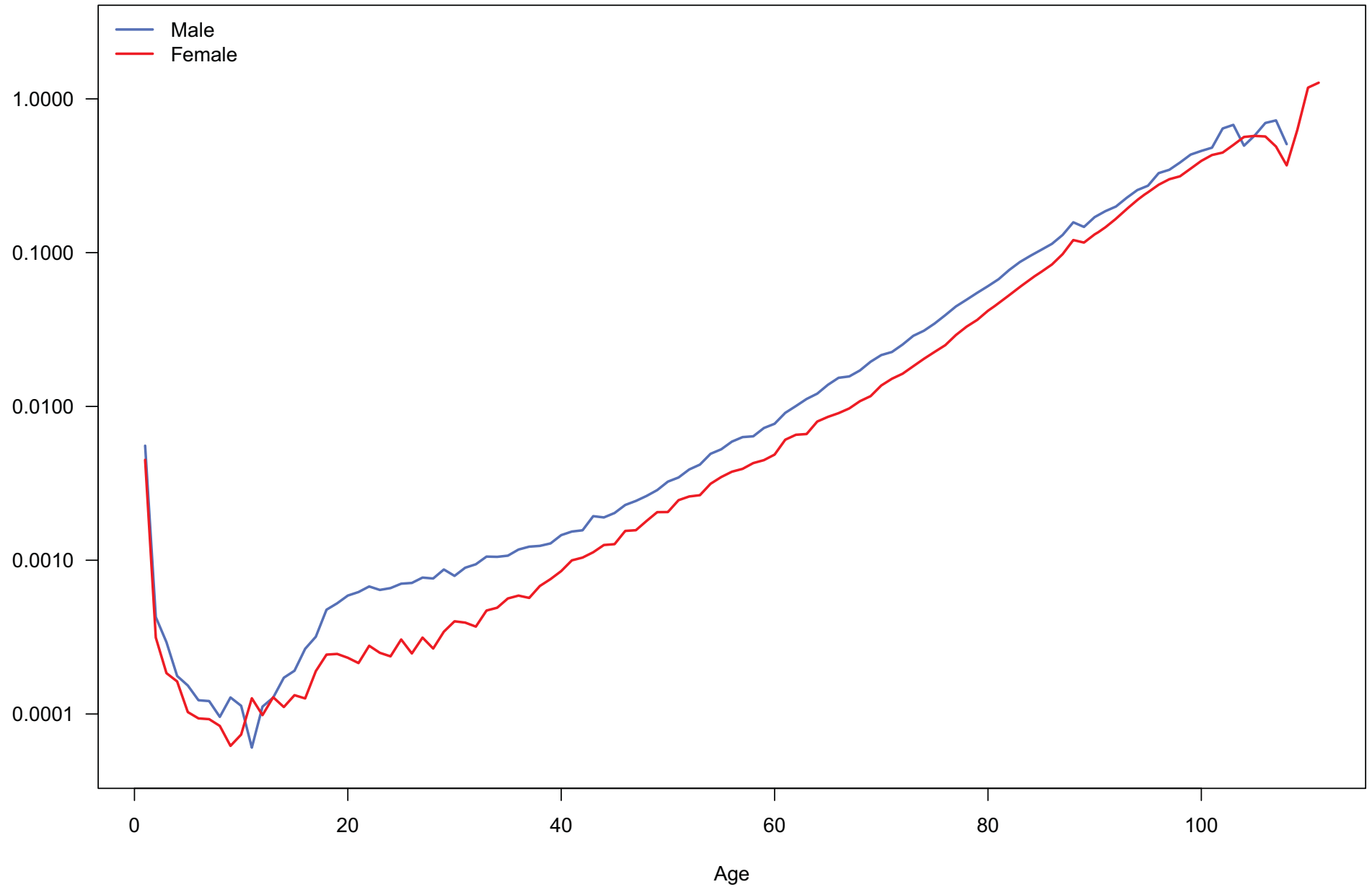


Figure 1. Multistate Schedules. Source: Rogers (1982).

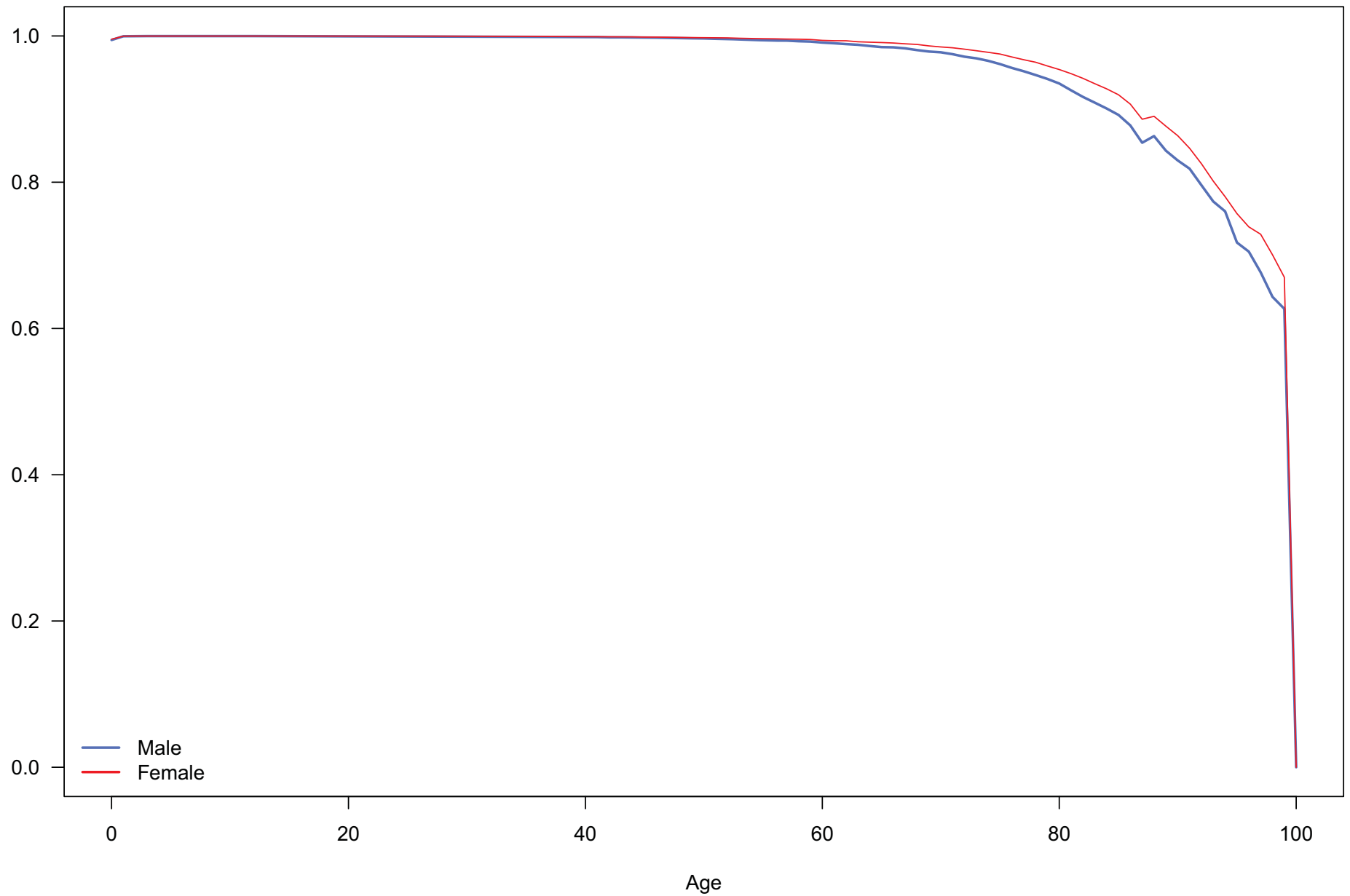
Number of Deaths (D_x) in England and Wales, 2007



Mortality Rate (m_x) in England and Wales, 2007



Survivorship (l_x) for England and Wales, 2007



- First demographic mathematical models developed for mortality.
- Mortality functions link the age-specific probability of death, m_x to age (x) using a function, $f(x)$
- Gompertz (1825) suggested that a *law of geometric progression pervades* in mortality after a certain age.

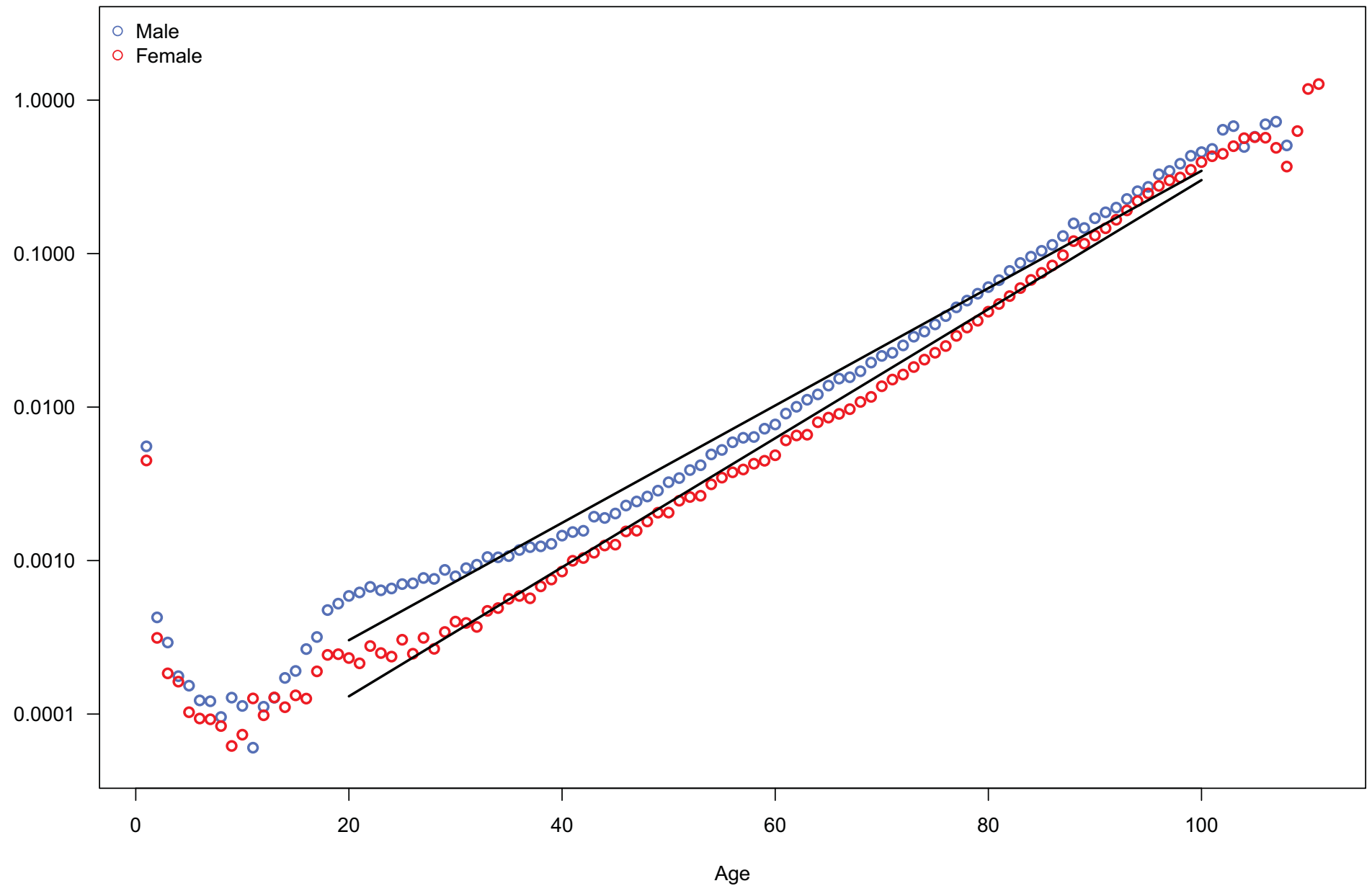
$$f(x) = \alpha e^{\beta x} \quad (1)$$

- For log-mortality this is a linear function of age

$$\log f(x) = \log(\alpha) + \beta x \quad (2)$$

Can use standard regression approach to calculate α and β

Fits of Gompertz Models for England and Wales, 2007

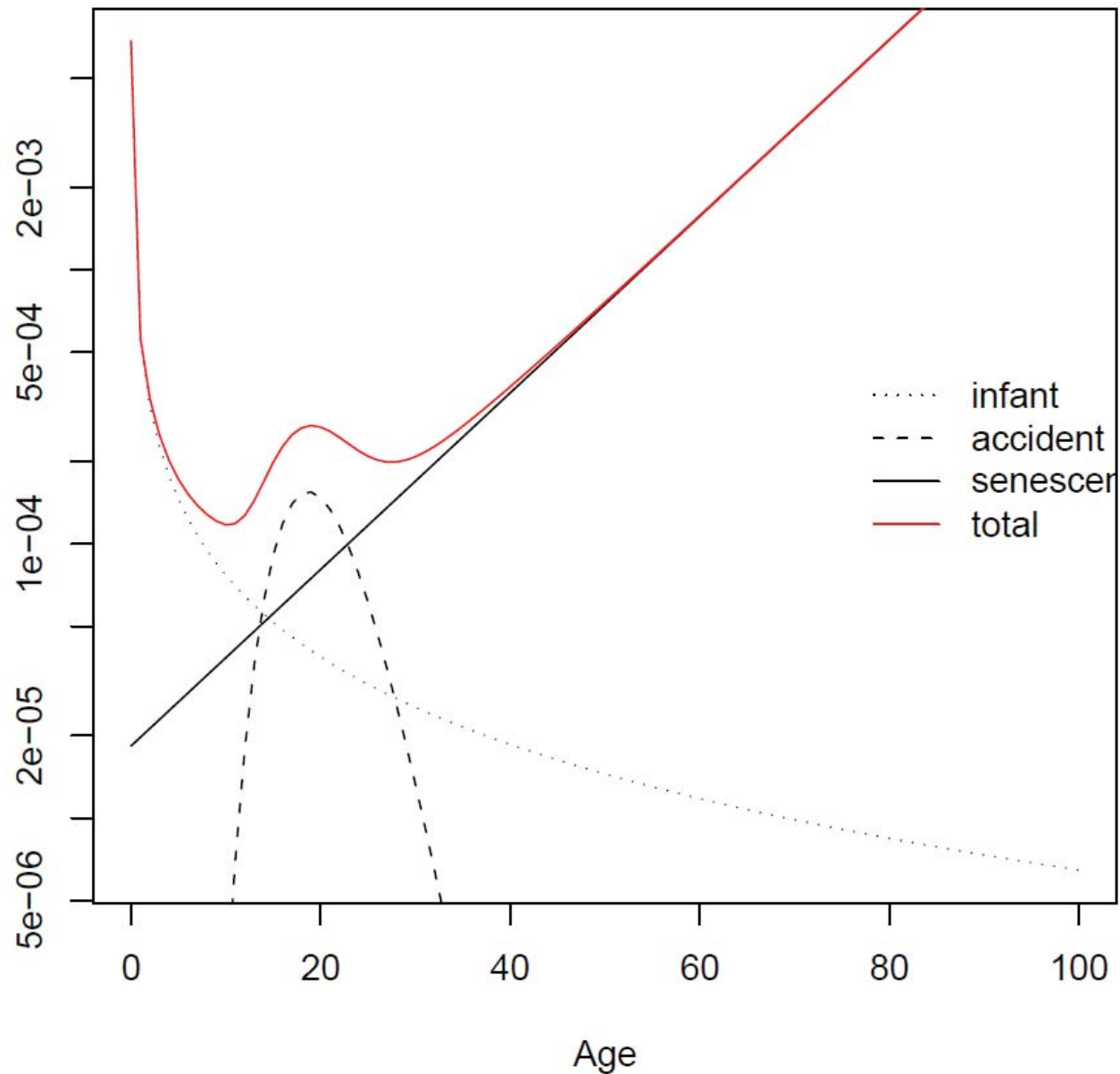


Competing Components Models

- Far more complex mathematical models have been developed.
- Most build together lines and curves for different components of the mortality (or fertility or migration) schedules.
- Heligman-Pollard is one such example for mortality:

$$f(x) = A^{(x+B)^C} + D \exp[-E \{\log \frac{x}{F}\}^2] + \frac{GH^x}{1 + GH^x} \quad (3)$$

Heligman-Pollard for 1995 UK mortality, source: Jones (2005)



Competing Components Models

- Double exponential model used for fertility (amongst others)

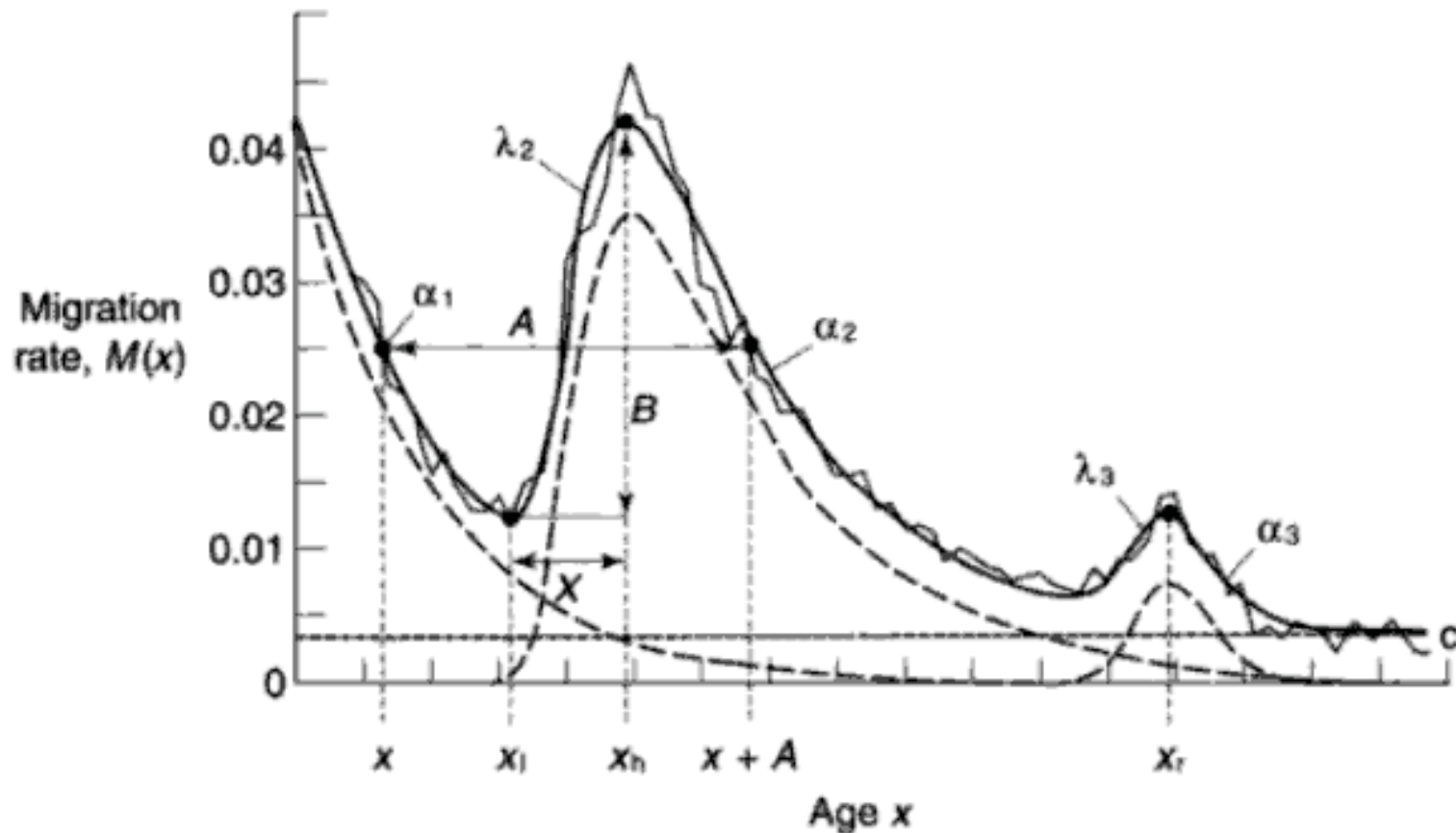
$$f(x) = a \exp(-\alpha(x - \mu) - \exp(-\lambda(x - \mu)))$$

- Coale-Trussel formula (and other formulations) also popular. These interact nuptiality and fertility schedules.
- Rogers and Castro (1981) propose a 7 or 11 parameter model for migration

$$m(x) = a_1 \exp(-\alpha_1 x) + a_2 \exp(-\alpha_2(x - \mu_2) - \exp(-\lambda_2(x - \mu_2))) + R$$

where $R = a_3 \exp(\alpha_3(x - \mu_3) - \exp(-\lambda_3(x - \mu_3)))$ if retirement peak.

Migration Models, source: Rogers and Castro (1981)



α_1 = rate of descent of pre-labor force component
 λ_2 = rate of ascent of labor force component
 α_2 = rate of descent of labor force component
 λ_3 = rate of ascent of post-labor force component
 α_3 = rate of descent of post-labor force component
 c = constant

x_l = low point
 x_h = high peak
 x_r = retirement peak
 x = labor force shift
 A = parental shift
 B = jump

Advantages:

- Reduce dimensions, small number of parameters relative to number of data points.
- Parameters can have nice interpretations.
- Good for comparisons.

Disadvantages

- Hard to fit.
- Starting values difficult to find.
- Estimates frequently unstable.
- Might not adapt to new sources of demographic change.

- Brass (1971) proposed a model to relate observed mortality data (which might be incomplete) to a standard schedule, $X^{(s)}$
- Requires logit transformation on our survivorship probabilities

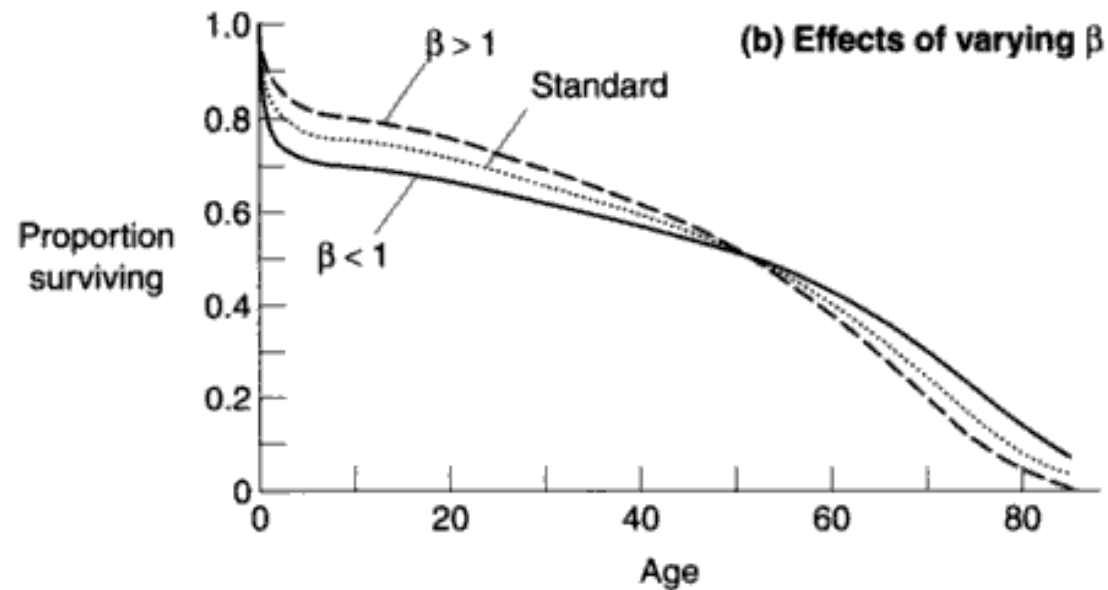
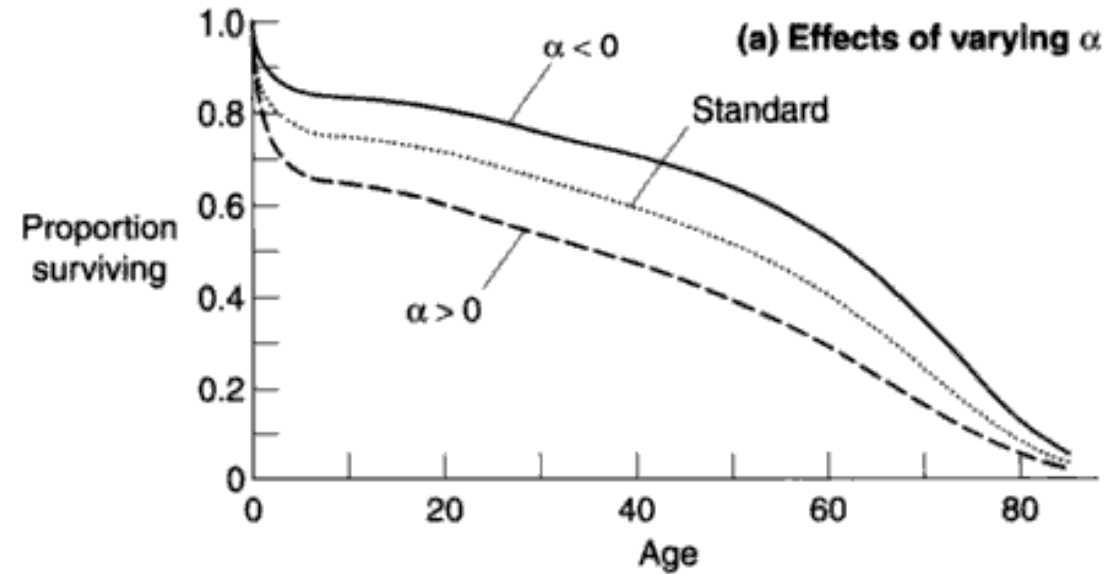
$$\hat{Y} = \text{logit}(x) = \alpha + \beta X^{(s)}$$

α is interpreted as the level of mortality

β is the shape of mortality

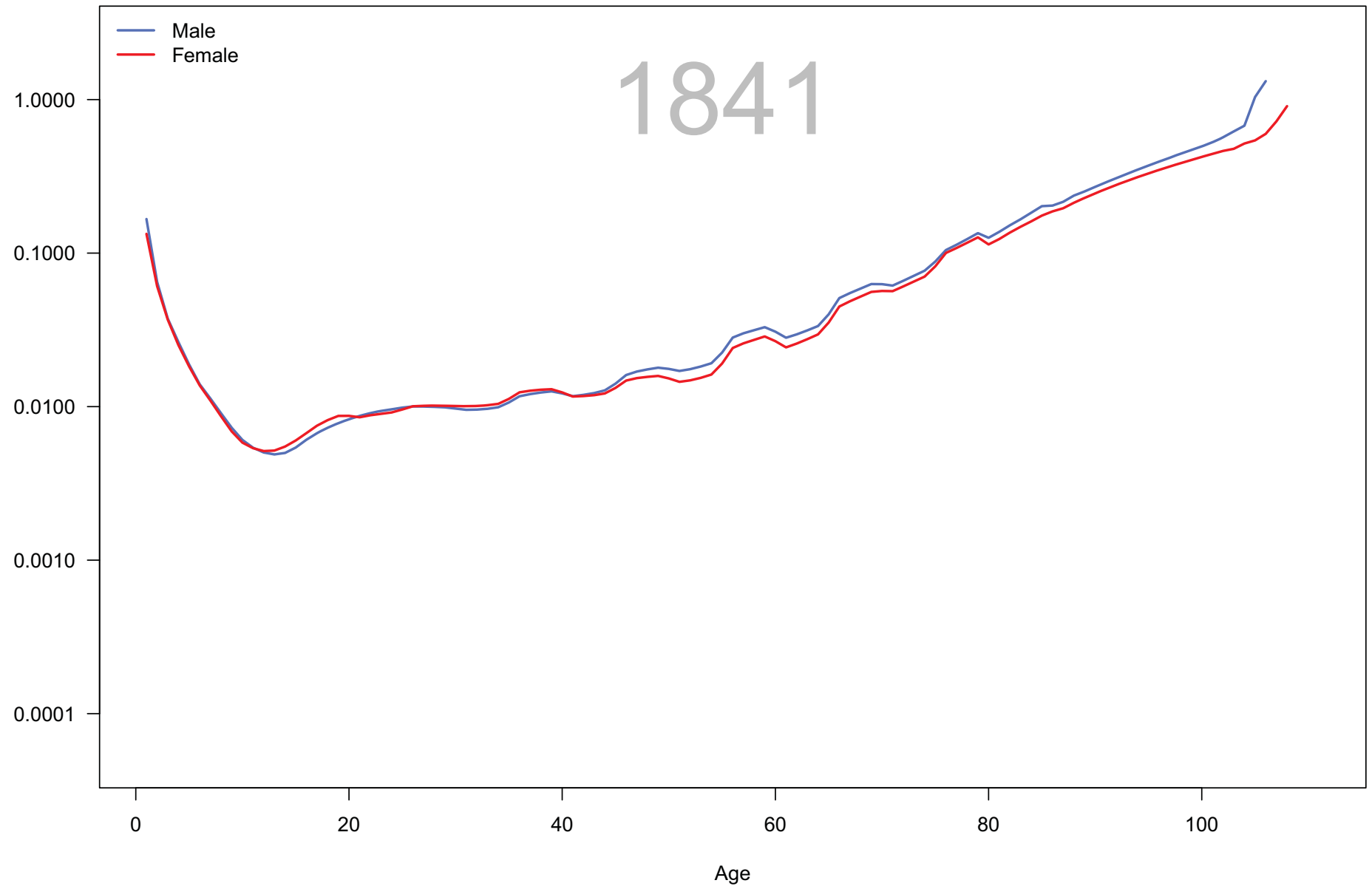
- Can apply to model any standard schedule of mortality, fertility or migration to adjusted using estimated parameters

Effect of Changing Brass Model Parameters, source: Preston et. al. (2001)

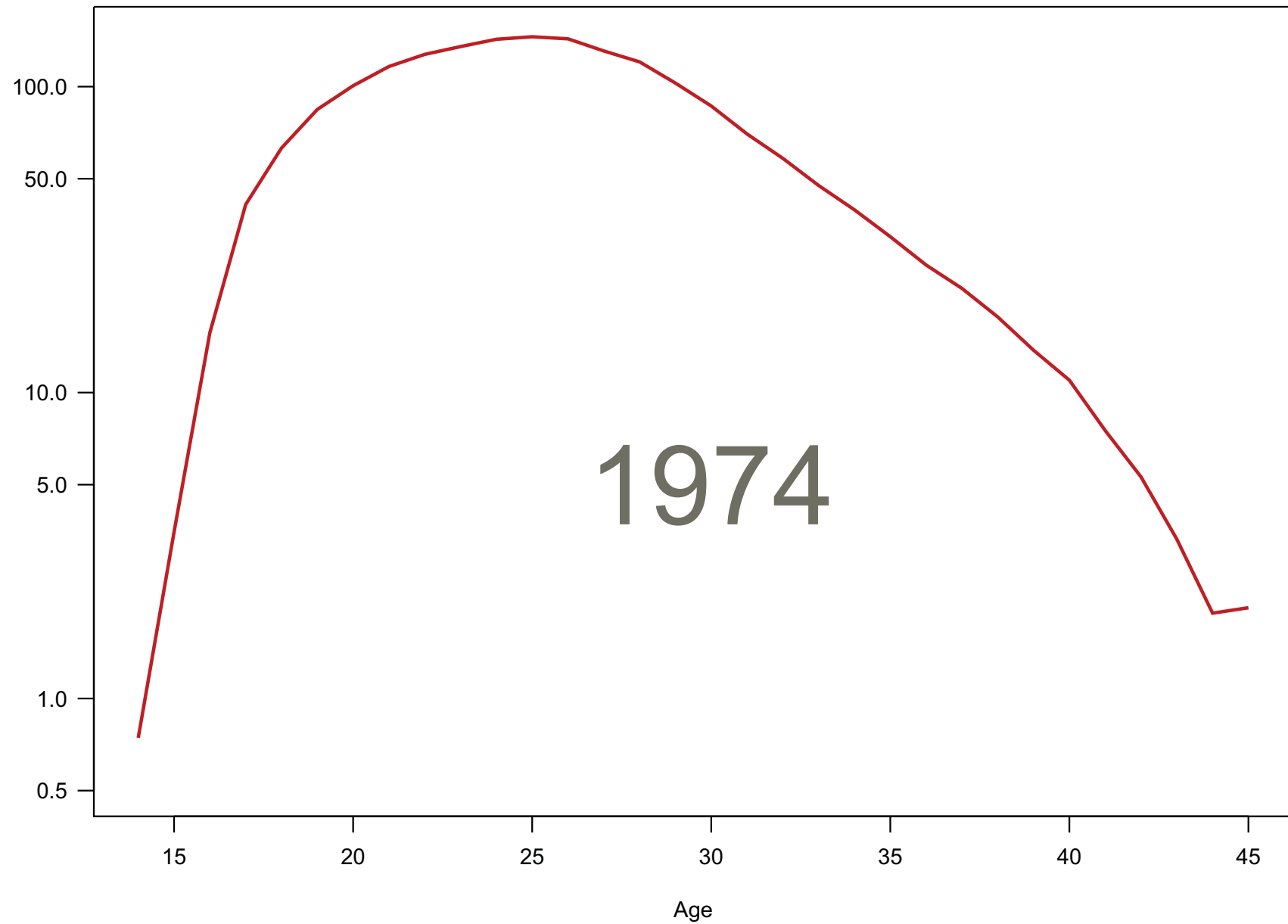


- Variety of life tables systems
- Based on a collection of real life tables, usually of high quality
- Use a multivariate analysis technique to find clusters
- Combine them to provide a standard
- e.g. Coale-Demeny (1966, 1983)
 - Divided countries into 4 regional models (North, South, East, West) based on mortality rates at different ages
 - Use available data to estimate life expectancy
 - Use available data to determine mortality pattern
 - Use standard table that fits your assumptions

Logarithm of m_x 1841 to 2007



UK F_x (000's) 1974 to 2007



- Lee Carter (1992) mortality model used in U.S. Census Bureau population forecasts

$$\log(m(x, t)) = a_x + K(t)b_x$$

a_x average of log mortality at age x

b_x age pattern of mortality change

$K(t)$ the level of mortality at time t

- Special case of log-linear model
 - Special case of PCA, using the first principal component of the log mortality matrix $m_x t$
 - Age profiles evolve in implausible ways.
- ARIMA on Model Parameters, see for example McNown and Rogers (1989)

- Recently authors have started to frame the modelling of profiles within the functional data paradigm.
- Hyndman and Ullah (2007) (below), Girosi and King (2008) and Ramsey et.al. (2009)
 - 1 Nonparametric smoothing of the each curve to obtain $f_t(x)$.
 - 2 Fit functional model to $f_t(x)$ to decompose curves into principal components

$$f_t(x) = \mu(x) + \sum_{k=1}^K \beta_{t,k} \phi_k(x) + e_t(x)$$

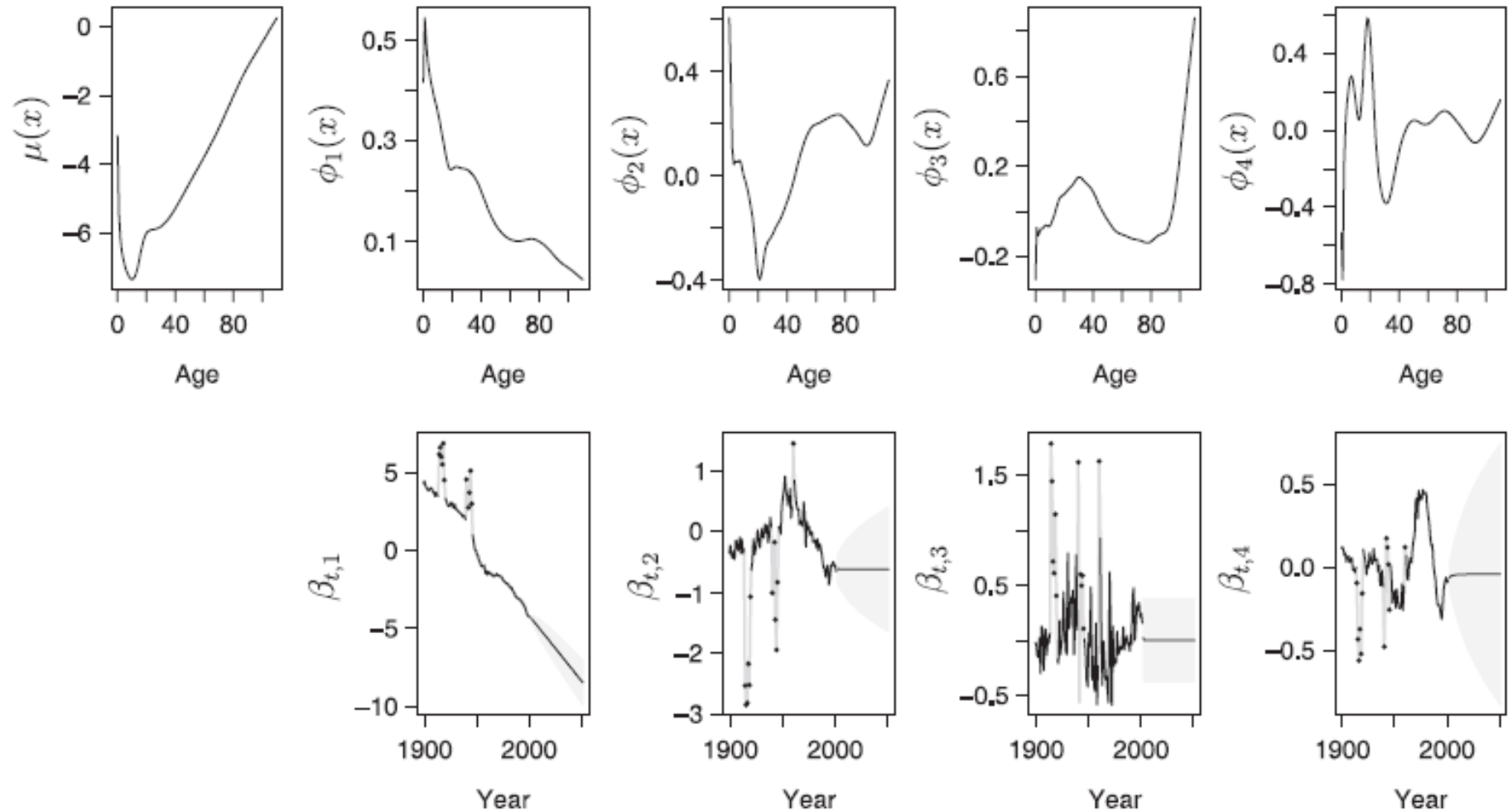
$\phi_k(x)$ is the k th principal component (basis) functions

$\beta_{t,k}$ are corresponding scores

$\mu(x)$ mean function

- 3 Univariate time series for each $\beta_{t,k}$
- 4 Forecasts of principal components scores multiplied with principal components to obtain estimated future curves.

Basis functions and coefficients for French Mortality, source: Hyndman and Ullah (2007)



Thank You

Thank you for listening!

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