

Forecasting functional time series

Rob J Hyndman



MONASH University

Mortality rates

Fertility rates

Some notation

Let $y_t(x)$ be the observed (smoothed) data in period t at age x , $t = 1, \dots, n$.

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

- Estimate $\mu(x)$ as mean $y_t(x)$ across years.

Some notation

Let $y_t(x)$ be the observed (smoothed) data in period t at age x , $t = 1, \dots, n$.

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

- Estimate $\mu(x)$ as mean $y_t(x)$ across years.
- Estimate $\beta_{t,k}$ and $\phi_k(x)$ using (robust) functional principal components.

Some notation

Let $y_t(x)$ be the observed (smoothed) data in period t at age x , $t = 1, \dots, n$.

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

- Estimate $\mu(x)$ as mean $y_t(x)$ across years.
- Estimate $\beta_{t,k}$ and $\phi_k(x)$ using (robust) functional principal components.
- $\beta_{t,k}$ are principal component scores

Some notation

Let $y_t(x)$ be the observed (smoothed) data in period t at age x , $t = 1, \dots, n$.

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

- Estimate $\mu(x)$ as mean $y_t(x)$ across years.
- Estimate $\beta_{t,k}$ and $\phi_k(x)$ using (robust) functional principal components.
- $\beta_{t,k}$ are principal component scores
- $\phi_k(x)$ are eigenfunctions

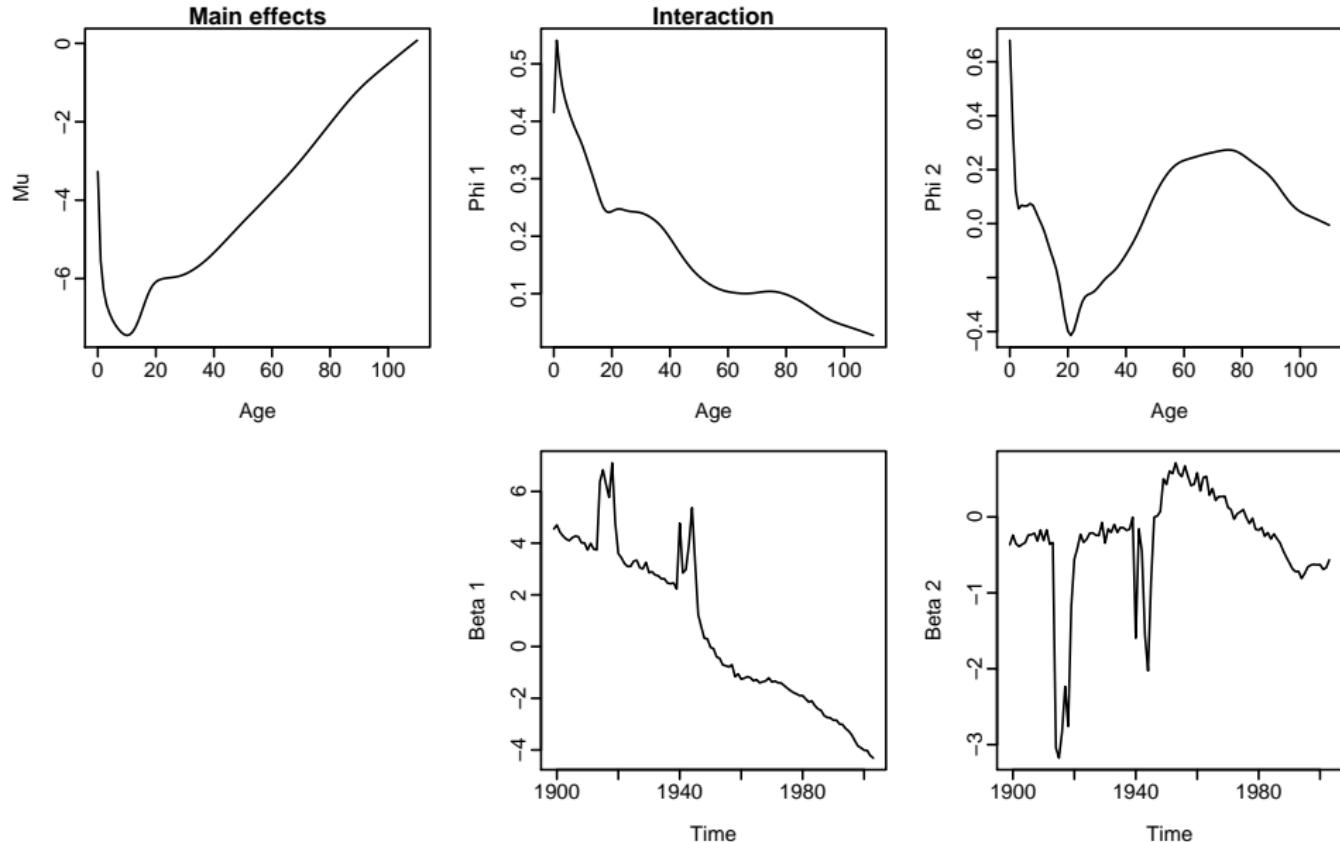
Some notation

Let $y_t(x)$ be the observed (smoothed) data in period t at age x , $t = 1, \dots, n$.

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

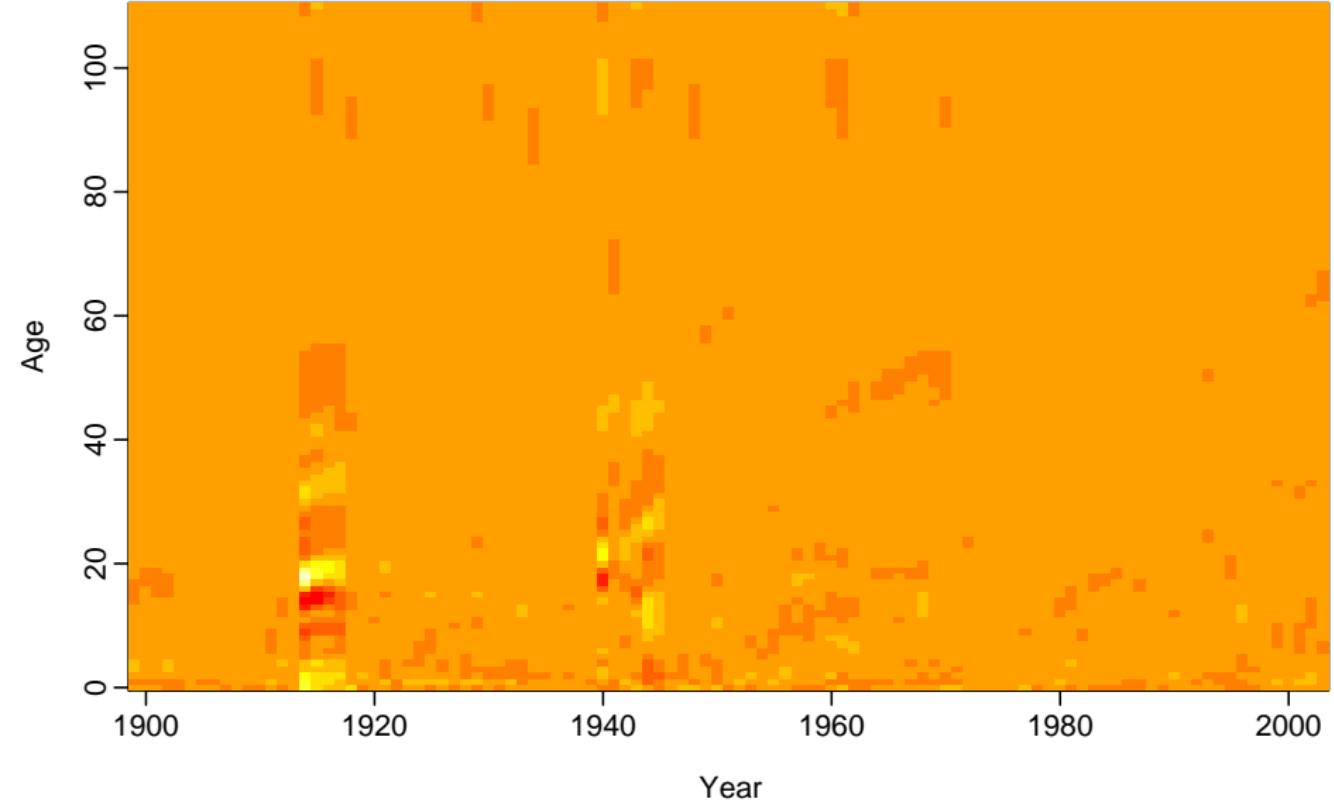
- Estimate $\mu(x)$ as mean $y_t(x)$ across years.
- Estimate $\beta_{t,k}$ and $\phi_k(x)$ using (robust) functional principal components.
- $\beta_{t,k}$ are principal component scores
- $\phi_k(x)$ are eigenfunctions
- $e_t(x) \stackrel{\text{iid}}{\sim} N(0, v(x))$ is random error

Functional PC



Functional PC

Residuals

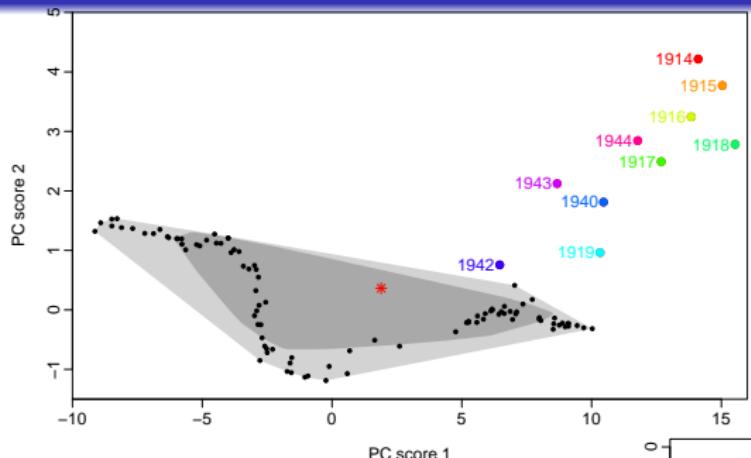


Outliers

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

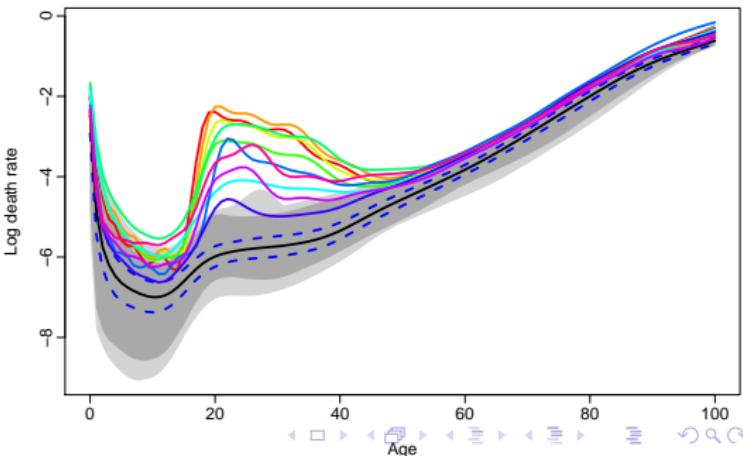
- Outliers in $\{y_t(x_i)\}$ will show up as outliers in $\{\beta_{t,1}, \beta_{t,2}, \dots\}$.

Bagplots

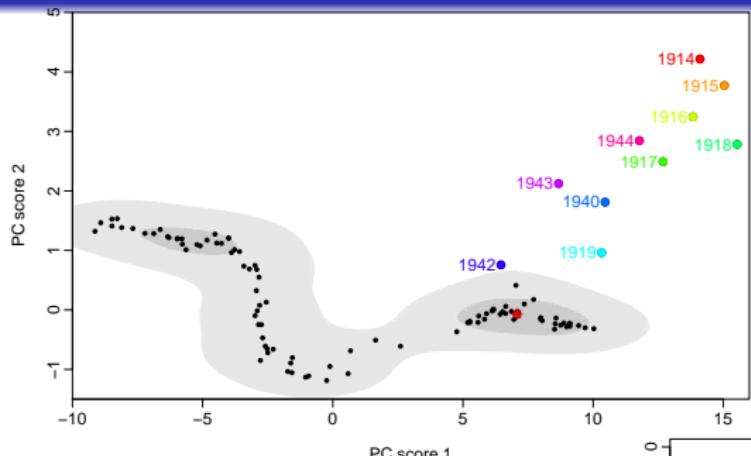


*Red asterisk: Median
Dark grey: 50% region
Light grey: outer region*

Bagplot for the French male mortality data.

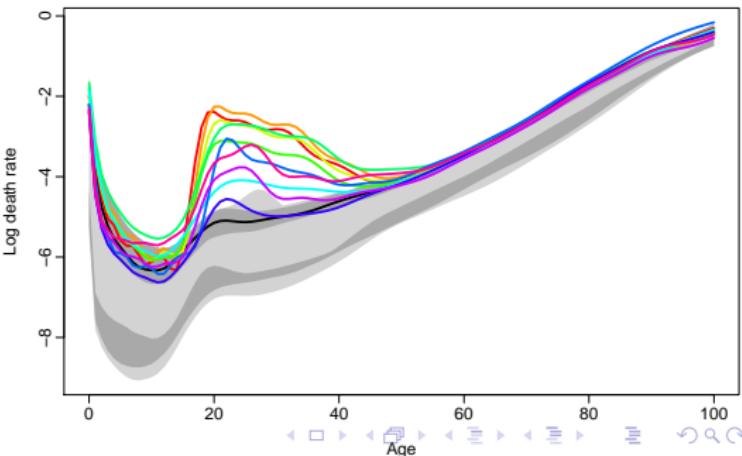


Functional HDR boxplot



*Red asterisk: Mode
Dark grey: 50% region
Light grey: outer region*

HDR boxplot for the French male mortality data.



Functional time series model

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

Functional time series model

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

- The eigenfunctions $\phi_k(x)$ show the main regions of variation.

Functional time series model

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

- The eigenfunctions $\phi_k(x)$ show the main regions of variation.
- The scores $\{\beta_{t,k}\}$ are uncorrelated by construction. So we can forecast each $\beta_{t,k}$ using a univariate time series model.

Functional time series model

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

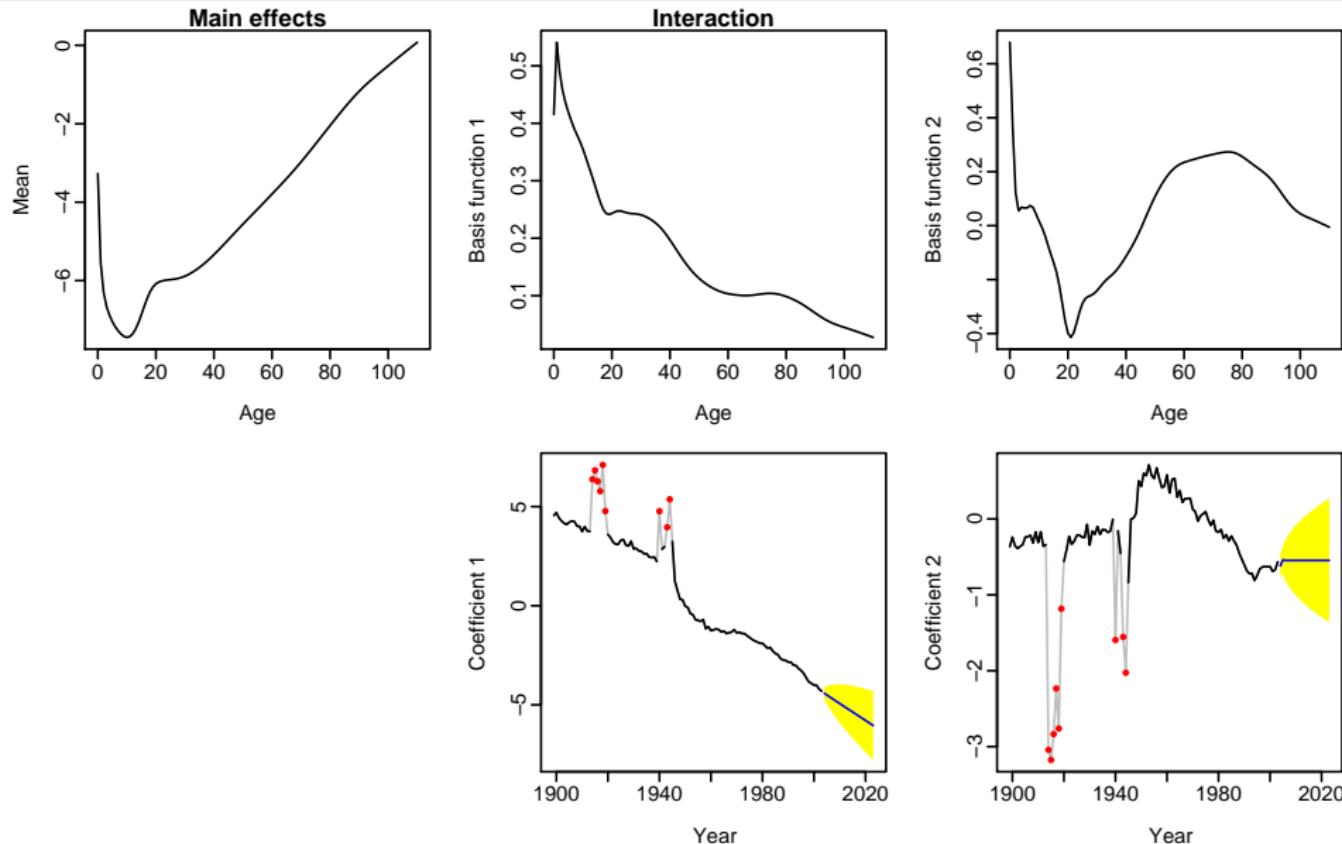
- The eigenfunctions $\phi_k(x)$ show the main regions of variation.
- The scores $\{\beta_{t,k}\}$ are uncorrelated by construction. So we can forecast each $\beta_{t,k}$ using a univariate time series model.
- Outliers are treated as missing values.

Functional time series model

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

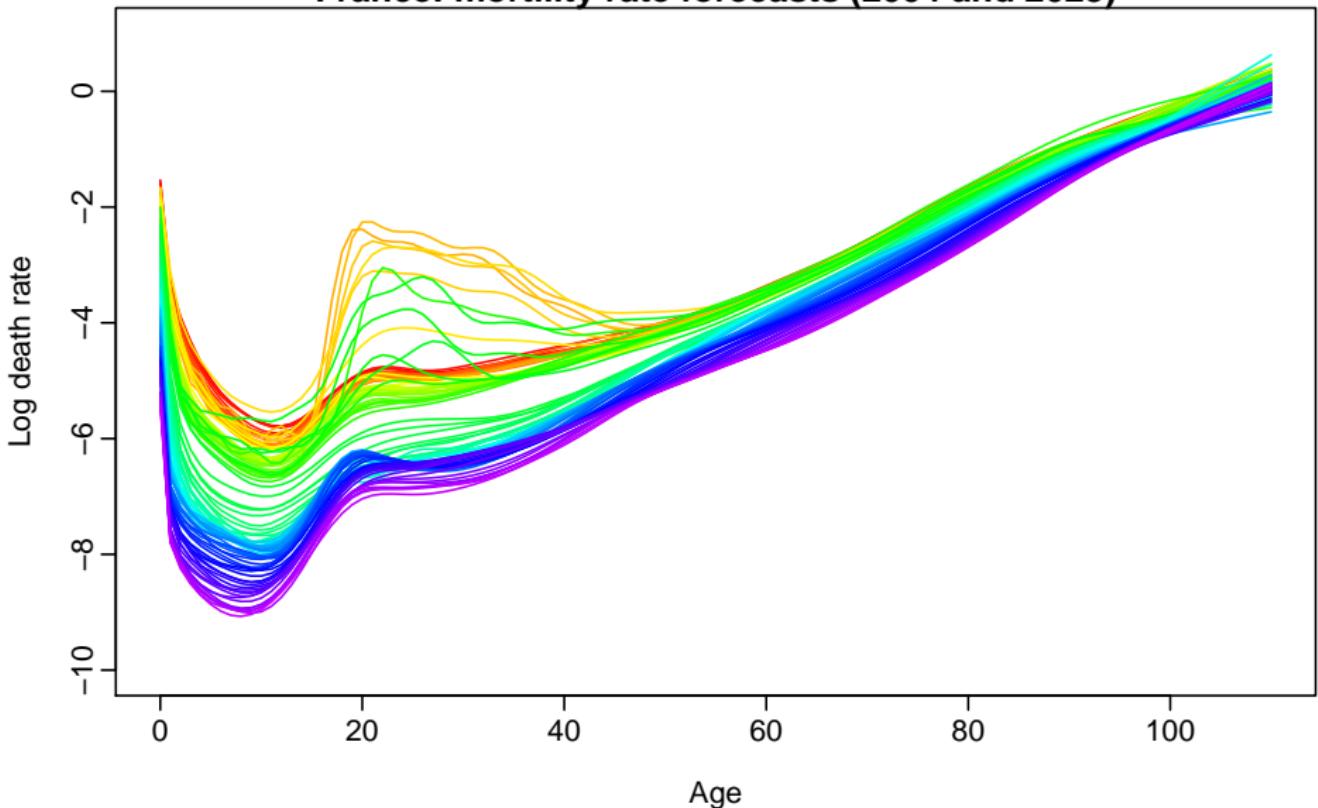
- The eigenfunctions $\phi_k(x)$ show the main regions of variation.
- The scores $\{\beta_{t,k}\}$ are uncorrelated by construction. So we can forecast each $\beta_{t,k}$ using a univariate time series model.
- Outliers are treated as missing values.
- **Univariate ARIMA models are used for forecasting.**

Forecasting the PC scores



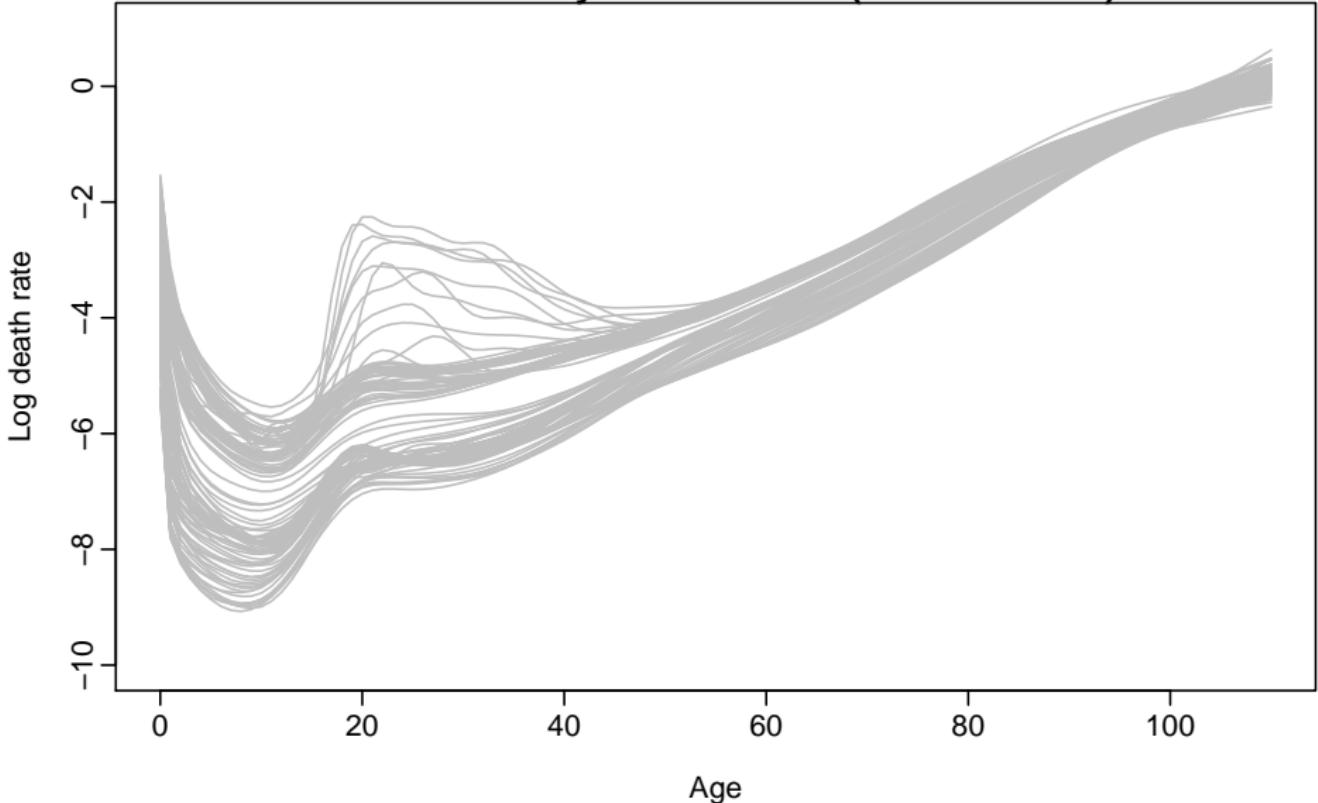
Forecasts of $y_t(x)$

France: mortality rate forecasts (2004 and 2023)



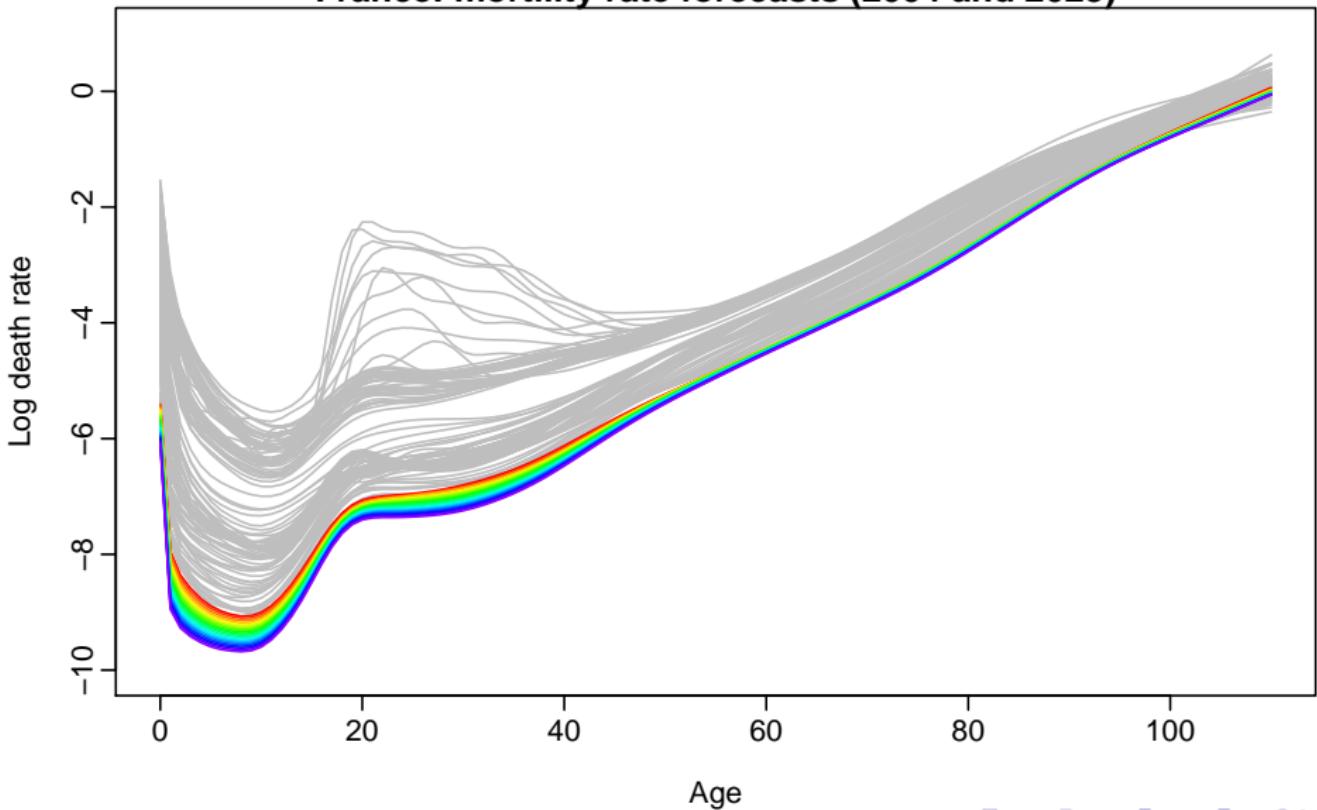
Forecasts of $y_t(x)$

France: mortality rate forecasts (2004 and 2023)



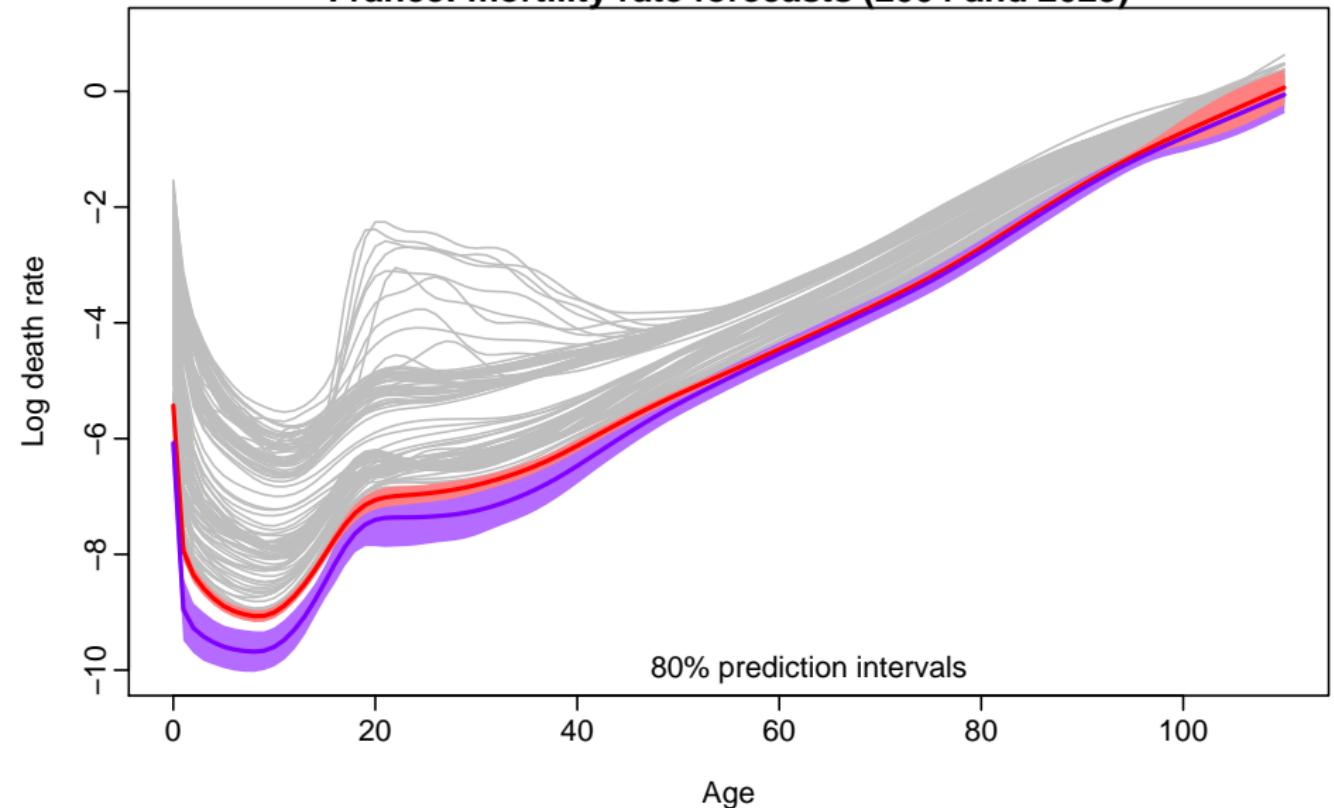
Forecasts of $y_t(x)$

France: mortality rate forecasts (2004 and 2023)



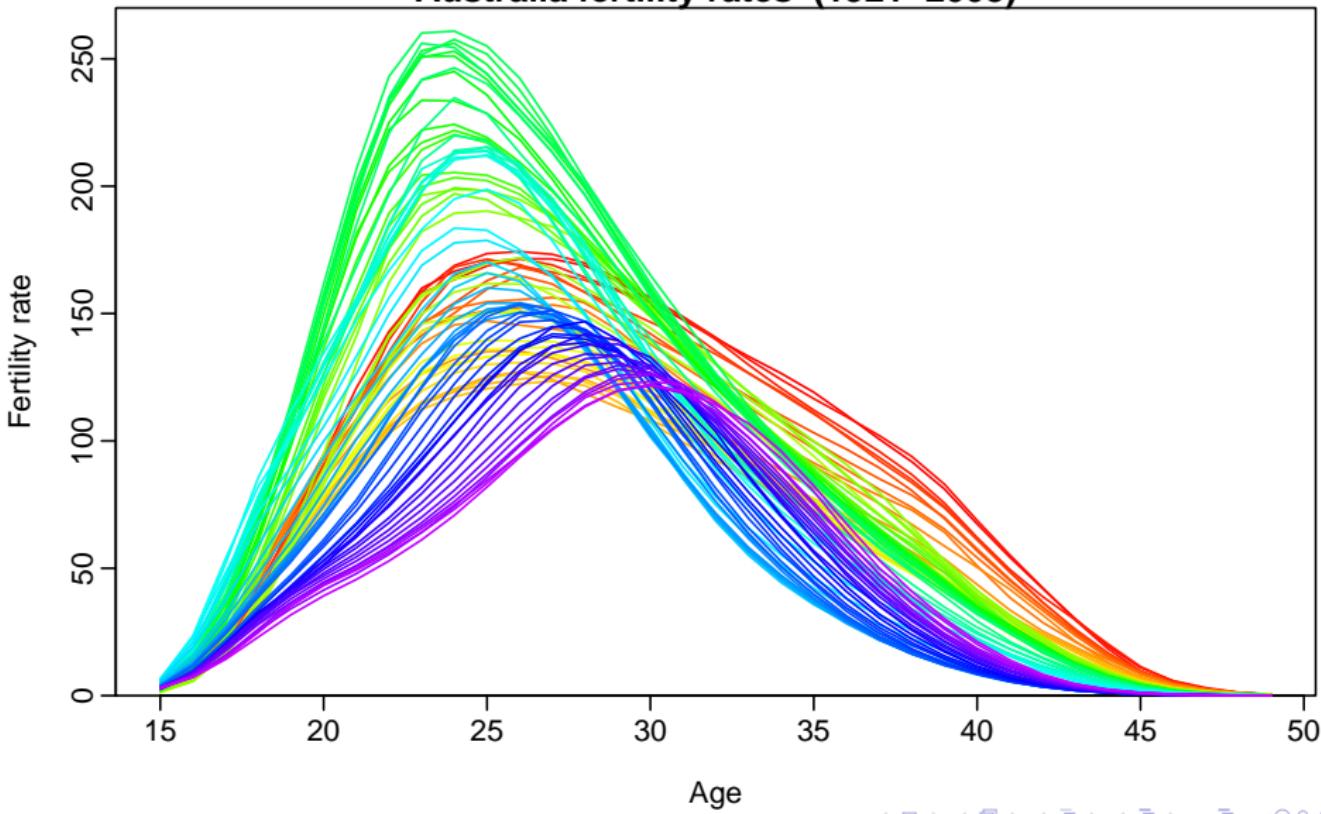
Forecasts of $y_t(x)$

France: mortality rate forecasts (2004 and 2023)

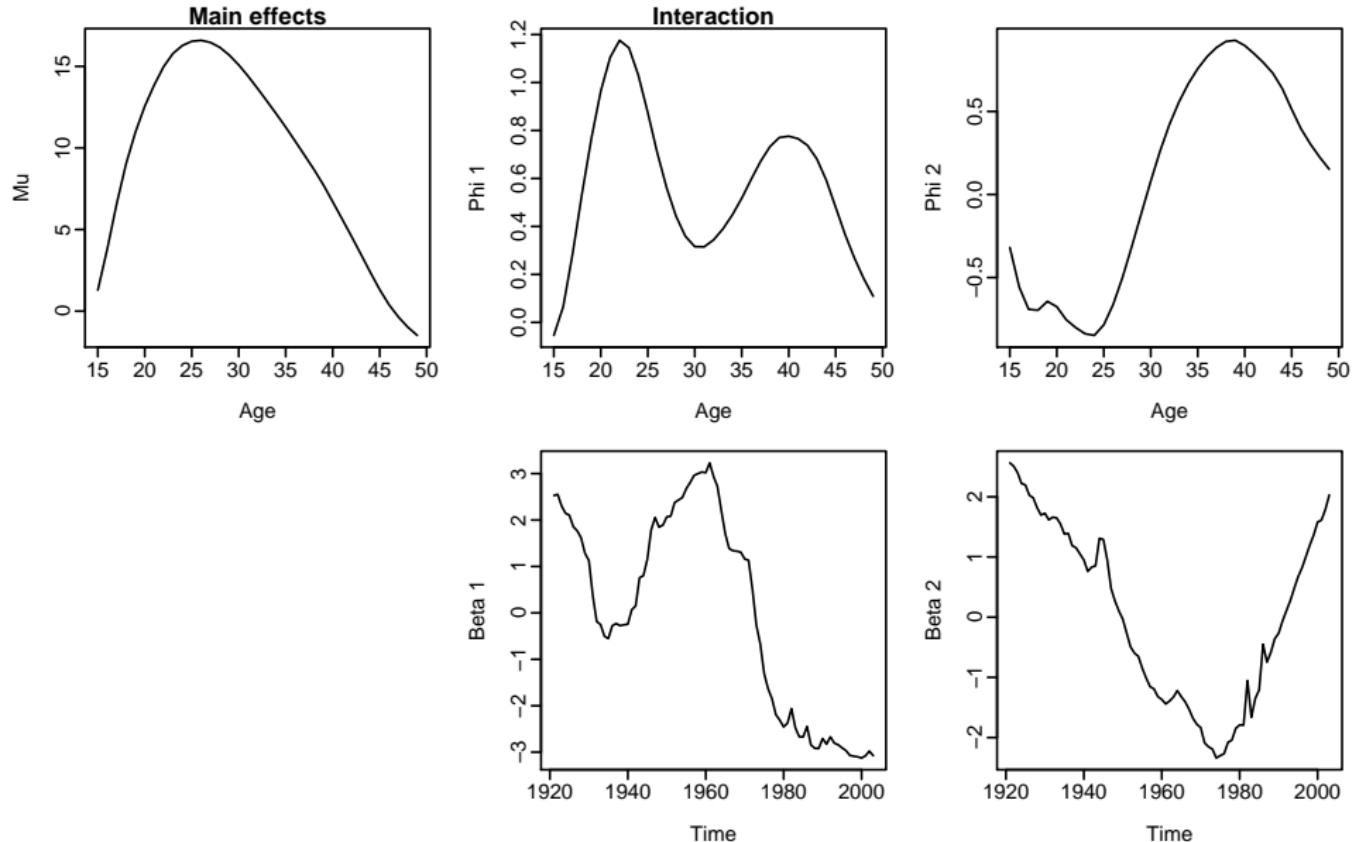


Fertility application

Australia fertility rates (1921–2003)

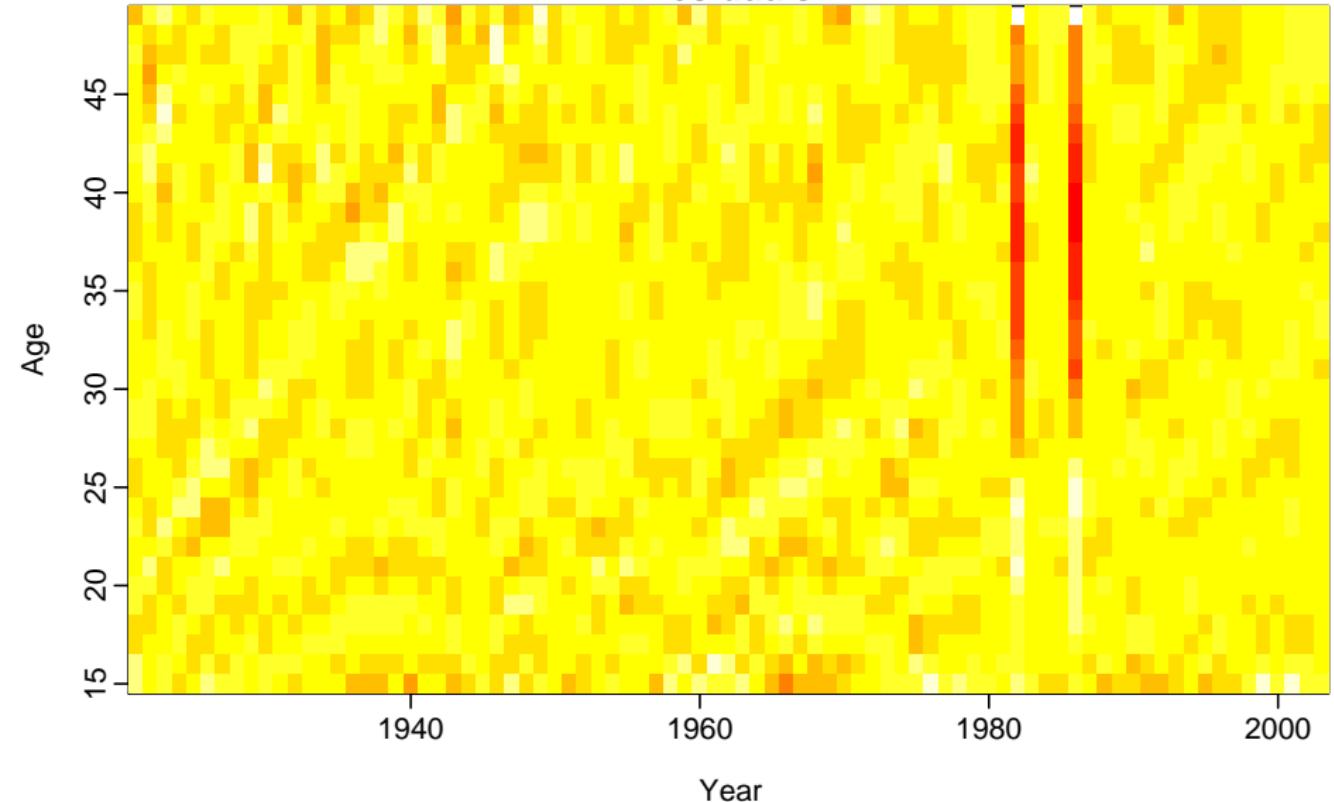


Fertility model

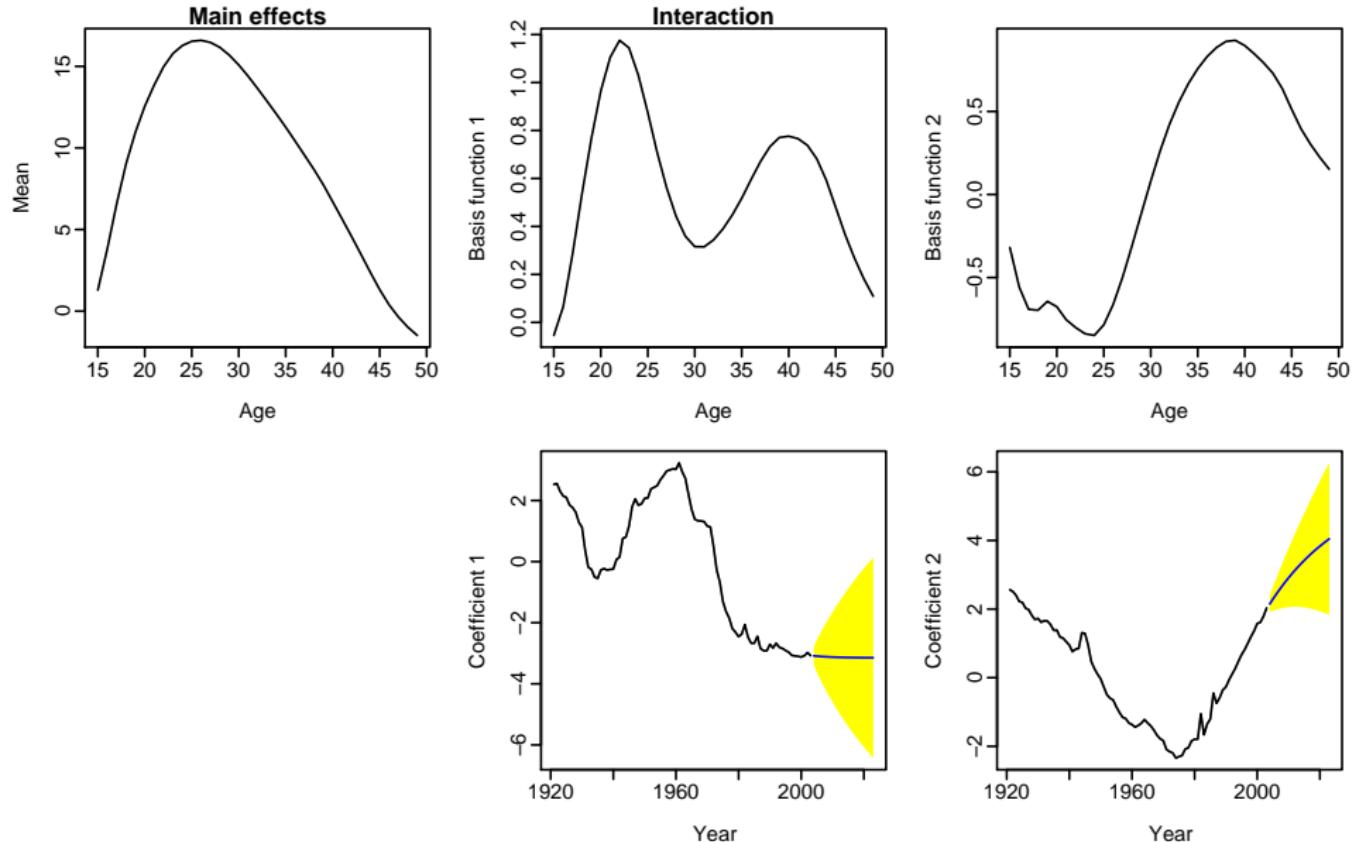


Fertility model

Residuals

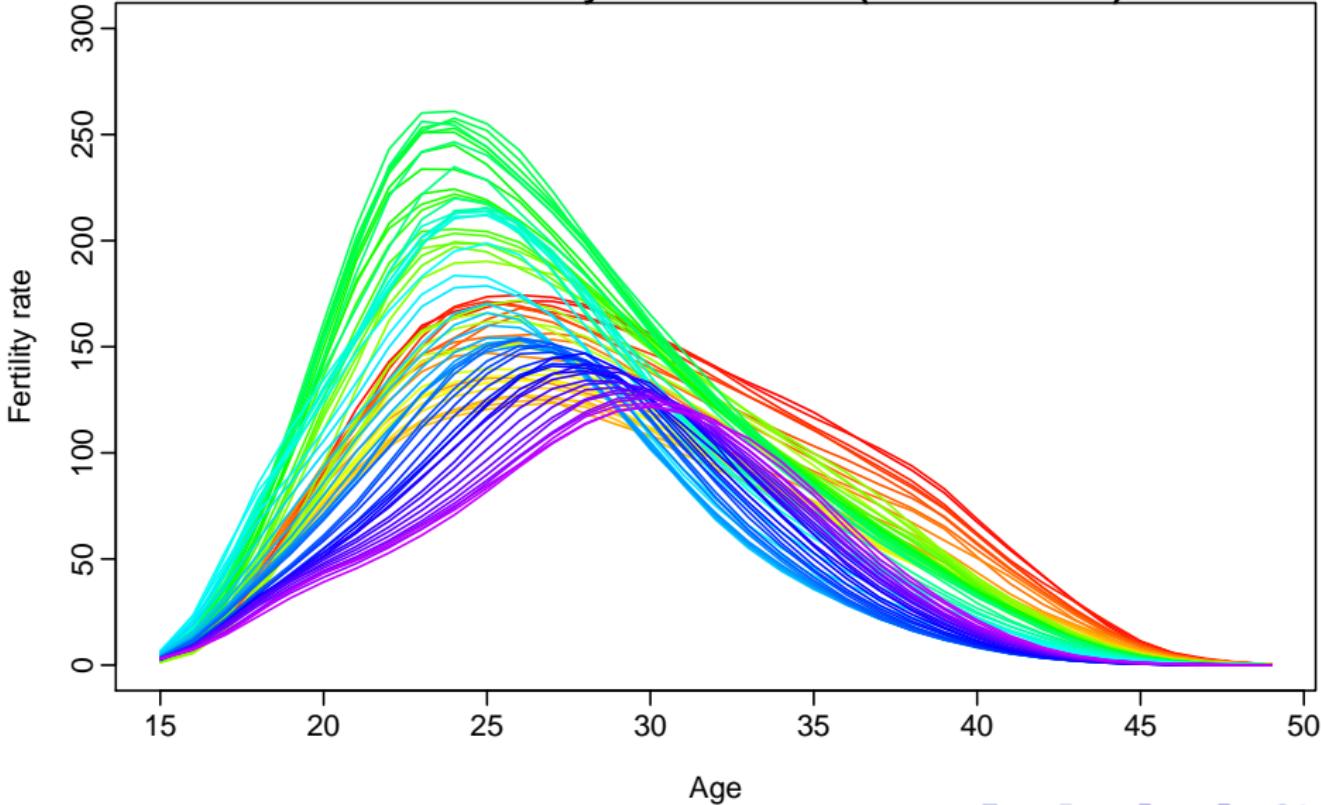


Fertility model



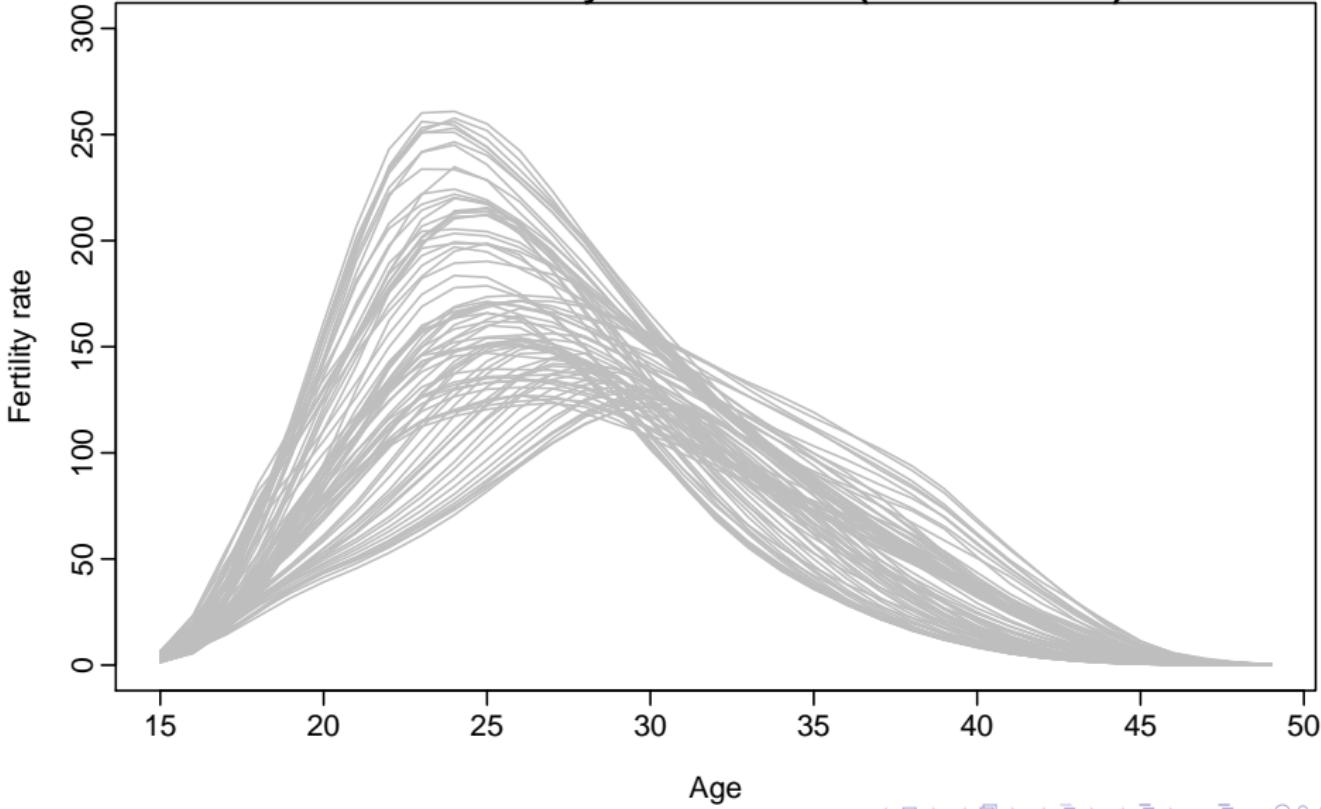
Forecasts of $y_t(x)$

Australia: fertility rate forecasts (2004 and 2023)



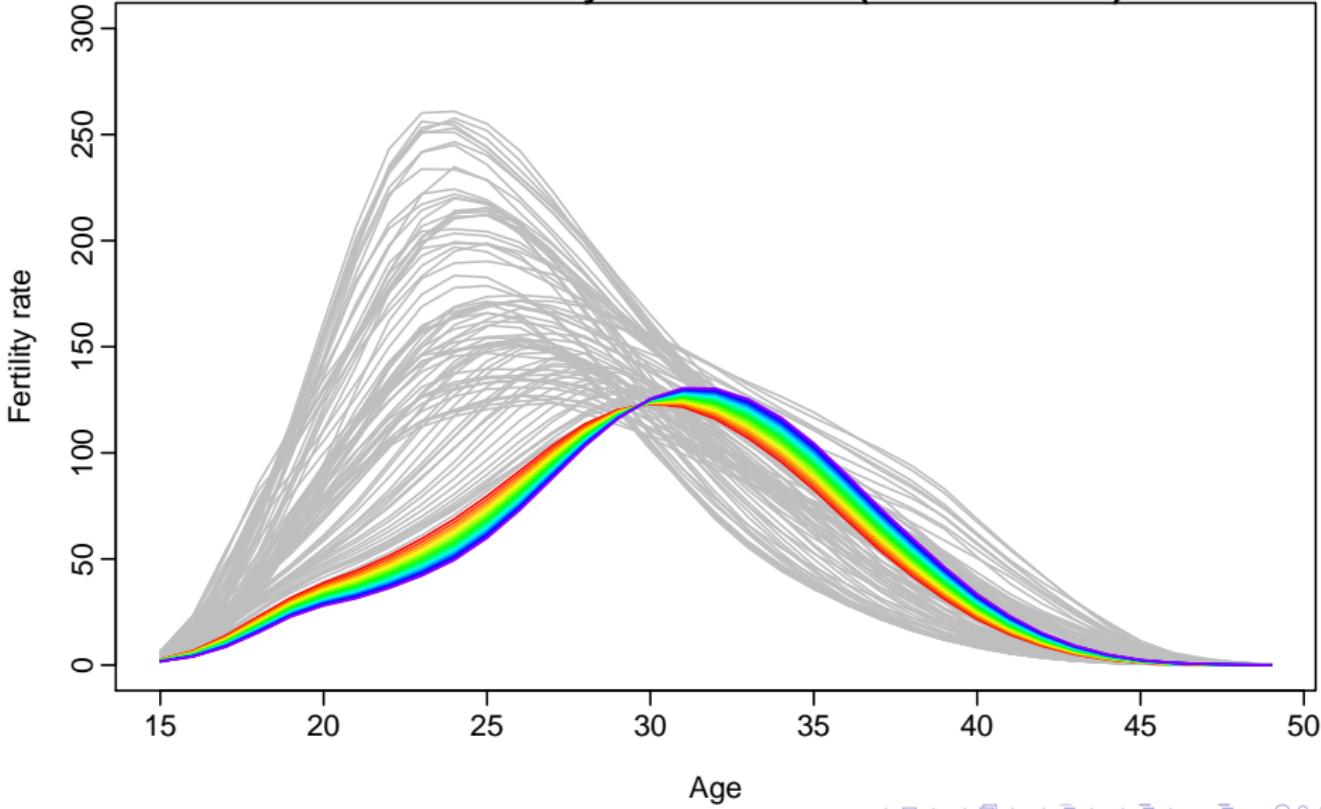
Forecasts of $y_t(x)$

Australia: fertility rate forecasts (2004 and 2023)



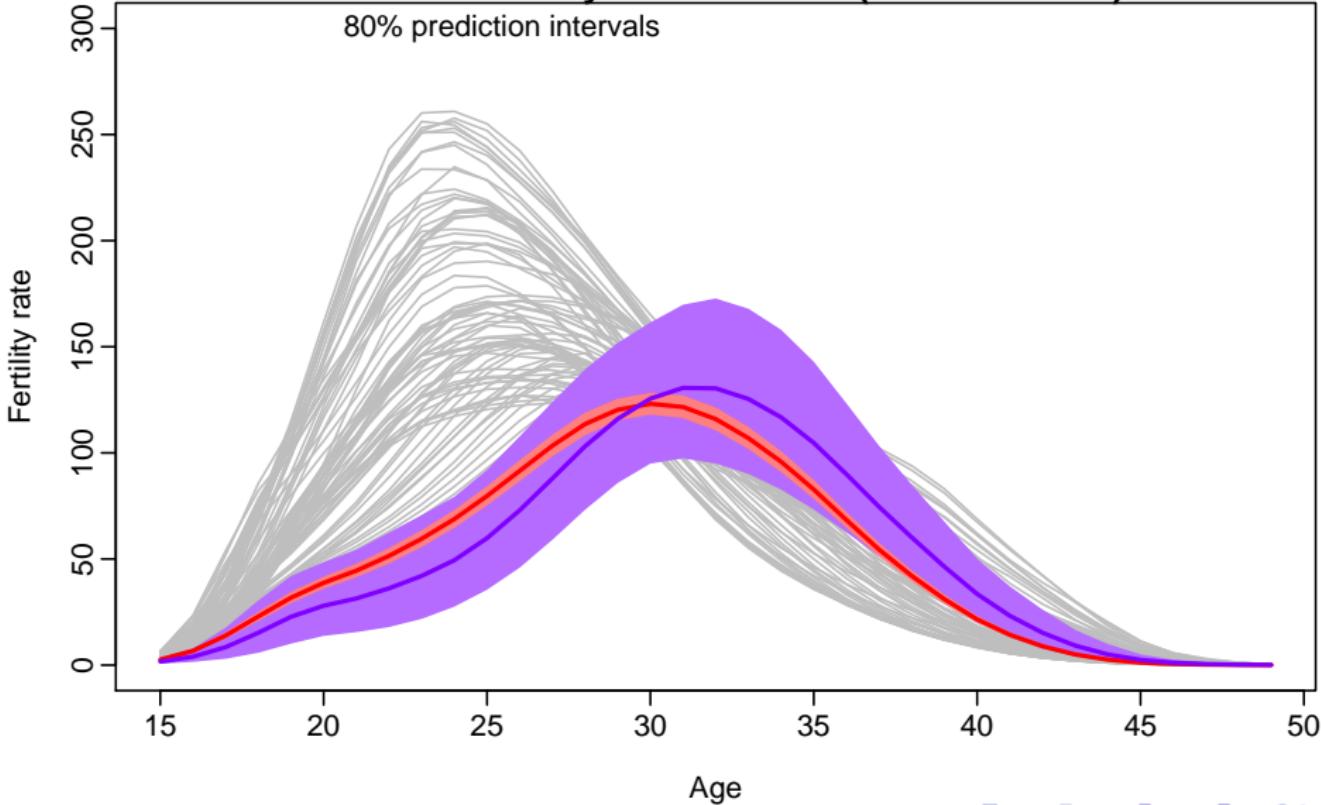
Forecasts of $y_t(x)$

Australia: fertility rate forecasts (2004 and 2023)



Forecasts of $y_t(x)$

Australia: fertility rate forecasts (2004 and 2023)



Research team



MONASH
University

- Bircan Erbas
- Han Lin Shang
- Farah Yasmeen
- Shahid Ullah
- Abdul-Aziz Hayat

Research team



MONASH
University

- Bircan Erbas
- Han Lin Shang
- Farah Yasmeen
- Shahid Ullah
- Abdul-Aziz Hayat

Things we are working on

➡ cohort effects

(our current models only allow period effects)

Research team



MONASH
University

- Bircan Erbas
- Han Lin Shang
- Farah Yasmeen
- Shahid Ullah
- Abdul-Aziz Hayat

Things we are working on

- ➡ cohort effects
(our current models only allow period effects)
- ➡ synergy and differences between groups
(e.g., multi-racial mortality, breast-screen effects)

Research team



MONASH
University

- Bircan Erbas
- Han Lin Shang
- Farah Yasmeen
- Shahid Ullah
- Abdul-Aziz Hayat

Things we are working on

- ➡ cohort effects
(our current models only allow period effects)
- ➡ synergy and differences between groups
(e.g., multi-racial mortality, breast-screen effects)
- ➡ choosing a basis better suited to
prediction (rather than explaining historical variation)

Applications so far

- Fertility rates

Applications so far

- Fertility rates
- All-cause mortality rates

Applications so far

- Fertility rates
- All-cause mortality rates
- Population

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates
- Yield curves in finance

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates
- Yield curves in finance
- Seasonal electricity demand

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates
- Yield curves in finance
- Seasonal electricity demand
- Seasonal El Niño sea-surface temperatures

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates
- Yield curves in finance
- Seasonal electricity demand
- Seasonal El Niño sea-surface temperatures

Applications so far

- Fertility rates
- All-cause mortality rates
- Population
- Cancer mortality and incidence rates
- Yield curves in finance
- Seasonal electricity demand
- Seasonal El Niño sea-surface temperatures

More information

Papers and R code available at

www.rohyndman.info