

# Modeling and forecasting mortality with economic growth: a multi-population approach

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# Background

- Human life expectancy has been increasing substantially over the past decades.
- Forecasting future life expectancy is crucial for various practices (demographics, finance, actuary, ...)
- Stochastic mortality modeling has been widely studied since 1990s.
- Modeling mortality in multiple populations has attracted increasing attention recently
  - ▶ Risk management, demographic analysis, capital market solution, ...;
  - ▶ Forecasts for individual populations may be improved by exploiting more data.



# Motivation

- Most stochastic mortality models up to date are “extrapolative”
  - ▶ They summarise historical mortality trends by latent factor(s);
  - ▶ Forecasts are obtained by extrapolating these factors.
- Extrapolative models achieved great success
  - ▶ Single-population: Lee and Carter (1992), Cairns-Blake-Dowd (2006), Renshaw and Haberman (2006) ...
  - ▶ Multi-population: Li and Lee (2005), Dowd et al. (2011), Hyndman et al. (2013) ...
- However, there are potential limitations
  - ▶ The interpretation of these latent factors are sometimes unclear;
  - ▶ It is not sure whether the historical latent trends will continue in the future.



# Motivation

- Can mortality trends be approximated by observable factors?
- Various candidates
  - ▶ Smoking pattern;
  - ▶ Temperature;
  - ▶ Unemployment rate;
  - ▶ ...
  - ▶ **Economic growth**;



# Literature review

- Cross-sectional — economic **level** and mortality **level**.
  - ▶ Countries with a higher level of higher national income have on average a higher life expectancy (Preston 1975).
- Time-series — economic **growth** and mortality **change**. Mixed results.
- Positive:
  - ▶ Economic growth positively affects mortality decline (Brenner 2005 ; Bloom and Canning 2005);
  - ▶ Health and life expectancy positively affects economic growth (Bloom and Canning 2005; Bhargava et al. 2011).
- Negative:
  - ▶ Mortality declines temporarily accelerates during economic recessions and slows down during economic expansions (Ruhm 2005; Tapia Granados 2005);
  - ▶ No evidence that increase in life expectancy causes economic growth (Acemoglu and Johnson 2007).



# Motivation

- However, few studies combined stochastic mortality modeling,
  - ▶ Hanewald (2011): correlation between  $k_t$  in the Lee-Carter model and macroeconomic fluctuations.
  - ▶ Niu and Melenberg (2014): include real GDP per capita in the Lee-Carter model.
- These studies are restricted to single-population contexts.



# Research question

- Are there long-term relationships between aggregate economic growths and aggregate mortality changes in a group of closely related populations?
- If so, can mortality forecasts be improved by exploiting these relationships?
- We extend the Li-Lee model to include GDP data, and forecasts future mortality by extrapolating the observable trends of GDP.
- Satisfying results are obtained in empirical studies.





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# The general specification

- $I$  populations. For population  $i$ , age  $x$ , year  $t$ ,

$$\log m_{i,x,t} = a_{i,x} + \sum_{\ell=1}^L \gamma_{\ell,x} g_{\ell,t} + B_x K_t + b_{i,x} k_{i,t} + \varepsilon_{i,x,t}$$

- $g_{\ell,t}$  the  $\ell$ -th principal component of the logarithm of real GDP per capita of the  $I$  populations.
- $K_t$  the common latent factor.
- $k_{i,t}$  the specific latent factor for population  $i$ .



# Normalization

- To identify the model, we propose

$$\sum_{x=1}^N B_x = 1,$$

$$\sum_{t=1}^T K_t = 0,$$

$$K = (K_1, \dots, K_T) \neq 0,$$

$$\sum_{t=1}^T K_t g_{\ell,t} = 0, \text{ for } \ell = 1, \dots, L,$$



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# Data

- Four groups
  - ▶ 14 low mortality countries (15-low in Li-Lee except West Germany), 1947 - 2010;
  - ▶ 6 Eastern European countries (as in Li-Lee), 1969 - 2009;
  - ▶ 6 former Soviet Union countries (Belarus, Estonia, Latvia, Lithuania, Russia, Ukraine), 1969 - 2010;
  - ▶ Female and male population of Sweden, 1947 - 2010.
- Data source
  - ▶ Uni-sex mortality rates from age 0 to 99;
    - ★ Human mortality database.
  - ▶ Real GDP per capita (purchasing power parity based).
    - ★ Maddison project on World Economy;
    - ★ ERS International Macroeconomic Data Set.



# Model specification

- We compare two models.
- The GDP model without common latent factor

$$\log m_{i,x,t} = a_{i,x} + \sum_{l=1}^L \gamma_{l,x} g_{l,t} + b_{i,x} k_{i,t} + \varepsilon_{i,x,t}.$$

- The Li-Lee model

$$\log m_{i,x,t} = a_{i,x} + B_x K_t + b_{i,x} k_{i,t} + \varepsilon_{i,x,t}$$



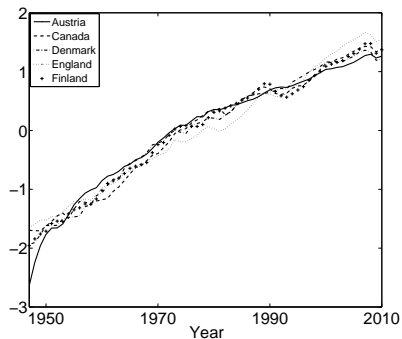
## Number of principal components

	1 PC	2 PCs
Group 1	98.70	99.57
Group 2	79.20	97.20
Group 3	77.81	98.01

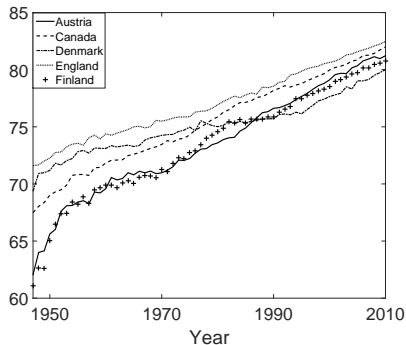
- $L = 1$  for Group 1 and 4;  $L = 2$  for Group 2 and 3.



# Group 1 Life expectancy and GDP



(a) GDP Group 1 (selected)

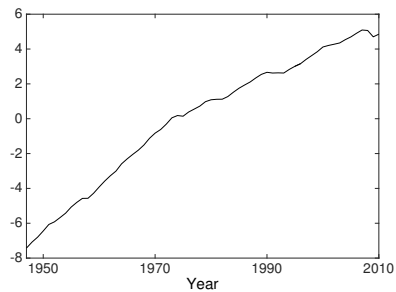


(b) Life Expectancy Group 1 (selected)

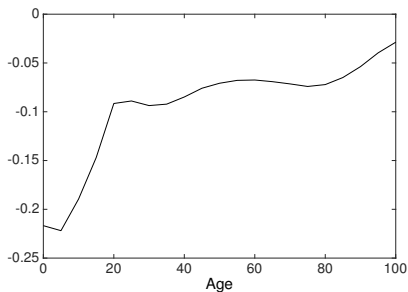




# Group 1 GDP model



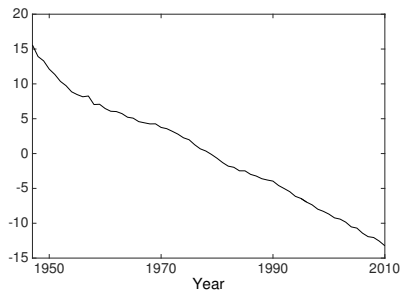
(a)  $g_t$



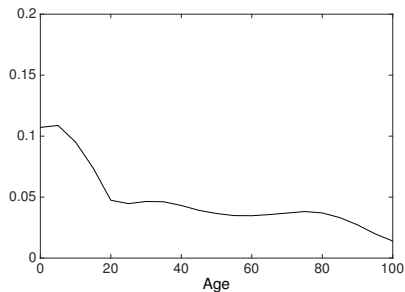
(b)  $\gamma_x$



# Group 1 Li-Lee model



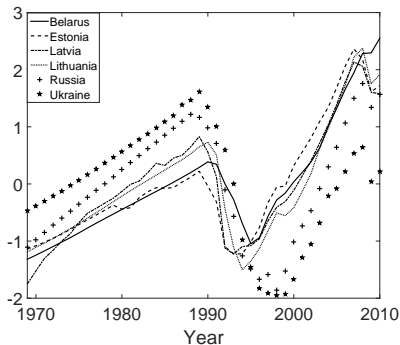
(a)  $K_t$



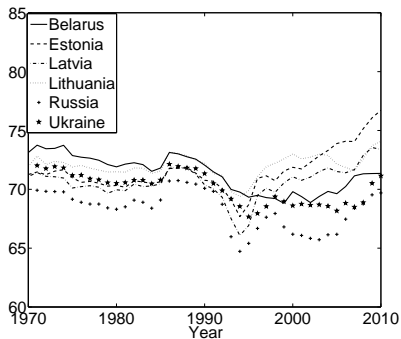
(b)  $B_x$



# Group 3 Life expectancy and GDP



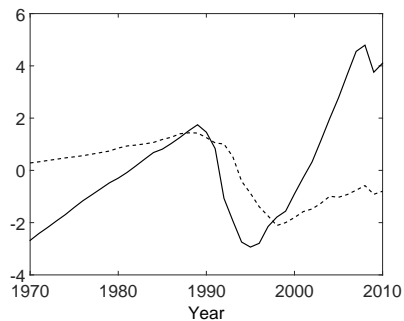
(a) GDP Group 3



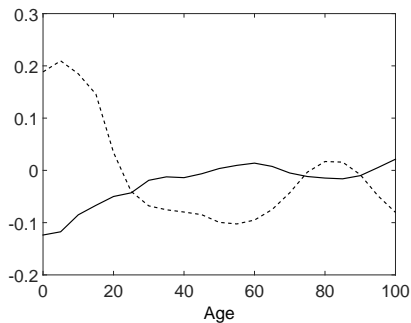
(b) Life Expectancy Group 3



# Group 3 GDP model



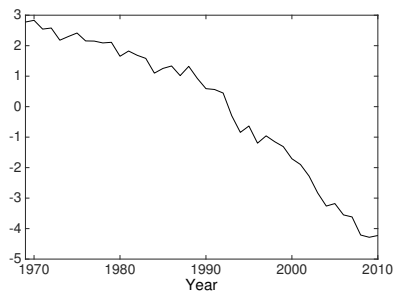
(a)  $g_{1,t}$  (solid line) and  $g_{2,t}$  (dashed line)



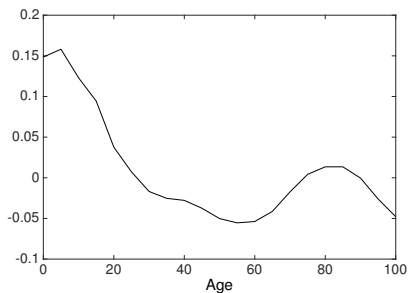
(b)  $\gamma_{1,x}$  (solid line) and  $\gamma_{2,x}$  (dashed line)



# Group 3 GDP model



(a)  $K_t$



(b)  $B_x$



# Variation explained by the common factors

- The ratio  $R_C$  used in Li and Lee (2015)

	Li-Lee	GDP
Group 1	0.90	0.85
Group 2	0.47	0.47
Group 3	0.53	0.55
Group 4	0.91	0.86



# AIC and BIC ratios

- AIC ratio

$$2 \log \hat{L} - 2m.$$

- BIC ratio

$$\log \hat{L} - \frac{1}{2}m \cdot \log M.$$

- $\hat{L}$  — likelihood;
- $m$  — number of free parameters;
- $M$  — sample size used for estimation.



# AIC and BIC ratios

	Li-Lee		GDP	
	BIC	AIC	BIC	AIC
Group 1	-15283	-18505	-14987	-18391
Group 2	-4152	-4735	-4064	-4676
Group 3	-4266	-4903	-4205	-4905
Group 4	-2250	-2789	-2015	-2680





# Model specification

- Common factors — random walk with drift

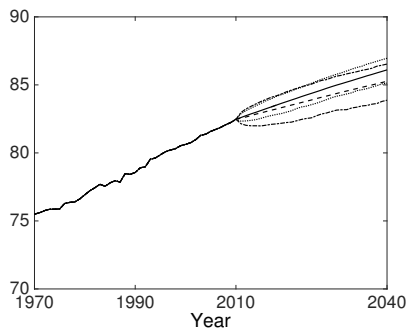
$$y_t = c + y_{t-1} + \eta_t, \quad \eta_t \stackrel{i.i.d.}{\sim} N(0, \sigma_y^2),$$

- Population-specific factors — AR(1)

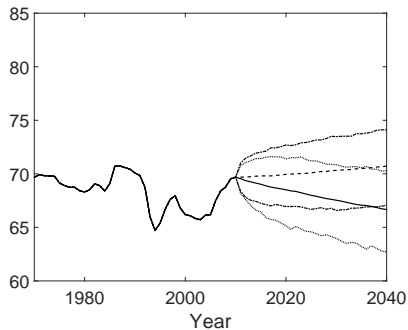
$$k_{i,t} = c_{i,0} + c_{i,1}k_{i,t-1} + \omega_{i,t}, \quad \omega_{i,t} \stackrel{i.i.d.}{\sim} N(0, \sigma_{\omega,i}^2).$$



# Model specification



(a) US



(b) Russia (Group 3)



## Out-of-sample forecast: RMSFE

- Using data up to 2000, and forecast from 2001 to the end of the sample.
- Relative root mean squared forecast error (RMSFE)

$$\sqrt{\frac{1}{N(U - \hat{u})} \sum_{u=\hat{u}}^U \sum_{x=1}^N \frac{(\log m_{i,x,u} - \widehat{\log m_{i,x,u}})^2}{|\log m_{i,x,u}|}}$$

- $\hat{u}$  — 2001;
- $U$  — end of sample.



# Choice of sample size

- The common factors do not seem linear in sample.
- Test for multiple-structural break (Van Berkum et al. 2016).
- Starting date for estimating the random walk with drift

	Li-Lee	GDP
Group 1	1956	1971
Group 2	1991	1993
Group 3	1989	1993
Group 4	1956	1949



# Performance

	Li-Lee	GDP
Group 1	0.07	0.07
Group 2	0.08	0.06
Group 3	0.08	0.07
Group 4	0.07	0.06

- Similar for Group 1.
- Better for other groups.



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# A more general framework

- The product-ratio model in Hyndman et al. (2013).
- It can be interpreted as an extension of the Li-Lee model.

$$\log f_{i,x,t} = \tilde{\mu}_{i,x} + \sum_{p=1}^P b_{p,x} \phi_{p,t} + \sum_{q=1}^Q b_{i,q,x} \Phi_{i,q,t} + \tilde{\varepsilon}_{i,x,t},$$

- $\log f_{i,x,t}$  — smoothed  $\log m_{i,x,t}$ ;
- $P$  common factors;
- $Q$  specific factors for each population.
- More general time series processes
  - ▶ ARIMA for  $\phi_{p,t}$ ;
  - ▶ ARFIMA for  $\Phi_{i,q,t}$ .



# GDP product-ratio

- Replace the first common latent factor by the mean of the logarithm of GDP

$$\phi_{1,t} = \frac{\sum_{i=1}^I \log G_{i,t}}{I}$$

- Replace the first population-specific factor for population  $i$  by the difference between its GDP and the common factor

$$\Phi_{1,t} = \log G_{i,t} - \phi_{1,t}$$





# Out-of-sample forecast

	PR	GDP
Group 1	0.05	0.05
Group 2	0.09	0.07
Group 3	0.10	0.08
Group 4	0.07	0.06

- $P = 3, Q = 3.$



# Conclusion

- We study the long-term relationship between the aggregate economic growth and the aggregate mortality change in a group of closely related populations.
- We extend the Li-Lee model by including the observable trend of real GDP per capita.
- The proposed model has better in-sample fit (AIC, BIC), and out-of-sample forecast performance than the Li-Lee model.
- The relationship seems to hold in a more general framework.



Thank You!

