

Multi-population mortality models

Fitting, Forecasting, Comparison and Applications

Vasil Enchev

Heriot-Watt University, Scotland

Joint work with Andrew J.G. Cairns and Torsten Kleinow

International Mortality and Longevity Symposium 2016

August 25, 2016



Actuarial
Research Centre

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Motivation and purpose

- ▶ Forecasting joint mortality rates
 - ▶ The projections are correlated
 - ▶ Learn from other populations

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 - ▶ Annuities
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 - ▶ Learn from other populations
- ▶ Multi-population risk assessment
 - ▶ Annuities
 - ▶ Reserving
 - ▶ Diversifying portfolio risk
 - ▶ SCR estimates
- ▶ Hedging risk
 - ▶ Q-forward contracts
 - ▶ S-forward contracts

Data specifics

Six populations considered:

<i>i</i>	Country	Exposure to risk at the age of 60 in 2010	
		Male	Female
1	Austria	47023	49526
2	Belgium	65344	66434
3	Czech Republic	71575	71575
4	Denmark	34420	35132
5	Sweden	59759	59742
6	Switzerland	46527	47078

Data range: 30 ages (from 60 to 89 years old) and 50 calendar years (from 1961 up to 2010)

Importance of data chosen

- ▶ Closely related countries
- ▶ Similar size and features

Multi-population mortality models

Model 0 : "*Li and Lee (2005)*"

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Common parameters

Country specific parameters

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The diagram illustrates the classification of parameters in the mortality model equation. The equation is $\log m(x, t, i) = \alpha(x, i) + B(x) K(t) + \beta(x, i) \kappa(t, i)$. The parameters $\alpha(x, i)$, $B(x)$, and $K(t)$ are grouped as "Common parameters" (indicated by red boxes and red arrows). The parameters $\beta(x, i)$ and $\kappa(t, i)$ are grouped as "Country specific parameters" (indicated by blue boxes and blue arrows).

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Common parameters Country specific parameters

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- ▶ Parameter estimation - *Maximum likelihood estimation*

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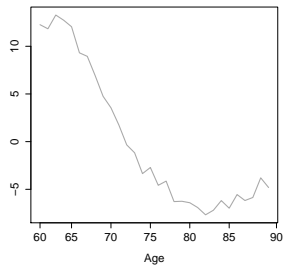
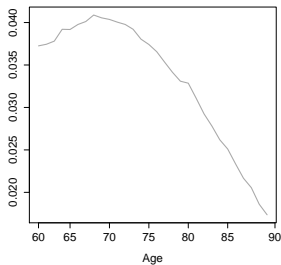
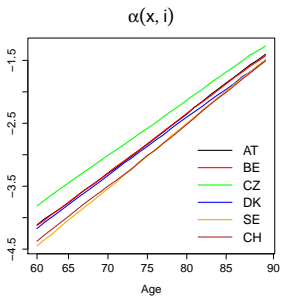
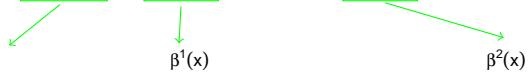
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- ▶ Model 1 and Model 2 - special cases of Model 0
- ▶ Identifiability issue - *Constraints implementation*
- ▶ Parameter estimation - *Maximum likelihood estimation*
- ▶ Model selection criterion - *BIC*

	BIC value	Rank
Model 0	100043.08	(2)
Model 1	101628.38	(4)
Model 2	101525.12	(3)
Model 3	99805.38	(1)

Estimated age dependent parameters in Model 3

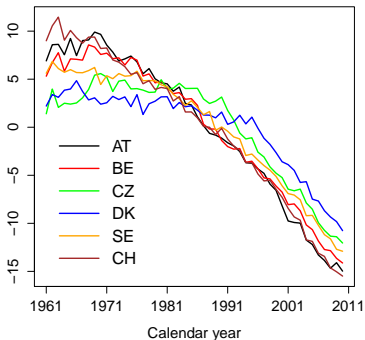
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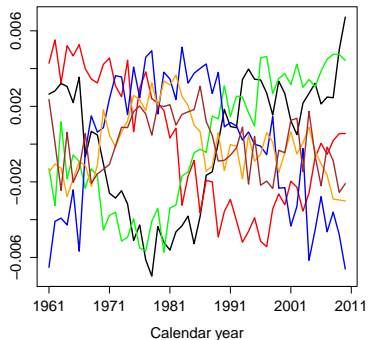
Estimated time dependent parameters in Model 3

$$\log m(x, t, i) = \alpha(x, i) + \beta^1(x) \kappa^1(t, i) + \beta^2(x) \kappa^2(t, i)$$

$\kappa^1(t, i)$



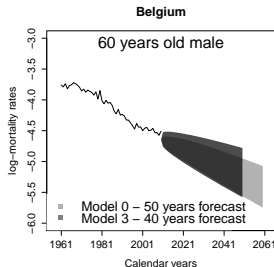
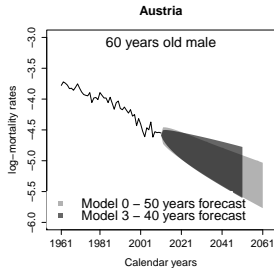
$\kappa^2(t, i)$



Forecasting joint mortality rates

Points to consider

- ▶ Shape of the estimated parameters
- ▶ Common time dependent parameters
 - ▶ Single variate time series processes
- ▶ Country specific time dependent parameters
 - ▶ Multivariate time series processes
- ▶ Projected joint mortality forecasts



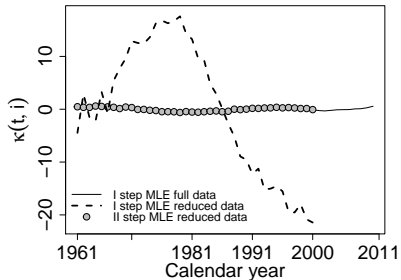
Robustness of the estimated parameters

- ▶ Full data set: 1961 - 2010
- ▶ Reduced data set: 1961 - 2000
 - ▶ One step MLE
 - ▶ Two step MLE

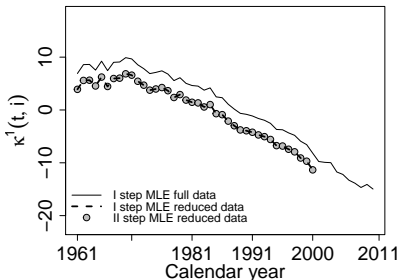
Robustness of the estimated parameters

- ▶ Full data set: 1961 - 2010
- ▶ Reduced data set: 1961 - 2000
 - ▶ One step MLE
 - ▶ Two step MLE
- ▶ Models robustness
 - ▶ Model 0 suffers from multi maxima problems
 - ▶ Model 3 is very robust
- ▶ Conclusion
 - ▶ Model 3 ranks as the best model

Model 0 – Austria



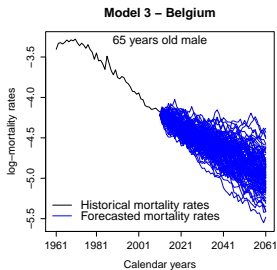
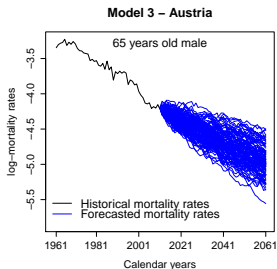
Model 3 – Austria



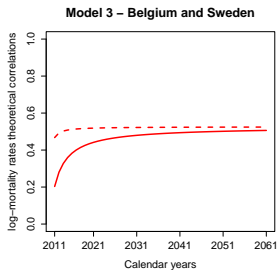
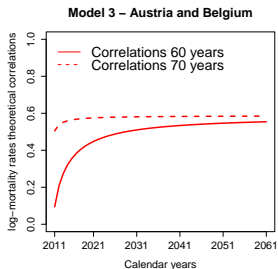
Multi-population mortality models applications

$$\log m(x, t, i) = \alpha(x, i) + \beta^1(x)\kappa^1(t, i) + \beta^2(x)\kappa^2(t, i)$$

Joint forecasts



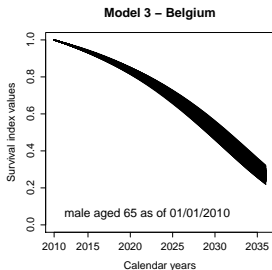
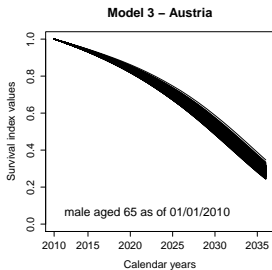
Theoretical correlations



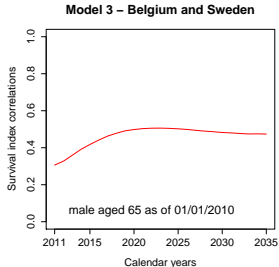
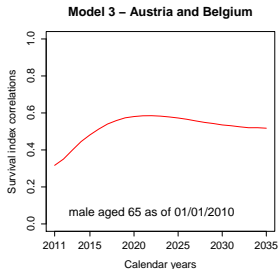
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Survival index values



Empirical correlations

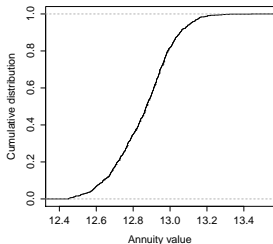


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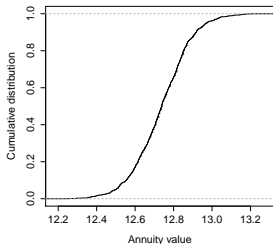
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Annuity ECDF

Model 3 – Austria

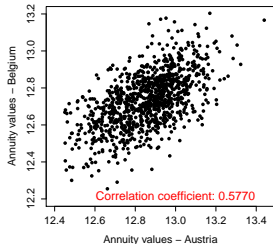


Model 3 – Belgium

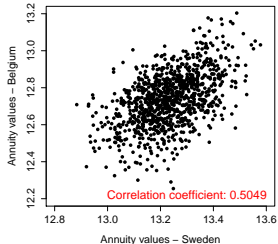


Annuity scatter plots

Model 3 – Annuity values



Model 3 – Annuity values



Multi-population mortality models applications

Other applications of multi-population mortality models include

- ▶ Reserving
- ▶ Assess benefits of diversification across countries
- ▶ SCR values: single and multi country
- ▶ Hedging portfolio risk
 - ▶ Q-forward contracts
 - ▶ S-forward contracts

<http://www.macs.hw.ac.uk/~andrewc/papers/Enchev2015.pdf>

▶ Paper link