



## Modeling Mortality Trend under Modern Solvency Regimes

**Matthias Börger**  
**Daniel Fleischer**  
**Nikita Kuksin**

**February 2012**

Helmholtzstraße 22  
D-89081 Ulm  
phone +49 (0) 731/50-31230  
fax +49 (0) 731/50-31239  
email ifa@ifa-ulm.de

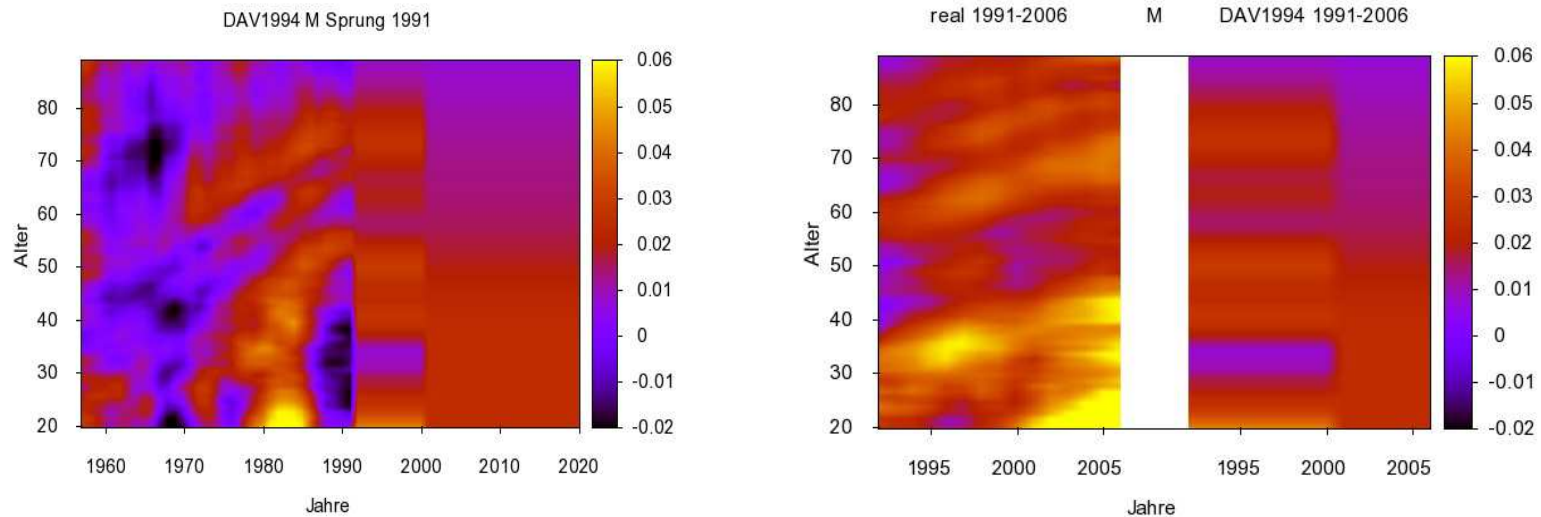
ifa

Institut für Finanz- und  
Aktuarwissenschaften



## Introduction

### What is longevity risk?



→ Longevity risk is the risk of underestimating future mortality improvements

- Trend risk
- Mortality risk has a trend risk and a catastrophe risk component
- Systematic and non-hedgeable risk
  - Explicitly accounted for under Solvency II and the Swiss Solvency Test (SST)

## Capital Requirements under Solvency II

- **General concept for Solvency Capital Requirement (SCR) under Solvency II**
  - SCR = 99.5% Value-at-Risk (VaR) of Available Capital over 1 year
  - „Capital necessary to cover losses over next year with at least 99.5% probability“
  - Overall risk is typically split into several modules, individual SCRs are finally aggregated
- **Stochastic mortality model is required for mortality/longevity trend risk under Solvency II**
- **In a 1-year setting, longevity/mortality trend risk consists of two components:**
  - Low/high realized mortality in the one year
  - Decrease/increase in expected future mortality, i.e. changes in the long-term mortality trend

## Mortality Trend Model Requirements

- **Goal: Specification and calibration of a mortality model with the following properties**
  - Simultaneous modeling of mortality and longevity risk
    - Exploiting of diversification effects
  - Full age range
    - 20 to 105 in our case
  - Consideration of several populations at the same time
    - Males and females in the same country
    - Populations from different countries
  - Quantification of risk over limited time horizons
    - One-year view of Solvency II and the SST particularly relevant
  - Plausible tail scenarios
    - 99.5% VaR
  - Conservative calibration

## Model Specification and Estimation

### We model the logit of mortality rates

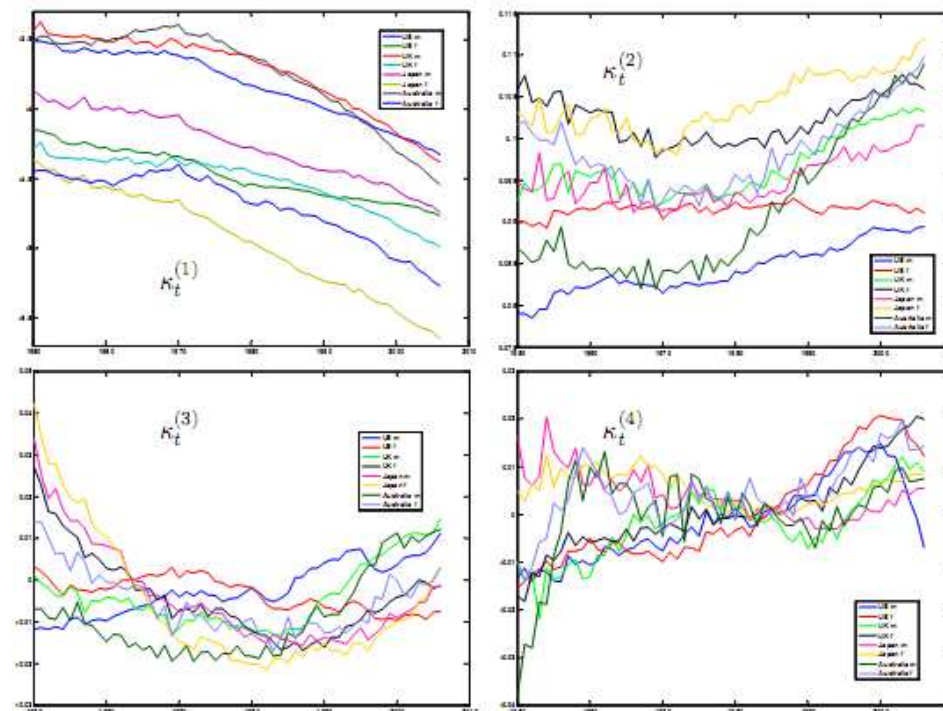
$$\text{logit}(q_{x,t}) = \alpha_x + \kappa_t^{(1)} + \kappa_t^{(2)}(x - x_{center}) + \kappa_t^{(3)}(x_{young} - x)^+ + \kappa_t^{(4)}(x - x_{old})^+ + \gamma_{t-x}$$

- $x_{center} = 60, x_{young} = 55, x_{old} = 85$

- $\kappa_t^{(1)}$  describes the general level of mortality,  $\kappa_t^{(2)}$  is the slope of the mortality curve,  $\kappa_t^{(3)}$  and  $\kappa_t^{(4)}$  describe additional effects in young and old age mortality, respectively

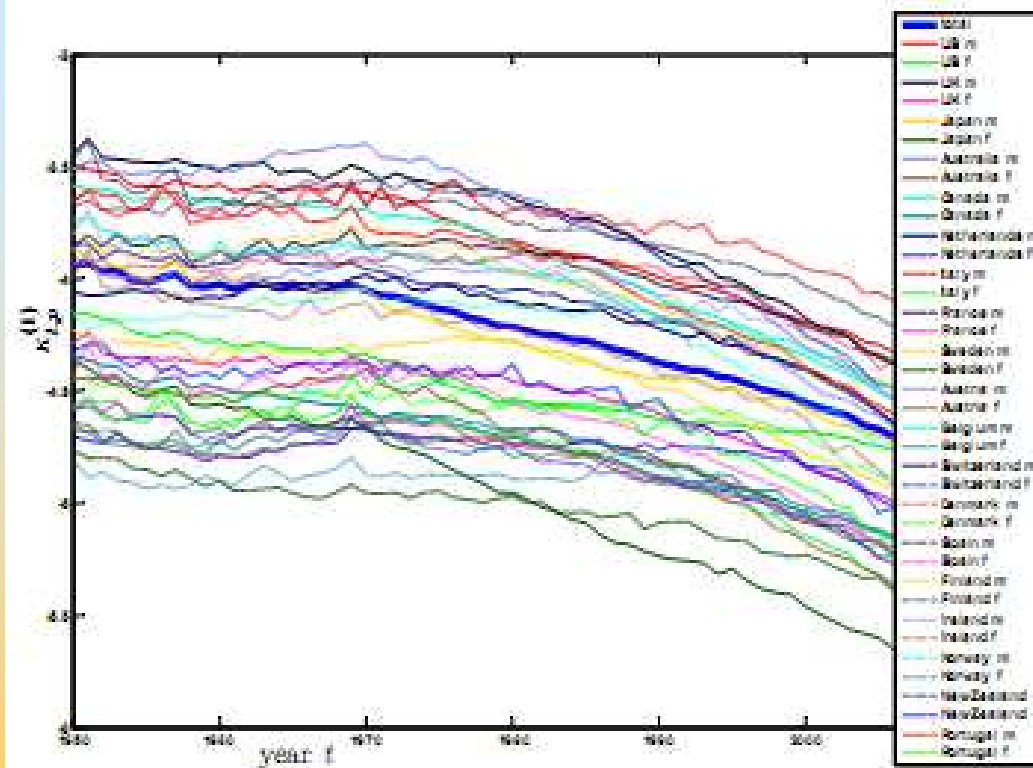
### Model estimation via Generalized Linear Model Theory

- Logit is canonical link function for Binomial distribution
- Number of deaths is binomially distributed given initial exposures
- Cohort parameters are fitted to residuals separately



ifa

## Multi-Population Setting



- There is clearly a common trend in  $K_t^{(1)}$
- A model for several populations must account for that
- We apply cointegration and an error correction model for deviations from the common trend

## Model Simulation

### Projection of $\kappa_{t,total}^{(1)}$ for the total population

- **Linear trends with breaks in the historical data**

- ▮ Commonly used random walk with drift does not allow for such trend breaks
- ▮ Trend breaks are particularly important under one-year view (change of best estimate trend)

- **Idea: Each year, fit regression line to historical data and forecast future best estimate mortality as**

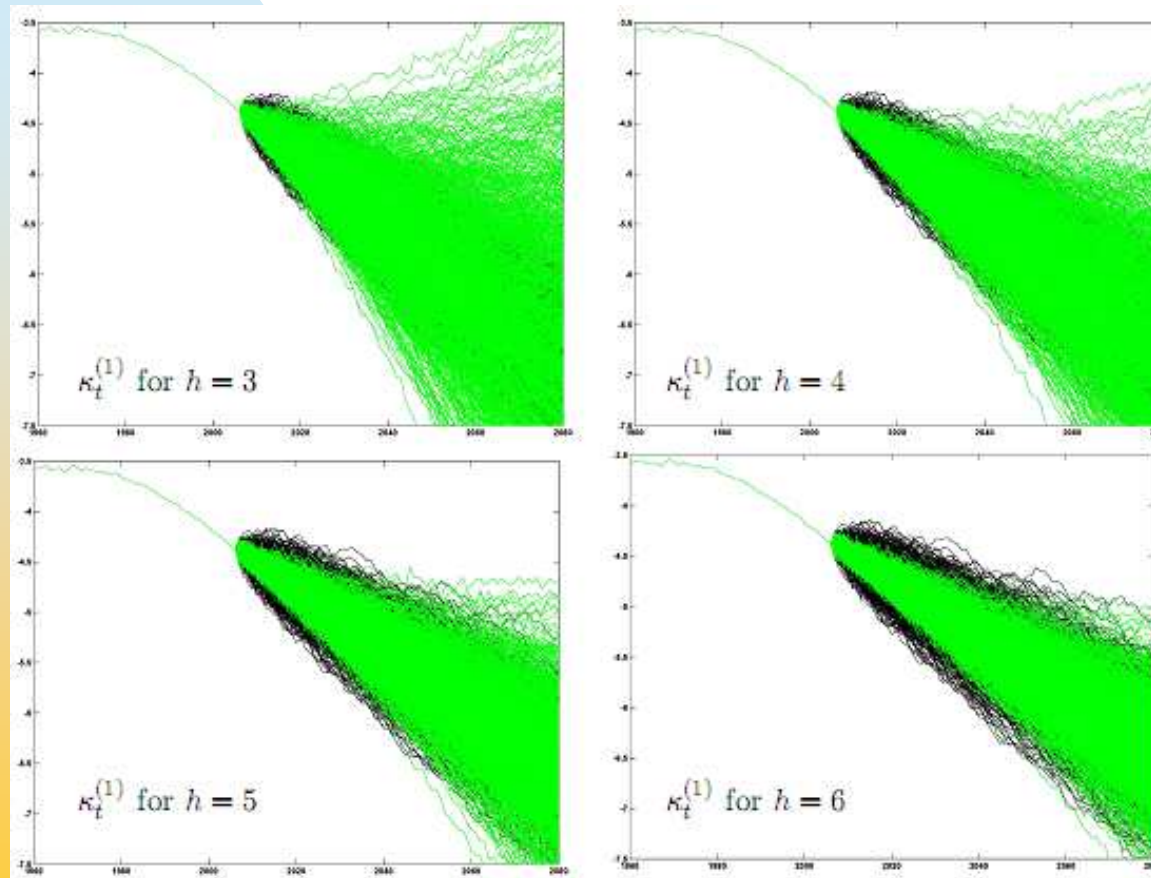
$$\kappa_{t+1,total}^{(1)} = l_t(t+1) + \varepsilon_{t+1}^{(1)}(\sigma^{(1)} + \bar{\sigma}^{(1)})$$

- ▮  $\bar{\sigma}^{(1)}$  is a volatility add-on
- ▮ volatility  $\sigma^{(1)}$  may be weighted to stress most recent past
- ▮ Implicit „re-calibration“ of the model with respect to the long-term trend
- ▮ To stress most recent mortality experience, the regression line is fitted with weights

$$w_s = \left(1 + \frac{1}{h}\right)^{s-t}$$

## Model Simulation (ctd.)

Iterative application of the trend forecasting:



- Weighting parameter  $h$  has massive impact
- Plausible one-year and run-off scenarios
- Each run-off scenario is a combination of one-year scenarios
- Disentangling of one-year noise and long-term trend uncertainty
- Possibly more plausible confidence bounds than for a random walk with drift



## Model Simulation (ctd.)

### Projection of $\kappa_{t,p}^{(1)}$ for individual populations

- For each individual population we project as

- $\kappa_{t,p}^{(1)} = \kappa_{t,total}^{(1)} + a_p + b_p (\kappa_{t-1,p}^{(1)} - \kappa_{t-1,total}^{(1)}) + \varepsilon_{t,p}$

- $b_p$  denotes the „mean reversion speed“ (absolute value should be smaller than 1)

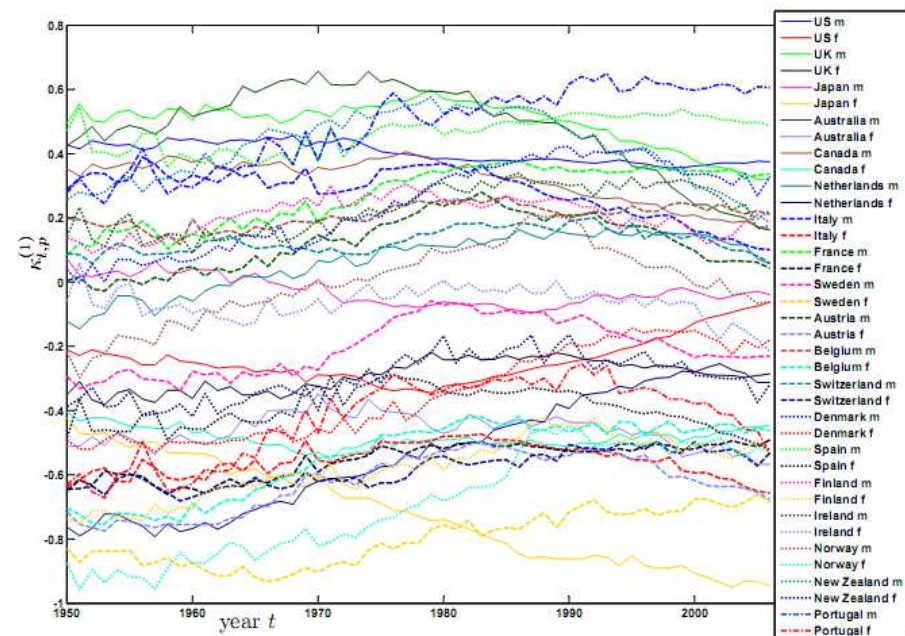
- $a_p / (1 - b_p)$  is the long-term difference between the total population and population p

- Different approaches of calibrating the long-term difference

- Fitting of an AR(1) process to historical differences

- Weighted/unweighted average of historical differences

- ...



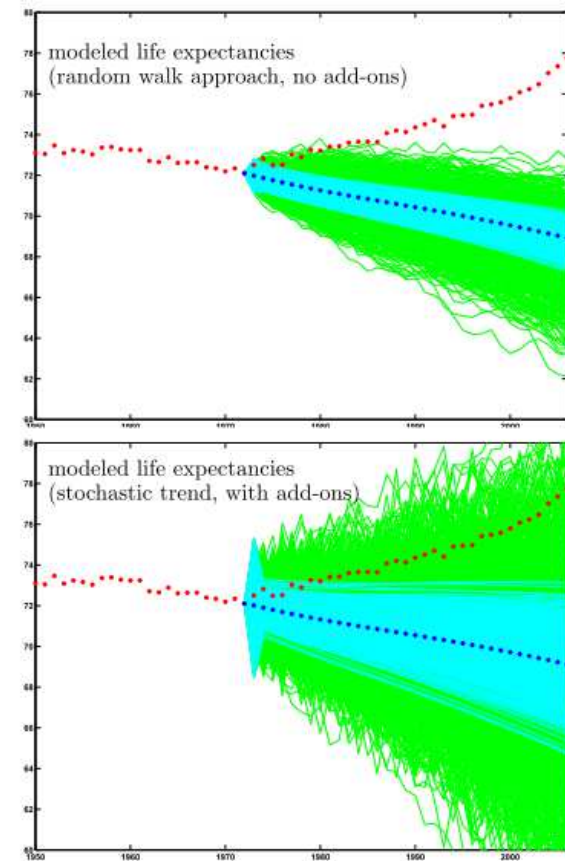
ifa

## Model Simulation (ctd.)

- **Projection of  $\kappa_t^{(2)}$ ,  $\kappa_t^{(3)}$ , and  $\kappa_t^{(4)}$  for the individual populations**
  - No substantial trend obvious in the historical data
  - Forecast as correlated 3-dimensional random walk
  - No substantial correlation with  $\kappa_t^{(1)}$
  - Volatility add-on  $\bar{\sigma}^{(2)}$  for  $\kappa_t^{(2)}$ 
    - The larger the changes in the slope of the mortality, the smaller the correlation between young and old ages
    - Thus the add-on affects diversification between mortality and longevity risk
  - Between populations, increments of  $\kappa_t^{(1)}$  and  $\kappa_t^{(2)}$  are correlated
    - This also implies slight correlation between the  $\kappa_t^{(3)}$  and  $\kappa_t^{(4)}$
    - Historical correlations should be checked carefully though and possibly adjusted
- **Projection of  $\gamma_{t-x}$** 
  - Cohort parameters should stay around zero
  - Forecast as Gaussian noise
  - Cohort parameters are rather irrelevant for short-term simulations as under Solvency II

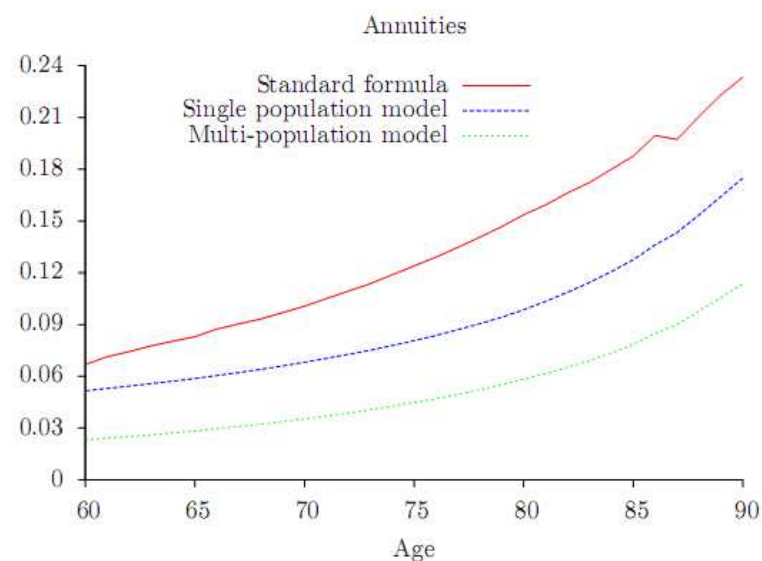
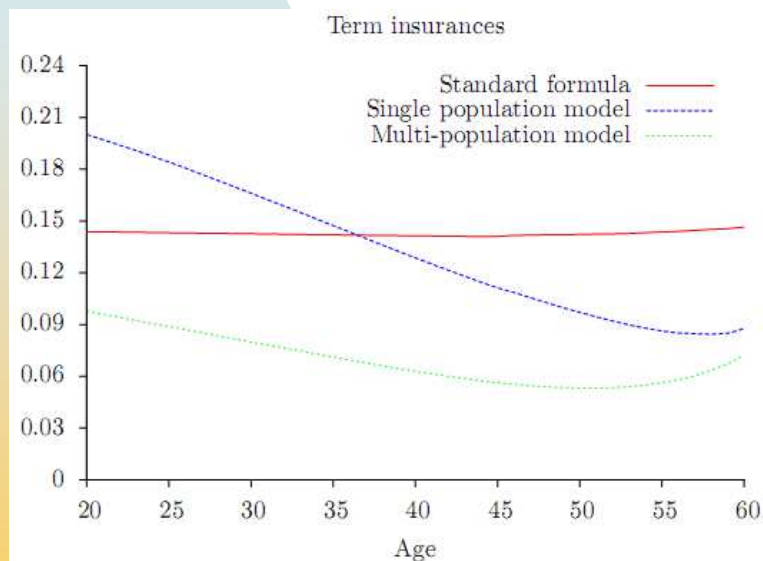
## Weighting Parameters and Volatility Add-ons

- Parameters  $h$ ,  $\bar{\sigma}^{(1)}$ , and  $\bar{\sigma}^{(2)}$  have a massive impact on simulation outcomes and thus SCRs
- Add-on  $\bar{\sigma}^{(1)}$  determines possible severity of short-term events
- Weighting parameter  $h$  determines trend changes over one year and width of confidence bounds
- Calibration is difficult but should be conservative
  - Fitting to most severe events/evolutions in the past
    - Example: Rapid increase in Dutch life expectancy gains starting from about 1970
    - Question: At which percentile should such extreme evolutions be observed?
- Calibration of  $\bar{\sigma}^{(2)}$ 
  - The larger the add-on the smaller the correlation between young and old ages thus limiting diversification
  - Choose  $\bar{\sigma}^{(2)}$  such that correlation between ages at the boundaries of the age range is (close to) zero for most populations



## Numerical comparisons

- **Standard formula approach**
  - Reduction/increase of all mortality rates by 20%/15%
  - Change in liabilities is about SCR for longevity/mortality risk
- **SCRs for term insurances (maturity at age 65) and whole life annuities:**



- **Multi-population model demands least capital**
  - Model is worthwhile even for only one population because of reduced trend uncertainty
- **Standard model seems to overstate the risk in general**

## Mortality/Longevity Threat Scenarios

- Available data contains only little information on tail scenarios which we are interested in
- Uncertainty remains whether model outcomes are severe enough
  - Incorporate epidemiological/demographic expert opinion
- **Specification of mortality/longevity threat scenarios**
  - Shock to mortality projection
  - Likely effects of finding of a cure for a certain illness
  - Scenarios which the statistical model cannot generate, e.g., diverging mortality trends between countries/regions
  - ...
- **Application of threat scenarios**
  - Check of model calibration: Adjustment of weighting parameter or volatility add-ons if the model outcomes should cover the threat scenarios but do not
  - Inclusion in SCR computations: SCR as a weighted average of model outcomes and threat scenarios

## Summary

- **Specification, calibration, and application of a mortality model for solvency purposes**
  - Full age range
  - Variability in simulation outcomes due to 5 stochastic drivers
  - Clear interpretation of the model parameters
  - Non-trivial correlation structure to allow for simultaneous modeling of mortality and longevity risk
  - Stochastic trend modeling spares full re-calibration of the model in each scenario
  - Plausible outcomes in one-year view and run-off view
  - Conservative calibration
  - Inclusion of expert opinion
  - Multi-population model allowing for diversification effects and risk reduction
- **Model can be applied for other purposes as well (with slightly different calibration)**
- **A stochastic trend process with several appealing features**
  - Quantification of trend risk over any desired time horizon
  - Advantages compared to the commonly used random walk with drift (e.g. long-term risk vs. short-term noise)
  - The trend process could easily be incorporated in other mortality models as well, e.g. Lee-Carter or Cairns-Blake-Dowd

- 
- 

## Contact Details

### Matthias Boerger

Institute of Insurance, Ulm University & Institute for Finance and Actuarial Sciences (ifa), Ulm  
Helmholtzstraße 22, 89081 Ulm, Germany

Phone: +49 731 50-31257, Fax: +49 731 50-31239

Email: [m.boerger@ifa-ulm.de](mailto:m.boerger@ifa-ulm.de)