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## Modeling Mortality Trend under Modern Solvency Regimes

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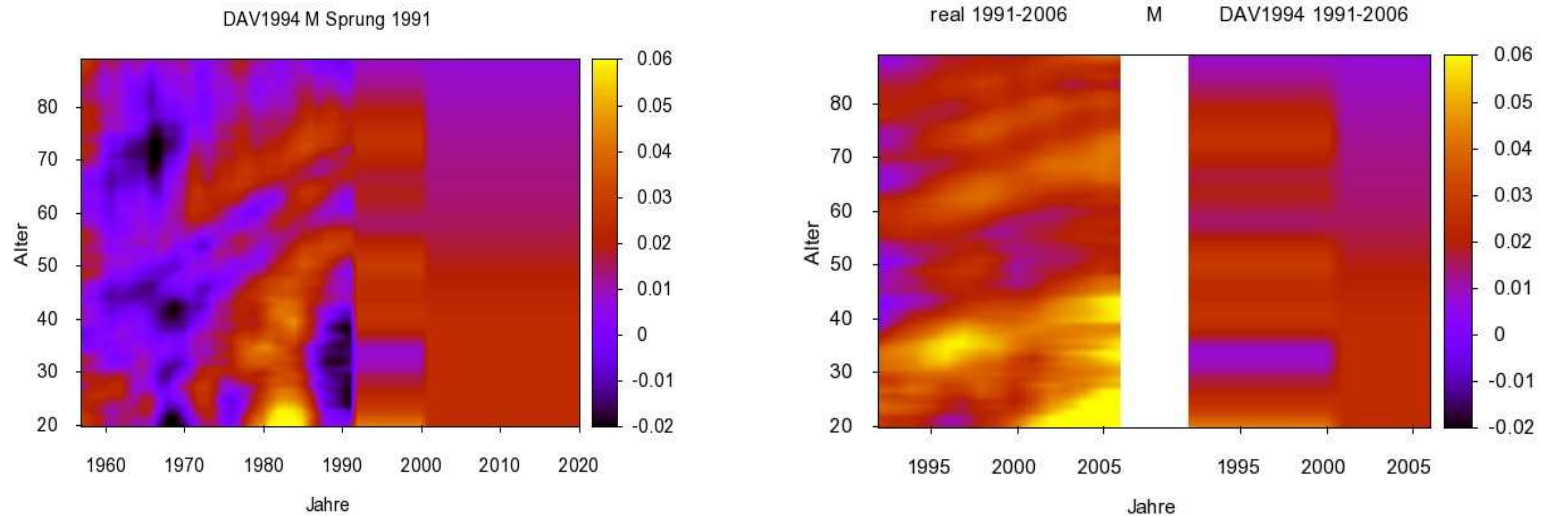
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## Introduction

### 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

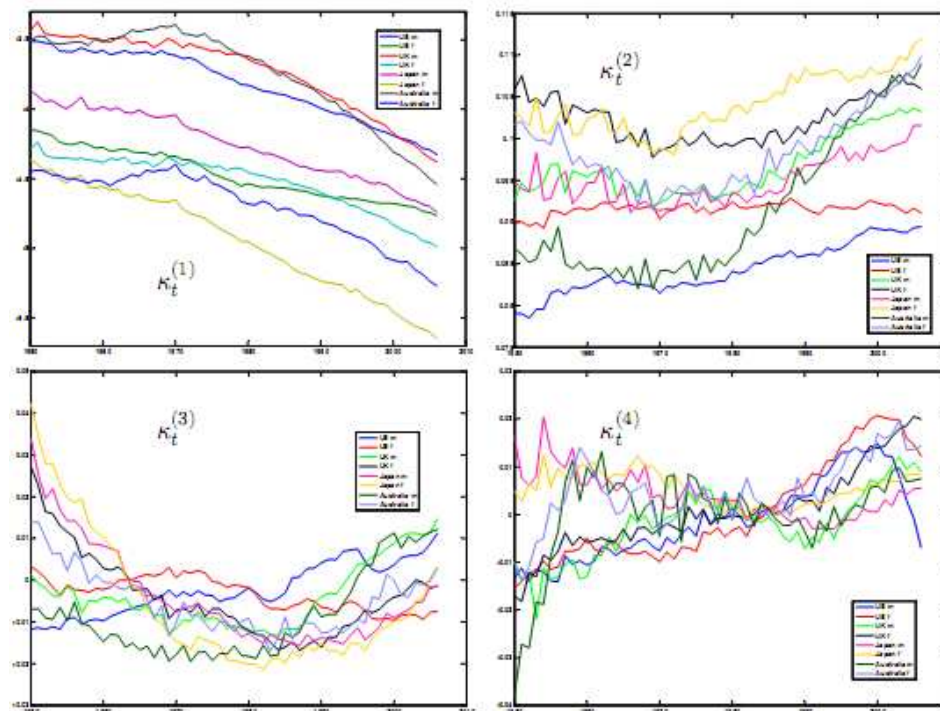
$$\text{logit}(q_{x,t}) = \alpha_x + \kappa_t^{(1)} + \kappa_t^{(2)}(x - x_{center}) + \kappa_t^{(3)}(x_{young} - x)^+ + \kappa_t^{(4)}(x - x_{old})^+ + \gamma_{t-x}$$

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

- $\kappa_t^{(1)}$  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

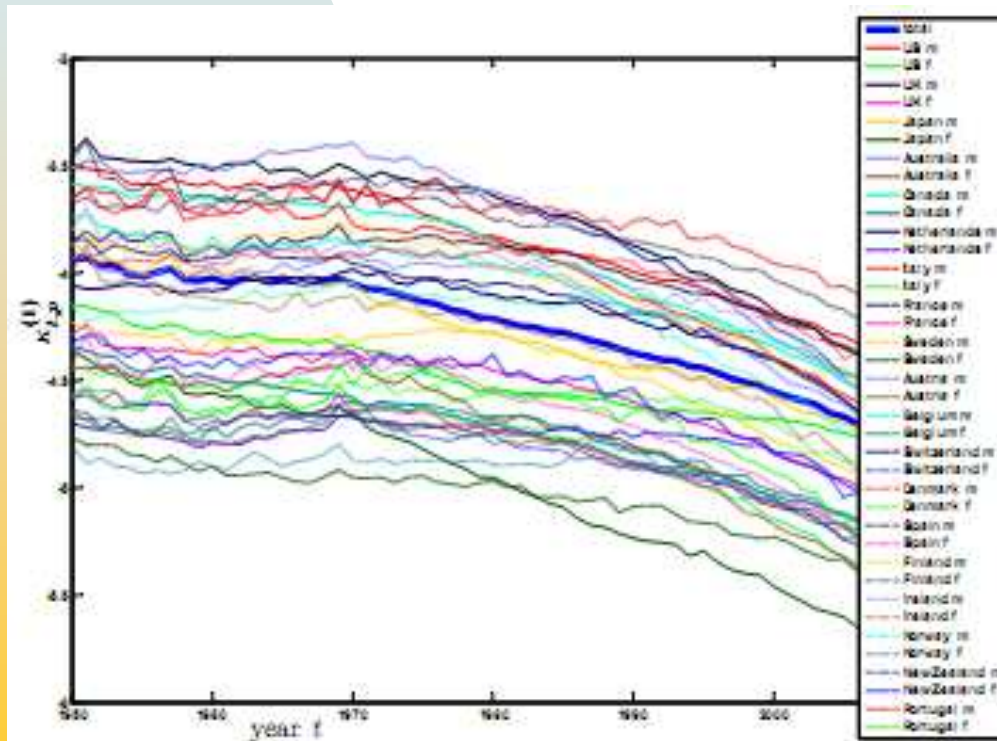
### Model estimation via Generalized Linear Model Theory

- Logit is canonical link function for Binomial distribution
- Number of deaths is binomially distributed given initial exposures



## Multi-Population Setting

- **Important note: Even if one is only interested in a single population considering several populations is worthwhile**
  - Trend uncertainty can be significantly reduced
  - We generally observe smaller SCRs in the multi-population model compared to the single population model



- **There is clearly a common trend in  $\kappa_t^{(1)}$**
- **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 $K_{t,total}^{(1)}$ for the total population

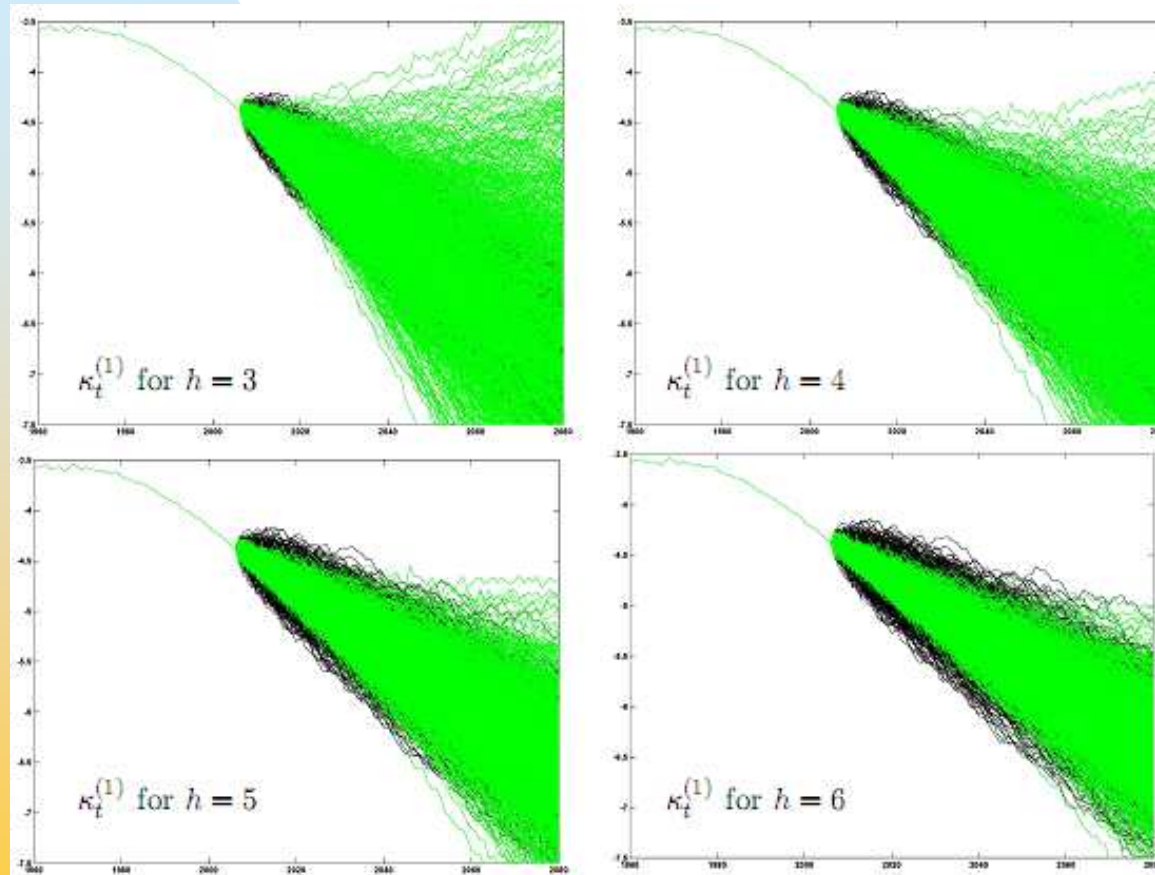
#### ■ Linear trends with breaks in the historical data

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

- $\bar{\sigma}^{(1)}$  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.)



- 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**

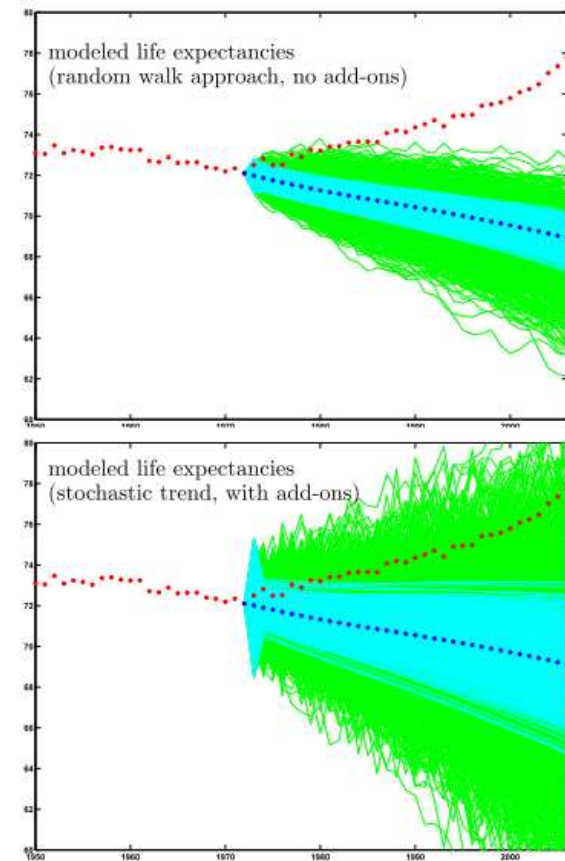
- Fitting of an AR(1) process to historical differences
- Weighted/unweighted average of historical differences

## 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 imposed stationary AR(1) process
  - 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

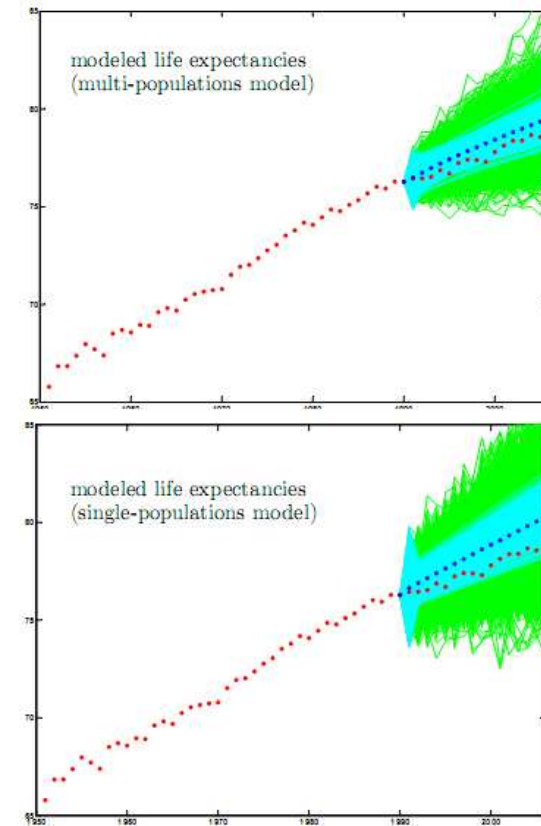


## Mortality/Longevity Threat Scenarios

- 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

- **We have specified and calibrated a mortality model with several appealing properties:**
  - 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 (could be applied in other models as well)
  - Plausible outcomes in one-year view and run-off view
  - Conservative calibration
  - Inclusion of expert opinion
  - Multi-population model allowing for diversification effects



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