



The Volatility of Mortality

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Taipei, July 2007

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Agenda

- Introduction
- Theoretical Framework
- Epidemiological Insights
- Specification of the Model
- Calibration
- Applications
- Summary and Outlook

Introduction

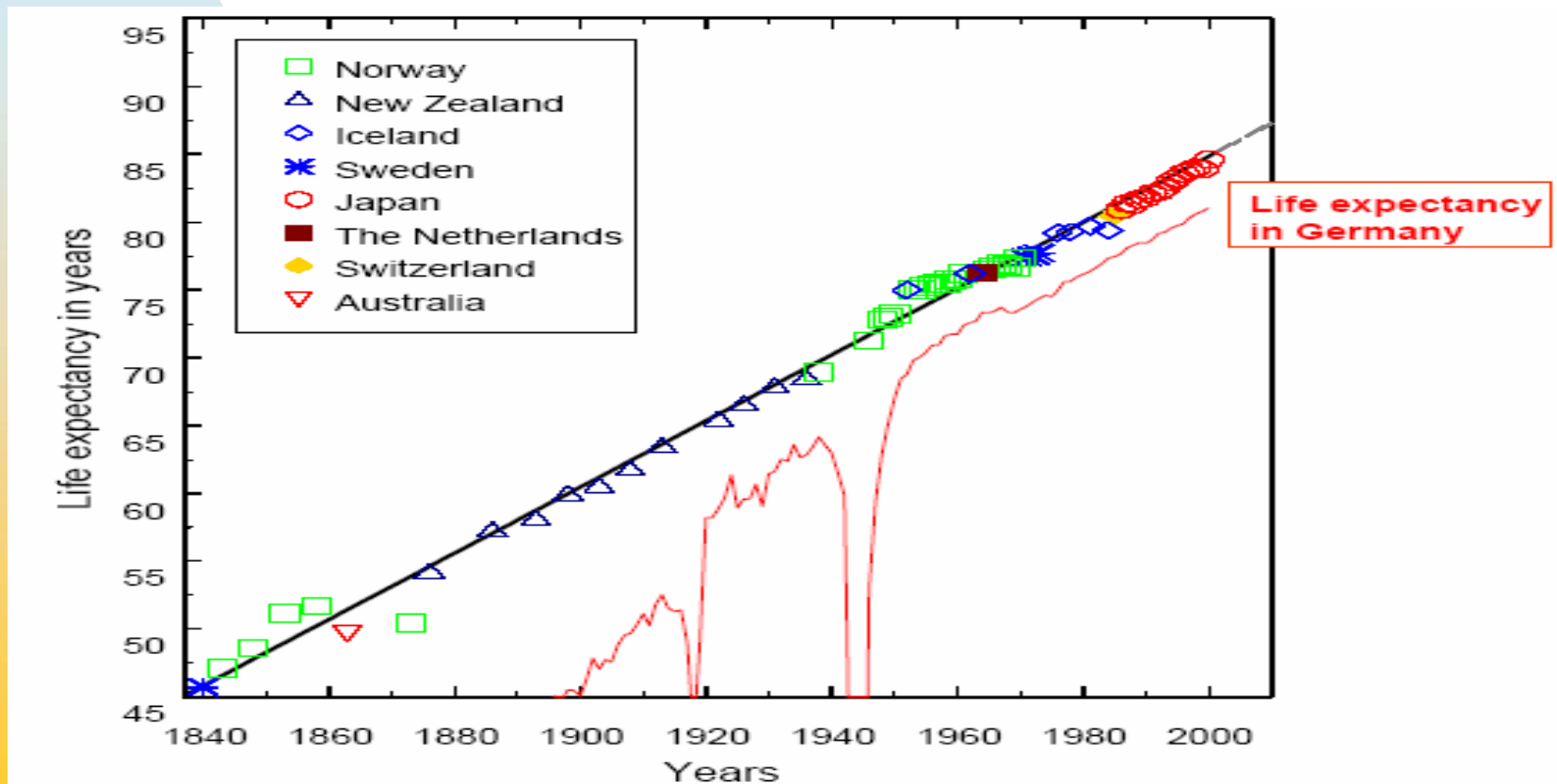
- **Traditional actuarial mathematics**
 - Future mortality probabilities are known
 - → Mortality and longevity risk diversifiable
- **Several recent studies**
 - Systematic mortality risk is an important risk factor for life and annuity insurers as well as pension providers
- **In particular, for annuity and pension providers**
 - Longevity risk
 - In many countries: very large longevity exposure in occupational pensions
 - Many countries provide tax incentives in case of annuitization
 - Longevity risk in the insurance industry will increase



Introduction

My favorite pictures

- 1. Increase in world-wide life expectancy is no temporary trend

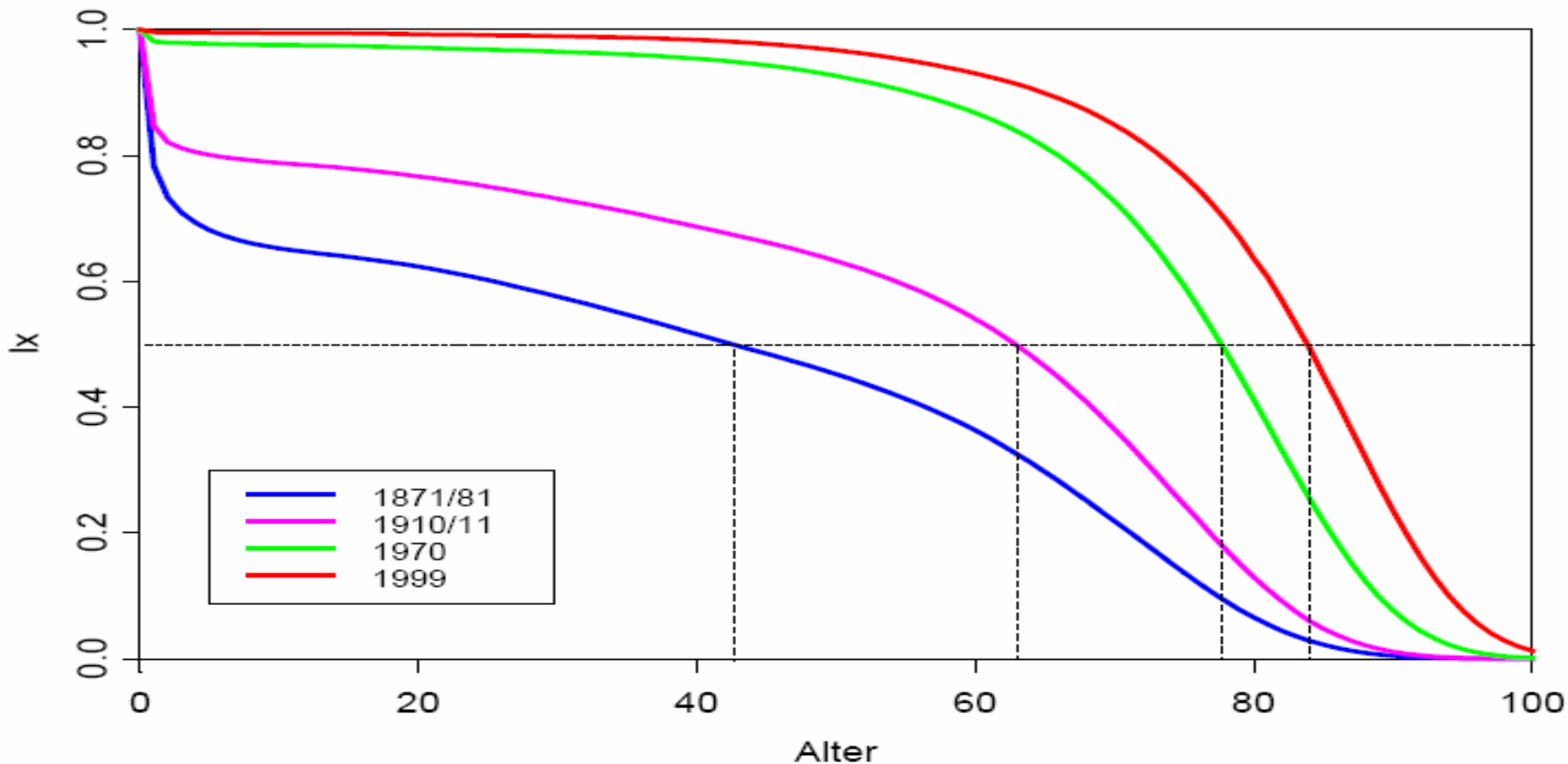


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Introduction

- My favorite pictures

- 2. Increase is not “uniform over the mortality curve”



Introduction

- **Since systematic mortality and longevity risk is not diversifiable, other methods of managing this risk have to be applied**
 - e.g. securitization (Cowley and Cummins (2005), Blake et al. (2006))
- **Target: quantify systematic mortality risk and model mortality linked securities**
 - e.g. longevity or survivor bonds
- **Required: Models which consider the stochastic characteristics of the mortality evolution.**
 - In recent literature, a number of such models have been presented
 - Overview in Cairns et al. (2005)

Introduction

- **Most of the models proposed so far are spot-force models**
 - Simplified: Stochastic versions of period life tables
- **In classical actuarial applications, annuities are usually priced based on generational tables**
 - → Trend is already embedded in the tables
 - Stochastic version of these: Forward mortality models
 - Model for the whole term structure of mortality
 - e.g. Cairns et al. (2005), Miltersen and Persson (2005)
- **However, no concrete forward models have been presented yet**
 - possible exception: stochastic extension of Smith/Olivieri model (Cairns)
- **In what follows: Presentation of two concrete models**

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Theoretical Framework

- In this presentation we focus on the main ideas and concepts and our results
- For more details on theory and formulae we refer to the full paper
 - available at www.mortalityrisk.org
- We use mortality bonds as a basis...
 - bonds with a payoff depending on the portion of a reference population of a certain age, that is still alive at some future point in time
- ... and assume that the forward force of mortality follows

$$d\tilde{\mu}_t(T, x_0) = \tilde{\alpha}(t, T, x_0) dt + \tilde{\sigma}(t, T, x_0) d\tilde{W}_t, \tilde{\mu}_0(T, x_0) > 0$$

for a finite dimensional Brownian Motion \tilde{W}

Theoretical Framework

- To specify a concrete model, we need
 - initial condition
 - drift term
 - volatility structure
- **Heath-Jarrow-Morton condition (HJM): Under the risk neutral measure Q , the volatility structure specifies the drift term**
 - → initial condition and volatility structure sufficient
- **Approach for calibrating volatility under Q (proposed by Bauer and Russ, 2006)**
 - Calibrate to prices of embedded guarantees in Variable Annuities with Guaranteed Minimum Benefits
 - However: Market is not (yet) sufficiently liquid
 - Therefore, in what follows, we work with observed mortality data, i.e. under the real-world measure P

Theoretical Framework

■ Some results under P

- We show that there is a P-HJM condition similar to the Q-HJM condition
- A model calibrated under P can be used to determine expected payoffs, percentiles, etc. of mortality contingent claims
- However, it is not immediately suitable as a pricing model
- Under the assumption of mortality risk-neutrality, the models coincide
- In general, in order to specify a corresponding pricing model, the market price of risk has to be specified
- In the Gaussian case and under the assumption of a deterministic market price of risk, the volatilities of the P and Q models coincide
 - The initial risk-neutral forward plane, the Q-HJM condition, and a P-calibrated volatility structure yield a pricing model

Theoretical Framework

- **The resulting pricing model can be used to**
 - determine the values of mortality securitization products, e.g., longevity bonds
 - determine values of mortality contingent options within insurance products such as Guaranteed Annuity Options or Guaranteed Minimum Benefits within Variable Annuities
 - etc.
- **Still open: Where do we find a suitable volatility structure under P?**

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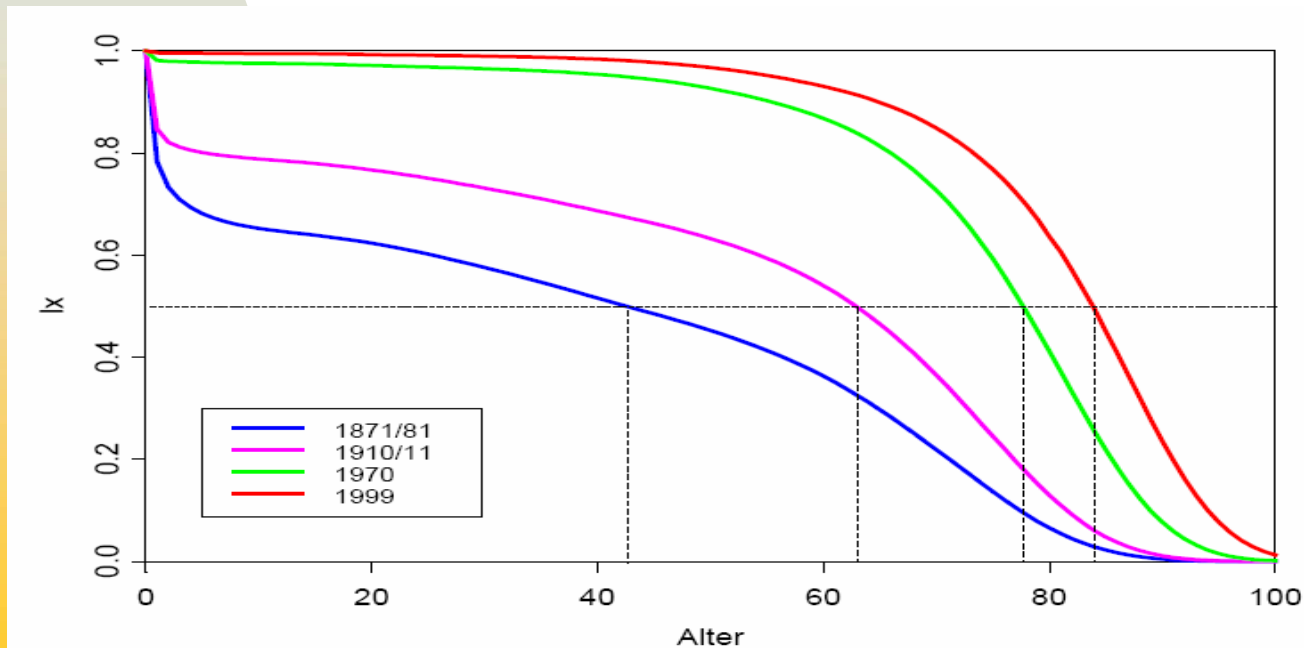
Epidemiological Insights

- Since mortality data are not collected frequently (unlike e.g. interest rate data), standard statistical methods like principal component analysis are not possible
- Furthermore, the question how mortality might evolve is not primarily a statistical/mathematical/actuarial question
- Thus, we tried to derive a suitable model structure from epidemiological/medical insights

- However, “typical” epidemiological approaches are not suitable in our framework
 - Cause of death models or risk factor models
- The resulting number of parameters would make the model unfeasible

Epidemiological Insights

- Idea: Combine effects that affect mortality for similar ages at similar points in time
- Qualitative epidemiological insights – age and term
 - Strong changes in mortality happen for different ages at different times



Epidemiological Insights

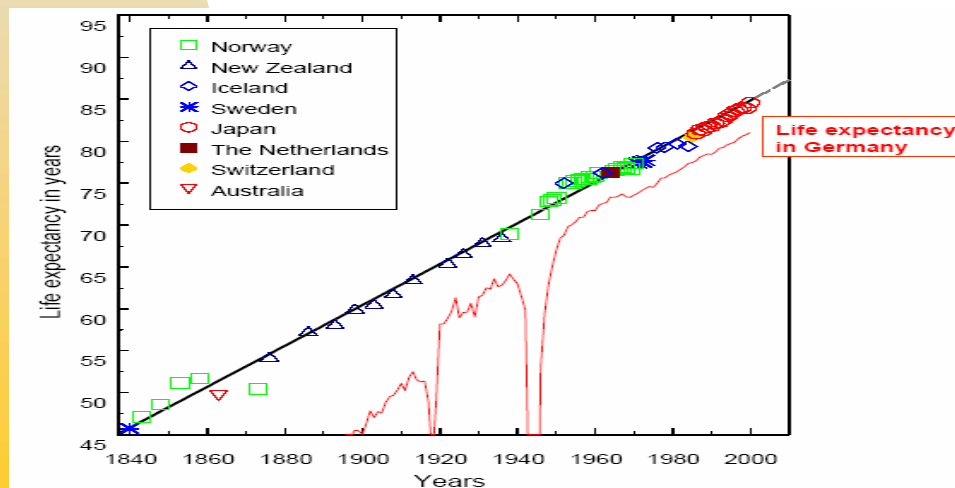
- **Idea: Combine effects that affect mortality for similar ages at similar points in time**
- **Qualitative epidemiological insights – age and term**
 - Strong changes in mortality happen for different ages at different times
 - → Grouping by age and time appears reasonable
 - → Use three age groups: young, mid, old
 - Old age starts at 80
 - Epidemiological studies show that currently the strongest improve in longevity happens at these ages
 - Although, e.g. ages 70+ are usually referred to as old, from a “mortality point of view” they are more similar to mid than old ages (Wiesner (2001), Boleslawski and Tabeau (2001))
 - Other groups: Young = age 20 to 55; mid = 55-80

Epidemiological Insights

- **Qualitative epidemiological insights – age and term**
 - We need a short term a mid term and a long term effect
 - Short-term effect affects only the “near future”
 - wave of influenza, bird flu,...
 - hits all ages “similarly” → not age dependent
 - One mid-term effect for each age group
 - One long term effect → not age dependent
 - When long term effect kicks in, all cohorts considered are rather old already
 - Finally, we introduce one overall effect (see below)
 - → 6 effects

Epidemiological Insights

- **Qualitative epidemiological insights – Distributional properties**
 - No limit to a potential temporary **increase** in mortality
 - Controversial discussion whether or not an upper limit exists for human life expectancy (**decrease** in mortality)
 - Some favor an upper limit
 - Other favor an “increasing boundary” which for our purpose can be treated as a “time dependent upper limit”



Epidemiological Insights

- **Qualitative epidemiological insights – Distributional properties**
 - No limit to a potential temporary **increase** in mortality
 - Controversial discussion whether or not an upper limit exists for human life expectancy (**decrease** in mortality)
 - Some favor an upper limit
 - Other favor an “increasing boundary” which for our purpose can be treated as a “time dependent upper limit”
 - → Forward force of mortality should have a positively skewed distribution with rather light tails.

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Specification of the model

The 6 effects

- exponential building blocks
- The age dependent mid-term effect: bell shaped with center at the middle of the age range

general: $\sigma_1(t, T, x_0) = c_1$

short-term: $\sigma_2(t, T, x_0) = c_2 \cdot \exp(-w_2(T - t))$

young age: $\sigma_3(t, T, x_0) = c_3 \cdot \exp(-w_{31}(T - t - 20)^2 - w_{32}(x_0 + T - 37.5)^2)$

middle age: $\sigma_4(t, T, x_0) = c_4 \cdot \exp(-w_{41}(T - t - 20)^2 - w_{42}(x_0 + T - 67.5)^2)$

old age: $\sigma_5(t, T, x_0) = c_5 \cdot \exp(-w_{51}(T - t - 20)^2 - w_{52}(x_0 + T - 110)^2)$

long-term: $\sigma_6(t, T, x_0) = c_6 \cdot \exp(-w_6(T - t - 120)^2)$

→ 14 parameters – that's a lot!

Specification of the model

- **Eliminating parameters for the sake of feasibility**
 - Short term effect: After one year reduced to 10% → $w_2 = \log(1/10)$
 - “Smooth transition” from one effect to the next
 - with respect to age and time
 - Each effect is reduced to 50% at the boundary to the next effect
 - Rather arbitrary – other restrictions yield different models
 - This transition and the overall effect, allow for correlation
- → Resulting (not yet final) model has 6 parameters

Specification of the model

- **Final adjustment: Distributional properties**
 - Model so far is Gaussian
 - Changes in forward mortality are absolute, i.e. volatility is decreasing in age
 - Moving to " $d\mu/\mu$ " would solve that but result in a log-normal-type distribution which is not desired
 - We therefore propose to use correction terms
 - multiply volatility derived so far with a suitable correction term

Specification of the model

Model 1: Gompertz correction term \rightarrow Gaussian model

- Assuming the Gompertz curve to be a reasonable approximation for the force of mortality μ , σ is roughly proportional to μ ; model remains Gaussian

general: $\sigma_1(t, T, x_0) = c_1 \cdot \exp(a(x_0 + T) + b)$

short-term: $\sigma_2(t, T, x_0) = c_2 \cdot \exp(a(x_0 + T) + b) \cdot \exp(\log(0.1)(T - t))$

young age: $\sigma_3(t, T, x_0) = c_3 \cdot \exp(a(x_0 + T) + b)$
 $\cdot \exp\left(\frac{\log(0.5)}{20^2}(T - t - 20)^2 + \frac{\log(0.5)}{17.5^2}(x_0 + T - 37.5)^2\right)$

middle age: $\sigma_4(t, T, x_0) = c_4 \cdot \exp(a(x_0 + T) + b)$
 $\cdot \exp\left(\frac{\log(0.5)}{20^2}(T - t - 20)^2 + \frac{\log(0.5)}{12.5^2}(x_0 + T - 67.5)^2\right)$

old age: $\sigma_5(t, T, x_0) = c_5 \cdot \exp(a(x_0 + T) + b)$
 $\cdot \exp\left(\frac{\log(0.5)}{20^2}(T - t - 20)^2 + \frac{\log(0.5)}{30^2}(x_0 + T - 110)^2\right)$

long-term: $\sigma_6(t, T, x_0) = c_6 \cdot \exp(a(x_0 + T) + b)$
 $\cdot \exp\left(\frac{\log(0.5)}{80^2}(T - t - 120)^2\right)$

Specification of the model

- **Model 2: correction term = SQRT(μ) \rightarrow general (or Square-root) model**
 - square-root Feller diffusion process
 - roughly X^2 distributed (cf. Glasserman (2004))
 - positively skewed and light tailed as desired

general: $\sigma_1(t, T, x_0) = c_1 \cdot \sqrt{\hat{\mu}_t(T, x_0)}$

short-term: $\sigma_2(t, T, x_0) = c_2 \cdot \sqrt{\hat{\mu}_t(T, x_0)} \cdot \exp(\log(0.1)(T - t))$

young age: $\sigma_3(t, T, x_0) = c_3 \cdot \sqrt{\hat{\mu}_t(T, x_0)}$
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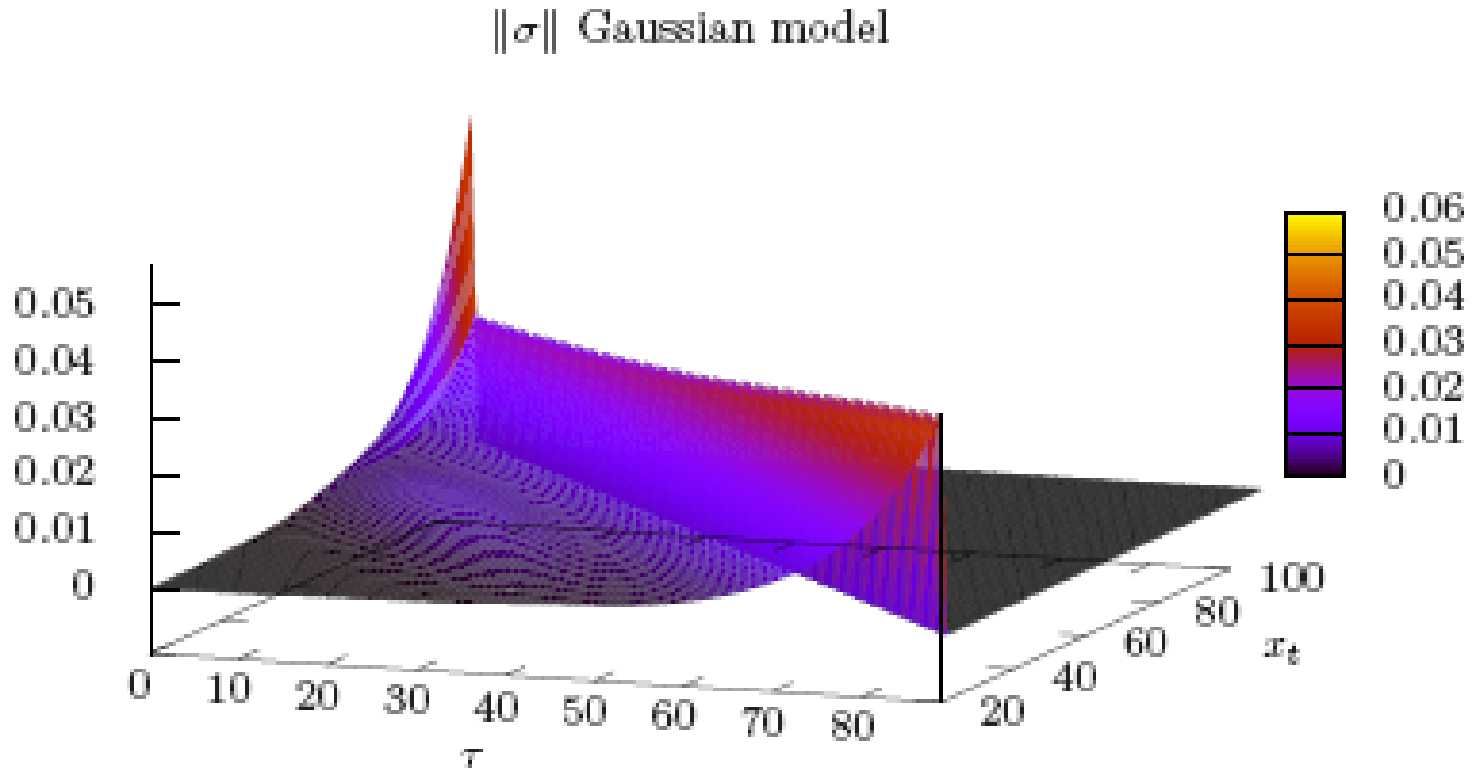
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Calibration

- **In the Gaussian case, we find close form solutions (up to the numerical solution of multidimensional integrals) for the moments of the distribution of the difference of the logs of the change in survival probabilities over time**
 - → Thus we know the probability density
 - This can be used in a maximum likelihood framework to perform a calibration
- **For the general model, we develop a pseudo-maximum likelihood methodology (based on Monte Carlo simulation)**
 - numerically complex
 - details see paper
- **We calibrate our models to the Group Annuity Mortality Tables (GAM tables) published by the Society of Actuaries**
 - 1951, 1971, 1983 and 1994 generation table
 - “artificial” creation of larger dataset seems possible

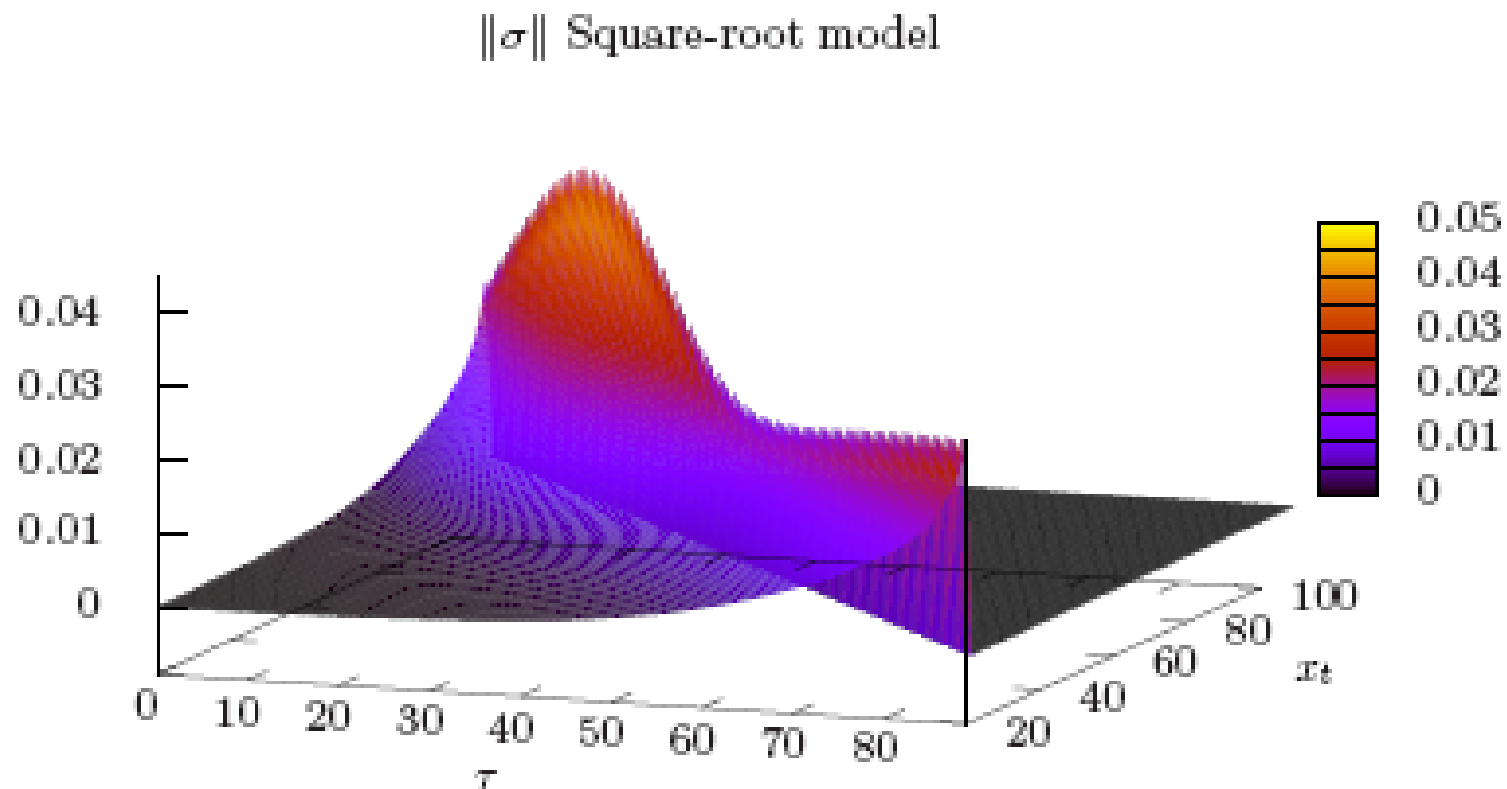
Calibration

- calibration results for the Gaussian model (absolute)
 - norm of $\sigma(t, T, x_0)$; $x_t := x_0 + t$ and $\tau = T - t$



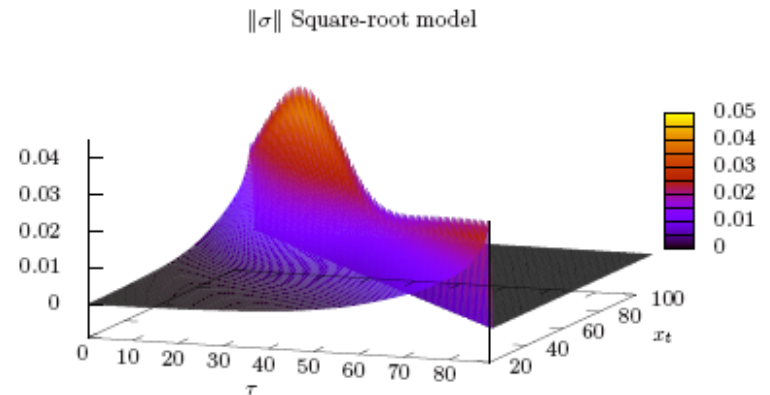
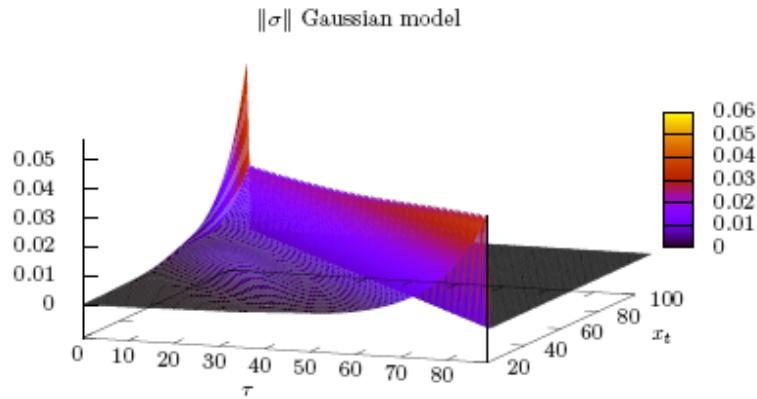
Calibration

- calibration results for the general model (absolute)
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Calibration

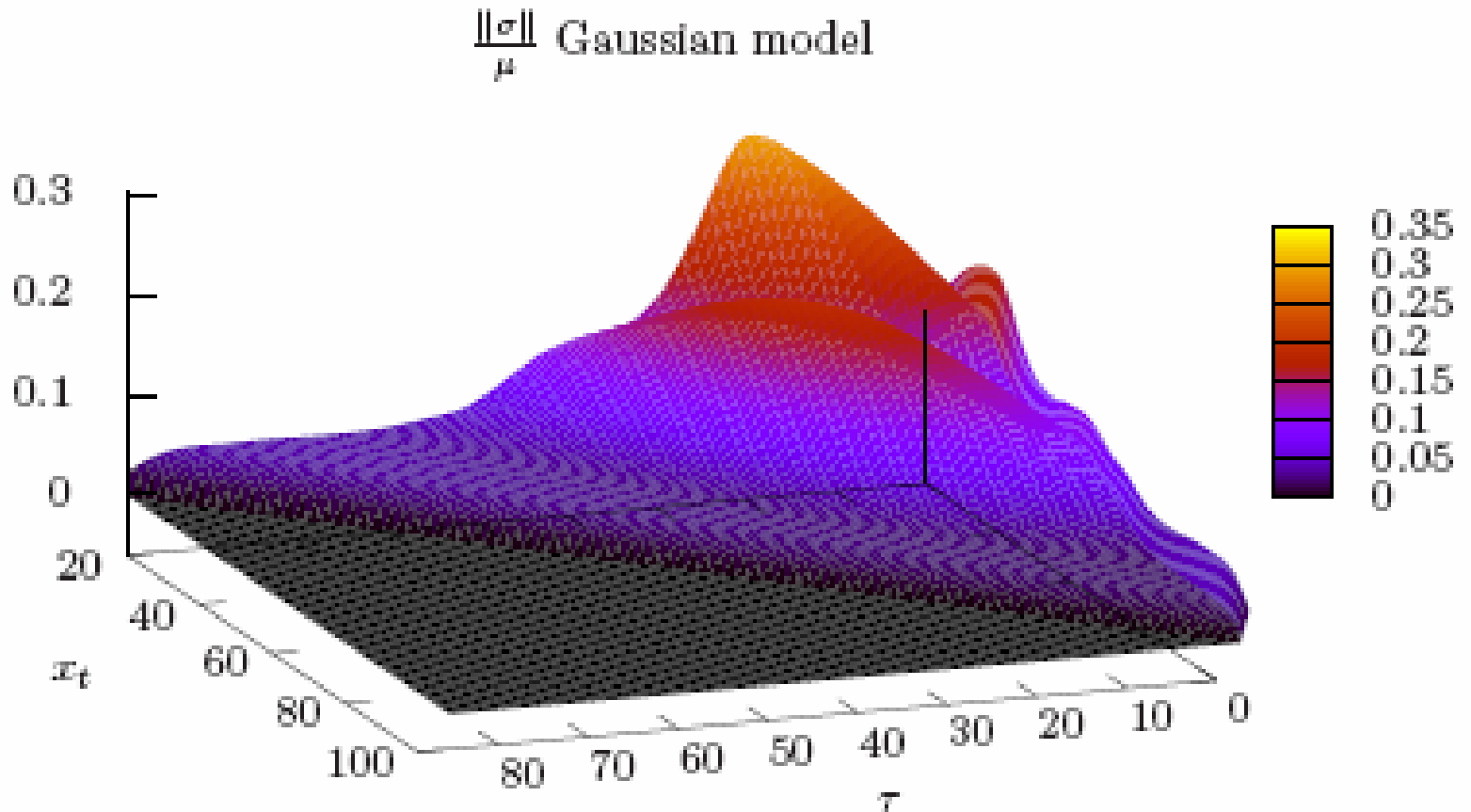
calibration results



- exponential trend in τ for each x_t and in x_t for each τ
- volatilities for the same age at maturity increase in the time to maturity \rightarrow the further in the future, the more uncertainty
- short term effect more pronounced in the Gaussian model; mid-term effect larger for the Square-root model
 - The comparison is somewhat misleading (different weightings)
 - \rightarrow also look at relative values

Calibration

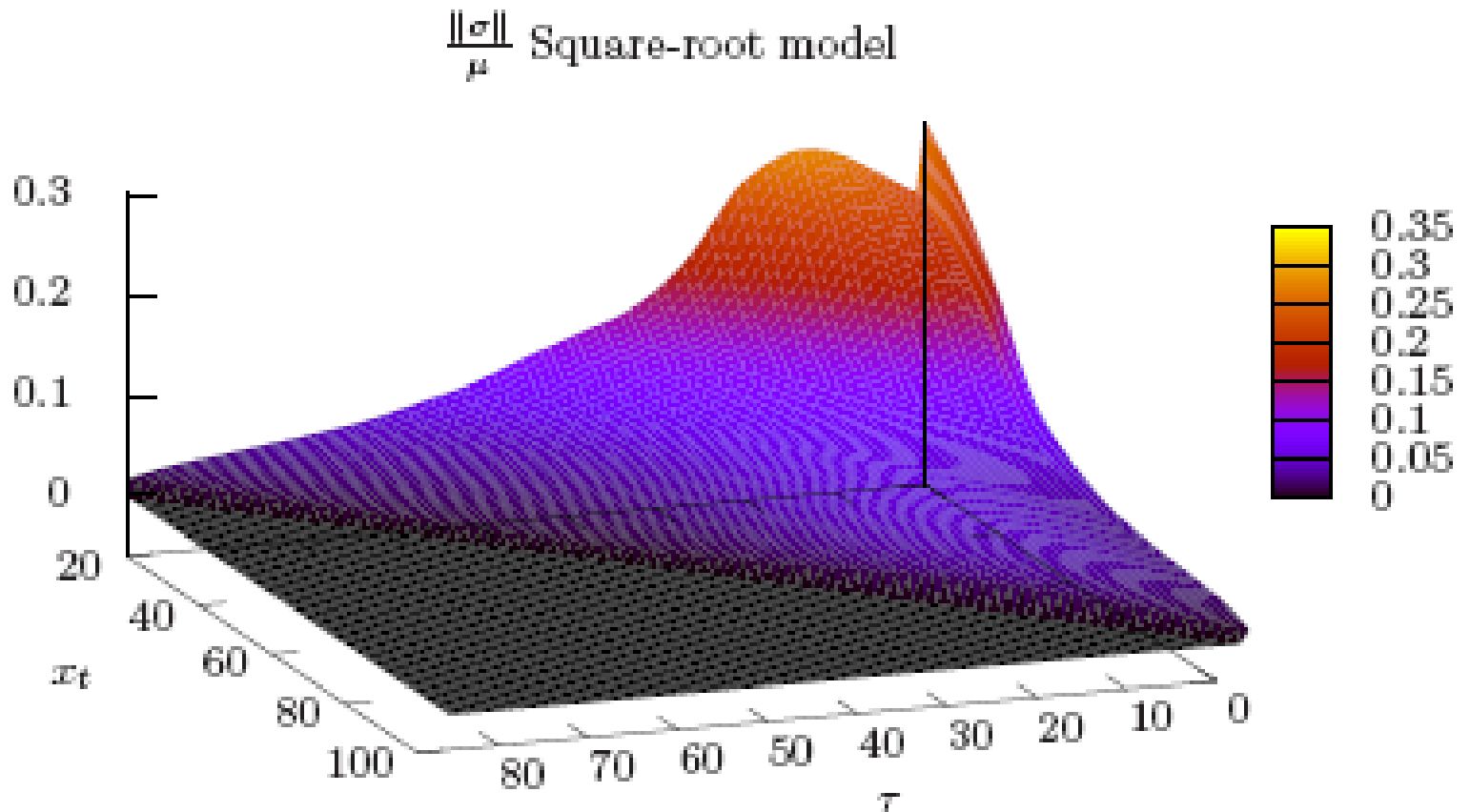
- calibration results for the Gaussian model (relative)



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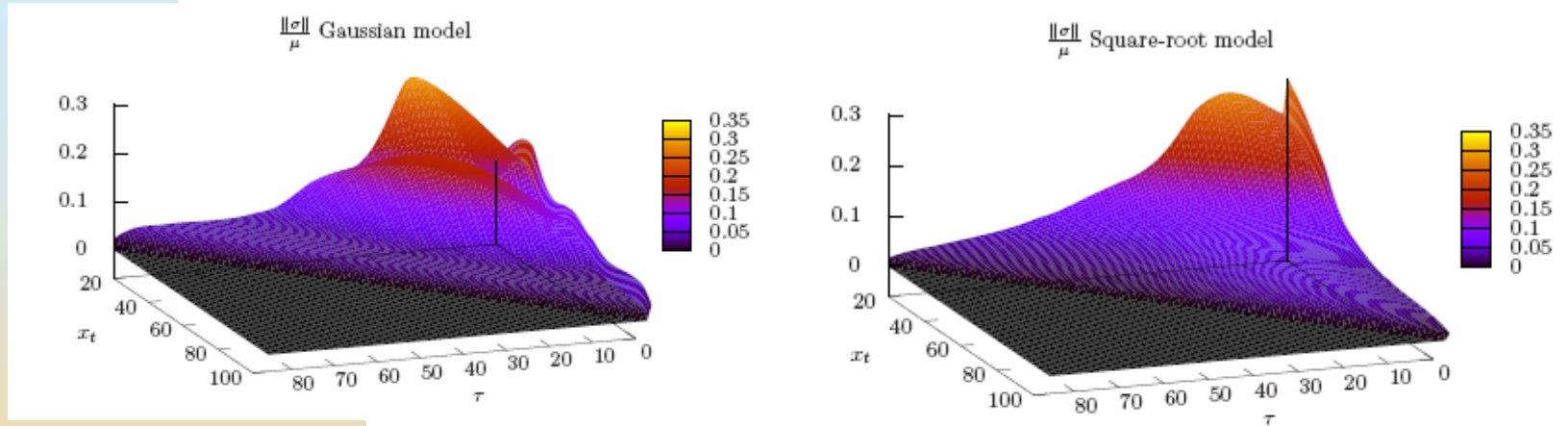
Calibration

- calibration results for the general model (relative)



Calibration

calibration results



- short-term effect is now more pronounced in the Square-root model
- young age effect is very large in both models
- mid age effect is hardly noticeable for the Square-root case but quite pronounced for the Gaussian calibration.
- overall relative volatility decreases in the age at maturity
 - this is different for the absolute volatility
- general structure and size of the volatilities are fairly similar

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Applications

- In the paper, we discuss several areas of application

- One example:

- Option with payoff $C_T = \left(TP_{x_0}^{(T)} - K \right)^+$,

with strike $K := (1 + \alpha) E_P \left[TP_{x_0}^{(T)} \right]$

- Trigger probabilities and expected payoff under P (T=20, $\alpha=2.5$)

Model	Gaussian model		Square-root model	
	$P \left(TP_{x_0}^{(T)} > K \right)$	$E \left[\left(TP_{x_0}^{(T)} - K \right)^+ \right]$	$P \left(TP_{x_0}^{(T)} > K \right)$	$E \left[\left(TP_{x_0}^{(T)} - K \right)^+ \right]$
$x_0 = 65$	0.4076	0.0491	0.4214	0.0200
$x_0 = 70$	0.4021	0.0334	0.4268	0.0194
$x_0 = 75$	0.3877	0.0174	0.4190	0.0122
$x_0 = 80$	0.3567	0.0058	0.3961	0.0042
$x_0 = 85$	0.3053	0.0011	0.3670	0.0008

Applications

- **Trigger probabilities slightly higher in the Square-root model**
 - consequence of the distributional differences of the models
 - Gaussian: mean and median coincide
 - Square-root: mean exceeds the median
 - higher probability for intensities just below the expected value
 - increase of the probability of survival probabilities slightly above the expected value
- **Expected payoff considerably larger in the Gaussian model**
 - relative difference between the expected payoffs increases in age
 - consequence of the far more pronounced middle age effect in the Gaussian model

Applications

- Using the market price of risk obtained by Lin and Cox (2005), we also priced this mortality option

	$p(0, T) E_P [{}_T P_{x_0}^{(T)}]$	$p(0, T) E_P [({}_T P_{x_0}^{(T)} - K)^+]$	$p(0, T) E_Q [({}_T P_{x_0}^{(T)} - K)^+]$
$x_0 = 65$	0.1006	0.0112	0.0214
$x_0 = 70$	0.058	0.0076	0.0168
$x_0 = 75$	0.0228	0.0040	0.0095
$x_0 = 80$	0.0053	0.0013	0.0033
$x_0 = 85$	0.0007	0.0003	0.0006

- Price is rather high
 - Insurer might combine this “long call on the survival probability” with a short put or chose a different strike.

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Summary

- **Whilst there seems to be consent that forward-type models are appropriate to assess stochastic mortality, so far no concrete model has been proposed**
 - Reasons: lack of data, in particular price data; medical questions
- **Therefore, we chose a different approach**
 - model with age and time effects
 - model structure based on epidemiological insight
 - model calibration based on observed mortality data (under P)
- **We derived several new theoretical results about the relation between P and Q models**
- **We specified two concrete models**
- **Model calibration**
 - semi-analytical for Gaussian model
 - new quasi maximum likelihood method developed for the general model
- **We showed several applications of the model**

Limitations and Outlook

- **Certain aspects of the approach are rather arbitrary**
 - age limits; exponential functions; parameter reduction; correction terms
 - other choices would lead to different results
 - empirical investigations based on “synthetic” data might yield further insights regarding the adequacy of the approach
- **Calibration Procedures:**
 - Problem: Data not necessarily coherent with possible structure implied by the model
 - → Gaussian case: Numerical instabilities due to over-determinedness of linear system of equation; only possible to incorporate “few” values leading to certain arbitrariness of the choice
 - → General case: Certain arbitrariness in the definition of some “acceptance level” within the used distance function
 - Solution: Smoothing the residuals – What smoothing function to use?
 - → “Consistency” Problem

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The Volatility of Mortality

Thank you very much!

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