

# The management of long term mortality risk at Swiss Re

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150 YEARS

#### Agenda

- Risk Management Framework
- Modeling of Influenza Pandemic Risk
- Mortality Trend Model
- Managing mortality exposure
- Causal Mortality Model





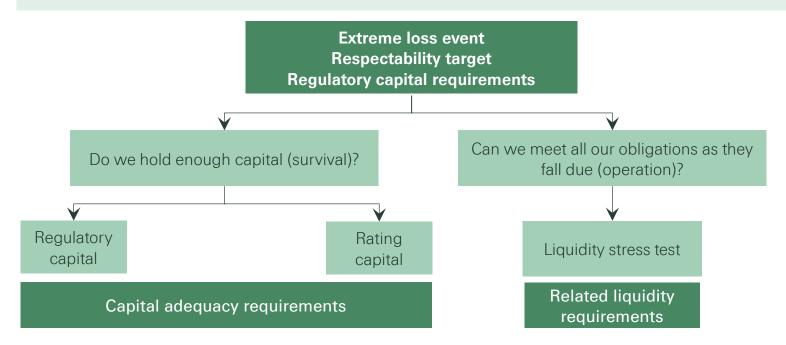
# Risk Management Framework



# Acceptance of mortality risk starts by deciding on the Company's Risk tolerance. This then acts as a basis for risk steering and limit setting.

The risk tolerance represents the amount of risk (mortality, property, liability, asset etc) Swiss Re is willing to accept within the constraints imposed by its capital and liquidity resources, its strategy, its risk appetite, and the regulatory and rating agency environment. It is based on the following objectives:

- Maintain capital and liquidity that are sufficiently attractive from a client perspective, and that
  meet regulatory requirements and expectations ("respectability criteria")
- Be able to **continue to operate following an extreme loss event** ("extreme loss criteria"):





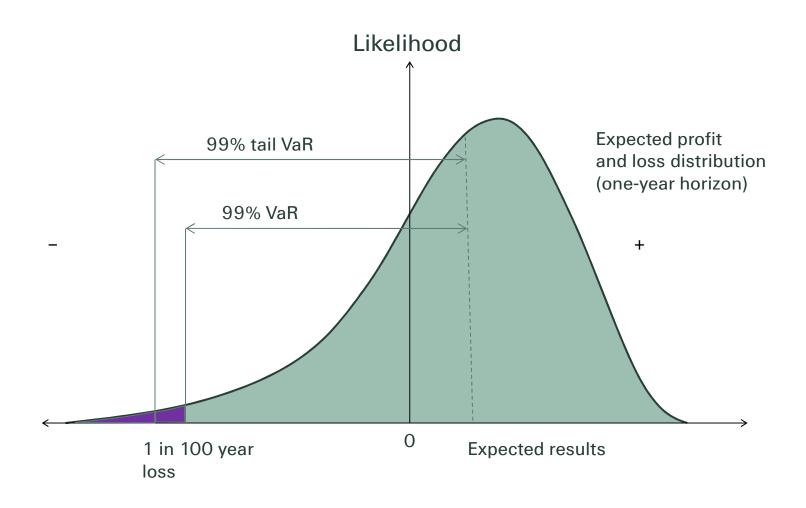
#### Life & Health Risks

- Life and Health Risk is defined as the unexpected economic impact from mortality, longevity or morbidity obligations as well as persistency rates deviating from the levels assumed at outset in costing or subsequently in reserving.
- For each risk factor, the underlying causes of uncertainty can be categorised into one of the following:
  - Shock Risk: an extreme, one-off fluctuation, e.g. mortality claims caused by a **lethal pandemic**.
  - Trend Risk: a permanent or cumulative deviation from the expected outcome, e.g. deviation in mortality/morbidity claims resulting from a medical advancement.
  - Parameter Risk: uncertainty related to pricing or reserving parameters, principally due to insufficient relevant information.
  - Volatility: a non-extreme, random fluctuation.

**Pandemic risk** and **trend risk** are the two major mortality related risks for Swiss Re and each will be discussed in detail later in the presentation.



# Swiss Re asseses its risk tolerance using a 99% tail VaR with a one year time horizon as its risk measure



### Risk Capacity Limits are set to keep the Company within its risk tolerance

- Risk capacity limits are established to control risk exposure accumulations at different levels.
- For Life and Health, three types of risk capacity limits are currently in place:

Type of limit	Description	
Top-level risk capacity limits	Aggregate limits based on Tail-VaR which govern the acceptance of all life and health risks, with separate individual limits for mortality and longevity risk	
Product capacity limits	To restrict risk-taking on certain non-core business lines. e.g. country level Catastrophe Excess of Loss and Stop Loss limits.	
Concentration limits	centration limits  To control concentration and volatility risk.  e.g. per life retention limits for individual business; accumulation limits o buildings in densely populated areas.	

• In addition, authority limits control risk origination by specifying the oversight required. All large, complex, or unusual transactions are reviewed and require approval from Risk Management.



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### Modeling of Influenza Pandemic Risk



#### Pandemic Risk

A major risk which can very quickly impact a mortality portfolio is an increase in mortality claims due to the influence of a pandemic. Significant pandemics are rare events of which the "1918 Spanish Flu" is the most well known. In today's world the particular features of such an event would be:

- fast human to human transmission spreading rapidly around the world due to air travel
- uncertainty with science's ability to quickly develop a cure / vaccine
- potential to impact infants, young or old adults or potentially all ages
- potential knock-on consequences such as fear of contact with other people, economic slowdown, asset falls etc.



#### Development of a Pandemic Model - Background

- 1918 was a unique event in 420 years with high mortality
  - 5.2 excess deaths per 1000 in the USA (vs. 0.4 in 1957 & 0.17 in 1968 when smaller pandemics occurred)
  - unusually, young adults were impacted most heavily
  - high incidence of viral pneumonia / cytokine storm
- 1957 / 1968 pandemics return to typical mortality intensity and pattern: most excess deaths confined to infants and elderly; use of antibiotics; better knowledge; behavioural changes etc
- 2009 pandemic infection risk highly skewed to young; low infectivity; low lethality; still uncertainty about serological attack rate by age
- Model
  - attempts to identify baseline variables and understand their importance and interaction (e.g. lethality, spread characteristics, age profile, proportion bacterial /viral, age-specific susceptibility)
  - incorporates most of these factors into an SIR model (susceptible, infected, recovered individuals, with defined rates of flow between groups)
  - uses 'event based' modelling which randomly selects certain key factors from a range of defined possibilities based on history and current conditions, e.g. basic reproduction number (R<sub>0</sub>), lethality, antiviral success



#### Factors that complicate comparison of pandemics

	Under- standing of viruses	Social distancing (available, even if not used)	Anti- biotics	Pandemic Vaccines	Anti- virals	R <sub>0</sub> value (spread capability)	Lethality (death per infection)
1918	X	√	X	X	X	2.1	1.1%
1957	-√	√	√	√ but too late	X	1.6	0.275%
1968	√	√	√	√ but too late	X	1.89	0.054%
2009	-√	√	√	small effect	√	1.45??	unknown – very low
today	-√	√	√	√ often too late	<b>√</b>	??	??

- Inferring current risk based on past total mortality is inappropriate
- Model must reliably imitate spread dynamics & changes that have occurred



#### Spread model – calculation process

- Model begins with infected people in any one of the 37 territorial entities
- Population cells: 5yr age groups, 37 countries, disease state (susceptible, infected by duration, recovered, dead, vaccinated)
- Model is based on daily iterations: cells change incrementally on a daily basis (i.e. gradually changing new infections, deaths)
- Specified within each population cell:
  - mixing between ages, mixing rates, mortality per infection, viral /bacterial disease progression, travel propensity, share of meds
- At each time, number of infections in each cell dependent on previous day's:
  - susceptibles; no. contacts (of both uninfected and infected); number infected in groups with which contact occurs; transmission prob. (affected by stage, antivirals, vaccines)
- Intervention affects one/some of the above, slowing infection and/or reducing mortality



#### Calculating spread

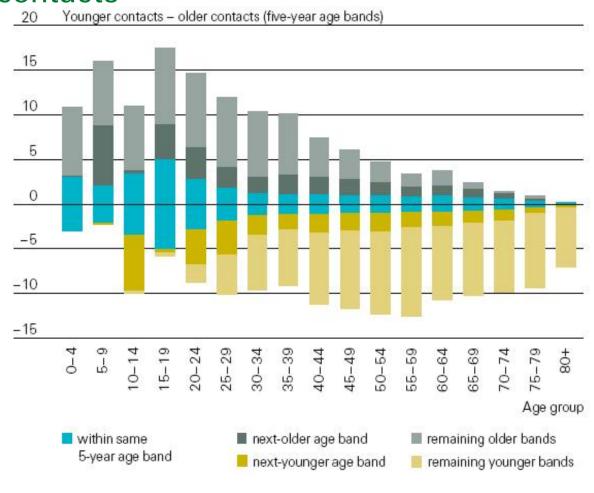
(process day by day)

 Using Susceptible/Infected/Recovered (SIR) cells the model calculates spread at discrete time intervals, each lasting 24 hours

Model run begins with a number of people infected in any chosen Day 1 country, and in any chosen age group Number infected by Day 2 is a mathematical function based on Day 2 underlying contagiousness of those already infected on Day 1 and the number of contacts they have with susceptible people As the days go by, more people become infected, fewer remain Day 3 susceptible, and the proportion of contacts with people who have recovered becomes increasingly large etc... Eventually the pool of susceptible people is sufficiently depleted (and the etc... pool of recovered people sufficiently large) that the number of new infections starts to decrease etc... etc... Pandemic has peaked; number of new infections drops rapidly

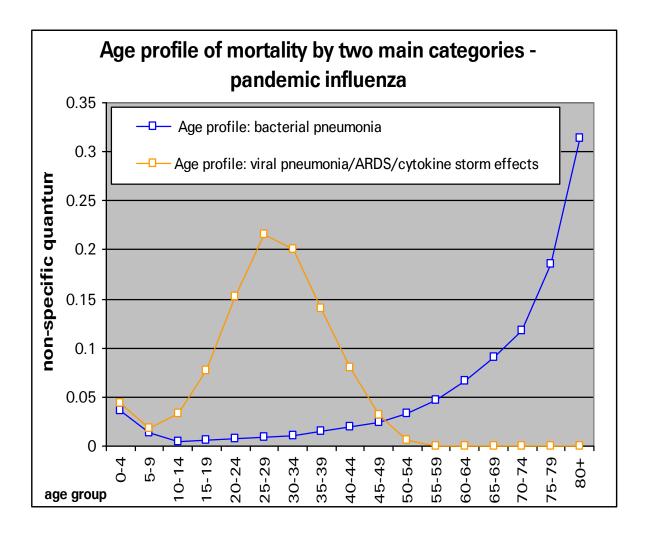
#### Demographic characteristics

## Number of contacts by age group and age profile of those contacts



- Young adults have more than 3x as many contacts with other people as the elderly
- Ageing populations tend to have lower spread values, especially as mixing rates peak among young adults

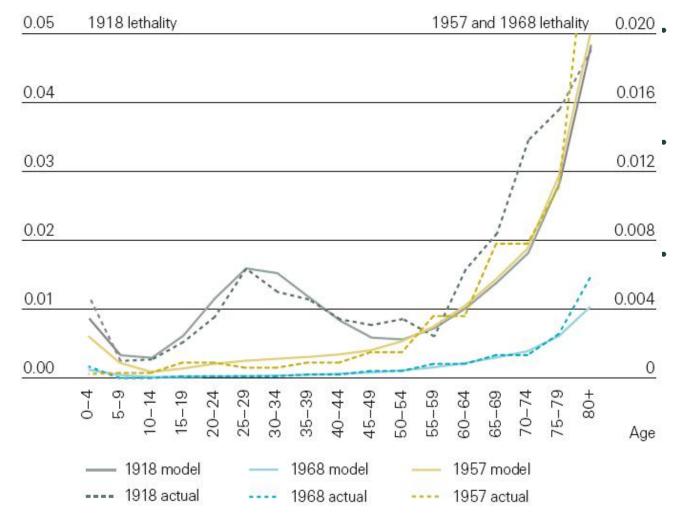
#### Age profile of lethality – bacterial and viral



	% of mortality due to bacterial pneumo nia	% of mortality other causes: viral pneumonia/ ARDS / cytokine storms
1918	57%	43%
1957	95%	5%
1968	100%	0%

#### One key parameter, for example:

#### Age profile of lethality



Low lethality pandemics (1957 and 1968) are assumed to be entirely due to bacterial pneumonia (Ushaped)

Mortality in high lethality pandemics (1918) is a combination of the two causes (W-shaped)

Impact of each of the two causes changes proportionally as lethality increases

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#### Other key parameters

- Age profile of lethality
- Contagiousness (by age & duration since infection)
- Behavioural factors (by age & clinical status)
- Susceptibility to infection (by age)
- Population age structure (for each geographical entity)
- Daily contacts (by age)



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#### Fit to historic events

- Frequency of pandemics: approximately 1 / 30 years on average
- 1918 using  $R_0$  = 2.1, and death-per-infection of 0.011 for USA
  - Fit numbers of deaths, lethalities, date of peak, etc
- 1957/ 1968 much less data, so tested lethality curves mainly, and used published data on R<sub>0</sub> values
- Algorithm developed to produce age profile of lethality:
  - model automatically produces age profile depending on lethality level
- In generating event set use 2009 demographic equivalent:
  - $-R_0: 1918=2.1; 1957=1.6; 1968=1.89; (2009=1.45?)$
  - Baseline lethality: 1889=0.00375; 1918=0.011; 1957=0.002753; 1968=0.00054



#### Moving 1918 to today

#### Demographics:

- Population age structure: rate of spread changes as populations age older people mix less, children have higher viral shedding
- Population density/ living conditions: tested effect of this on R<sub>0</sub>
- Underlying health status
  - higher life expectancy = better underlying health status, especially in developing world
- Antibiotics:
  - approximately 57% of deaths in 1918 assumed mainly due to bacterial pneumonia
  - bacterial pneumonia deaths predominantly in elderly and very young
  - antibiotics reduce bacterial pneumonia deaths by 60-80%
  - access varies by country



#### Moving 1918 to today

#### Antivirals:

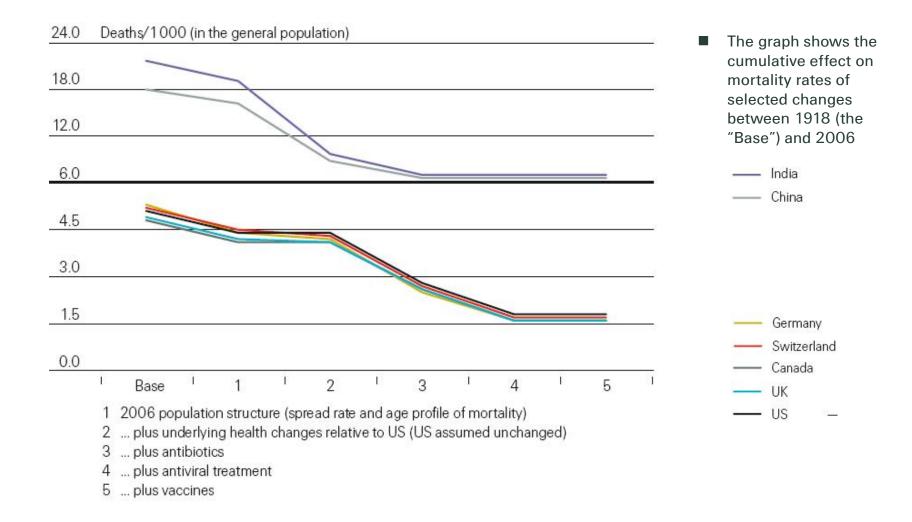
- makes infected people less infectious to others: slower spread, reduced peak, lower serological attack rate
- lower lethality: when effective and available assume 38% reduction in viral pneumonia mortality & 67% reduction in bacterial pneumonia mortality
- potential usage varies by country
  - maximum access is 65%, accounting for need for rapid administration
  - further constrained by country stockpile and supply rates
- antivirals assumed to be effective in 3/4 of pandemics

#### ■ Travel:

- has been tested and has minimal impact
- travel restrictions delay entry of the virus into countries, but given the growth patterns when it does eventually reach a country (which it almost always will), the impact on final mortality outcomes is small



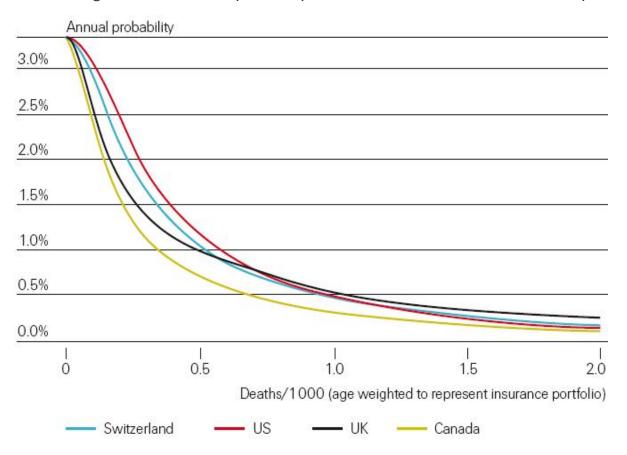
#### Simulated 1918 pandemic, then & now...





#### Modelling results: Selected developed countries

Insured-age excess mortality due to pandemic influenza, selected developed countries

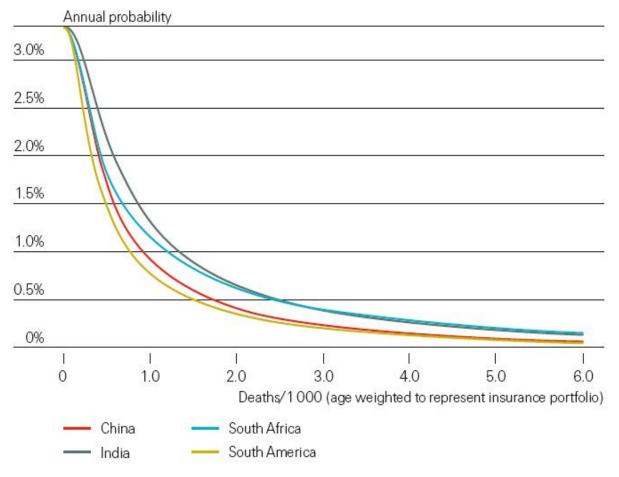


Canada is among the countries appearing to be least impacted, with estimated 1-in-200-year excess mortality at around 0.7‰ in an insurance-age population



#### Modelling results: Selected developing countries

Insured-age excess mortality due to pandemic influenza, selected developing countries

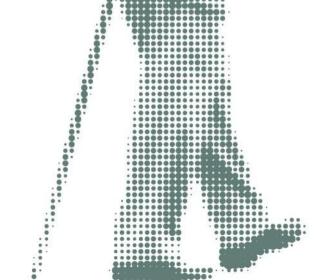


Countries expected to experience higher levels of mortality include India, Pakistan and Indonesia, due to high population density, along with a weak capacity to reduce contact rates (India shown)

The healthcare systems of these countries are also weaker than in developed countries, almost no antivirals are available



# Mortality Trend Model



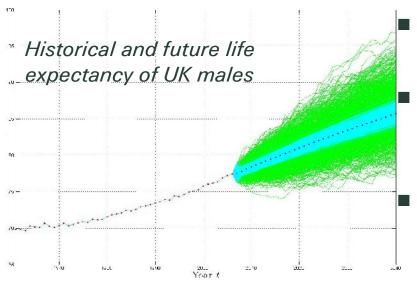
150 YEARS

# Mortality trend model is required for both internal use as well as to meet regulatory requirements such as Solvency II and Swiss Solvency Test

• All these measures define the capital required to meet the expected loss event as ...  $(AC_{T+1} - AC_T | F_T)$ , where

AC  $_T = A_T - L_T$  denotes Available Capital (Assets minus Liabilities) today at time T,  $F_T$  denotes a filtration representing today's information,

" denotes a risk measure (VaR at 99.5%, expected shortfall at 99%), and AC  $_{T+1} = A_{T+1} - L_{T+1}$  denotes Available Capital one year into the future



Liabilities are affected by best estimate assumptions on future mortality rates  $\hat{q}_{x,t}$ 

Mortality Trend Risk in the one-year view of the above definition is the potential deviation of next year's best estimates from today's best estimates

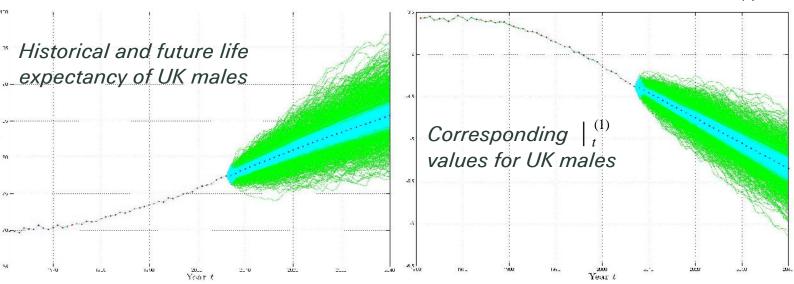
Focus on systematic Mortality Trend Risk of given populations, e.g. UK males, US females, etc.

#### Ultimate view

• Usually, mortality models simulate an *ultimate view* of future mortality rates, e.g.

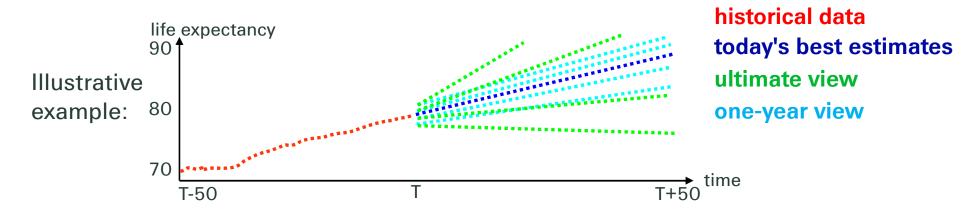
Lee-Carter model Swiss Re's model

Lee-Carter model 
$$\ln(m_{x,t}) = \Gamma_x + \kappa_t^{(1)} S_x$$
 Age-Period-Cohort model 
$$\ln(m_{x,t}) = \Gamma_x + \kappa_t^{(1)} + \lambda_{t-x}$$
 Cairns-Blake-Dowd model 
$$\log it(q_{x,t}) = \Gamma_x + \kappa_t^{(1)} + \binom{2}{t}(\overline{x} - x) + \lambda_{t-x}$$
 Plat model 
$$\ln(m_{x,t}) = \Gamma_x + \kappa_t^{(1)} + \binom{2}{t}(\overline{x} - x) + \binom{3}{t}(\overline{x} - x)^+ + \lambda_{t-x}$$
 Swiss Re's model 
$$\log it(q_{x,t}) = \Gamma_x + \kappa_t^{(1)} + \binom{2}{t}(x - x_{center}) + \binom{3}{t}(x_{young} - x)^+ + \binom{4}{t}(x - x_{old})^+ + \lambda_{t-x}$$



#### One-year view and ultimate view

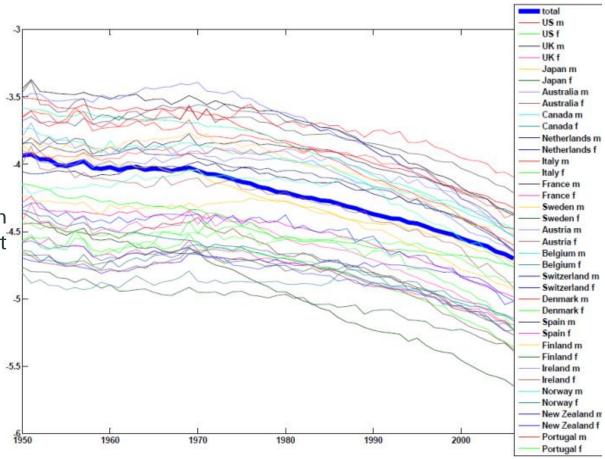
One-year view can be constructed by (semi-)nested simulations as follows



- Run model to determine today's best estimates based on historical mortality rates and to simulate future mortality rates, say, 10000 realizations
- Use each of the 10000 realizations of next year's mortality rates as additional historical data and re-run the model to get 10000 best-estimate mortality rates
- One-year view imposes new challenges, e.g. fast algorithms to estimate and re-estimate parameters, consistency between one-year and ultimate view required

#### Multi-population model

- Two step approach
  - Model mortality trend of the total population (blue line)
  - 2. For each population the difference to the mortality trend of the total population is modeled (main ingredient is to model differences of | (1) as AR(1) process)

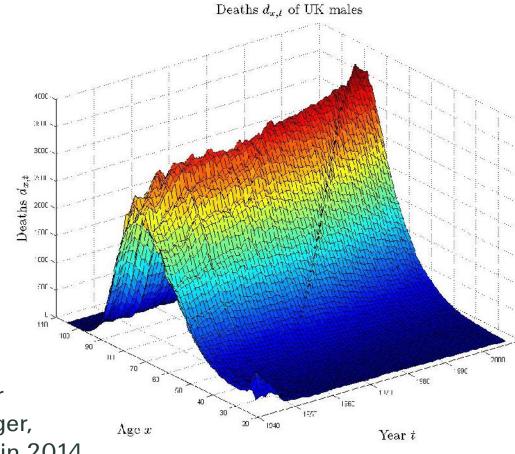




#### Summary

- Focus on one-year view
- Consistency between one-year and ultimate view
- Calibration of model's ultimate view to historical data
- Fast algorithms allow (semi-)nested simulations
- Multi-population model
- Details can be found in the article "Modeling the Mortality Trend under Modern Solvency Regimes", M. Börger,

D. Fleischer, N. Kuksin, ASTIN Bulletin 2014

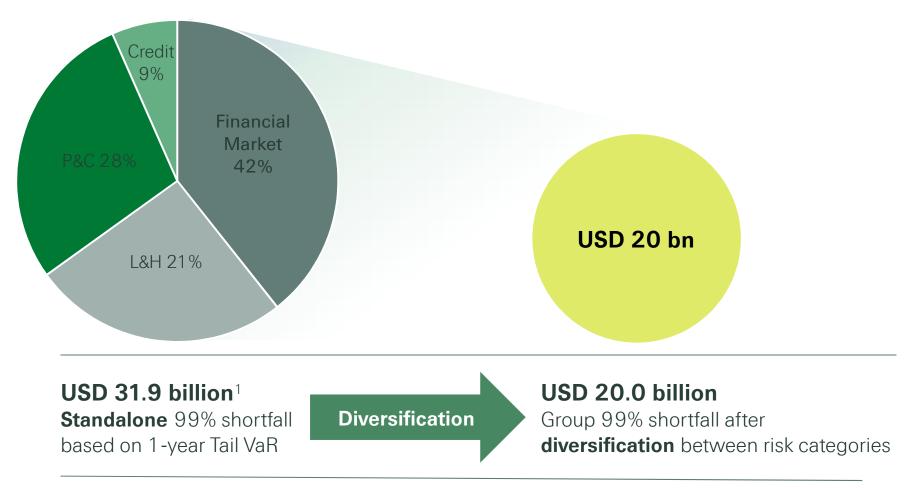




# Managing mortality exposure



#### Risks are diversified across the Swiss Re Group

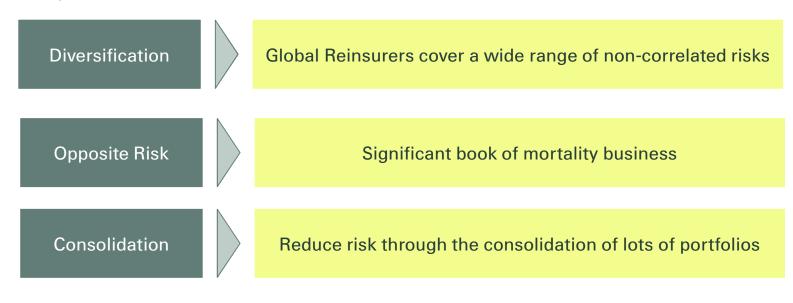


<sup>&</sup>lt;sup>1</sup> Simple sum, as of 31.12.2013, prior to diversification; both calculations are based on internal model figures, as disclosed in 2013 annual report



#### Longevity Risk

Longevity risk provides a hedge against Swiss Re's mortality business. Swiss Re also holds other risks which are not correlated with longevity risk. It is unlikely that several unrelated extreme events will occur at the same time. This reduces the capital that is required for a well diversified reinsurer.



Large diversified global reinsurers have significant expertise and a strong rationale to write and hold longevity risks



#### External retrocession and ILS

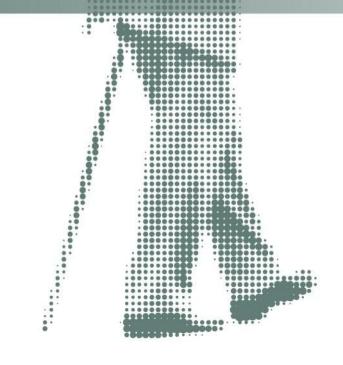
- Swiss Re uses retrocession to other reinsurers to managed the risk exposure in the life and health book. (More generally insurance companies are reinsuring their own exposures to manage their mortality risks).
- In addition, Insurance-linked securities (ILS) are a key part of Swiss Re's overall strategy to reduce "peak" risk exposures.
- Extreme mortality events such as pandemic influenza and terrorism events in the U.S, Canada or Australia are some of Swiss Re's "peak" perils. Hedging these risks in the capital markets is an important tool for Swiss Re in managing its risk capital efficiently.
- We have developed capabilities in risk assessment, structuring, transformation, and distribution that enable us to transfer life insurance risks to the capital markets to complement our traditional retrocession instruments.
- Since pioneering the securitization of extreme mortality risk in 2003, Swiss Re has regularly issued ILS, such as VITA bonds, to transfer risk to the capital markets.





# Causal Mortality Model





#### Visions of the future

**GENES** 



**HEALTHCARE** 



INDIVIDUAL BEHAVIOUR



CONNECTIVITY

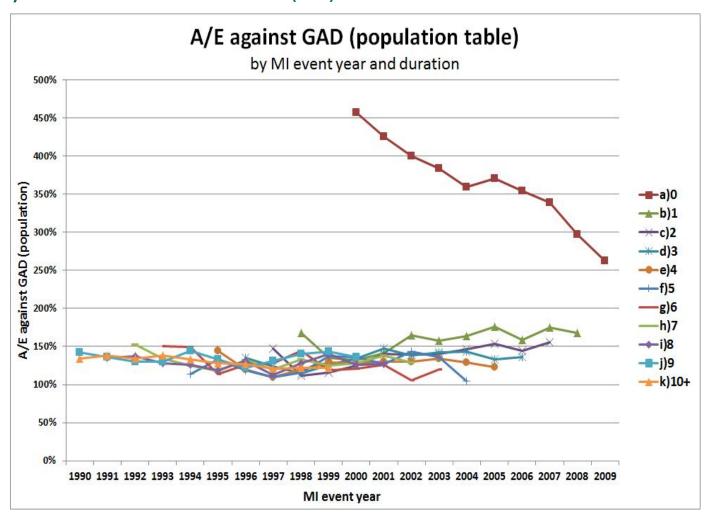


### Developing predictive models of longevity Integrated Risk factors and Impairment Scenarios

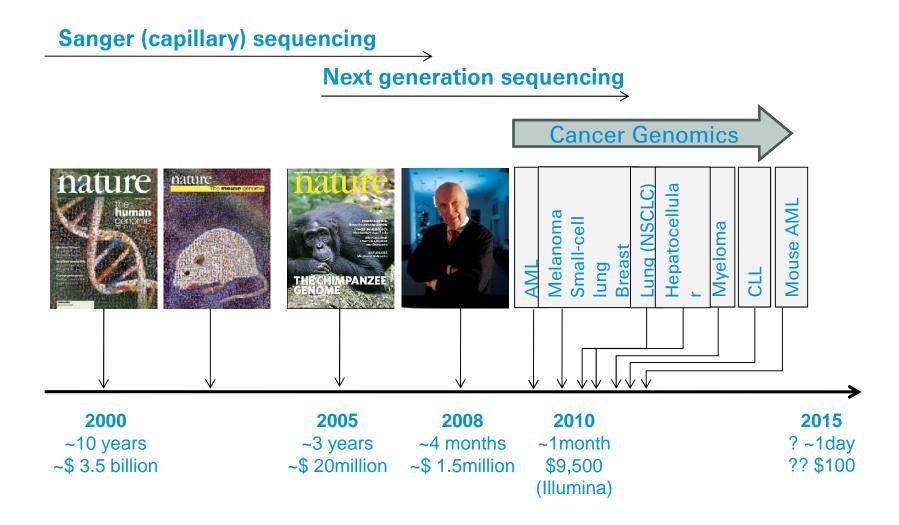
- Complementary approach to
  - stochastic mortality models
  - blending between current mortality improvements and long-term assumptions over defined horizons
- Bringing together:
  - Swiss Re experience (data and expert knowledge)
  - Large patient medical databases in different countries
  - External networks capturing expert opinion
- Causal-based mortality predictions, evaluating factors such as:
  - Promotion and adoption of healthy lifestyle choices
  - Advances in screening and diagnostic technology
  - Pharmaceutical pipeline and its likely impact



#### Deep analysis of mortality experience Patient medical data (GPRD) mortality split by duration and calendar year since heart attack (MI)



#### Advances in DNA sequencing

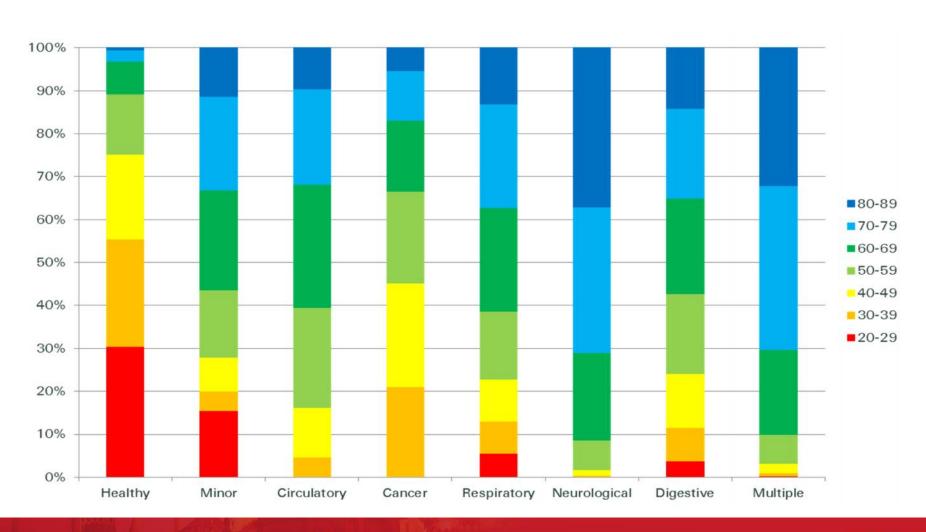


#### IRIS – multi-state model of mortality Global and disease-specific factors to consider

Individual risk factors Patient interaction Research & development Healthcare funding General drivers to Public vs private funding Age, gender, diet, smoking - Health awareness Public vs commercial smoking considerations: Disease-based patient Trust and confidence in sponsors diagnosis and Taxes and restrictions Regulators' attitude to advocacy groups' advice given survival Current treatments influence Use of clinical guidelines developments Allocation of resources (buproprion) to improve quality of care • **Disease-**focused approach Future treatments (vaccines) towards cure vs vs global impact of ageing prevention Circulatory Respiratory Stroke, angina, Chronic obstructive Multiple heart attack pulmonary disease diseases Disease types and **Healthy** Death disease progression Cancer Neurological Lung, colorectal Dementia, prostate breast Alzheimer's, **Current approaches Risk factors** Parkinson's Targeting DCIS Family history Surgery with node follow-up Obesity Adjuvant radiotherapy • Having children later in life Herceptin, Tamoxifen **Breast cancer**  Not breast feeding Clinical trials pipeline **Early detection** Phase II (230 trials\*) Digital mammography Factors involved in Medical innovations • Phase III (56) • MRI for high-risk Growth factor inhibition • eg pertuzumab (limits cancer assessing specific • Gail algorithm (own factors) Future of personalised growth) Klaus algorithm (family history) example disease medicine (eg tumour profiling)



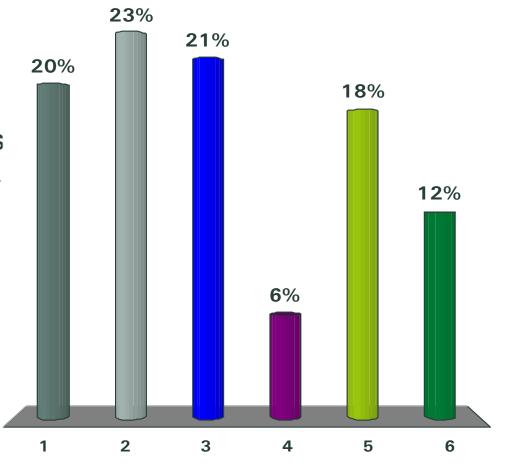
# Output from scenario testing in IRIS Distribution of health/disease across age groups in 20 years





### Expert opinion from Swiss Re conferences Which will have the highest impact on future longevity?

- 1. Stem cell therapy
- 2. Genetic testing
- 3. Vaccines
- 4. Monoclonal antibodies
- 5. Monitoring technology
- 6. Nanomedicine



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