

# Natural Hedging between Longevity Risk and Mortality Risk

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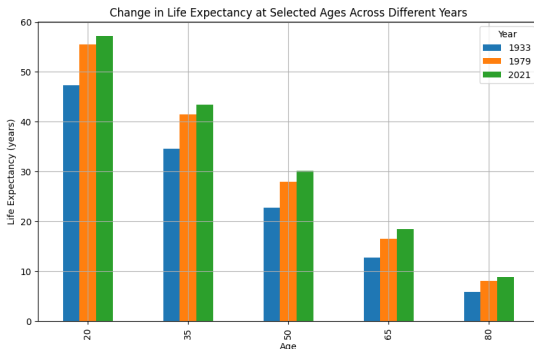
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# Motivation

## Global Life Expectancy at Birth

46.5 yr in 1950  $\Rightarrow$  71.7 yr in 2022

Source: United Nation



# Pension & Life Insurance

## Pension

- pays  $x$  at the beginning of each year
- participants accrued rights based on their age

## Life Insurance

- receive  $F$  at the end of death year

# Natural Hedging

Similar to financial risk, we can find derivatives such as mortality swaps to hedge it. However, due to its **over-counter, complex, modelling difficulty** which makes it **expensive**.

Hence, Cox and Lin propose the idea of natural hedging in 2007. Their empirical findings suggest that companies engaged in selling both life insurance and annuity policies typically offer annuities at **lower prices** compared to firms that operate solely in one line of business by markedly reduce the sensitivity of an insurance portfolio to mortality risk.

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# Why do we need stochastic mortality models?

## Stochastic (LC) vs Static

- mortality at  $t = T$  is only known by time  $T + 1$
  - Robust risk management
  - affect contracts with embedded options
- 
- Data Source: US Life table from [mortality.org](http://mortality.org)
  - Base model: Lee Carter model (LC model)

# Lee Carter Model

## Model Setting

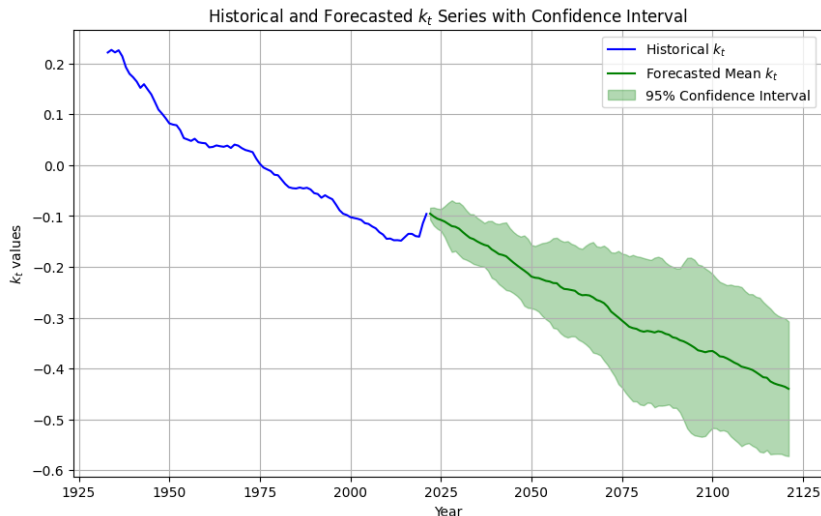
$$\ln m_{x,t} = \alpha_x + \beta_x K_t + \varepsilon_{x,t} \quad (1)$$

$$K_{t+1} = K_t + c + \sigma_k Z_t \quad (2)$$

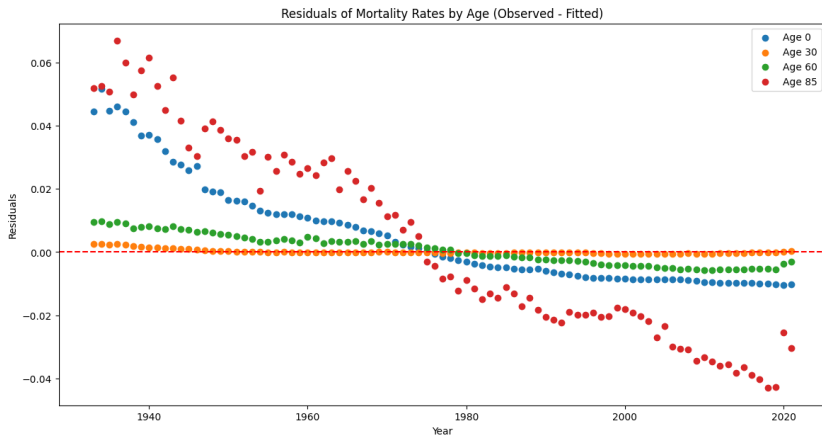
- $m_{t,x}$ : central death rate in year  $t$  at age  $x$
- $\alpha_x$ : mean of the central mortality rates after logarithm transformation across age  $x$ .
- $\beta_x$ : age-specific effect on mortality improvement
- $K_t$ : stochastic mortality improvement
- $\varepsilon_{x,t} \stackrel{\text{iid}}{\sim} N(0, \sigma_x)^2$



# Modeling of $K_t$



# Evaluate the Fit



# Kannisto Method

Motivation: The data in high ages is sparse and highly incredible.

## Kannisto Method

$$\text{logit}(q_x) = \ln \frac{q_x}{1 - q_x} = \ln a + bx \quad (3)$$

We use  $\{m_x : x \in [75, 85]\}$  to fit and extend the lifetable up to age 100.

$$m_x = -\ln\left(-c + \frac{1}{1 + ae^{bx}}\right) \quad (4)$$

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# Assumptions

- 1 Mortality data is based on the US Census Bureau HMD mortality data from 1933 to 2021.
- 2 The face value of the whole life insurance policy issued at age 45 is \$10 payable at the end of the year of death.
- 3 The pension fund provides an annual benefit of 1 starting from age 65. The present value of this liability is similar to the life one. The accrual of pension right increases linearly to a normalized value of 1 at age 65. For instance, for age 20,  $\delta_x = \min(\frac{x-19}{46}, 1)$
- 4 Constant interest rate at  $r = 4\%$ . Denote  $v = (1 + r)^{-1}$  as the discounting factor.
- 5 No one survive above age 100.

# Age Distribution

Due to limited access in data, we assume the age composition portfolio of insurer and pension fund is same as the demographic portfolio.



# Present Value of Life Insurance and Pension

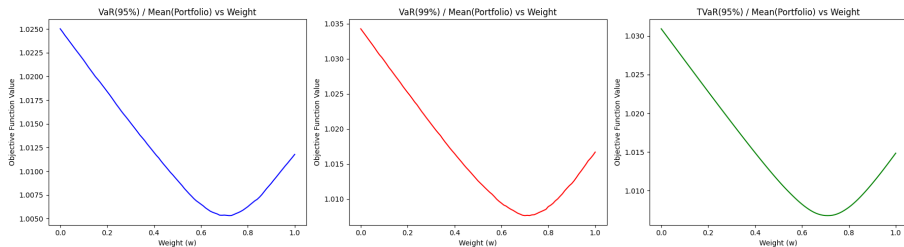
Let  $L$  be the present value of the loss.

	PV Life	PV Pension
$\mathbb{E}[L]$	3.7544	3.8167
$\sigma_L$	0.0263	0.0582
Minimum	3.6507	3.5972
$Q_{25\%}$	3.7364	3.7775
Median	3.7539	3.8170
$Q_{75\%}$	3.7719	3.8566
Maximum	3.8594	4.0155
$\rho$	-0.78	

**Table:** Key Summary Statistics of PV Life and PV Pension

# Value at Risk

Motivated by the risk capital regime, we consider  $VaR_{95\%}(L)/\mathbb{E}[L]$ ,  $VaR_{99\%}(L)/\mathbb{E}[L]$ ,  $TVaR_{95\%}(L)/\mathbb{E}[L]$ .

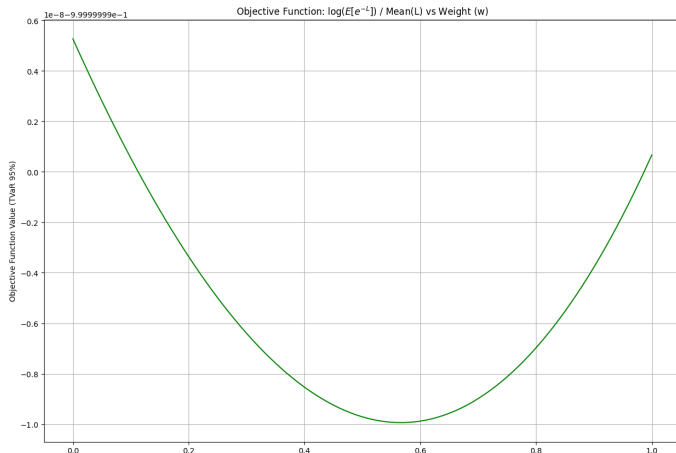


The optimal weights are (0.72, 0.70, 0.71).



# Entropic Utility Function

As insurers are risk averse, we consider the entropic utility function  $\log(\mathbb{E}[e^{-L}])$  to reflect insurers' risk appetite. The optimal weight is 0.71.



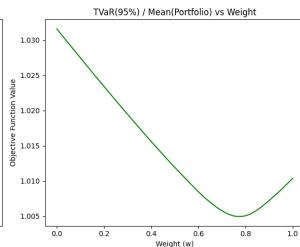
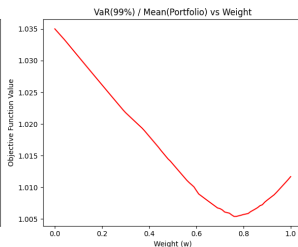
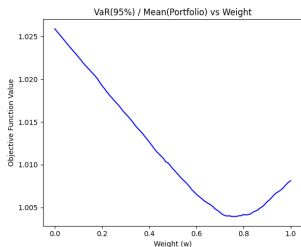
# Summary

Objective Function ( $J$ )	$w^*$
$VaR_{95\%}(L)$	0.72
$VaR_{99\%}(L)$	0.70
$TVaR_{95\%}(L)$	0.71
$\log(\mathbb{E}[e^{-L}])$	0.71

**Table:** Optimal Ratio under Different Objective Function  $J$

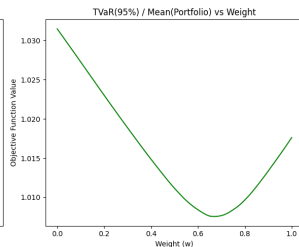
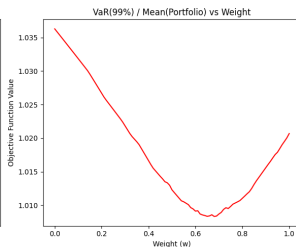
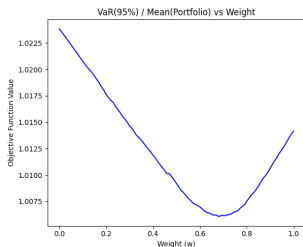
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# Interest Rate Down ( $4\% \rightarrow 3\%$ )



The optimal weights are (0.76, 0.77, 0.76).

# Interest Rate Up (4% $\rightarrow$ 5%)



The optimal weights are (0.68, 0.68, 0.67).

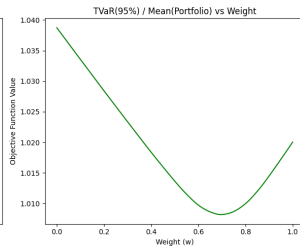
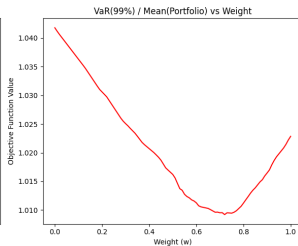
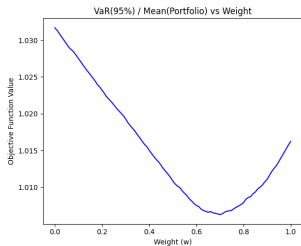
# Optimal Weight under Interest Rate Shock

Objective Function ( $J$ )	$w^*(3\%)$	$w^*(4\%)$	$w^*(5\%)$
$VaR_{95\%}(L)$	0.76	0.72	0.68
$VaR_{99\%}(L)$	0.77	0.70	0.68
$TVaR_{95\%}(L)$	0.76	0.71	0.67
$\log(e^{-L})$	0.77	0.71	0.68

**Table:** Optimal Ratio under Different Objective Function  $J$

# Change of Population Basis

To show that it's effective in different population group. Same analysis is done to population in the Netherlands.



The optimal weights are (0.70, 0.72, 0.70).

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# Conclusion

## Recommendation

- Recommend insurer to start a new line of business in annuity such as Qualifying Deferred Annuity Policy (QDAP)
- Recommend pension fund to absorb life insurance exposure
- Recommend to keep the ratio of life insurance to annuity at 3:1 or keep the weight of life insurance within 0.7 to 0.8

# Conclusion

## Further Research

- Lee Carter model is often criticized by under estimating the variation due to its normality assumption on error. More sophisticated mortality model such as CBD model can be considered.
- Parameter uncertainty can be embedded in the model projection as well.