



Mean Variance Optimization for Participating Life Insurance Contracts

Motivation

▶ Investment problem: “ $\max_{\text{strategies}} \text{Mean}(Y) - \text{Variance}(Y)$ ”

→ find optimal strategy

▶ Y : payoff of insurer

▶ Aim: find optimal terminal wealth and optimal investment strategy

▶ on top: show existence of all parameter

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Motivation

Participating Life Insurance Contracts:

- crucial role in the Life Sector

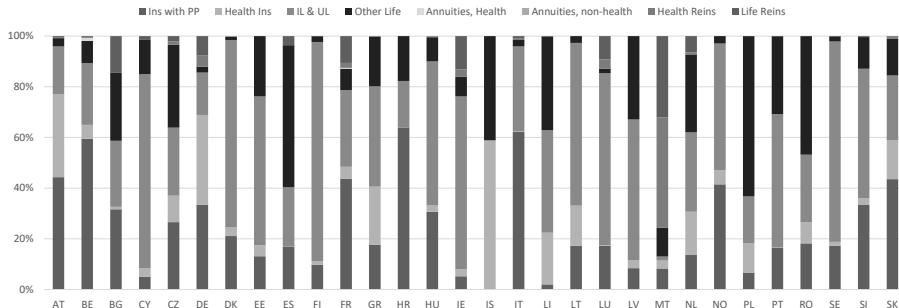


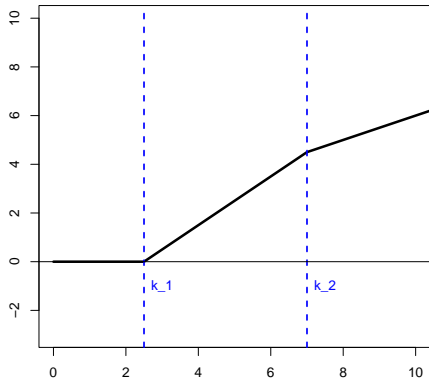
Figure: Market share in 2022 of the gross premium separated by the line of business in the life sector. Data source: European Insurance Overview from the EIOPA (2023)

Motivation

Participating Life Insurance Contracts:

- ▶ 2 main products: without or with guarantee for the policyholders

Non-protected participating life insurance



Protected participating life insurance

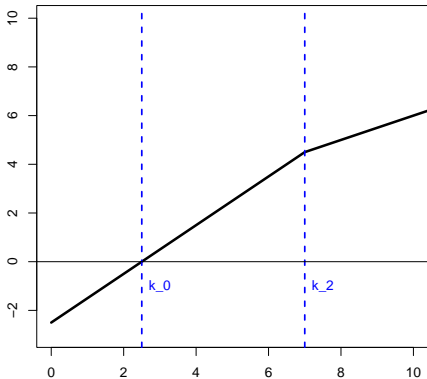


Figure: Payoffs of the insurer.

Motivation

▶ Main difficulty: non-linearity in $\text{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$

→ give an equivalent problem as in Zhou & Li (2000)

▶ Lagrangian optimization, e.g., Basak & Shapiro (2001)

▶ optimize Participating Life Insurance Contracts:

▶ Problem: non-convexity resp. non-concavity of the payoff function

▶ Lin *et al.* (2017), Nguyen & Stajda (2020) for S-shaped utility functions

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Structure

Model Setup

Optimization in the Black-Scholes market

Numerical Results

Optimization in an Incomplete Market

Structure

Model Setup

Optimization in the Black-Scholes market

Numerical Results

Optimization in an Incomplete Market

d -dimensional Black-Scholes market

- ▶ $T > 0$ finite time horizon
- ▶ 1 risk-free asset: $dB_t = B_t r_t dt$
- ▶ d risky assets: $dS_t^i = S_t^i \mu_t^i dt + S_t^i \sigma_t^i dW_t$
- ▶ strategies: u^i denotes the fraction of wealth invested in risky asset i , progressively measurable and square-integrable
- ▶ wealth: $dX_t = X_t [r_t + u_t^T (\mu_t - r)] dt + X_t u_t^T \sigma_t dW_t$ with $X_0 = x_0$
- ▶ price density: $d\xi_t = -\xi_t r_t dt - \xi_t \kappa_t^T dW_t$ with $\xi_0 = 1$ and the Sharpe ratio process $\kappa_t = (\sigma_t)^{-1} (\mu_t - r_t)$
- ▶ Interpretation: $\xi_T(\omega)$ Arrow-Debreu value per probability unit in state ω at time T
- ▶ Assume: μ and σ deterministic, σ bounded, bounded away from zero, invertible
- ▶ Assume: r, μ integrable, σ, κ square integrable over $[0, T]$

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Optimization Functional

- ▶ main target: $J(0, T, \hat{u}, x_0) = \sup_{u \in \mathcal{U}} J(0, T, u, x_0)$
- ▶ functional J : $J(0, T, u, x_0) := \mathbb{E}[F(0, T, u, x_0)] - \gamma \text{Var}(F(0, T, u, x_0))$ with risk aversion parameter $\gamma > 0$
- ▶ Function F :

$$\begin{aligned}
 F(s, t, u, x) &:= \alpha ((X_t - k_1)_+ - k_0) - \alpha_2 (X_t - k_2)_+ \\
 &= \begin{cases} -\alpha k_0 & X_t < k_1 \\ \alpha (X_t - k_1 - k_0) & k_1 \leq X_t < k_2 \\ \tilde{\alpha} (X_t - k_2) + \alpha (k_2 - k_1 - k_0) & X_t \geq k_2 \end{cases}
 \end{aligned}$$

where $X_s = x$, $0 \leq k_0, k_1 \leq k_2 < \infty$ with $k_0 + k_1 \leq k_2$, $0 \leq \alpha_2 < \alpha < \infty$ with $\tilde{\alpha} := \alpha - \alpha_2$ (Note: $0 < \tilde{\alpha} \leq \alpha$)

- ▶ Notation: $F(X_T)$ instead of $F(0, T, u, x)$ if u is clear

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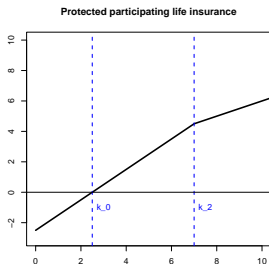
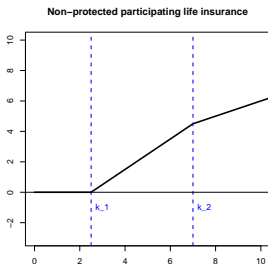
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► non-protected contract: $\alpha = 1$, $k_0 = 0$, k_1 is the guarantee

► protected contract: $\alpha = 1$, k_0 is the guarantee, $k_1 = 0$



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Equivalent Problem

- ▶ Problem: non-linearity in $\text{Var}(X) = \mathbb{E}[X^2] - (\mathbb{E}[X])^2$
- ▶ Solution:
 - ▶ value functional \tilde{J} : $\tilde{J}(0, T, u, x_0) := \mathbb{E}[\lambda F(0, T, u, x_0) - \gamma F(0, T, u, x_0)^2]$
with $\lambda = 1 + 2\gamma \mathbb{E}[F(0, T, \hat{u}, x_0)]$ where \hat{u} is the optimal strategy
 - ▶ solve with λ as a parameter

Lemma

If \hat{u} is an optimal strategy for J , it is also an optimal strategy for \tilde{J} .

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If \hat{u} is an optimal strategy for J , it is also an optimal strategy for \tilde{J} .

Optimal Terminal Wealth

Theorem

The optimal terminal wealth \hat{X}_T is given by:

$$\hat{X}_T := \begin{cases} k_2 + \frac{\lambda\tilde{\alpha} - y\xi_T}{2\gamma\tilde{\alpha}^2} - \frac{\alpha}{\tilde{\alpha}}(k_2 - k_1 - k_0) & \xi_T \in (0, \xi_1^*] \\ k_2 & \xi_T \in (\tilde{\alpha}\hat{\xi}, \xi_2^*] \\ k_0 + k_1 + \frac{\lambda\alpha - y\xi_T}{2\gamma\alpha^2} & \xi_T \in (\alpha\hat{\xi}, \xi_3^*] \\ 0 & \text{else} \end{cases},$$

where y is the Lagrangian multiplier which solves $\mathbb{E}[\xi_T \hat{X}_T(y)] = \xi_0 x_0$.

- ▶ \hat{X}_T is the optimal wealth of the portfolio, i.e., before distributing
- ▶ insurer: $F(\hat{X}_T) = \alpha((\hat{X}_T - k_1)_+ - k_0) - \alpha_2(\hat{X}_T - k_2)_+$

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$$\hat{\xi} := \max \left\{ 0, \frac{\lambda - 2\gamma\alpha(k_2 - k_1 - k_0)}{y} \right\},$$

$$\bar{\xi} := \frac{\lambda\alpha}{y} + \frac{2\gamma\alpha^2 k_0}{y},$$

$$\tilde{\xi}_1^* := \tilde{\alpha}\hat{\xi} - \frac{2\gamma\tilde{\alpha}}{y} \left(\sqrt{\max \left\{ 0, (\alpha(k_0 + k_1) - \alpha_2 k_2)^2 - \alpha^2 k_0^2 + \frac{\lambda}{\gamma}(\alpha k_1 - \alpha_2 k_2) \right\}} - \tilde{\alpha}k_2 \right),$$

$$\xi_1^* := \max \left\{ 0, \min \left\{ \tilde{\alpha}\hat{\xi}, \tilde{\xi}_1^* \right\} \right\},$$

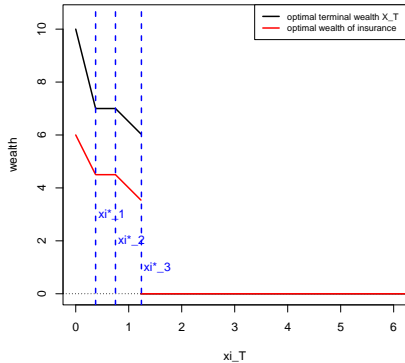
$$\tilde{\xi}_2^* := \frac{\alpha\lambda}{y} - \frac{\gamma\alpha^2(k_2 - k_1)^2 - 2\gamma\alpha^2 k_0(k_2 - k_1) + \lambda\alpha k_1}{yk_2},$$

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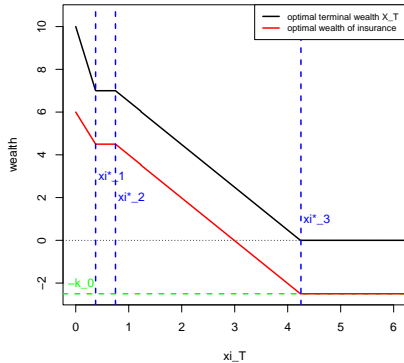
$$\xi_3^* := \max \left\{ \alpha\hat{\xi}, \bar{\xi} - \frac{2\gamma\alpha^2}{y} \left(\sqrt{k_1^2 + k_1 \left(2k_0 + \frac{\lambda}{\gamma\alpha} \right) - k_1} \right) \right\}.$$

Optimal Terminal Wealth

Non-protected participating life insurance



Protected participating life insurance



Optimal Terminal Wealth

Theorem

In particular, the Lagrange multiplier exists. Note that we suppress the dependence on λ and y for the sake of simplicity in notation unless we state it otherwise in some proofs. Moreover, let ξ^ be as follows:*

$$\xi^* := \begin{cases} \xi_3^* & , \text{if } \xi_3^* > \alpha \hat{\xi} \\ \xi_2^* & , \text{if } \xi_3^* = \alpha \hat{\xi}, \xi_2^* > \tilde{\alpha} \hat{\xi} . \\ \xi_1^* & , \text{if } \xi_3^* = \alpha \hat{\xi}, \xi_2^* = \tilde{\alpha} \hat{\xi} \end{cases}$$

Then, it holds that $\xi^ > 0$ and $\hat{X}_T > 0$ for $\xi \in (0, \xi^*)$ and $\hat{X}_T = 0$ for $\xi > \xi^*$.*

- ▶ $(0, \xi_1^*] \cup (\tilde{\alpha} \hat{\xi}, \xi_2^*] \cup (\alpha \hat{\xi}, \xi_3^*] = (0, \xi^*]$, i.e., these three intervals are connected
- ▶ \hat{X}_T as a function of ξ is continuous and non-increasing in $(0, \xi^*) \cup (\xi^*, \infty)$
- ▶ There exists always a solution for λ and y and an equation system which can be numerically solved to determine them.

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Optimal Terminal Wealth of the Insurer

Corollary

The optimal payoff of the insurer is given by:

$$F(\hat{X}_T) = \begin{cases} \frac{\lambda\tilde{\alpha} - y\xi_T}{2\gamma\tilde{\alpha}} & \xi_T \in (0, \xi_1^*] \\ \alpha(k_2 - k_1 - k_0) & \xi_T \in (\tilde{\alpha}\hat{\xi}, \xi_2^*] \\ \frac{\lambda\alpha - y\xi_T}{2\gamma\alpha} & \xi_T \in (\alpha\hat{\xi}, \xi_3^*] \\ -\alpha k_0 & \text{else} \end{cases},$$

where $\hat{\xi}$, ξ_1^* , ξ_2^* , and ξ_3^* are as before.

Optimal Strategy

Theorem

The optimal solution \hat{u} is given by:

$$\hat{u}_t = (\sigma_t^T)^{-1} \kappa_t \frac{v_t}{\hat{X}_t},$$

Optimal Strategy

Theorem

$$\begin{aligned}
 v_t = & \left(k_2 + \frac{\lambda}{2\gamma\tilde{\alpha}} - \frac{\alpha}{\tilde{\alpha}}(k_2 - k_1 - k_0) \right) \frac{e^{-\int_t^T r_s ds}}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} \varphi(d_1(\xi_1^*, t)) \\
 & + \frac{y}{2\gamma\tilde{\alpha}^2} \xi_t e^{\int_t^T -(2r_s - \|\kappa_s\|^2) ds} \left[\Phi(d_2(\xi_1^*, t)) - \frac{1}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} \varphi(d_2(\xi_1^*, t)) \right] \\
 & + k_2 \frac{e^{-\int_t^T r_s ds}}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} \left(\varphi(d_1(\xi_2^*, t)) - \varphi(d_1(\tilde{\alpha}\hat{\xi}, t)) \right) \\
 & + \left(k_0 + k_1 + \frac{\lambda}{2\gamma\alpha} \right) \frac{e^{-\int_t^T r_s ds}}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} \left(\varphi(d_1(\xi_3^*, t)) - \varphi(d_1(\alpha\hat{\xi}, t)) \right) \\
 & + \frac{y}{2\gamma\alpha^2} \xi_t e^{\int_t^T -(2r_s - \|\kappa_s\|^2) ds} \left[\left(\Phi(d_2(\xi_3^*, t)) - \Phi(d_2(\alpha\hat{\xi}, t)) \right) \right. \\
 & \quad \left. - \frac{1}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} \left(\varphi(d_2(\xi_3^*, t)) - \varphi(d_2(\alpha\hat{\xi}, t)) \right) \right]
 \end{aligned}$$

Optimal Strategy

Theorem

$$\begin{aligned}
 \hat{X}_t &= \left(k_2 + \frac{\lambda}{2\gamma\tilde{\alpha}} - \frac{\alpha}{\tilde{\alpha}}(k_2 - k_1 - k_0) \right) e^{-\int_t^T r_s ds} \Phi(d_1(\xi_1^*, t)) \\
 &\quad - \frac{y}{2\gamma\tilde{\alpha}^2} \xi_t e^{\int_t^T -(2r_s - \|\kappa_s\|^2) ds} \Phi(d_2(\xi_1^*, t)) \\
 &\quad + k_2 e^{-\int_t^T r_s ds} \left(\Phi(d_1(\xi_2^*, t)) - \Phi(d_1(\tilde{\alpha}\hat{\xi}, t)) \right) \\
 &\quad + \left(k_0 + k_1 + \frac{\lambda}{2\gamma\alpha} \right) e^{-\int_t^T r_s ds} \left(\Phi(d_1(\xi_3^*, t)) - \Phi(d_1(\alpha\hat{\xi}, t)) \right) \\
 &\quad - \frac{y}{2\gamma\alpha^2} \xi_t e^{\int_t^T -(2r_s - \|\kappa_s\|^2) ds} \left(\Phi(d_2(\xi_3^*, t)) - \Phi(d_2(\alpha\hat{\xi}, t)) \right), \\
 d_1(x, t) &= \frac{\ln x - \ln \xi_t + \int_t^T r_s - \frac{\|\kappa_s\|^2}{2} ds}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}}, \\
 d_2(x, t) &= \frac{\ln x - \ln \xi_t + \int_t^T r_s - \frac{3\|\kappa_s\|^2}{2} ds}{\sqrt{\int_t^T \|\kappa_s\|^2 ds}} = d_1(x, t) - \sqrt{\int_t^T \|\kappa_s\|^2 ds}
 \end{aligned}$$

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Parametrization

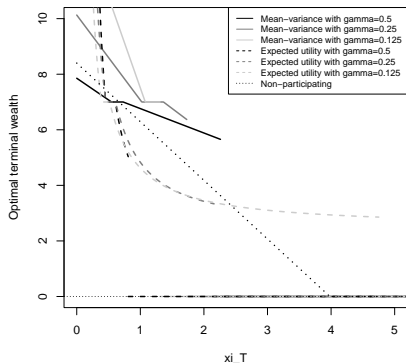
- ▶ $k_0 = 0$ / $k_0 = 2.5$
- ▶ $k_1 = 2.5$ / $k_1 = 0$
- ▶ $k_2 = 7$
- ▶ $\alpha = 1$
- ▶ $\alpha_2 = 0.25$
- ▶ $\gamma = 0.25$
- ▶ $x_0 = 4$
- ▶ $T = 10$
- ▶ $d = 1$
- ▶ $r = 0.02$
- ▶ $\mu = 0.08$
- ▶ $\sigma = 0.2$
- ▶ $\kappa = 0.3$
- ▶ $\delta = 0.01$
- ▶ $N = 1000$

For a comparison:

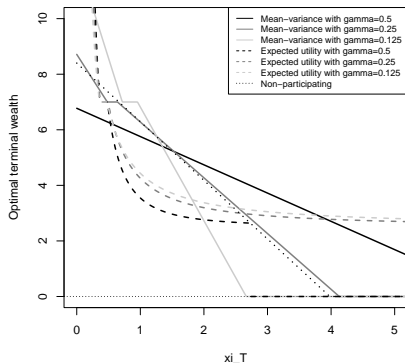
- ▶ S-shaped utility function:
$$U(x) = \begin{cases} x^{\tilde{\gamma}} & x \geq 0 \\ -\tilde{\lambda}(-x)^{-\tilde{\gamma}} & x < 0 \end{cases} \text{ with } \tilde{\lambda} = 2 \text{ and}$$
 different values for $\tilde{\gamma}$

- ▶ comparison with results from expected utility of Lin *et al.* (2017)
- ▶ different risk aversion levels

Optimal terminal wealth – Non-protected

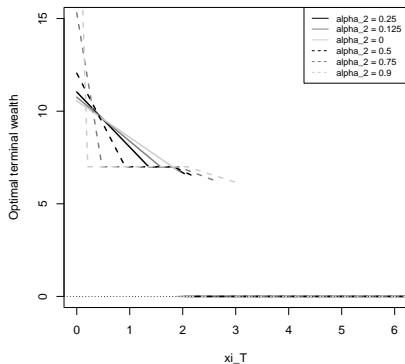


Optimal terminal wealth – Protected

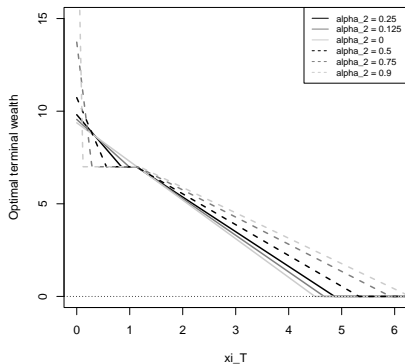


► influence of participation rate α_2

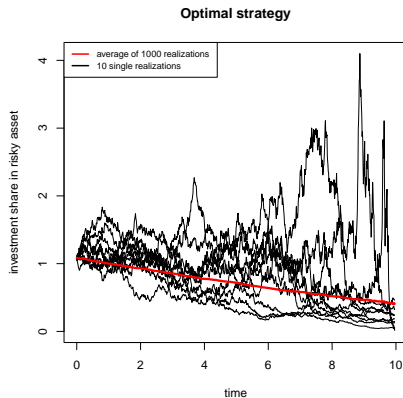
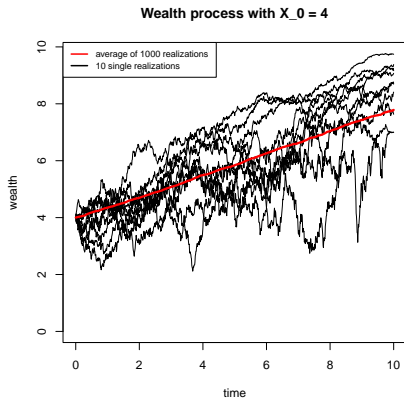
Optimal terminal wealth – Non-protected



Optimal terminal wealth – Protected

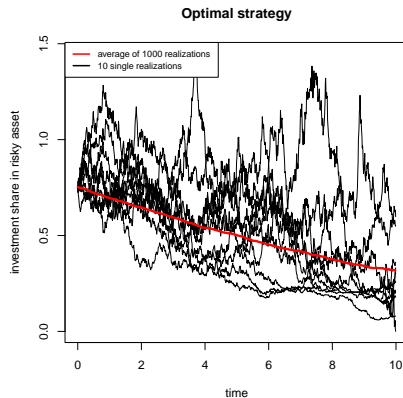
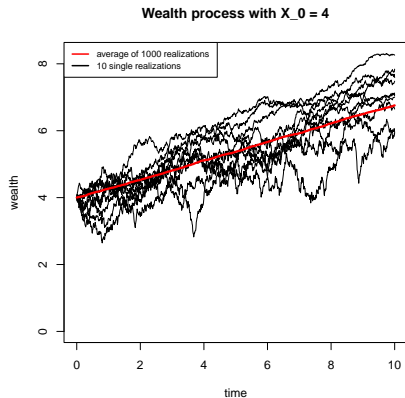


► Non-protected participating life insurance contract



- comparably riskier strategies if economy evolves bad
- comparably safer strategies if economy evolves extremely good

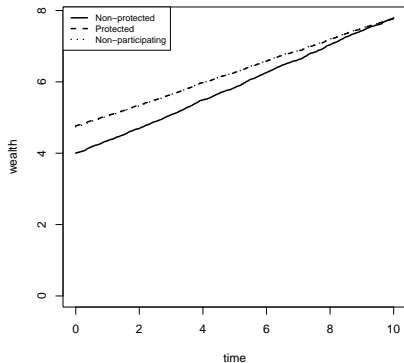
► Protected participating life insurance contract



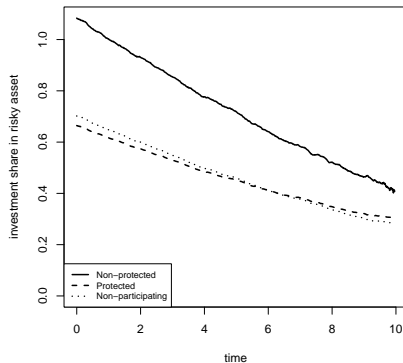
► extreme strategy changes if final value is close to k_2

- ▶ Comparison between both participating contracts and a non-participation contract
- ▶ x_0 is chosen such that \hat{X}_T is approximately equal in these cases

Wealth process – same expected value

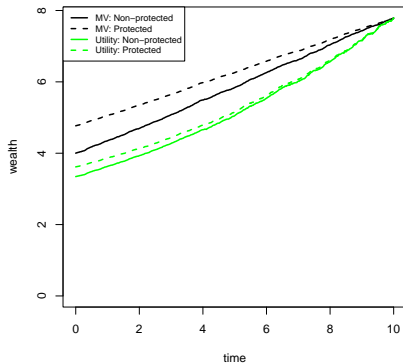


Optimal strategy

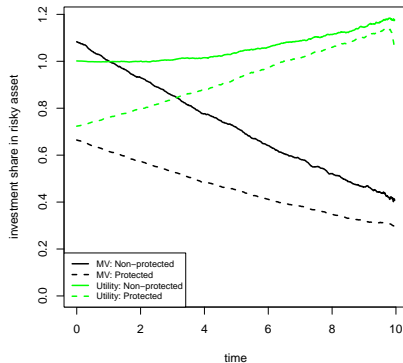


- ▶ Comparison between both participating contracts with mean-variance and expected utility
- ▶ x_0 is chosen such that \hat{X}_T is approximately equal in these cases

Wealth process – same expected value



Optimal strategy



Structure

Model Setup

Optimization in the Black-Scholes market

Numerical Results

Optimization in an Incomplete Market

Setting adaptations

- ▶ wealth process: $dX_t = b(X_t, u_t)dt + \sigma(X_t, u_t)dW_t$ where b and σ are measurable functions which satisfy a uniform Lipschitz condition
- ▶ value functional: $V(t, x) = \sup_{u \in \mathcal{U}(t, x)} \tilde{J}(t, T, u, x) = \sup_{u \in \mathcal{U}(t, x)} \mathbb{E}[\lambda F(0, T, u, x_0) - \gamma F(0, T, u, x_0)^2]$ where $\mathcal{U}(t, x)$ denotes the subset of \mathcal{U} with processes starting at t and $X_t = x$
- ▶ remember: $\lambda = 1 + 2\gamma \mathbb{E}[F(0, T, \hat{u}, x_0)]$

Theorem

If $V \in C^{1,2}$, then for every $\lambda \geq 0$ the optimal value functional V is the solution of the following SDE for all $(t, x) \in [0, T) \times \mathbb{R}$:

$$-\frac{dV}{dt}(t, x) - \sup_{u \in \mathcal{U}} \mathcal{L}^u V(t, x) = 0,$$

$$V(T, x) = F(T, T, 0, x),$$

where the operator \mathcal{L}^u is defined as

$$\mathcal{L}^u v(t, x) := b(x, u) \frac{dV}{dx}(t, x) + \frac{1}{2} \sigma(x, u) \sigma^T(x, u) \frac{d^2 V}{dx^2}(t, x),$$

where tr denotes the trace of a matrix.

Theorem

Let the control space \mathcal{U} be compact and the Hamiltonian H defined as usual, i.e.:

$$H : [0, T) \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R},$$

$$H(t, x, p, M) := \sup_{u \in \mathcal{U}} \left[b(x, u) \frac{dV}{dx}(t, x) + \frac{1}{2} \sigma(x, u) \sigma^T(x, u) \frac{d^2V}{dx^2}(t, x) \right].$$

If V is locally bounded on $[0, T) \times \mathbb{R}$, then for every $\lambda \geq 0$ V is a viscosity solution of the following Hamilton-Jacobi-Bellman (HJB) equation for $(t, x) \in [0, T) \times \mathbb{R}$:

$$-\frac{dV}{dt}(t, x) - H \left(t, x, \frac{dV}{dx}(t, x), \frac{d^2V}{dx^2}(t, x) \right) = 0,$$

$$V(T, x) = F(T, T, 0, x).$$

- can relax the assumption of \mathcal{U} being compact

Thank you for your attention!

Preprint available: <https://arxiv.org/pdf/2407.11761>

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