

Success and failure of the financial regulation on a surplus-driven financial company

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October 4, 2021

Outline:

- Motivation
- Model setup and optimization problems
- Solutions to the optimization problems
- Conclusion

Motivation

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- secure the stability of the financial system as a whole.



- Banking regulation: **Basel III**

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- Insurance regulation: Solvency II
 - Three-pillar framework;
 - Pillar I: the quantitative requirements to ensure the solvency of the banks/insurance companies;



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 - a measure of the **maximum loss** of a portfolio over a **given horizon** and a pre-determined **confidence level**;

- Quantitative requirements: **risk-based metrics** to indicate the **adequacy capital** that the financial institution needs to hold in order to meet the **solvency requirement**;
- e.g., Value-at-Risk (VaR):
 - a measure of the **maximum loss** of a portfolio over a **given horizon** and a pre-determined **confidence level**;
- Average Value-at-Risk (aka, Expected Shortfall)
 - a measure of the **average** of all potential losses **exceeding the VaR** at a **given confidence level**;

A stream of literatures focus on the impact of these risk measures on the investment strategies of the financial institutions.

- [Basak and Shapiro, 2001]
optimal asset allocation under the VaR/expected discounted shortfall constraint;
- [Cuoco et al., 2008]
optimal asset allocation under multiple VaR constraints;
- [Chen et al., 2018]
optimal asset allocation under VaR and minimum insurance;
- [Nguyen and Stadjé, 2020]
nonconcave optimal investment with VaR constraints;

Considering the **surplus-driven** characteristic of the financial company, we are interested in how these risk measures protect the **liability holders** of the financial institution.

Therefore, we explicitly **distinguish the liability and surplus** in the asset structure, and investigate the optimal investment problem under risk constraints.



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 - The company has **limited liability**;
- payoff of the equity holders:
 $\varphi_E(X_T) = X_T - \varphi_L(X_T) = \max(X_T - D_T, 0) = (X_T - D_T)^+$

Let $\rho(X_T)$ denote a risk measure of the terminal wealth, we formulate the following optimization problem:

$$\max_{X_T \in \mathcal{X}} \mathbb{E}[U((X_T - D_T)^+)] \quad (1)$$

$$\text{s.t. } \mathbb{E}[X_T \xi_T] \leq x_0, \quad \rho(X_T) \geq \leq.$$

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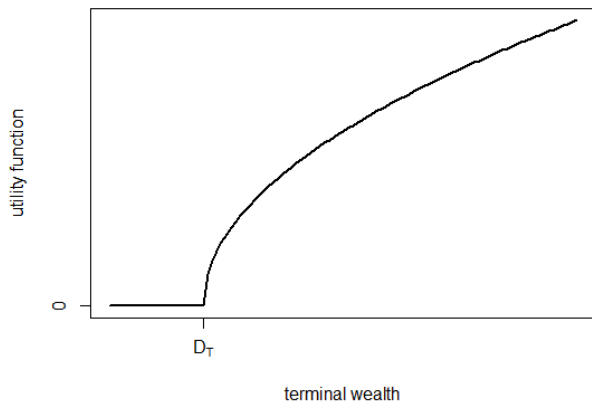
$$\text{s.t. } \mathbb{E}[X_T \xi_T] \leq x_0, \quad \rho(X_T) \geq \leq.$$

Remark:

- ξ_T is the state price density or stochastic discount factor;

$$\mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(s)ds} X_T] \leq x_0 \iff \mathbb{E}^{\mathbb{P}}[\xi_T X_T] \leq x_0.$$

- In a **complete** financial market, each state-contingent claim (X_T) can be replicated by a self-financing trading strategy;



The formal definition of the risk constraint $\rho(X_T)$

quantile-based risk constraints:

- Value-at-Risk:

$$VaR_{X_T}(\alpha) := \sup\{x | \mathbb{P}(X_T < x) \leq \alpha\}. \quad (2)$$

As a constraint, $VaR_{X_T}(\alpha) \geq L$ is equivalent to $\mathbb{P}(X_T \geq L) \geq 1 - \alpha$.

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- average Value-at-Risk:

$$AVaR_{X_T}(\alpha) := \frac{1}{\alpha} \int_0^\alpha VaR_{X_T}(\beta) d\beta \geq L'. \quad (3)$$

Shortfall-based risk constraints:

- Expected Shortfall under the physical measure $\mathbb{P}$

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- Expected Shortfall under the risk-neutral measure $\mathbb{Q}$

$$ES_{\mathbb{Q}} := \mathbb{E}[\xi_T(L - X_T)^+] \leq \epsilon_{\mathbb{Q}}. \quad (5)$$

$$(\mathbb{E}^{\mathbb{Q}}[e^{-\int_0^T r(s)ds}(L - X_T)^+] \equiv \mathbb{E}^{\mathbb{P}}[\xi_T(L - X_T)^+])$$

Essentially, we formulate four optimization problems:

$$\max_{X_T \in \mathcal{X}} \mathbb{E}[U((X_T - D_T)^+)]$$

$$\text{s.t. } \mathbb{E}[X_T \xi_T] \leq x_0, \quad \rho(X_T) \in \{\text{VaR}, \text{AVaR}, ES_{\mathbb{P}}, ES_{\mathbb{Q}}\}.$$

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In comparison, the **benchmark** problem is

$$\max_{X_T \in \mathcal{X}} \mathbb{E}[U((X_T - D_T)^+)]$$

$$\mathbb{E}[X_T \xi_T] \leq x_0.$$

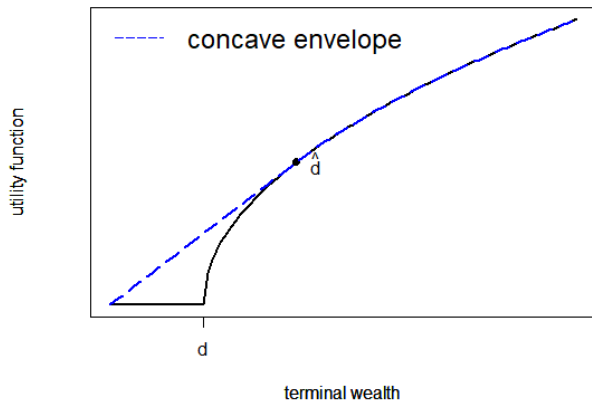
Solutions to the optimization problems

The benchmark solution:

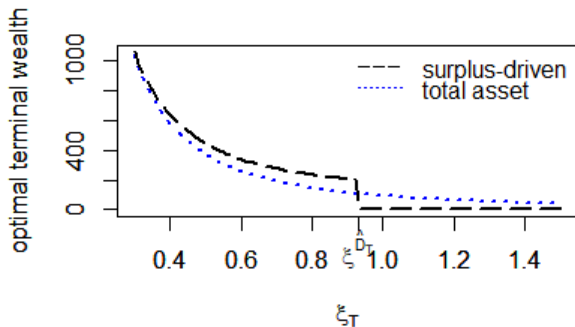
$$X_T^B = (I(\lambda_B \xi_T) + D_T) \mathbb{1}_{\xi_T < \xi^{\hat{D}_T}}, \quad (6)$$

where $\hat{D}_T$ is the **tangent point** with respect to D_T ,
 $\xi^{\hat{D}_T} = U'(\hat{D}_T - D_T)/\lambda_B$, and λ_B is obtained by solving $\mathbb{E}[X_T^B \xi_T] = x_0$.

Concave envelope: the smallest concave function that dominates the original non-concave function;



The benchmark wealth will either be above the tangent point $\hat{D}_T$ or jump to zero.



ES_Q constraint:

- a) If $L \leq D_T$, the optimal solution to is:

$$X^{ES_Q} = (I(\lambda_{\epsilon_Q} \xi_T) + D_T) \mathbb{1}_{\xi_T < \xi_{\epsilon_Q}} \quad \text{if } \xi_{\epsilon_Q} \leq \xi_{\epsilon_Q}^{\tilde{L}}, \quad (7)$$

$$X^{ES_Q} = (I(\lambda_{\epsilon_Q} \xi_T) + D_T) \mathbb{1}_{\xi_T < \xi_{\epsilon_Q}^{\tilde{L}}} + L \mathbb{1}_{\xi_{\epsilon_Q}^{\tilde{L}} \leq \xi_T < \xi_{\epsilon_Q}} \quad \text{if } \xi_{\epsilon_Q} > \xi_{\epsilon_Q}^{\tilde{L}}, \quad (8)$$

where $\xi_{\epsilon_Q}^{\tilde{L}} = U'(\tilde{L} - (D_T - L))/\lambda_{\epsilon_Q}$, $\tilde{L}$ is the tangent point with respect to $D_T - L$, ξ_{ϵ_Q} is defined through $\mathbb{E}[L \xi_T \mathbb{1}_{\xi_T \geq \xi_{\epsilon_Q}}] = \epsilon_Q$, $\xi^{\hat{D}_T} = U'(\hat{D}_T - D_T)/\lambda_{\epsilon_Q}$ and λ_{ϵ_Q} is obtained by solving $\mathbb{E}[\xi_T X_T^{ES_Q}] = x_0$.

- b) If $D_T < L \leq \hat{D}_T$, the optimal solution to is:

$$X_T^{ES_Q} = (I(\lambda_{\epsilon_Q} \xi_T) + D_T) \mathbb{1}_{\xi_T < \xi_{\epsilon_Q}} \quad \text{if } \xi_{\epsilon_Q} < \underline{\xi}_{\epsilon_Q}, \quad (9)$$

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where $\underline{\xi}_{\epsilon_Q}$ is defined through $U'(L - D_T)/\lambda_{\epsilon_Q}$.

ES_ℙ constraint:

- a) If $L \leq D_T$, the optimal solution to is:

$$X_T^{ES_{\mathbb{P}}} = (I(\lambda_{\epsilon_{\mathbb{P}}} \xi_T) + D_T) \mathbb{1}_{\xi_T < \xi_{\epsilon_{\mathbb{P}}}} \quad \text{if } \xi_{\epsilon_{\mathbb{P}}} \leq \tilde{\xi}_{\epsilon_{\mathbb{P}}}, \quad (11)$$

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where $\tilde{\xi}_{\epsilon_{\mathbb{P}}} = U'(\tilde{L} - (D_T - L)) / \lambda_{\epsilon_{\mathbb{P}}}$, $\tilde{L}$ is the tangent point with respect to $D_T - L$, $\xi_{\epsilon_{\mathbb{P}}}$ is defined through $\mathbb{E}[L \mathbb{1}_{\xi_T \geq \xi_{\epsilon_{\mathbb{P}}}}] = \epsilon_{\mathbb{P}}$, and $\lambda_{\epsilon_{\mathbb{P}}}$ is obtained by solving $\mathbb{E}[\xi_T X_T^{ES_{\mathbb{P}}}] = x_0$.

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where $\underline{\xi}_{\epsilon_{\mathbb{P}}}$ is defined by $U'(L - D_T) / \lambda_{\epsilon_{\mathbb{P}}}$.

VaR constraint:

a) If $L \leq D_T$, the optimal solution is:

$$X_T^{VaR} = (I(\lambda_\alpha \xi_T) + D_T) \mathbb{1}_{\xi_T < \bar{\xi}_\alpha} \quad \text{if } \xi^{\tilde{L}} \geq \bar{\xi}_\alpha, \quad (15)$$

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where $\xi^{\tilde{L}} = U'(\tilde{L} - (D_T - L))/\lambda_\alpha$, $\tilde{L}$ is the tangent point with respect to $D_T - L$, $\xi^{\hat{D}T} = U'(\hat{D}_T - D_T)/\lambda_\alpha$ and λ_α is defined via the budget constraint $\mathbb{E}[X_T^{VaR} \xi_T] = x_0$.

b) If $D_T < L \leq \hat{D}_T$, the optimal solution is:

$$X_T^{VaR} = (I(\lambda_\alpha \xi_T) + D_T) \mathbb{1}_{\xi_T < \underline{\xi}_\alpha} + L \mathbb{1}_{\underline{\xi}_\alpha \leq \xi_T < \bar{\xi}_\alpha} \quad \text{if } \bar{\xi}_\alpha \geq \underline{\xi}_\alpha, \quad (17)$$

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where $\underline{\xi}_\alpha = U(L - D_T)/\lambda_\alpha$, $\xi^{\hat{D}T} = U'(\hat{D}_T - D_T)/\lambda_\alpha$ and λ_α is obtained by solving $\mathbb{E}[X_T^{VaR} \xi_T] = x_0$.

AVaR constraint:

A useful lemma:

i)

$$\frac{1}{\alpha} \int_0^\alpha \text{VaR}_{X_T}(\beta) d\beta \geq \epsilon_0 \iff \mathbb{E}[(L - X)^+] \leq \alpha(L - \epsilon_0), \quad L \equiv q_X(\alpha)$$

where $q_X(\alpha)$ denotes the α -quantile of the terminal wealth, i.e., $q_X(\alpha) := \sup\{x | \mathbb{P}(X_T < x) < \alpha\}$.

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The function $f : L \rightarrow \mathbb{R}$, $f(L) = \mathbb{E}[(L - X)^+] - \alpha(L - \epsilon_0)$ attains its minimum at $L = q_X(\alpha)$.

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iii)

$$\{X_T | \mathbb{E}[X_T | \text{AVaR}_{X_T}(\alpha)] \geq \epsilon_0\} = \bigcup_{L \geq \epsilon_0} \{X_T | \mathbb{E}[(L - X_T)^+] \leq \alpha(L - \epsilon_0)\}.$$

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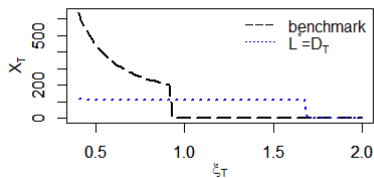
Theorem

Let $X_T^{ES,L}$ denote the optimal solution of the ES constrained problem

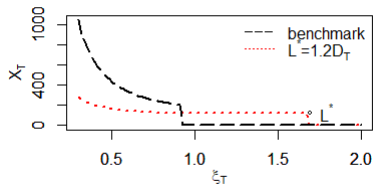
$$\left[\max_{X_T} \mathbb{E}[U(X_T - D_T)^+], \quad \text{s.t.} \quad \mathbb{E}[\xi_T X_T] \leq x_0, \quad \mathbb{E}[(L - X_T)^+] \leq \alpha(L - L') \right].$$

For each given initial wealth x_0 , there exists $L^* \geq L'$ such that the optimal solution of the AVaR constrained problem X_T^{AVaR} coincides with the optimal solution of the ES-constrained problem, i.e.,

$$X_T^{AVaR} = X_T^{ES,L^*}.$$



(a) $L^* = D_T$.



(b) $L^* = 1.2D_T$.

Figure: The optimal wealth under the risk constraint.

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- The optimal wealth shows a two-region/three-region distribution depending on the parameters;
- The probability of default is effectively reduced by the regulation;
- All four risk constraints can lead to the same optimal wealth;

Example: A Black Scholes market

Assuming $\alpha = 1\%$ and $L' = 0.9D_T$, the AVaR constraint is given by

$$\frac{1}{0.01} \int_0^{0.01} \text{VaR}_{X_T}(\beta) d\beta \geq 0.9D_T.$$

The optimal solution under the AVaR constraint is the same as the optimal solution under the $ES_{\mathbb{P}}$ constraint:

$$\mathbb{E}[(L^* - X_T)^+] \leq (L^* - L')\alpha.$$

The magnitude of L^* depends on the initial wealth. We consider two possible values to illustrate the equivalence among the risk constraints.

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Suppose $L^* = D_T$,

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$$\mathbb{E}[(L^* - X_T)^+] \leq \epsilon_{\mathbb{P}} = 0.1\% D_T;$$

- the equivalent $ES_{\mathbb{Q}}$ constraint is

$$\mathbb{E}[\xi_T (L^* - X_T)^+] \leq \epsilon_{\mathbb{Q}} = 0.00022\% D_T.$$

Example

	$\mathbb{P}(X_T \geq L^*) \geq \alpha^*$	$\epsilon_{\mathbb{P}}$	$\epsilon_{\mathbb{Q}}$
$L^* = D_T$	0.1%	$0.1\% D_T$	$0.00022\% D_T$
$L^* = 1.2 D_T$	0.25%	$0.3\% D_T$	$0.0015\% D_T$



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Considering the **surplus-driven** characteristic of the financial company,

- **all** the current risk-based metrics have the **same** regulatory effect;
- the **probability of default** is effectively **reduced** by the regulation but the liability holders **cannot be fully protected**.



Thanks for your attention!



https://papers.ssrn.com/sol3/papers.cfm?abstract_id=3920338

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