# Minimizing Probability of Lifetime Ruin Under Stochastic Volatility

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BFS2010, Toronto, June 25

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#### Outline

- 1 Lifetime Ruin Problem
- 2 Stochastic Volatility Model
- 3 Mathematical Tools
- 4 Main Results with Numerical Examples
- 5 Conclusion

#### Introduction

An optimal investment problem:

- Individual can invest in a market with
  - Risk less account:  $dB_t = rB_t dt$ ;
  - Risky asset /stock:  $dS_t = \mu S_t dt + \sigma_t S_t dB_t$
- She earns income A and has a minimal consumption c;
- Her future lifetime is random.

Question: How should she invest in order to minimize the probability of outliving her wealth, i.e, the probability of lifetime ruin?

Wealth dynamic:  $dW_t = [\mu \pi_t + r(W_t - \pi_t) + A - c]dt + \sigma_t \pi_t dB_t$ , with investing strategy  $\pi_t$  denoting the amount of money invested in risky asset at time *t*.

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#### Lifetime Ruin Problem

- Ruin level:  $w_a = 0$
- Safe level:  $w_s = \frac{c-A}{r}$
- Ruin time:  $\tau_0 = \inf \{t : W_t \le w_a\}$
- Death time:  $au_d$ , random, depends on the hazard rate  $\lambda$
- Minimum probability of ruin:

$$\psi(w, t) = \inf_{\pi_t \in \mathcal{A}} \mathbb{P}(\tau_0 < \tau_d | W_t = w, t < \tau_d)$$
  
= 
$$\inf_{\pi_t \in \mathcal{A}} \mathbb{E}^w[e^{-\int_t^{\tau_0} \lambda ds}]$$
(1)

• Boundary conditions:

$$\psi(w, t) = 1, \quad \text{ for } w \leq w_a, \ \psi(w, t) = 0, \quad \text{ for } w \geq w_s,$$

- Young [2004] obtained explicit formula for the optimal strategy and minimum lifetime ruin probability, when the stock price follows standard Black-Scholes model;
- Bayraktar et al. [2008] modeled the consumption as an increasing function of wealth, and considers random consumption;
- Young and Moore [2006] considered varying hazard rate.
- However, all the above work model the stock as a geometric Brownian motion with constant volatility.

More realistic stock model: Stochastic Volatility Model.

#### Stochastic Volatility Model

• Stock price:

$$dS_{t} = \mu S_{t} dt + \sigma_{t} S_{t} dB_{t}^{0}$$
  

$$\sigma_{t} = f(Y_{t}, Z_{t}),$$
  

$$dY_{t} = \frac{1}{\epsilon} (m - Y_{t}) dt + \frac{\sqrt{2}\nu}{\sqrt{\epsilon}} dB_{t}^{1}, \quad 0 < \epsilon << 1,$$
  

$$dZ_{t} = \delta c(Z_{t}) dt + \sqrt{\delta} d(Z_{t}) dB_{t}^{2}, \quad 0 < \delta << 1.$$
(2)

- Two volatility factors
  - Y<sub>t</sub>: fast volatility factor
  - Z<sub>t</sub>: slow volatility factor
- Reasons to use this model
  - can fit the implied volatility smile well;
  - we can obtain analytical approximation using multi-scale analysis (Fouque et al. [2000])).

#### Minimizing Probability of Ruin Under Stochastic Volatility

- Minimum probability of ruin:  $\psi(w, y, z) = \inf_{\pi_t} \mathbb{E}^{w, y, z}[e^{-\lambda \tau_0}].$
- Ito's formula and Dynamic Programming Principle give HJB equation for ψ:

$$\inf_{\pi\in\mathcal{A}}\mathcal{D}^{\pi}\psi=\mathbf{0}$$

where

$$\begin{aligned} \mathcal{D}^{\pi}\psi &= -\lambda\psi + (rw - c)\psi_{w} + \frac{1}{\epsilon}(m - y)\psi_{y} + \delta c(z)\psi_{z} \\ &+ \frac{1}{\epsilon}\nu^{2}\psi_{yy} + \frac{1}{2}\delta g^{2}(z)\psi_{zz} + \rho_{23}\sqrt{2}\nu\frac{\sqrt{\delta}g(z)}{\sqrt{\epsilon}}\psi_{yz} \\ &+ \left[\pi(\mu - r)\psi_{w} + \frac{1}{2}f^{2}(y, z)\pi^{2}\psi_{ww} + \frac{\rho_{12}f(y, z)\pi\nu\sqrt{2}}{\sqrt{\epsilon}}\psi_{wy} + \sqrt{\delta}\rho_{13}\pi f(y, z)g(z)\psi_{wz}\right] \end{aligned}$$

#### Mathematical Tools

- Verification Theorem: to validate a candidate solution.
- Legendre Transform: to obtain duality relationship between  $\psi$  and a concave function  $\hat{\psi}$  satisfying a PDE with free boundary condition.
- Asymptotic Analysis (Fouque et al. [2000]): to asymptotically expand  $\hat{\psi}$  in power of  $\sqrt{\epsilon}$  and  $\sqrt{\delta}$ , then compute explicit formula for each component.
- Markov Chain Approximation Method (MCAM) (Kushner and Dupuis [2001]): alternatively, we can approach the original problem directly and obtain numerical approximation.

#### Verification Theorem

Theorem 3.1. Suppose  $v : \mathbf{D} \to \mathbb{R}$  is a bounded, continuous function that satisfies the following conditions:

- $v(\cdot, y, z) \in C^2$  is a non-increasing, convex function;
- 2  $v(w, \cdot, \cdot) \in C^{2,2};$
- v(0, y, z) = 1;
- v(c/r, y, z) = 0;
- $\ \, {\mathfrak O}^{\beta} v \geq 0 \ \text{for all} \ \beta \in {\mathbf R}.$

Then,  $v \leq \psi$  on **D**.

#### Dual of $\psi$

• A related optimal controller-stopper problem:

$$dX_t^{\gamma} = -(r-\lambda)X_t^{\gamma}dt - \frac{\mu - r}{f(Y_t, Z_t)}X_t^{\gamma}d\tilde{B}_t^{(0)} + \gamma_t^{(1)}d\tilde{B}_t^{(1)} + \gamma_t^{(2)}d\tilde{B}_t^{(2)}$$

Define

$$\hat{\psi}(x,y,z) = \inf_{\tau} \sup_{\gamma} \mathbf{E}^{x,y,z} \left[ \int_{0}^{\tau} e^{-\lambda t} c X_{t}^{\gamma} dt + e^{-\lambda \tau} \min\left((c/r) X_{\tau}^{\gamma}, 1\right) \right].$$
(3)

Convex Legendre Dual:

$$\Psi(w, y, z) = \max_{x} \left( \hat{\psi}(x, y, z) - wx \right).$$
(4)

Theorem 4.1. Ψ equals the minimum prob of lifetime ruin ψ, and the optimal strategy π\* is given by the first order condition.

#### Asymptotic Approximation

• Asymptotic approximation result for  $\hat{\psi}$ :

$$\begin{split} \hat{\psi}(x,z) &= \hat{\psi}_{0,0}(x,z) + \sqrt{\epsilon} \, \hat{\psi}_{0,1}(x,z) + \sqrt{\delta} \, \hat{\psi}_{1,0}(x,z) + \mathcal{O}(\epsilon,\delta,\sqrt{\epsilon\,\delta}) \\ &= -\frac{1}{B_1(z) - 1} \left(\frac{c}{r} \cdot \frac{B_1(z) - 1}{B_1(z)} \cdot x\right)^{B_1(z)} + \frac{c}{r} \, x \\ &+ \sqrt{\epsilon} \, A(z) \, x^{B_1(z)} \, \log\left(x \cdot \frac{B_1(z) - 1}{B_1(z)} \cdot \frac{c}{r}\right) \\ &+ \sqrt{\delta} \, x^{B_1(z)} \, \log\left(x \cdot \frac{B_1(z) - 1}{B_1(z)} \cdot \frac{c}{r}\right) \left[A_1(z) + A_2(z) \log\left(x \cdot \frac{B_1(z)}{B_1(z) - 1} \cdot \frac{r}{c}\right)\right] \\ &+ \mathcal{O}(\epsilon, \delta, \sqrt{\epsilon\,\delta}), \end{split}$$
(5)

• Asymptotic approximation result for optimal strategy

$$\hat{\pi^{*}}(x, y, z) = -\frac{\mu - r}{f^{2}(y, z)} \times \hat{\psi}_{0,0,xx} + \sqrt{\epsilon} \left( -\frac{\mu - r}{f^{2}(y, z)} \times \hat{\psi}_{0,1,xx} + \rho_{12} \frac{\nu \sqrt{2}}{f(y, z)} \hat{\psi}_{0,2,xy} \right) + \sqrt{\delta} \left( -\frac{\mu - r}{f^{2}(y, z)} \times \hat{\psi}_{1,0,xx} + \rho_{13} \frac{h(z)}{f(y, z)} \hat{\psi}_{0,0,xz} \right) + \mathcal{O}(\epsilon, \delta, \sqrt{\epsilon \delta}).$$
(6)

#### Numerical Example

- Question 1: What do the optimal solutions look like?
- Question 2: How does the stochastic environment affect our strategy?
- Question 3: How do different strategies perform in stochastic environment?

Main Results with Numerical Examples

# Q1: Minimum Ruin Probability



(a) Fast volatility factor  $\epsilon = 0.004$  (b) Slow volatility factor  $\delta = 0.02$  (reverting speed =  $1/\epsilon = 250$ ) (reverting speed =  $\delta = 0.02$ )

FIGURE 1: Minimum Probability of Ruin

Main Results with Numerical Examples

#### Q1: Optimal Strategy



(a) Fast volatility factor  $\epsilon = 0.004$  (b) Slow volatility factor  $\delta = 0.02$ 

FIGURE 2: Optimal Strategy

#### Question 2: How does the stochastic environment affect our strategy?

• Recall that in constant volatility environment, (Young[2004]),

- Optimal strategy  $\tilde{\pi}(w; \sigma) = \frac{\mu r}{\sigma^2} \frac{c wr}{(p-1)r}$ ;
- Minimum ruin probability  $\tilde{\psi}(w) = (1 \frac{rw}{c})^p$ . where  $p = \frac{1}{2r}[(r + \lambda + s) + \sqrt{(r + \lambda + s)^2 - 4r\lambda}]$ , and  $s = \frac{1}{2} \left(\frac{\mu - r}{\sigma}\right)^2$ .

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#### **Optimal Strategy:**



(a) Fast volatility factor  $\epsilon = 0.004$  (b) Slow volatility factor  $\delta = 0.02$  (reverting speed =  $1/\epsilon = 250$ ) (reverting speed =  $\delta = 0.02$ )

FIGURE 3: Minimum Probability of Ruin

#### Optimal Strategy(Cont):



(a) Medium reverting speed = 0.2

FIGURE 4: Minimum Probability of Ruin

# Question 3: How do different strategies perform in stochastic environment?

• Consider the following strategies:

• 
$$\pi^a$$
:  $\pi^a(w) = \tilde{\pi}(w; \sigma_0);$ 

• 
$$\pi^b$$
:  $\pi^a(w) = \tilde{\pi}(w; \sigma_m);$ 

• 
$$\pi^c$$
:  $\pi^a(w) = \tilde{\pi}(w; f(y, z));$ 

•  $\pi^M$ : invest only in money market.

- $\pi^{\epsilon}, \pi^{\delta}$  : strategy obtained by asymptotic approximation.
- $\pi^*$ : optimal strategy;

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Ruin Probability: (volatility factor reverts with medium speed 0.2)

Figure 3.1: σ<sub>0</sub>=0.6;

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• Figure 3.3:  $\sigma_0 = 0.1$ .



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#### Conclusion

We obtained

- closed-form asymptotic approximation to optimal investment strategy and minimum probability of ruin;
- effects of the stochastic volatility environment;
- an easy-to-implement rule for nearly optimal lifetime ruin probability.

# Thanks for your attention!

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