

Risk Minimizing Static-Dynamic Hedges for Exotic Options

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Static-Dynamic Hedging

- ⊗ Suppose we have written an exotic derivative with payoff G^e
- ⊗ Want to **minimize risk** from possible loss at time T
- ⊗ Find optimal hedging strategy for **dynamically** trading in stocks and **statically** trading in options

$$-G^e + V_0 + \int_0^T \theta \cdot dS + \lambda \cdot G - \lambda \cdot p$$

- ⊗ Justification: options have higher transaction costs.
- ⊗ Only meaningful if option cannot be replicated by dynamic trading strategy (\Rightarrow market incomplete).
- ⊗ Alternatively **maximize utility**.

What is meant by “utility” or “risk”?

- ⊙ “Utility” is measured as **expected** utility from wealth. Utility function $U(v)$ is concave in v . Examples:
 - $U(v) = -\exp(-\gamma v)$, $\gamma > 0$
 - $U(v) = v^\beta$, $0 < \beta < 1$, $v \geq 0$
 - $U(v) = \log v$, $v > 0$
- ⊙ \mathcal{X} = vector space of bounded (\mathcal{F}_T -measurable) random variables (financial position at time T)
- ⊙ “Risk” is measured by a **convex risk measure**. A convex risk measure is a function $\rho : \mathcal{X} \rightarrow \mathbb{R}$ satisfying the following axioms:
 - *monotonicity*:

$$X \geq Y \quad \Rightarrow \quad \rho(X) \leq \rho(Y).$$

– *convexity*: if $0 < \lambda < 1$, then

$$\rho(\lambda X + (1 - \lambda)Y) \leq \lambda\rho(X) + (1 - \lambda)\rho(Y).$$

– *monetary invariance*: if $m \in \mathbb{R}$, then

$$\rho(X + m) = \rho(X) - m.$$

⊙ A convex risk measure is **coherent** if

$$\rho(tX) = t\rho(X) \quad \text{for } t \geq 0.$$

⊙ A convex risk measure ρ is determined by its **acceptance set**:

$$A_\rho := \{X \mid \rho(X) \leq 0\}$$

⊙ “utility” is measured by $-\{\text{convex risk measure}\}$, “utility” does not need to satisfy monetary invariance.

Examples of Risk Measures

- ✗ *Value at Risk.*

$$\text{VaR}_\lambda(X) = \min\{m \in \mathbb{R} \mid P\{X + m < 0\} \leq \lambda\}$$

does *not* define a convex risk measure in general.

- ✗ *Average Value at Risk.*

$$\text{AVaR}_\lambda(X) = \frac{1}{\lambda} \int_0^\lambda \text{VaR}_\mu(X) d\mu$$

does define a (coherent) convex risk measure.

- ✗ Any convex subset \mathcal{A} of X satisfying

$$X \in \mathcal{A}, Y \geq X \quad \Rightarrow \quad Y \in \mathcal{A}$$

gives rise to a convex risk measure.

$$\rho_{\mathcal{A}}(X) = \inf\{m \in \mathbb{R} \mid m + X \in \mathcal{A}\}$$

- ✘ *Exponential Utility or Entropic risk measure.*

$$\rho(X) = \gamma^{-1} \log E^P \{e^{-\gamma X}\}$$

- ✘ *Shortfall risk.* If $l : \mathbb{R} \rightarrow \mathbb{R}$ is a convex increasing function and P is a prior probability measure then

$$\mathcal{A} = \{X \mid E^P \{l(-X)\} \leq l_0\}$$

is the acceptance set of a convex risk measure.

Example: $l(x) = x^p/p$, for $x \geq 0$, and $l(x) = 0$ otherwise, where $p > 1$.

Example: $l(x) = e^{\gamma x}$ gives exponential utility.

References

- ▷ Coherent risk measures: Artzner, Delbaen, Eber, Heath (1999).
- ▷ Convex risk measures: Föllmer, Schied (2002).

Convex Duality

- ⊖ Convex risk measures can be understood through convex duality:

$$\rho(X) = \sup_Q (E^Q\{-X\} - \alpha(Q)),$$

where Q runs over all probability measures and α is a **penalty** function (the convex dual of ρ).

- ⊖ Strict convexity of ρ corresponds to differentiability of α and vice versa.
- ⊖ Example: $\rho(X) = \log E\{e^{-\gamma X}\}$ gives entropic penalty $\alpha(Q) = H(Q|P) = E\left\{\frac{dQ}{dP} \log \frac{dQ}{dP}\right\}$.
- ⊖ Example: shortfall with loss function $l(x) = x^p/p$ gives $\alpha(Q) = E\left\{\left(\frac{dQ}{dP}\right)^q\right\}^{1/q}, p^{-1} + q^{-1} = 1$.

Convolution of Risk Measures

- ↪ Want to optimize over static *and* dynamic strategies.
- ↪ Can incorporate dynamic part of optimization problem into the risk measure by [inf-convolution](#).
- ↪ Given the convex risk measure ρ define new convex risk measure $\rho^{\mathcal{H}}$ by

$$\rho^{\mathcal{H}}(X) = \inf_{\theta \in \mathcal{H}} \rho(X + V_T^\theta).$$

- ↪ The convolution in going from ρ to $\rho^{\mathcal{H}}$ corresponds to going from α to $\alpha^{\mathcal{H}} = \alpha + \infty \cdot \mathbf{1}_{\mathcal{M}}$, where \mathcal{M} is the set of [martingale measures for \$S\$](#) .
- ↪ Reference: Barrieu, El Karoui (2004).

Indifference Pricing

≡ Recall the static hedging problem:

$$\lambda^* = \arg \min_{\lambda} \rho^{\mathcal{H}}(\lambda \cdot G - G^e) + \lambda p.$$

≡ Define a new convex risk measure, the **indifference price** given an initial liability G^e , by

$$\rho^{e, \mathcal{H}}(\lambda \cdot G) = \rho^{e, \mathcal{H}}(\lambda \cdot G - G^e) - \rho^{\mathcal{H}}(-G^e).$$

≡ Recasting the static hedging problem:

$$\lambda^* = \arg \min_{\lambda} \rho^{e, \mathcal{H}}(\lambda \cdot G) + \lambda p.$$

≡ Reference: Mingxin Xu (2005).

Optimal strategies in Finite State Spaces

⇔ **Theorem.** Assume that

- (i) the convex risk measure ρ is *differentiable* and *strictly convex* on \mathcal{X} ,
- (ii) no nontrivial linear combination $\lambda \cdot G$ of the derivatives can be replicated by a dynamic trading strategy in the stocks and bond, then

$$\begin{aligned}\lambda \mapsto \rho^{\mathcal{H}}(\lambda \cdot G - G^e) &= \inf_{\theta} \rho(V^{\theta} + \lambda \cdot G - G^e) \\ &= \sup_{Q \in \mathcal{M}} \{E^Q\{-\lambda \cdot G + G^e\} - \alpha(Q)\}\end{aligned}$$

is strictly concave.

- ⇔ Proof: if the function was affine on the segment joining λ_1, λ_2 , then the optimal measures are the same, $Q_1^* = Q_2^*$, and the option $(\lambda_2 - \lambda_1) \cdot G$ is redundant.
- ⇔ Therefore, as long as p is a no-arbitrage price vector for G , there exists a unique minimizer $\lambda^* = \lambda^*(p)$ minimizing

$$\rho^{\mathcal{H}}(\lambda \cdot G - G^e) + \lambda \cdot p.$$

Exponential Utility

- ◇ Assume stock price S is a locally bounded semimartingale in the “usual” continuous-time setup, option payoffs are bounded, and there exists a martingale measure $Q \sim P$ of finite relative entropy.
- ◇ Let \mathcal{P}_f be the set of martingale measures for S with finite relative entropy with respect to P .
- ◇ **Theorem.**

$$\inf_{\theta \in \mathcal{H}} \log E\{e^{-\gamma(G+V_T^\theta)}\} = \sup_{Q \in \mathcal{P}_f} E^Q\{-G\} - \frac{1}{\gamma} H(Q|P),$$

where the supremum and infimum are achieved by Q^* and θ^* such that

$$\frac{dQ^*}{dP} = ce^{-\gamma(V_T^{\theta^*} + G)}.$$

DGRSSS (2002).

◇ Let

$$b(\lambda) = \inf_{\theta \in \mathcal{H}} \log E \left\{ e^{-\gamma \left(\int_0^T \theta dS + \lambda \cdot G - G^e \right)} \right\}.$$

Lemma. Set of no-arbitrage option prices = $\{\nabla b(\lambda) | \lambda \in \mathbb{R}^N\}$.

◇ **Theorem.** If no nontrivial linear combination $\lambda \cdot G$ of the derivatives is replicable by a dynamic trading strategy in the stocks and bond, then, for any no-arbitrage option price p , there exists a unique static options position $\lambda^*(p)$ minimizing

$$b(\lambda) + \lambda \cdot p.$$

◇ Can include semi-static hedging in options, i.e., rolling over option positions.

◇ Extension to other risk measures (e.g. shortfall risk) is current work.

Implementation

- ↪ Would like to find the optimal strategies in an efficient way.
- ↪ Seems to be a nontrivial problem in general.
- ↪ Can try tree methods or PDE methods by exploiting convex duality.

Trees

- ⊢ Special **recursive structure** available in exponential utility case. Essentially stems from, for a suitable measure $\tilde{P} \sim P$,

$$H(Q|P) + E^Q[G] = H(Q|\tilde{P}) \quad \text{and} \quad H(Q|P) = \sum_t H_t(Q|P).$$

- Musiela, Zariphopoulou (2004)
- Lim (2005)
- ⊢ Recursive structure not present in general.

Expected Shortfall

—○ Non-traded asset model

$$\begin{aligned}dS_t &= \mu S_t dt + \sigma S_t dW_t^1, & S_0 &= S, \\dY_t &= bY_t dt + aY_t(\rho dW_t^1 + \sqrt{1-\rho^2} dW_t^2), & Y_0 &= y.\end{aligned}$$

—○ Shortfall risk with short derivative position $g(Y_T)$

$$\begin{aligned}\rho^{\mathcal{H}}(g(Y_T)) &= \inf_{\theta \in \mathcal{H}} \inf\{m \in \mathbb{R} \mid E\{\ell(-m + g(Y_T) - V_T^\theta - v)\} \leq \rho_0\}, \\ &= \inf\{m \in \mathbb{R} \mid \inf_{\theta \in \mathcal{H}} E\{\ell(g(Y_T) - m - V_T^\theta - v)\} \leq \rho_0\}, \\ &= \inf\{m \in \mathbb{R} \mid \inf_{\theta \in \mathcal{H}} E\{\ell(g(Y_T) - V_T^\theta - m)\} \leq \rho_0\} - v.\end{aligned}$$

— Take $l(x) = x^p/p$ for $x \geq 0$, where $p > 1$.

— Introduce the value function

$$H(t, v, y) = \inf_{\theta} \frac{1}{p} \mathbb{E} \left\{ ((g(Y_T) - V_T^\theta)^+)^p \mid V_t = v, Y_t = y \right\},$$

— the “inverse” of H

$$H(t, u(t, v, y, \xi), y) = \xi,$$

— and the Fenchel-Legendre transform of u

$$\hat{u}(t, y, z) = \inf_{\xi \geq 0} (u(t, y, \xi) + \xi z).$$

—o The PDE that \hat{u} solves:

$$\hat{u}_t + \mathcal{L}_y \hat{u} + \frac{\rho\mu a}{\sigma} (z\hat{u}_{zy} - \hat{u}_y) + \frac{\mu^2}{2\sigma^2} z^2 \hat{u}_{zz} - \frac{1}{2} a^2 (1 - \rho^2) \frac{(z\hat{u}_{zy} - \hat{u}_y)^2}{z^2 u_{zz}} = 0,$$

with

$$\hat{u}(t, y, z) = g(y) - \left(\frac{p-1}{p} \right) z^{1/(1-p)}.$$

—o This is the HJB equation related to another stochastic control problem:

$$\sup_{\gamma} E^{Q^{\gamma}} \left\{ g(Y_T) - \left(\frac{p-1}{p} \right) Z_T^{1/1-p} \mid Y_t = y, Z_t = z \right\},$$

where under Q^{γ}

$$dZ_t = Z_t \left(\frac{\mu}{\sigma} dW_t^{\gamma,1} + \gamma_t dW_t^{\gamma,2} \right), \quad Z_t = z,$$

$$dY_t = \left(b - \rho \frac{\mu}{\sigma} a - \sqrt{1 - \rho^2} a \gamma_t \right) Y_t dt + a Y_t (\rho dW_t^{\gamma,1} + \sqrt{1 - \rho^2} dW_t^{\gamma,2}).$$

- ★ For exponential utility, the nonlinear PDE that we need to solve:

$$\begin{aligned}\phi_t + \mathcal{L}_y^0 \phi + \frac{1}{2}(1 - \rho^2)a^2 y^2 (\phi_y)^2 &= \frac{\mu^2}{2\gamma\sigma^2}, \\ \phi(T, x, y) &= g(y).\end{aligned}$$

- ★ Long-Maturity Call and Put Example:

$$\begin{aligned}G^e &= -(Y_{T_c} - K_c)^+, \\ G &= -(K_p - Y_{T_p})^+.\end{aligned}$$

Parameters for numerics: $K_c = K_p = 100$, $\mu = 0.075$, $\sigma = 0.22$,
 $a = 0.45$, $b = \mu a / \sigma$, $\rho = 0.5$, $T_p = 1$, $T_c = 3$, $\gamma = 1$.

Numerical Solution

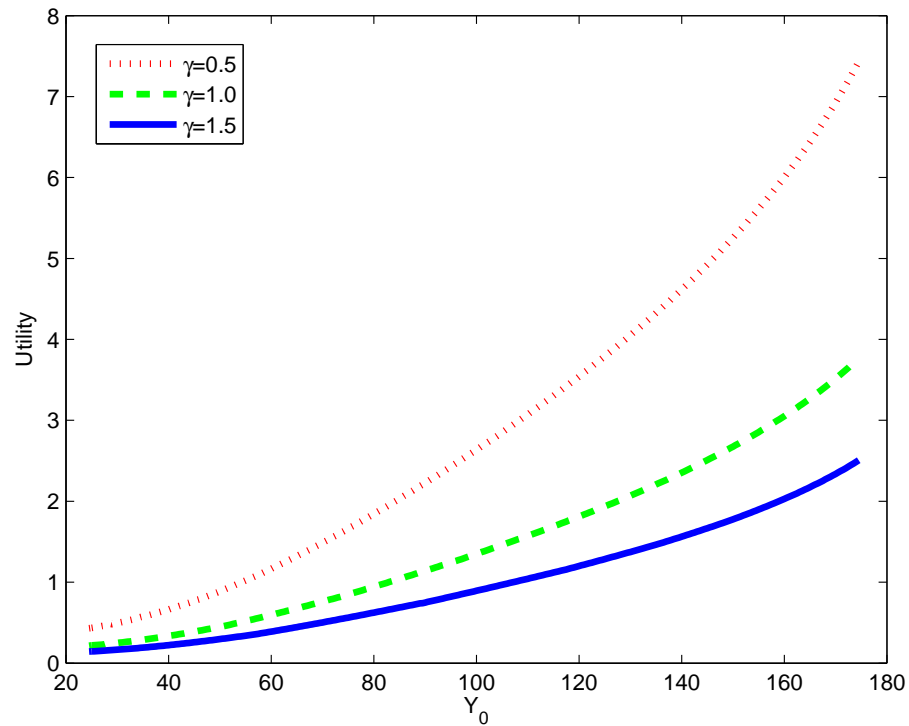


Figure 1: Negative of the risk measure with optimal static hedging in the puts.

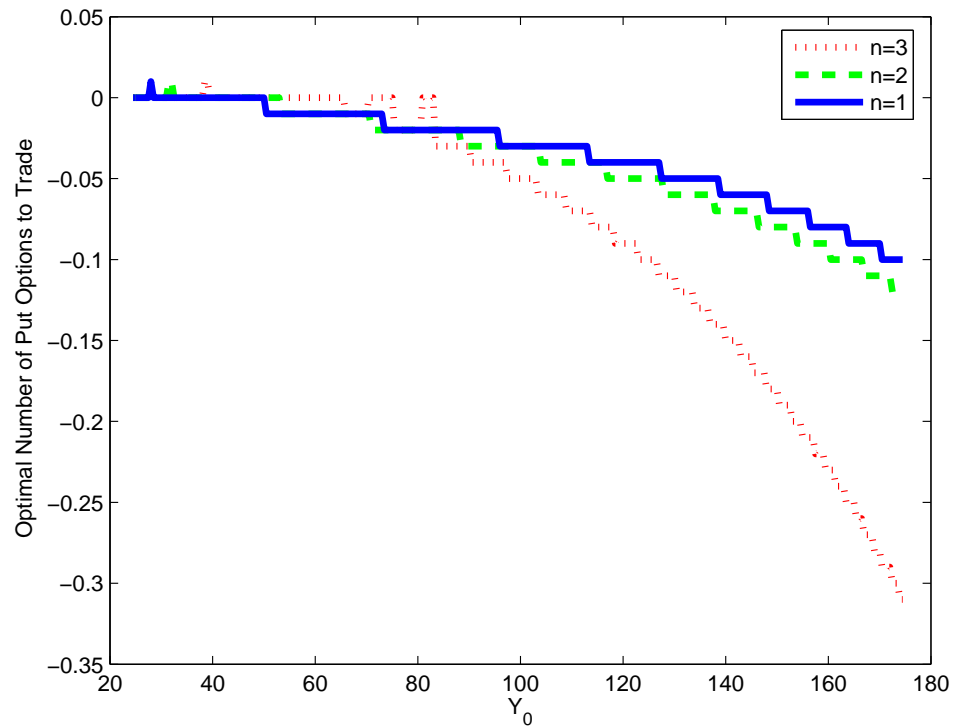


Figure 2: Optimal number of one year put options to trade each year.

Conclusion

- ↪ Given a risk measure ρ , try to find optimal static-dynamic trading (or hedging) strategies.
- ↪ Results in:
 - finite state spaces
 - continuous time with exponential utility
- ↪ Current work:
 - continuous time and other risk measures
 - Efficient numerical implementation.