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Mean Variance Optimization with State Dependent Risk Aversion

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The Classical Mean Variance Problem

- X_t wealth process
- *u_t* amount invested in the risky asset

$$\max E_{t,x}[X_T] - \frac{\gamma}{2} Var_{t,x}[X_T]$$

• time inconsistency

$$\max E_{t,x}[F(X_T)] + G(E_{t,x}[X_T])$$

• the control does not depend on the current wealth

$$u(t,x)=h(t)$$

Why State Dependent Risk Aversion

- u(t,x) = h(t) since the risk aversion parameter γ is constant
- it does not matter if the wealth is 100 USD, or 100 000 000 USD, we invest the **same** amount of dollars in the risky asset
- conceptual difference between the 1-period model and the multi-period model.
- make γ explicitely dependent on X_t

$$\max E_{t,x}[X_T] - \frac{\gamma(x)}{2} Var_{t,x}[X_T]$$

where $X_t = x$

The Problem

riskless asset

$$dB_t = rB_t dt$$

risky asset

$$dS_t = \alpha S_t dt + \sigma S_t dW_t$$

wealth portfolio

$$dX_t = [rX_t + (\alpha - r)u_t]dt + \sigma u_t dW_t$$

- u_t amount of money invested in the stock
- $J(t, x, \mathbf{u}) = E_{t,x}[X_T] \frac{\gamma(x)}{2} Var_{t,x}[X_T]$

Time Inconsistency

$$E_{t,x}[X_T] - \frac{\gamma(x)}{2} Var_{t,x}[X_T] = E_{t,x}[F(x, X_T)] + G(x, E_{t,x}[X_T])$$

- standard dynamic programming problem $\max E_{t,x}[F(X_T)]$
- here $\max E_{t,x}[F(x, X_T)] + G(x, E_{t,x}[X_T])$
- conceptual problem what is optimal?
- computational problem how we compute it?

- **Pre-commitment:** Solve (somehow) the problem at 0, x₀ and ignore the fact that later on, your "optimal" control will no longer be viewed as optimal.
- **Game theory:** Take the time inconsistency seriously. View the problems as a game and look for a Nash equilibrium point.

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- Ekeland & Lazrak (2006); Ekeland & Pirvu (2007)
- Basak & Chabakauri (2008)
- Björk & Murgoci (2009)

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The Game Theoretic Approach

- We view this as a game where there is one player for each t.
- Player No t chooses the control function u(t, ·) at time t, and applies the control u(t, X_t)
- The value, to player No t, if all players use the control law u is

$$J(t,x;u) = E_{t,x}[x,F(X_T^u)] + G(x,E_{t,x}[X_T^u])$$

Subperfect Nash Equilibrium

Definition

The strategy \hat{u} is a **Nash subgame perfect equilibrium** if the following holds for all *t*:

- Assume that all players No s with s > t use the control $\hat{u}(s, X_s)$.
- Then it is optimal for player No t also to use $\hat{u}(t, X_t)$.

Note!

• this leads to an extension of the HJB equation as a PDE system with an embedded fixed point problem.

Motivation	Our Problem	Time Inconsistency	Numerical Results	Conclusions
Notation				

$$V(T,x) = F(x,x) + G(x,x)$$

$$F(x,y) = y - \frac{\gamma(x)}{2}y^{2},$$

$$G(x,y) = \frac{\gamma(x)}{2}y^{2}.$$

• Probabilistic interpretation

$$f(t, x, y) = E_{t,x} \left[F(y, X_T^{\hat{\mathbf{u}}}) \right]$$
$$g(t, x) = E_{t,x} \left[X_T^{\hat{\mathbf{u}}} \right]$$

Note! $V(t,x) = f(t,x,x) + \frac{\gamma(x)}{2}g^2(t,x)$

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Fixed Point PDE System

$$\begin{split} \sup_{u \in \mathcal{U}} \left\{ (\mathcal{A}^{u}V)(t,x) - (\mathcal{A}^{u}f)(t,x,x) + (\mathcal{A}^{u}f^{x})(t,x) \\ - \mathcal{A}^{u}G(x,g(t,x)) + G_{y}(x,g(t,x)) \cdot \mathcal{A}^{u}g(t,x) \right\} &= 0, \\ \mathcal{A}^{\hat{u}}f^{y}(t,x) &= 0, \\ \mathcal{A}^{\hat{u}}g(t,x) &= 0, \\ V(T,x) &= F(x,x) + G(x,x) \\ f(T,x,y) &= F(y,x), \\ g(T,x) &= x. \end{split}$$

Remember! $V(t,x) = f(t,x,x) + \frac{\gamma(x)}{2}g^2(t,x)$

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Solving the PDE system

Optimal control

$$\hat{u}(t,x) = -\frac{\alpha - r}{\sigma^2} \frac{f_x(t,x,x) + \gamma(x)g(t,x)g_x(t,x)}{f_{xx}(t,x,x) + \gamma(x)g(t,x)g_{xx}(t,x)}$$

New PDE system

$$f_t + [rx + (\alpha - r)\hat{u}]f_x + \frac{1}{2}\sigma^2 f_{xx} = 0$$

$$g_t + [rx + (\alpha - r)\hat{u}]g_x + \frac{1}{2}\sigma^2 g_{xx} = 0$$

with f and g evaluated at (t, x, y) and

$$f(T, x, y) = x - \frac{\gamma(y)}{2}x^2$$

$$g(T, x, y) = x$$

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One Possible Solution

for

$$\gamma(x) = \frac{\gamma}{x}$$

we show that

• $\hat{u}(t, x) = c(t)x$ is a solution to the PDE system • $c(t) = \frac{\beta}{\gamma\sigma^2} \left[\frac{a(t)}{b(t)} + \gamma \left(\frac{a^2(t)}{b(t)} - 1 \right) \right]$ where $\beta = \alpha - r$ $a(t) = e^{\int_t^T [r + \beta c(s)] ds}$ $b(t) = e^{2\int_t^T [r + \beta c(s) + \frac{1}{2}\sigma^2 c^2(s)] ds}$

• $V(t,x) = \{a(t) + \frac{\gamma}{2}[a^2(t) - b(t)]\}x$

Existence for
$$c(t)$$

$$c_{0}(t) = 1$$

$$c_{n+1}(t) = \frac{\beta}{\gamma\sigma^{2}} \left[e^{-\int_{t}^{T} [r+\beta c_{n}(s)+\sigma^{2}c_{n}^{2}(s)]ds} + \gamma e^{-\int_{t}^{T} \sigma^{2}c_{n}^{2}(s)ds} - \gamma \right],$$

$$n = 0, 1, 2...$$

• Step 1. $\{c_n(\cdot)\}$ uniformly bounded in $\mathcal{C}([0, T])$ • Step 2. $\{\dot{c}_n(\cdot)\}$ uniformly bounded in $\mathcal{C}([0, T])$. • Step 3. For any $t_1, t_2 \in [0, T]$, we have

$$|c_n(t_2) - c_n(t_1)| = \int_0^1 \dot{c}_n(t_1 + \theta(t_2 - t_1)) d\theta(t_2 - t_1)$$

 $\leq k |t_2 - t_1| \quad \forall n$

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where k is a constant independent of n.

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Existence for c(t)

$$c_{0}(t) = 1$$

$$c_{n+1}(t) = \frac{\beta}{\gamma\sigma^{2}} \left[e^{-\int_{t}^{T} [r+\beta c_{n}(s)+\sigma^{2}c_{n}^{2}(s)]ds} + \gamma e^{-\int_{t}^{T} \sigma^{2}c_{n}^{2}(s)ds} - \gamma \right],$$

$$n = 0, 1, 2...$$

Step 1+2+3 \Rightarrow there is a $c(\cdot) \in \mathcal{C}([0, T]$ such that

$$c_{n_i}(\cdot) \stackrel{n \to \infty}{\to} c(\cdot) \in \mathcal{C}([0, T])$$

Uniqueness can be proved easily

Proportion of Money invested in the Stock for Various γ



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Proportion of Money invested in the Stock for Various Time Horizons



Motivation

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THANK YOU!