## Portfolio Optimization in Incomplete Markets

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Shuenn-Jyi Sheu Portfolio Optimization

## Introduction-3

- Dynamic programming approach and martingale method are often used in the study of optimization problems in finance.
- The dynamic programming approach is useful for Markovian model.
- We can obtain an optimal portfolio by solving Hamilton-Jacobi-Bellman (HJB) equation.
- Martingale method is very useful when the market is complete.
   It avoids solving nonlinear (complicated) HJB equation.
- Martingale method can be applied in general (complete) diffusion models and general utility functions.

In this note we want to show the possibility to use dynamic programming approach together with martingale method to solve the problem.

- We use dynamic programming approach to derive HJB equation.
- Martingale method suggests a stochastic differential game problem.
- This suggests to rewrite HJB equation as Hamilton-Jacobi-Isaacs' equation of a game problem.

- An iteration procedure is used to solve the portfolio optimization problem.
- The market in each step is complete. Martigale method can be used.
- The limit is a solution.

The talk is based on discussions (and on going project) with Prof. Fleming (Brown University).

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### **General Utility Functions**

(C1) 
$$\sup_{(c,\pi)\in\mathcal{A}} E[\int_0^T e^{-\rho t} u(c_t X_t^{c,\pi}) dt + e^{-\rho T} u(X_T^{c,\pi})]$$

*u* is a utility function: *u* is increasing and concave.

- $X_t^{c,\pi}$  is the wealth with strategy  $(c,\pi)$ .
- $\pi_t$  is a trading strategy.  $c_t X_t^{c,\pi}$  is the rate of consumption.
- $\mathcal{A}$  is a family of admissible strategies.
- $\rho$  is the discount factor.

**Purpose:** To find an optimal strategy  $(\pi_t, c_t)$ .

### **Optimal Consumption Problem-3**

#### **CRRA Utility Functions:**

• 
$$u(c_t X_t^{c,\pi}) = \frac{1}{\gamma} (c_t X_t^{c,\pi})^{\gamma}, \ \gamma < 0, \gamma \neq 0.$$

• 
$$u(c_t X_t^{c,\pi}) = \log(c_t X_t^{c,\pi}), \gamma = 0.$$

We optimize

(C2) 
$$\sup_{(c,\pi)\in\mathcal{A}} E[\int_0^T e^{-\rho t} \frac{1}{\gamma} (c_t X_t^{c,\pi})^{\gamma} dt + e^{-\rho T} \frac{1}{\gamma} (X_T^{c,\pi})^{\gamma}],$$

### **Optimal Consumption Problem-2**

### **Expected Utility of Final Wealth**

No consumption c = 0.  $X_t^{\pi}$  is the wealth process with trading strategy  $\pi_t$ .

$$(E) \qquad \sup_{\pi \in \mathcal{A}} E[u(X_T^{\pi})]$$

### **Risk Sensitive Portfolio Optimization**

(RS) 
$$\sup_{\pi \in \mathcal{A}} E[\frac{1}{\gamma}(X_T^{\pi})^{\gamma}],$$

The consumption problem (C2) on infinite time horizon and the problem (RS) on infinite time has close relation.

Relation between consumption problems and rsik-sensitive protfolio optimization problems can be found in the papers.

#### **References:**

H. Hata and S. J. Sheu (2012), On the HJB eqution for an optimal consumption problem: I. Existence of Solution, Vol 50, 2373-2400. H. Hata and S. J. Sheu (2012), On the HJB eqution for an optimal consumption problem: II. Verification Theorem, Vol 50, 2401-2430.

Invest on N risky assets and a bank account.

• Prices of risky assets:  $S_i(t)$ ,  $i = 1, 2, \dots, N$ ,

$$dS_i(t) = S_i(t)(\mu^i(Y(t))dt + \sigma_P^{ij}(Y(t))dB_j(t)),$$

- Bank account: interest rate is r(Y(t)).
- Factor process:  $Y(t) = (y_1(t), \dots, y_n(t))$ . Dynamics:

$$dY(t) = b(Y(t))dt + \sigma_F(Y(t))dB(t).$$

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### Wealth Process

• Investment strategy  $\pi_t, c_t$ :

$$\pi(t)=(\pi_1(t),\pi_2(t),\cdots,\pi_N(t))$$

 $\pi_i(t)$  is the proportion of wealth on asset *i*.  $c_t X^{\pi,c}(t)$  is the consumption rate.  $X^{\pi,c}(t)$  is the wealth process.

• Dynamics of  $X^{\pi,c}(t)$ :

$$dX^{\pi,c}(t) = X^{\pi,c}(t) \{ \sum_{i=1}^{N} \pi_i(t) \frac{dS_i(t)}{S_i(t)} + (1 - \sum_{i=1}^{N} \pi_i(t)) \frac{dS_0(t)}{S_0(t)} - c_t dt \}.$$

### **Dynamics of Wealth Process**

$$dX^{\pi,c}(t) = X^{\pi,c}(t) \{ (\sum_{i} \pi_{i}(t)(\mu^{i}(Y(t)) - r(Y(t)) + r(Y(t)) - c(t))dt + \pi_{i}(t)\sigma_{P}^{ij}(Y_{t})dB_{j}(t) \}.$$

### **Dynamics of Factor Process**

$$dY(t) = b(Y(t))dt + \sigma_F(Y(t))dB(t).$$

### **Optimal Consumption Problem:**

Find 
$$(c, \pi)$$
 to optimize  
 $(C1)'$   
 $V(T, x, y) = \sup_{(c,\pi)\in\mathcal{A}} E[\int_0^T e^{-\rho t} u(c_t X_t^{c,\pi}) dt + e^{-\rho T} u(X_T^{c,\pi})].$ 

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Thaleia Zariphopoulou(2009), Optimal asset allocation in a stochastic factor model: an overview and open problems.

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- Dynamic programming approach has been extensively used in control theory.
- HJB (Hamilton-Jacobi-Bellman) equation can de derived. HJB is a nonlinear partial differential equation.
- A solution of HJB equation gives a candidate of optimal control.
- Merton (1970) implements the idea in continuous time models to derive an optimal trading strategy for an investment problem.

• Hamilton-Jacobi-Bellman (HJB) equation for  $V(\cdot)$ : (HJB)  $\frac{\partial V}{\partial t} = \frac{1}{2} \sum_{ij} a_F^{ij}(y) D_{ij} V + b(y)^* D_y V + \sup_{c,\pi} [u(cx) - \rho V + x\pi^* \sigma_F(y) \sigma_P(y)^* D_{xy} V + \frac{1}{2} x^2 \pi^* \sigma_P(y) \sigma_P^*(y) \pi D_{xx} V + xD_x V(x, y) \{r(y) + \pi^*(\mu(y) - r(y)\mathbf{1}) - c\}],$ 

 $c, \pi$  are taken  $c \in [0, \infty), \pi \in \mathbb{R}^m$ .

• From V(t, x, y), we define  $c^*(t, x, y)$ ,  $\pi^*(t, x, y)$  by

 $u'(cx) = D_x V(t, x, y)$ 

 $x^2 D_{xx} V \sigma_P(y) \sigma_P(y)^* \pi + x(\sigma_F(y) \sigma_P(y) D_{xy} V(t, x, y)$  $+ D_x V(t, x, y)(\mu(y) - r(y)\mathbf{1}) = 0.$ 

c\*(T - t, X(t), Y(t)), π\*(T - t, X(t), Y(t)) gives a candidate of optimal strategy.

The verification theorem is a theorem to verify this optimality.

The final form of the HJB Equation becomes:

$$\frac{\partial V}{\partial T} = \frac{1}{2} a_F^{ij}(y) D_{y_i y_j} V + g(y)^* D_{y_i} V + rx D_X V + u^* (D_X V) - \rho V$$
  
$$- \frac{1}{2} \frac{1}{D_{xx} V} (D_X V(\mu(y) - r(y)\mathbf{1}) + \sigma_P(y) \sigma_F^*(y) D_{xy} V)^*$$
  
$$\cdot a_P(y)^{-1} (D_X V(\mu(y) - r(y)\mathbf{1}) + \sigma_P(y) \sigma_F^*(y) D_{xy} V).$$

The initial condition is V(0, x, y) = u(x).

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### **CRRA Utility Function**

- We consider  $u(x) = \frac{1}{\gamma} x^{\gamma}, \gamma < 1, \ \gamma \neq 0.$
- Assuming

$$V(T, x, y) = \frac{x^{\gamma}}{\gamma} e^{W(T, y)}.$$

The equation for W(T, y):

$$(HJB') \quad \frac{\partial W}{\partial t} = \frac{1}{2} \sum_{ij} a_F^{ij}(y) D_{ij}W + H_{\gamma}(y, W(y), DW(y)).$$

• Candidate of optimal strategy:  $\pi^*(T - t, Y(t)), c^*(T - t, Y(t)),$ 

$$\pi^*(t, \mathbf{y}) = \frac{1}{1-\gamma} \mathbf{a}_P^{-1}(\mathbf{y})((\mu(\mathbf{y}) - \mathbf{r}(\mathbf{y})\mathbf{1}) + \sigma_P(\mathbf{y})\sigma_F^*(\mathbf{y})\nabla W(t, \mathbf{y})),$$

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$$c^*(t, y) = \exp(-\frac{W(t, y)}{1-\gamma}),$$

### Notations:

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$$H_{\gamma}(\boldsymbol{y}, \boldsymbol{w}, \boldsymbol{p}) = \hat{H}_{\gamma}(\boldsymbol{y}, \boldsymbol{p}) + (1 - \gamma) \exp(-\frac{\boldsymbol{w}}{1 - \gamma}) - \rho.$$

$$\hat{H}_{\gamma}(\boldsymbol{y},\boldsymbol{p}) = rac{1}{2}\sum_{ij=1}^{m}a_{\gamma}^{ij}(\boldsymbol{y})\boldsymbol{p}_{i}\boldsymbol{p}_{j} + \sum_{i=1}^{m}b_{\gamma}^{i}(\boldsymbol{y})\boldsymbol{p}_{i} + U_{\gamma}(\boldsymbol{y}),$$

$$a_F(y) = \sigma_F(y)\sigma_F^*(y), \quad a_P(y) = \sigma_P(y)\sigma_P^*(y).$$

$$b_{\gamma}(y) = b(y) + \frac{\gamma}{1-\gamma}\sigma_F(y)\sigma_P(y)^*a_P(y)^{-1}(\mu(y) - r(y)\mathbf{1}),$$

$$a_{\gamma}(y) = a_F(y) + \frac{\gamma}{1-\gamma}\sigma_F(y)\sigma_P(y)^*a_P(y)^{-1}\sigma_P(y)\sigma_F(y)^*,$$

$$\gamma(y) = \frac{\gamma}{2(1-\gamma)}(\mu(y) - r(y)\mathbf{1})^*a_P(y)^{-1}(\mu(y) - r(y)\mathbf{1}) + \gamma r(y)$$

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### **General Untility Functions**

- HJB equations for general utility functions are very complicated.
- How to study the equation is not well understood.
- Existence of the solution? Regularity of the solution?
- The candidate of an optimal strategy is a function of the derivatives of the solution.

The regularity of the solution is an important issue.

• For the existence of the solution, we may need to use the concept of viscosity solution.

### **CRRA Utility Functions**

There are several recent studies:

Nagai [1996, 2003], Fleming-Sheu[1999, 2000, 2002], Fleming-Hernandez[2003], Fleming-Pang[2004], Kaise-Sheu[2004,2006], Hata-Sheu[2012a, b], Hata-Nagai-Sheu[2013], Nagai[2013].

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There is a particular case that the equation is very simple.

- $\sigma_P(y)$  is invertible. The market is complete.
- HJB equation in the case of CRRA utility function becomes

$$\begin{array}{ll} \frac{\partial W}{\partial t} = & \frac{1}{2} \sum_{ij} a_F^{ij}(y) D_{ij}W + b_{\gamma}(y) DW + \frac{1}{2(1-\gamma)} DW^* a_F(y) DW \\ & + (1-\gamma) e^{-\frac{W}{1-\gamma}} + (U_{\gamma}(y) - \rho). \end{array}$$

•  $\psi(t, y) = \exp(\frac{W}{1-\gamma})$  satisfies a linear equation:

$$\frac{\partial \psi}{\partial t} = \frac{1}{2} \sum_{ij} a_F^{ij}(y) D_{ij} \psi + b_{\gamma}(y) D \psi \\ + \frac{1}{1-\gamma} (U_{\gamma} - \rho) \psi + 1.$$

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The solution of this linear equation is given by:

$$\psi(T, y) = E_{Y}[\int_{0}^{T} \exp(\frac{1}{1-\gamma} \int_{0}^{t} (U_{\gamma}(Y_{\gamma}(s)) - \rho) ds) dt \\ + \exp(\frac{1}{1-\gamma} \int_{0}^{T} (U_{\gamma}(Y_{\gamma}(s)) - \rho) ds)].$$
$$dY_{\gamma}(t) = b_{\gamma}(Y_{\gamma}(t)) dt + \sigma_{F}(Y_{\gamma}(t)) dB(t).$$

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- The problem is difficult to solve when the market is incomplete.
- The idea of the market completion is developed in Pages[1987], He-Pearson[1990], Karatzas-Lehoczky-Shreve-Xu[1991].
- The idea of "fictitious" completion is discussed in Karatzas-Lehoczky-Shreve-Xu[1991].

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Consider the wealth process

$$\frac{dX^{\pi,c}(t)}{X^{\pi,c}(t)} = \{r(Y(t)) + \pi(t)^*(\mu(Y(t)) - r(Y(t))\mathbf{1}) \\ -c(t))dt\} + \pi(t)^*\sigma_P(Y_t)dB_j(t).$$

• We write 
$$\bar{\pi}(t) = \sigma_P(Y(t))^* \pi(t)$$
.

- We define  $\bar{\mu}(y) r(y)\mathbf{1} = \sigma_P(y)^* a_P(y)^{-1} (\mu(y) r(y)\mathbf{1}).$
- The wealth process has equation

$$\frac{dX^{\pi,c}(t)}{X^{\pi,c}(t)} = \{r(Y(t)) + \bar{\pi}(t)^*(\bar{\mu}(Y(t)) - r(Y(t))\mathbf{1}) \\ -c(t)\}dt + \bar{\pi}(t)^*dB(t).$$

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From

$$\bar{\pi}(t) = \sigma_P(Y(t))^* \pi(t),$$

we have

$$\sigma_{\mathcal{P}}(\mathbf{Y}(t))\bar{\pi}(t) = \sigma_{\mathcal{P}}(\mathbf{Y}(t))\sigma_{\mathcal{P}}(\mathbf{Y}(t))\pi(t) = \mathbf{a}_{\mathcal{P}}(\mathbf{Y}(t))\pi(t).$$

That is,

$$\pi(t) = a_P(Y(t))^{-1} \sigma_P(Y(t)) \overline{\pi}(t).$$

We also have

$$\pi(t)^{*}(\mu(Y(t)) - r(Y(t))\mathbf{1}) = \bar{\pi}(t)^{*}\sigma_{P}(Y(t))^{*}a_{P}(Y(t))^{-1}(\mu(Y(t)) - r(Y(t))\mathbf{1}) = \bar{\pi}(t)^{*}(\bar{\mu}(Y(t)) - r(Y(t))\mathbf{1}).$$

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The weath process

(1) 
$$\frac{dX(t)}{X(t)} = \{r(Y(t)) + \bar{\pi}(t)^* (\bar{\mu}(Y(t)) - r(Y(t))\mathbf{1}) - c(t)\}dt + \bar{\pi}(t)^* dB(t).$$

This corresponds to market with the stocks

(2) 
$$\frac{dS^{j}(t)}{S^{j}(t)} = \bar{\mu}^{j}(Y(t))dt + dB_{j}(t), \ j = 1, 2, \cdots, d.$$

- The market with price of stocks in (2) is complete.
- The original price processes are special cases of (1) with  $\bar{\pi}(t) = \sigma_P(Y(t))^* \pi(t)$ .
- In some sense, we complete the market from (S) to (S).
   Then there are more choices of the portfolio.

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- We show such complete market is equivalent to a market completion by Karatzas-Lehoczky-Shreve-Xu[1991].
- They introduce the "fictitious" stocks:

(3) 
$$\frac{d\tilde{S}^{j}(t)}{\tilde{S}^{j}(t)} = r(Y(t))dt + \sigma_{f}^{jk}(Y(t))dB_{k}(t), \ j = 1, 2, \cdots, d,$$

•  $\sigma_f(y) = I - \sigma_P(y)^* (\sigma_P(y)\sigma_P(y)^*)^{-1} \sigma_P(y).$ 

• Special properties of  $\sigma_f(y)$ :

(i) 
$$\sigma_f(y)\sigma_f(y) = \sigma_f(y).$$
  
(ii)  $\sigma_p(y)\sigma_f(y) = 0, \ \sigma_f(y)\sigma_P(y)^* = 0.$ 

- We have *m* stocks  $S^{i}(t)$  and *d* stocks  $\tilde{S}^{j}(t)$ .
- A portfolio is given by c, π, π̃.
   π<sup>i</sup>(t) is the proportion of the wealth in S<sup>i</sup>(t).
   π<sup>j</sup>(t) is the proportion of the wealth in Š<sup>j</sup>(t).

• The wealth process  $\tilde{X}(t) = \tilde{X}^{x,c,\pi,\tilde{\pi}}(t)$  has the dynamics,

$$\frac{d\tilde{X}(t)}{\tilde{X}(t)} = \{r(Y(t)) + \pi(t)^* (\mu(Y(t)) - r(Y(t))\mathbf{1}) - c(t)\} dt \\ + (\sigma_P(Y(t))^* \pi(t) + \sigma_f(Y(t))\tilde{\pi}(t))^* dB_t$$

- There is one-one correspondence for trading strategies for  $(\overline{S}^{i}(t))$  and for  $(S^{i}(t), \tilde{S}^{j}(t))$  with the same wealth process.
- Hence two markets are equivalent.
- For each  $(\pi(t), \tilde{\pi}(t))$ , we have

$$\bar{\pi}(t) = \sigma_{\mathcal{P}}(\mathbf{Y}(t))^* \pi(t) + \sigma_f(\mathbf{Y}(t))\tilde{\pi}(t).$$

• Using properties of  $\sigma_f$  above, we have

 $\sigma_P(Y(t))\overline{\pi}(t) = \sigma_P(Y(t))\sigma_P(Y(t))^*\pi(t) = a_P(Y(t))\pi(t).$ 

Then

$$\pi(t) = a_P(Y(t))^{-1} \sigma_P(Y(t)) \bar{\pi}(t).$$

We can see

$$ilde{X}^{x,c,\pi, ilde{\pi}}(t)=ar{X}^{x,c, ilde{\pi}}(t)$$
 is a set of the set

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Portfolio Optimization

### The First Observation:

- Market with price (S
   <sup>S</sup>) (or (S, S
   <sup>S</sup>)) is complete. The wealth process is X
   <sup>π,c</sup> (or X
   <sup>π,π,c</sup>).
- Market with price (S) is not complete.
   The wealth process is X<sup>π,c</sup>.
- For each  $\pi$ , we take  $\bar{\pi}(t) = \sigma_P(Y(t))\pi(t)$ , we have

$$X^{\pi,c}=\bar{X}^{\bar{\pi},c}.$$

• The value function satisfies  $V(T, x, y) \leq \overline{V}(T, x, y) = \widetilde{V}(T, x, y).$ 

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### The Second Observation:

- (S, S̃) is a special case of "fictitious" completion of the market in Karatzas-Lehoczky-Shreve-Xu[1991].
- Let  $\theta(t)$  be a  $R^d$ -valued progressive measurable process.
- Using this, we can construct additional stocks for trading.

$$\frac{d\tilde{S}_{\theta}^{j}(t)}{\tilde{S}_{\theta}^{j}(t)} = (r(Y(t)) + (\sigma_{f}(Y(t))\theta(t))^{j})dt + \sigma_{f}(Y(t))^{jk}dB_{k}(t)),$$

$$j=1,2,\cdots,d.$$

•  $\pi^{i}(t)$  is the proportion of wealth on  $S^{i}$ stock.  $\tilde{\pi}^{j}(t)$  the proportion of the wealth on the  $\tilde{S}_{\theta}^{j}$  stock.

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• The wealth process  $ilde{X}_{ heta}(t) = ilde{X}^{x,c,\pi, ilde{\pi}}(t)$ , has the dynamics,

$$\frac{d\tilde{X}_{\theta}(t)}{\tilde{X}_{\theta}(t)} = \{r(Y(t)) + (\pi(t)^*(\mu(Y(t)) - r(Y(t))\mathbf{1}) + \tilde{\pi}(t)^*\sigma_f(Y(t))\theta_t \\ -c(t)\}dt + (\sigma_P(Y(t))^*\pi_t + \sigma_f(Y(t))\tilde{\pi}(t))^*dB(t)$$

- Denote  $\bar{\pi}(t) = \sigma_P(Y(t))^* \pi(t) + \sigma_f(Y(t))\tilde{\pi}(t)$ .
- The dynamics is rewritten as

$$\frac{dX_{\theta}(t)}{\tilde{X}_{\theta}(t)} = \{ r(Y(t)) + (\bar{\pi}(t)^*(\bar{\mu}(Y(t)) - r(Y(t))\mathbf{1}) \\ + \bar{\pi}(t)^*\sigma_f(Y(t))\theta(t) - c(t) \} dt + \bar{\pi}(t)^* dB(t)$$

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• The dynamices for the wealth process  $\tilde{X}_{\theta}(t) = \tilde{X}^{x,c,\pi,\tilde{\pi}}(t)$ :

$$\frac{d\tilde{X}_{\theta}(t)}{\tilde{X}_{\theta}(t)} = \{r(Y(t)) + (\pi(t)^*(\mu(Y(t)) - r(Y(t))\mathbf{1}) + \tilde{\pi}(t)^*\sigma_f(Y(t))\theta_t \\ -c(t)\}dt + (\sigma_P(Y(t))^*\pi_t + \sigma_f(Y(t))\tilde{\pi}(t))^*dB(t)$$

 This is the wealth process for the market with stock price dynamics

$$\frac{d\bar{S}_{\theta}^{j}(t)}{\bar{S}_{\theta}^{j}(t)} = (\bar{\mu}^{j}(\boldsymbol{Y}(t)) + (\sigma_{f}(\boldsymbol{Y}(t))\theta(t))^{j})dt + dB_{j}(t), \ j = 1, 2, \cdots, d.$$

• Markets with stock prices  $(S, \tilde{S}_{\theta})$  and  $(\bar{S}_{\theta})$  are equivalent.

- We denote  $\bar{V}^{\theta}$  the value function for the consumption problem in  $\theta$ -market.
- Karatzas-Lehoczky-Shreve-Xu[1991] uses the duality argument to derive the fundamental relation:

$$V(T, x, y) = \inf_{\theta} \overline{V}^{\theta}(T, x, y).$$

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## **Issacs' Equation-7**

The relation

$$V(T, x, y) = \inf_{\theta} \overline{V}^{\theta}(T, x, y).$$

suggests the following result.

**Lemma 1.** The HJB equation for V(T, x, y) can be rewritten as

$$\frac{\partial V}{\partial T} = \frac{1}{2} a_F(y)^{ij} D_{y_i y_j} V + b(y)^* D_{y_i} V + xr(y) D_x V \\ + \inf_{\theta} sup_{\overline{\pi},c} \{ u(xc) - \rho V + \frac{1}{2} |\overline{\pi}|^2 x^2 D_{xx} V \\ + x(\sigma_F(y)\overline{\pi})^* D_{xy} V + x(\overline{\pi}^*(\overline{\mu}(y) - r(y)\mathbf{1} + \sigma_f(y)\theta) - c) D_x V \}.$$

### HJB equation:

$$\frac{\partial V}{\partial t} = \frac{1}{2} \sum_{ij} a_F^{ij}(y) D_{ij} V + b(y)^* D_y V + \sup_{c,\pi} \left[ u(cx) - \rho V \right]$$

 $(HJB) + x\pi^*\sigma_F(y)\sigma_P(y)^*D_{xy}V + \frac{1}{2}x^2\pi^*\sigma_P(y)\sigma_P^*(y)\pi D_{xx}V$ 

$$+xD_xV(x,y)\{r(y)+\pi^*(\mu(y)-r(y)\mathbf{1})-c\}],$$

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- This is the Issacs' equation for a stochastic differential game.
- We can also verify Issacs' condition. That is,

$$\inf_{\theta} \sup_{c,\pi} \{\cdots\} = \sup_{c,\pi} \inf_{\theta} \{\cdots\}.$$

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## **Issacs' Equation-5**

The  $\theta$  to take inf will be important for the following analysis, we give a proof of Lemma 1.

• In  $\inf_{\theta} \sup_{c,\pi} \{\cdots\}$ , we fix  $\theta$ . Take supremum on  $c, \pi$ . The result is

$$\frac{1}{-2D_{xx}V}|\sigma_F(y)^*D_{xy}V + ((\bar{\mu}(y) - r(y)\mathbf{1}) + \sigma_f(y)\theta)D_xV|^2$$

$$+xr(y)D_xV+u^*(D_xV)-\rho V.$$

• The  $\bar{\pi}$  taking supremum is equal to

$$\bar{\pi} = -\frac{1}{D_{xx}V}(\sigma_F(y)^* D_{xy}V + ((\bar{\mu}(y) - r(y)\mathbf{1}) + \sigma_f(y)\theta)D_xV).$$

• We need condition  $D_{xx} V < 0$ . V(T, x, y) is concave in x.

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## **Issacs' Equation-4**

• We now choose  $\theta$  to minimize

$$\frac{1}{-2D_{xx}V}|\sigma_F(y)^*D_{xy}V + ((\bar{\mu}(y) - r(y)\mathbf{1}) + \sigma_f(y)\theta)D_xV|^2 + xr(y)D_xV + u^*(D_xV) - \rho V.$$

We write

 $\sigma_F(\mathbf{y})^* D_{\mathbf{x}\mathbf{y}} \mathbf{V} + (\bar{\mu}(\mathbf{y}) - \mathbf{r}(\mathbf{y})\mathbf{1}) D_{\mathbf{x}} \mathbf{V} = \sigma_P(\mathbf{y})^* \pi + \sigma_f(\mathbf{y}) \tilde{\pi}.$ 

 $\pi$  and  $\tilde{\pi}$  will be given later.

We need to minimize

$$\begin{aligned} &|\sigma_{\mathcal{P}}(\boldsymbol{y})^*\pi + \sigma_f(\boldsymbol{y})(\tilde{\pi} + \theta D_X \boldsymbol{V})|^2 \\ &= |\sigma_{\mathcal{P}}(\boldsymbol{y})^*\pi|^2 + |\sigma_f(\boldsymbol{y})(\tilde{\pi} + \theta D_X \boldsymbol{V})|^2 \end{aligned}$$

Here we use  $\sigma_P \sigma_f = \mathbf{0} = \sigma_f \sigma_P^*$ .

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# **Issacs' Equation-3**

• The minimum satisfies  $\tilde{\pi} + \theta D_X V = 0$ .

$$\theta = -\frac{1}{D_x V} \tilde{\pi}$$

•  $\inf_{\theta} \sup_{c,\pi} \{\cdots\}$  is equal to

$$\frac{1}{-2D_{xx}V}|\sigma_P(y)^*\pi|^2+xr(y)D_xV+u^*(D_xV)-\rho V.$$

•  $\pi, \tilde{\pi}$  solves

$$\sigma_F(y)^* D_{xy} V + (\bar{\mu}(y) - r(y)\mathbf{1}) D_x V = \sigma_P(y)^* \pi + \sigma_f(y) \tilde{\pi}.$$

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# Issacs' Equation-2

• We now solve the following relation for  $\pi, \tilde{\pi}$ :

 $\sigma_{\mathsf{F}}(y)^* D_{xy} V + (\bar{\mu}(y) - r(y)\mathbf{1}) D_x V = \sigma_{\mathsf{P}}(y)^* \pi + \sigma_f(y) \tilde{\pi}.$ 

We have

$$\pi = a_P(y)^{-1} \sigma_P(y) (\sigma_F(y)^* D_{xy} V + (\bar{\mu}(y) - r(y)\mathbf{1}) D_x V)$$
  
$$\sigma_f(y) \tilde{\pi} = \sigma_f(y) (\sigma_F(y)^* D_{xy} V + (\bar{\mu}(y) - r(y)\mathbf{1}) D_x V)$$
  
$$= \sigma_f(y) \sigma_F(y)^* D_{xy} V.$$

• In the second relation, we use

$$\sigma_f(y)(\bar{\mu}(y)-r(y)\mathbf{1})=\mathbf{0}.$$

This follows from the definition of

$$\bar{\mu}(\mathbf{y}) - \mathbf{r}(\mathbf{y})\mathbf{1} = \sigma_{P}(\mathbf{y})\mathbf{a}_{P}(\mathbf{y})^{-1}(\mu(\mathbf{y}) - \mathbf{r}(\mathbf{y})\mathbf{1})$$

and  $\sigma_f \sigma_P = \mathbf{0} = \sigma_P^* \sigma_f$ .

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- We use the expression of  $\pi$  to check Issacs' equation.
- We use the expression of  $\tilde{\pi}$  to obtain

$$\theta = -\frac{1}{D_x V} \sigma_f(y) \sigma_F(y)^* D_{xy} V.$$

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In the following, we want to show a use of this game interpretation of HJB equation.

• In the rest we consider CRRA utility functions:

$$u(x) = u_{\gamma}(y) = \frac{1}{\gamma}x^{\gamma},$$

where  $\gamma < 1$  and for  $\gamma = 0$  we understand  $u_0(x) = \log x$ .

 In the following, we consider 0 < γ < 1. The case γ < 0 can be similarly treated.</li>

 In this case the value function of the consumption problem has the form

$$V(T, x, y) = \frac{1}{\gamma} x^{\gamma} \exp(W(T, y)).$$

• The equation for W(T, y) is given by

$$\frac{\partial W}{\partial T} = \frac{1}{2} a_{\mathsf{F}}(y)^{ij} D_{y_i y_j} W + \frac{1}{2} Q_{\gamma}(y, DW) \\ + U_{\gamma}(y) - \rho + (1 - \gamma) \exp(-\frac{W}{1 - \gamma})$$

W(0, y) = 1.

Here

$$Q_{\gamma}(\boldsymbol{y},\boldsymbol{p}) = \boldsymbol{p}^* \sigma_{\boldsymbol{F}}(\boldsymbol{y}) (\boldsymbol{I} + \frac{\gamma}{1-\gamma} \sigma_{\boldsymbol{P}}(\boldsymbol{y})^* \boldsymbol{a}_{\boldsymbol{P}}(\boldsymbol{y})^{-1} \sigma_{\boldsymbol{P}}(\boldsymbol{y})) \sigma_{\boldsymbol{F}}(\boldsymbol{y})^* \boldsymbol{p},$$

 $U_{\gamma}(y) = \frac{\gamma}{2(1-\gamma)}(\mu(y) - r(y)\mathbf{1})^* a_{\mathcal{P}}(y)^{-1}(\mu(y) - r(y)\mathbf{1}) + \gamma r(y).$ 

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- We return to the fundamental relation  $V(T, x, y) = \inf_{\theta} \overline{V}(T, x, y).$
- Is it possible to choose θ in a smart way that the value is close to V(T, x, y)?
- We shall try an iteration procedure so that we can gradually reach this purpose.

- We now consider the Markovian fictitious completion,  $\theta(t) = \theta(t, Y(t))$ .
- The stock price dynamics

$$\frac{d\bar{S}_{\theta}^{j}(t)}{\bar{S}_{\theta}^{j}(t)} = (\bar{\mu}^{j}(Y(t)) + (\sigma_{f}(Y(t))\theta(t,Y(t)))^{j})dt + dB_{j}(t), j = 1, 2, \cdots, d.$$

Factor process dynamics

$$dY(t) = b(Y(t))dt + \sigma_F(Y(t))dB(t).$$

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#### • The wealth process dynamics :

$$\frac{d\tilde{X}_{\theta}(t)}{\tilde{X}_{\theta}(t)} = \{ r(Y(t)) + (\bar{\pi}(t)^*(\bar{\mu}(Y(t)) - r(Y(t))\mathbf{1}) \\ + \bar{\pi}(t)^*\sigma_f(Y(t))\theta(t) - c(t) \} dt + \bar{\pi}(t)^* dB(t).$$

The value function is

$$\overline{V}^{(\theta)}(t, x, y) = \sup_{\pi, c} E_{t, y} [\int_t^T \exp(-\rho(t-s)) \frac{1}{\gamma} (c(s) X^{c, \overline{\pi}}(s))^{\gamma} ds \\ + \exp(-\rho(T-t)) \frac{1}{\gamma} (X^{c, \overline{\pi}}(T))^{\gamma}].$$

• Here  $E_{t,y}[\cdots]$  indicates Y(t) = y.

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Assume

$$\overline{V}^{ heta}(t, x, y) = rac{1}{\gamma} x^{\gamma} \exp(\overline{W}^{ heta}(t, y)).$$

The HJB equation: (HJB-θ)

$$\frac{\partial \overline{W}^{\theta}}{\partial t} + \frac{1}{2} a_{F}(y)^{ij} D_{y_{i}y_{j}} \overline{W}^{\theta} + \frac{1}{2(1-\gamma)} a_{F}(y)^{ij} D_{y_{i}} \overline{W}^{\theta} D_{y_{j}} \overline{W}^{\theta} \\ + b_{\gamma}^{\theta}(t, y)^{*} D_{y} \overline{W}^{\theta} + U_{\gamma}^{\theta}(t, y) - \rho + (1-\gamma) \exp(-\frac{\overline{W}^{\theta}}{1-\gamma}) = 0$$

Here

$$\begin{aligned} b_{\gamma}^{\theta}(t,y) &= b(y) + \frac{\gamma}{1-\gamma} \sigma_{\mathsf{F}}(y) (\overline{\mu}(y) - r(y)\mathbf{1} + \sigma_{\mathsf{f}}(y)\theta(t,y)) \\ U_{\gamma}^{\theta}(t,y) &= \frac{\gamma}{2(1-\gamma)} |\overline{\mu}(y) - r(y)\mathbf{1} + \sigma_{\mathsf{f}}(y)\theta(t,y)|^2 + \gamma r(y). \end{aligned}$$

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#### Define

$$\overline{\psi}^{\theta}(t, y) = \exp(\frac{1}{1 - \gamma} \overline{W}^{\theta}(t, y)).$$

We have

$$\begin{aligned} & \frac{\partial \overline{\psi}^{\theta}}{\partial t} + \frac{1}{2} a_{F}(y)^{ij} \mathcal{D}_{y_{i}y_{j}} \overline{\psi}^{\theta} + b_{\gamma}^{\theta}(t, y)^{*} \mathcal{D}_{y} \overline{\psi}^{\theta} \\ & + \frac{1}{1 - \gamma} (\mathcal{U}_{\gamma}^{\theta}(t, y) - \rho) \overline{\psi}^{\theta} + 1 = 0. \end{aligned}$$

$$\overline{\psi}^{\theta}(T, y) = 1.$$

A solution is given by

$$\overline{\psi}^{\theta}(t, y) = E_{t,y} [\int_t^T \exp(\frac{1}{1-\gamma} \int_t^s (U_{\gamma}(u, Y^{\gamma}(u)) - \rho) du) ds \\ + \exp(\frac{1}{1-\gamma} \int_t^T (U_{\gamma}(u, Y^{\gamma}(u)) - \rho) du)].$$

Here

$$dY^{\gamma}(s) = b^{\theta}_{\gamma}(s, Y^{\gamma}(s))ds + \sigma_{F}(Y^{\gamma}(s))dB(s), t < s, \ Y^{\gamma}(t) = y.$$

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Assume a smooth solution ψ<sup>θ</sup>(t, y).
 A candidate of the optimal strategy is given by

$$\overline{c}^{\theta}(t) = \exp(-\frac{1}{1-\gamma}\overline{W}^{\theta}(t, Y(t)))$$

$$\overline{\pi}^{\theta}(t) = \frac{1}{1-\gamma} ((\overline{\mu}(Y(t)) - r(Y(t))\mathbf{1}) + \sigma_f(Y(t))\theta(t, Y(t)) + \sigma_F(Y(t))^* D\overline{W}^{\theta}(t, Y(t))).$$

 Now we want to use the solution for θ-market to switch to another θ so that we can get a better value (for the approximation of the original problem).

Issacs' equation suggests to change to

$$\theta^{(1)}(t, y) = -\sigma_f(y)\sigma_F(y)^* D\overline{W}^{\theta}(t, y).$$

• With this particular  $\theta(t, y) = \theta^{(1)}(t, y)$ , the value function is denoted by

$$\overline{V}_1(t,x,y) = \frac{1}{\gamma} x^{\gamma} \exp(\overline{W}_1(t,y)).$$

•  $\overline{W}_1(t,y) := \overline{W}^{\theta^{(1)}}(t,y).$ 

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•  $\overline{W}_1$  satisfies (HJB- $\theta^{(1)}$ ):

$$\frac{\partial \overline{W}_1}{\partial t} + \frac{1}{2} a_F(y)^{ij} D_{y_i y_j} \overline{W}_1 + \frac{1}{2(1-\gamma)} a_F(y)^{ij} D_{y_i} \overline{W}_1 D_{y_j} \overline{W}_1 \\ + b^{(1)}(t, y)^* D_y \overline{W}_1 + U^{(1)}(t, y) - \rho + (1-\gamma) \exp(-\frac{\overline{W}_1}{1-\gamma}) = 0.$$

#### Here

$$b^{(1)}(t,y)=b^{ heta}_{\gamma}(y), \ U^{(1)}(t,y)=U^{ heta}_{\gamma}(y)$$

with  $\theta = \theta^{(1)}(t, y)$ .

**Lemma 2.**  $\overline{W}^{\theta}(t, y)$  is a supersolution (subsolution) of (HJB- $\theta^{(1)}$ ) if  $\gamma > 0$  (if  $\gamma < 0$ ).

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#### Assumption (A)

(A1)  $\sigma_F^{ij}(\cdot), \sigma_P^{ij}(\cdot), b^i(\cdot), \mu^i(\cdot), r(\cdot)$  are smooth functions.

(A2) For any r > 0, there are  $c_1(r), c_2(r) > 0$  such that for all  $y \in B_r$ , we have

$$\begin{aligned} \mathbf{c}_1(r)\mathbf{I}_m &\leq \sigma_P(y)\sigma_P(y)^* \leq \mathbf{c}_2(r)\mathbf{I}_m, \\ \mathbf{c}_1(r)\mathbf{I}_n &\leq \sigma_F(y)\sigma_F(y)^* \leq \mathbf{c}_2(r)\mathbf{I}_n. \end{aligned}$$

(A3) r(y) > 0 for all y.

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**Theorem 2.** Assume (A),  $\theta(t, y)$  is a smooth function such that  $\theta(T, y) = 0$  for all *y*. Assume HJB equation has smooth subsolution  $\underline{W}$  and (HJB- $\theta$ ) has a smooth solution  $\overline{W}^{\theta}$  such that

$$\underline{W}(T-t,y) \leq \overline{W}^{\theta}(t,y), \ t \in [0,T].$$

We take

$$\theta^{(1)}(t,y) = -\sigma_f(y)\sigma_F(y)^* D\overline{W}^{\theta}(T-t,y).$$

Then (HJB- $\theta^{(1)}$ ) has a smooth solution  $\overline{W}^{\theta^{(1)}}$  satisfying

$$\underline{W}(T-t,y) \leq \overline{W}^{\theta^{(1)}}(t,y) \leq \overline{W}^{\theta}(t,y).$$

**Theorem 3.** Assume (A),  $\theta(t, y)$  is a smooth function such that  $\theta(T, y) = 0$  for all *y*. Assume (HJB) has subsolution  $\underline{W}$  and (HJB- $\theta$ ) has a solution  $\overline{W}^{\theta}$  such that

$$\underline{W}(T-t,y) \leq \overline{W}^{\theta}(t,y), \ t \in [0,T].$$

Then (HJB) has a solution

$$\underline{W}(T-t,y) \leq W(t,y) \leq \overline{W}^{\theta}(t,y).$$

#### Assumption (A)'

(A1)'  $\sigma_F^{ij}(\cdot), \sigma_P^{ij}(\cdot), b^i(\cdot), \mu^i(\cdot), r(\cdot)$  are smooth functions. (A2)' There are  $c_1, c_2 > 0$  such that for all  $y \in \mathbb{R}^n$ 

$$c_1 I_m \leq \sigma_P(y) \sigma_P(y)^* \leq c_2 I_m, \\ c_1 I_n \leq \sigma_F(y) \sigma_F(y)^* \leq c_2 I_n.$$

- (A3)' There is  $r_0 > 0$  such that  $r(y) \ge r_0$  for all y.
- (A4)' The functions in (A1)' are bounded with bounded first order derivatives.

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- We consider  $\theta = 0$ .
- The (HJB-0) is given by

$$\frac{\partial W}{\partial T} = \frac{1}{2} a_F(y)^{ij} D_{y_i y_j} W + \frac{1}{2(1-\gamma)} a_F(y)^{ij} D_{y_i} W D_{y_j} W \\ + b_{\gamma}(y)^* D_y W + U_{\gamma}(y) - \rho(1-\gamma) \exp(-\frac{W}{1-\gamma}).$$

#### Here

$$b_{\gamma}(y) = b(y) + \frac{\gamma}{1-\gamma}\sigma_{F}(y)(\overline{\mu}(y) - r(y)\mathbf{1})$$
  
$$U_{\gamma}(y) = \frac{\gamma}{2(1-\gamma)}|\overline{\mu}(y) - r(y)\mathbf{1}|^{2} + \gamma r(y).$$

**Lemma 4.** We assume the condition (A)'. Then (HJB-0) has a solution given by  $W_0(T, y) = (1 - \gamma) \log \psi_0(T, y)$ , where

$$\begin{split} \psi_0(\mathcal{T}, \mathbf{y}) = & \mathcal{E}[\int_0^{\mathcal{T}} \exp(\frac{1}{1-\gamma} \int_0^t (\mathcal{U}_{\gamma}(\mathbf{Y}(s)) - \rho) ds) \\ & + \exp(\frac{1}{1-\gamma} \int_0^{\mathcal{T}} (\mathcal{U}_{\gamma}(\mathbf{Y}(s)) - \rho) ds)]. \end{split}$$

 $W_0(T, y)$  satisfies

$$W_0(T, y) \leq (1 - \gamma)(\log(\frac{1 + c}{c} + cT)),$$

where  $c = \sup_{y} \{ \frac{1}{1-\gamma} U_{\gamma}(y) \}$ . (HJB) has a subsolution given by

$$\underline{W}(T, y) = (1 - \gamma) \log \left\{ \frac{1 - \gamma}{\rho} (1 - \exp(-\frac{\rho}{1 - \gamma}T)) + \exp(-\frac{\rho}{1 - \gamma}T) \right\}.$$

Moreover, we have  $\underline{W}(T, y) \leq W_0(T, y), y \in \mathbb{R}^n$ ,  $T \geq 0$ .

**Theorem 5.** Assume (A)', we can define recursively  $\theta_n(t, y)$  and  $W_n$  such that  $\theta_0 = 0$ ,

$$\begin{aligned} \theta_{n+1}(t,y) &= -\sigma_f(y)\sigma_F(y)^* DW_n(t,y), \\ W_n(t,y) &= E_{t,y}[\int_t^T \exp(\frac{1}{1-\gamma}\int_t^s (U^{(n)}(u,Y^{(n)}(u)) - \rho) du) ds \\ &+ \exp(\frac{1}{1-\gamma}\int_t^T (U^{(n)}(u,Y^{(n)}(u)) - \rho) du)], \end{aligned}$$

where

$$egin{aligned} &U^{(n)}(s,y) = U^{ heta_n}_{\gamma}(s,y), b^{(n)}(s,y) = b^{ heta_n}_{\gamma}(s,y) \ &dY^{(n)}(s) = b^{(n)}(s,Y^{(n)}(s)) ds + \sigma_F(Y^{(n)}(s)) dB(s). \end{aligned}$$

Then  $W_n$  converges to W. W is the unique solution of (HJB) such that W is bounded.

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## Remarks

- Hata-Nagai-Sheu [2013] proves the existence and uniqueness of HJB equation under some condition of the cofficients.
- Similar ideas may be applied for general utility functions.
- In such case, the equation (HJB-θ) may not have smooth solution even if θ is Markovian and smooth.
   We may need to use martingale method (for complete market) to find solution for each Markovian θ.
- The detail analysis shall be interesting.
- The incompleteness due to the portfolio constrains or transaction costs may be also interesting cases to implement such idea.