

5.4.4 Uniqueness of the Risk-Neutral Measure

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Outline:

- 1 Completeness of a market model

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- 2 Second fundamental theorem of asset pricing

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- 3 Lemma of Second fundamental theorem

Completeness of a market model

Theorem 5.4.8

A market model is complete if every derivative security can be hedged.

Suppose we have a market model with :

- 1 A filtration generated by a d -dimensional Brownian motion
- 2 A risk-neutral measure \tilde{P}
 - We have solved the market price of risk equations

$$\alpha_i(t) - R(t) = \sum_{j=1}^d \sigma_{ij}(t) \Theta_j(t), \quad i = 1, \dots, m$$

- Using the resulting market prices of risk $\Theta_1(t), \dots, \Theta_d(t)$ to define the Radon-Nikodym derivative process $Z(t)$.

- 2 A risk-neutral measure \tilde{P}
- have changed to the measure \tilde{P} under which $\tilde{W}(t)$ defined by

$$\tilde{W}(t) = W(t) + \int_0^t \Theta(u) du$$

is a d-dimensional Brownian motion.

- 3 A $F(T)$ -measurable random variable $V(T)$, which is the payoff of some derivative security.

- We would like to be sure we can hedge a short position in the derivative security whose payoff at time T is $V(T)$
- We can define $V(t)$

$$V(t) = \tilde{E} \left[e^{-\int_t^T R(u)du} V(T) | F(t) \right]$$

so that $D(t)V(t)$ satisfies

$$D(t)V(t) = \tilde{E} [D(T)V(T) | F(t)]$$

and just as in

$$\begin{aligned} \tilde{E} [D(t)V(t) | F(s)] &= D(s)V(s) \\ 0 \leq s \leq t \leq T \end{aligned}$$

- We see that $D(t)V(t)$ is a martingale under \tilde{P} .
- According to the Martingale Representation , there are processes $\tilde{\Gamma}_1(u), \dots, \tilde{\Gamma}_d(u)$ such that

$$D(t)V(t) = V(0) + \sum_{j=1}^d \int_0^T \tilde{\Gamma}_j(u) d\tilde{W}(u)$$

- Consider a portfolio value process that begins at $X(0)$. According to

$$d(D(t)X(t)) = \sum_{i=1}^m \Delta_i(t) d(D(t)S_i(t))$$

and

$$d(D(t)S_i(t)) = D(t)S_i(t) \sum_{j=1}^d \sigma_{ij} d\widetilde{W}_j(t)$$

- A portfolio value process would be

$$\begin{aligned}d(D(t)X(t)) &= \sum_{i=1}^m \Delta_i(t) d(D(t)S_i(t)) \\ &= \sum_{j=1}^d \sum_{i=1}^m \Delta_i(t) D(t) S_i(t) \sigma_{ij} d\widetilde{W}_j(t)\end{aligned}$$

Or, equivalently,

$$D(t)X(t) = X(0) + \sum_{j=1}^d \int_0^T \sum_{i=1}^m \Delta_i(t) D(t) S_i(t) \sigma_{ij} d\widetilde{W}_j(t)$$

- Comparing

$$D(t)V(t) = V(0) + \sum_{j=1}^d \int_0^T \tilde{\Gamma}_j(u) d\tilde{W}(u)$$

and

$$D(t)X(t) = X(0) + \sum_{j=1}^d \int_0^T \sum_{i=1}^m \Delta_i(t) D(t) S_i(t) \sigma_{ij} d\tilde{W}_j(t)$$

- To hedge the short position, we should take $X(0) = V(0)$ and choose the portfolio processes $\Delta_1(t), \dots, \Delta_m(t)$, so that the hedging equations

$$\frac{\tilde{\Gamma}_j(t)}{D(t)} = \sum_{i=1}^m \Delta_i(t) S_i(t) \sigma_{ij} d\tilde{W}_j(t)$$

are satisfied.

- These are d equations in m unknown processes

$$\Delta_1(t), \dots, \Delta_m(t)$$

2nd fundamental theorem of asset pricing

Second fundamental theorem of asset pricing

Consider a market model that has a risk-neutral probability measure.

- The model is complete if and only if the risk-neutral probability measure is unique.

- Assume that the model is complete.
- We wish to show that there can be only one risk-neutral measure.

Model is complete $\implies Q$ measure is unique

- Suppose the model has two risk-neutral measures, \tilde{P}_1 and \tilde{P}_2 .
- Let A be a set in F , which we assumed at the beginning of this section is the same as $F(T)$.
- Consider the derivative security with payoff

$$V(T) = I_A \frac{1}{D(T)}$$

(Cont.) Model is complete $\implies Q$ measure is unique

- Because the model is complete, a short position in this derivative security can be hedged.
(there is a portfolio value process with some initial condition $X(0)$ that satisfies $X(T) = V(T)$.)
- Since both \tilde{P}_1 and \tilde{P}_2 are risk-neutral, the discounted portfolio value process $D(t)X(t)$ is a martingale under both \tilde{P}_1 and \tilde{P}_2 .

(Cont.) Model is complete $\implies Q$ measure is unique

- It follows that

$$\begin{aligned}\tilde{P}_1(A) &= \tilde{E}_1[D(T)V(T)] = \tilde{E}_1[D(T)X(T)] \\ &= X(0) \\ &= \tilde{E}_2[D(T)X(T)] = \tilde{E}_2[D(T)V(T)] \\ &= \tilde{P}_2(A)\end{aligned}$$

- Since A is an arbitrary set in F and $\tilde{P}_1(A) = \tilde{P}_2(A)$, these two risk-neutral measures are really the same.

(Cont.) Model is complete \iff Q measure is unique

- Suppose there is only one risk-neutral measure.
- This means first of all that the filtration for the model is generated by the d -dimensional Brownian motion driving the assets.

(Cont.) Model is complete $\iff Q$ measure is unique

- Uniqueness of the risk-neutral measure implies that the market price of risk equations

$$\alpha_i(t) - R(t) = \sum_{j=1}^d \sigma_{ij}(t) \Theta_j(t), \quad i = 1, \dots, m$$

have only one solution $\Theta_1(t), \dots, \Theta_d(t)$.

- For fixed t and w , these equations are of the form

$$Ax = b$$

(Cont.) Model is complete \iff Q measure is unique

- For fixed t and w , these equations are of the form

$$Ax = b$$

Where A is the $m \times d$ -dimensional matrix

$$A = \begin{bmatrix} \sigma_{11}(t) & \sigma_{12}(t) & \cdots & \sigma_{1d}(t) \\ \sigma_{21}(t) & \sigma_{22}(t) & \cdots & \sigma_{2d}(t) \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{m1}(t) & \sigma_{m2}(t) & \cdots & \sigma_{md}(t) \end{bmatrix}$$

(Cont.) Model is complete $\iff Q$ measure is unique

- x is the d -dimensional column vector
and b is the m -dimensional column vector

$$x = \begin{bmatrix} \Theta_1(t) \\ \Theta_2(t) \\ \vdots \\ \Theta_d(t) \end{bmatrix}, \quad b = \begin{bmatrix} \alpha_1(t) - R(t) \\ \alpha_2(t) - R(t) \\ \vdots \\ \alpha_m(t) - R(t) \end{bmatrix}$$

- Our assumption that there is only one risk-neutral measure means that the system of $Ax = b$ has a unique solution x .

(Cont.) Model is complete \iff Q measure is unique

- In order to be assured that every derivative security can be hedged, we must be able to solve the hedging equations $j = 1, \dots, d$

$$\frac{\tilde{\Gamma}_j(t)}{D(t)} = \sum_{i=1}^m \Delta_i(t) S_i(t) \sigma_{ij} d\tilde{W}_j(t)$$

for $\Delta_1(t), \dots, \Delta_m(t)$ no matter what values of $\frac{\tilde{\Gamma}_j(t)}{D(t)}$ appear on the left-hand side.

(Cont.) Model is complete \iff Q measure is unique

- For fixed t and w , the hedging equations are of the form

$$A^t y = c$$

Where A^t is the transpose of the matrix A , and y is the m -dimensional vector

$$y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix} = \begin{bmatrix} \Delta_1(t) S_1(t) \\ \Delta_2(t) S_2(t) \\ \vdots \\ \Delta_m(t) S_m(t) \end{bmatrix}$$

(Cont.) Model is complete \iff Q measure is unique

- And c is the d -dimensional vector

$$c = \begin{bmatrix} \frac{\widetilde{\Gamma}_1(t)}{D(t)} \\ \frac{\widetilde{\Gamma}_2(t)}{D(t)} \\ \vdots \\ \frac{\widetilde{\Gamma}_d(t)}{D(t)} \end{bmatrix}$$

- In order to be assured that the market is complete, there must be a solution y to the system of $A^t y = c$, no matter what vector c appears on the right-hand side.

(Cont.) Model is complete $\iff Q$ measure is unique

- If there is always a solution y_1, \dots, y_m , then there are portfolio processes $\Delta_i(t) = \frac{y_i}{S_i(t)}$ satisfying the hedging equations

$$\frac{\tilde{\Gamma}_j(t)}{D(t)} = \sum_{i=1}^m \Delta_i(t) S_i(t) \sigma_{ij} d\tilde{W}_j(t)$$

no matter what processes appear on the left-hand side of those equations.

- We could conclude that a short position in an arbitrary derivative security can be hedged.

(Cont.) Model is complete $\iff Q$ measure is unique

- By the **Lemma of Second fundamental theorem**
 \Rightarrow The uniqueness of the solution x to $Ax = b$
implies the existence of a solution y to
 $A^t y = c$.
- Consequently, uniqueness of the risk-neutral
measure implies that the market model is
complete.

Lemma of Second fundamental theorem

Lemma

Let A be an $m \times d$ -dimensional matrix, b an m -dimensional vector, and c a d -dimensional vector. If the equation

$$Ax = b \quad (1)$$

has a unique solution x_0 , then the equation

$$A^t y = c \quad (2)$$

has at least one solution y_0 .

PROOF:

- We regard A as a mapping from \mathbb{R}^d to \mathbb{R}^m and define the *kernel* of A to be

$$K(A) = \{x \in \mathbb{R}^d : Ax = 0\}$$

- If x_0 solves (1) and $x \in K(A)$ then $x_0 + x$ also solves (1).
- Thus, the assumption of a unique solution to (1) implies that $K(A)$ contains only the d -dimensional zero vector.

- The rank of A is defined to be the number of **linearly independent columns** of A .
- Because $K(A)$ contains only the d -dimensional zero vector, the rank of A must be d .
- Otherwise, we could find a linear combination of these columns that would be the m -dimensional zero vector, and the coefficients in this linear combination would give us a non-zero vector in $K(A)$.

Dimension Theorem

Let A is a linear transformation $A : V \rightarrow W$

If $\dim(V) < \infty$

then $\dim(\text{kernel}(A)) + \text{rank}(A) = \dim(V)$

- By Dimension Theorem, $\text{rank}(A) = d$

$$\dim(\text{kernel}(A)) + \text{rank}(A) = \dim(V)$$

$$\Rightarrow 0 + \text{rank}(A) = d$$

Example: If rank of A is not d

- Let $A = \{a_1, \dots, a_{d-1}, a_d\}$, where $\{a_i\}$ are the column vectors of A
- If the rank of A is $d - 1$ (It exists $\{w_i\}$ such that $a_d = \sum_{i=1}^{d-1} w_i a_i$)

(Cont.) Example: If rank of A is not d

- Then

$$\begin{aligned} Ax &= [a_1, \dots, a_d] \begin{bmatrix} x_1 \\ \vdots \\ x_d \end{bmatrix} \\ &= [a_1x_1 + a_2x_2 + \dots + a_dx_d] \\ &= \left[a_1x_1 + a_2x_2 + \dots + x_d \sum_{i=1}^{d-1} w_i a_i \right] \\ &= [(x_1 + w_1x_d)a_1 + (x_2 + w_2x_d)a_2 \\ &\quad + \dots + (x_{d-1} - w_{d-1}x_d)a_d] \end{aligned}$$

(Cont.) Example: If rank of A is not d

- By solve above equation

$$s.t. \quad x = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad or \quad \begin{bmatrix} -w_1 c \\ -w_2 c \\ \vdots \\ c \end{bmatrix}, \quad \forall c \in \mathbb{R}$$

- Therefore, the dimension of $K(A)$ would not equal 0.

Theorem

Any matrix and its transpose have the same rank.

$$\dim(R(A)) = \dim(R(A^t))$$

- The rank of A^t is d as well.
- The rank of a matrix is also the dimension of its range space. The range space of A^t is

$$R(A^t) = \{z \in \mathbb{R}^d : z = A^t y, \quad \exists y \in \mathbb{R}^m\}$$

Theorem

Let W is a subspace of a vector space V , where $\dim(V) < \infty$.

if $\dim(W) = \dim(V)$, then $W = V$.

- The dimension of $R(A^t)$ is also d and it is a subspace of \mathbb{R}^d , it must in fact equal to \mathbb{R}^d
- In other words, for every $z \in \mathbb{R}^d$, there is some $y \in \mathbb{R}^m$ such that $z = A^t y$.
- Hence (2) has at least one solution $y_0 \in \mathbb{R}^m$.

Thank you !