Financial Engineering Lecture 7

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Outline

- Martingale measures and Risk Neutral Measures.
- First Fundamental Theorem of Asset Pricing: Statement, consequences and its proof.

First Fundamental Theorem of Asset Pricing

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• We only need to specify the probabilities for each ω_i : There are $0 < p_i < 1, i = 1, ..., N$ such that $\sum_{i=1}^{N} p_i = 1$ and

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• Thus, if $A \in \mathcal{F}$, then there is $I \subseteq \{1, ..., N\}$ such that

$$A = \bigcup_{i \in I} \{\omega_i\} \Longrightarrow \mathbb{P}(A) = \sum_{i \in I} p_i$$
 (See Lecture 1).

• The set of information will be given by the filtration $(\mathscr{F}_t)_{0 \le t \le T}$ with $\mathscr{F}_0 = \{\emptyset, \Omega\}$ and

$$\mathscr{F}_{t-1}\subseteq\mathscr{F}_t\subseteq\mathscr{F},\ 1\leq t\leq T.$$

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• Note that in this framework, any random variable or vector ξ in $(\Omega, \mathscr{F}, \mathbb{P})$ has only N possible outcomes, namely

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This means that our price process

$$P_t = (B_t, S_t^{(1)}, \dots, S_t^{(d)}),$$

can take **up to** N values.



 As example of the previous set-up, consider the Three-Step Binomial model discussed in Lecture 5:

$$\Omega = \{uuu, uud, udu, udd, duu, dud, ddu, ddd\}, \quad N = 8$$

but the stock price has only two possible outcomes.

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Recall that in this model returns are i.i.d. and satisfy that

$$K_S(t) = egin{cases} R_u & ext{with probability } p; \ R_d & ext{with probability } 1-p. \end{cases}$$

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Thus,

$$\mathbb{P}(\{udu\}) = \mathbb{P}(K_S(1) = R_u, K_S(2) = R_d, K_S(3) = R_u) = p(1-p)p.$$



• The market with information is given:

$$\mathfrak{M} = \left\{ (\Omega, \mathscr{F}, (\mathscr{F}_t)_{0 \le t \le T}, \mathbb{P}), P = (B_t, S_t^{(1)}, \dots, S_t^{(d)})_{0 \le t \le T} \right\}. \quad (1)$$

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, $\forall 1 \leq t \leq T$.

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, $\forall 1 \leq t \leq T$.

• A typical example corresponds to the price of a zero-coupon bond which pays a yearly interest rate r > 0, i.e.

$$B_t = (1+r)^t, \ \forall 0 \le t \le T.$$



Admissible Strategies

Definition (Admissible Strategies)

Let $\Theta = (\varphi_t, \theta_t^{(1)}, \dots, \theta_t^{(d)})_{0 \le t \le T}$ be a strategy on the market \mathfrak{M} . We will say that Θ is admissible if:

1 It is self-financed: It only requires an initial capital. In symbols

$$\varphi_t B_t + \sum_{j=1}^d \theta_t^{(j)} S_t^{(j)} =: V_t^{\Theta} = \varphi_{t+1} B_t + \sum_{j=1}^d \theta_{t+1}^{(j)} S_t^{(j)}, \ \forall \ 0 \le t \le T - 1.$$

- **2** Non-anticipative: It is build up only on current market information, i.e. Θ_{t+1} , the portfolio created at time t, is \mathscr{F}_t -measurable.
- **3** It has a limited credit line: There is a non-random constant C > 0, such that

$$V_t^{\Theta} \geq -C, \ \forall \ 0 \leq t \leq T.$$



Arbitrage

Definition (Arbitrage)

An arbitrage on $\mathfrak M$ is an admissible strategy $\Theta = (\varphi_t, \theta_t^{(1)}, \dots, \theta_t^{(d)})_{0 \le t \le T}$ such that

1 It has zero initial capital:

$$V_0^{\Theta} = \varphi_0 B_0 + \sum_{i=1}^d \theta_0^{(i)} \cdot S_0^{(i)} = 0.$$

2 At time t = T, we are out of debts with 100% certainty: Almost surely

$$V_T^{\Theta} = \varphi_T B_T + \sum_{i=1}^d \theta_T^{(i)} \cdot S_T^{(i)} \geq 0.$$

3 We have a chance to make a profit:

$$\mathbb{P}\left(\varphi_T B_T + \sum_{j=1}^d \theta_T^{(j)} \cdot S_T^{(j)} > 0\right) > 0.$$

The Main Assumptions

Assumption (A.3)

The market \mathfrak{M} given by (1) does not admit arbitrage opportunities.

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- In the One-Step Binomial model, we proved that the market does not allow arbitrage if and only if there is $0 < q^* < 1$ such that

$$\mathbb{E}_*(K_S(1)) := R_u q^* + R_d(1 - q^*) = r.$$

- The First Fundamental Theorem of Asset Pricing (FFTAP from now on) gives necessary and sufficient conditions for the absence of arbitrage in the market described in the previous assumption.
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$$\mathbb{E}_*(K_S(1)) := R_u q^* + R_d(1 - q^*) = r.$$

• Such a $0 < q^* < 1$ exists (and it is unique) provided that $R_d < r < R_u$. In such situation

$$q^* = \frac{R_u - r}{R_u - R_d}.$$



 Moreover, in the previous lecture we checked that in the Multi-Step Binomial model if we let the returns of S be i.i.d. with

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$$K_S(t) = egin{cases} R_u & ext{with probability } q^*; \ R_d & ext{with probability } 1-q^*. \end{cases}$$

• Then the discounted price

$$\tilde{S}_t = \frac{S_t}{(1+r)^t}, \ 0 \le t \le T,$$

is a martingale on the filtration

$$\mathscr{F}_0 = \{\emptyset, \Omega\}, \ \mathscr{F}_t = \sigma(K_S(1), \dots, K_S(t)), \ 1 \le t \le T.$$

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 - 1 It transforms the real-world probabilities (the one observed in the market) (p, 1-p) satisfying that

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into a risk-neutral world in which the "new" distribution for the returns

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- 2 It transforms the discounted prices into a martingale!
- In this model these two situations are equivalent.

Discounted prices

Let

$$\mathfrak{M} = \left\{ (\Omega, \mathscr{F}, (\mathscr{F}_t)_{0 \leq t \leq T}, \mathbb{P}), P = (B_t, S_t^{(1)} \dots, S_t^{(d)})_{0 \leq t \leq T} \right\}$$

be as in (1).

Given a process X, we will define and denote the discounted version
of X as

$$\tilde{X}_t := \frac{X_t}{B_t}, \ \ 0 \le t \le T.$$

 This notation will be mainly applied to price processes as well as wealth processes.

Martingale Measures

Definition (Martingale Measures)

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be the market described in (1). Let $\mathbb Q$ be a probability measure on $(\Omega,\mathscr F)$, that is, there are $0< q_i^*<1,\ i=1,\ldots,N$ such that $\sum_{i=1}^N q_i^*=1$ and

$$\mathbb{Q}(\{\omega_i\})=q_i^*,\ i=1,\ldots,N.$$

We will say that \mathbb{Q} is a martingale measure if under \mathbb{Q} the discounted price $(\tilde{P}_t^{(j)})_{0 \leq t \leq T}$ is a martingale w.r.t. $(\mathscr{F}_t)_{0 \leq t \leq T}$, for every $j = 1, 2, \ldots, d+1$.

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Remark: $\tilde{P}_t^{(1)} = \tilde{B}_t = 1$, so it is always a martingale under any filtered probability space.

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Remark: $\tilde{P}_t^{(1)} = \tilde{B}_t = 1$, so it is always a martingale under any filtered probability space.

Notation: Expectations and conditional expectations w.r.t. \mathbb{Q} are gonna be denoted as \mathbb{E}_* and \mathbb{E}_* ($|\mathscr{H}|$), respectively.

Martingale Measures are Risk-Neutral

• By the martingale property, for all j = 1, 2, ..., d

$$\mathbb{E}_* \left(\frac{S_{t+1}^{(j)}}{\underbrace{\mathcal{B}_{t+1}}_{\text{Determ.}}} \middle| \mathscr{F}_t \right) = \frac{\underbrace{S_t^{(j)}}_{S_t^{(j)}}}{B_t} \Leftrightarrow \mathbb{E}_* \left(\frac{S_{t+1}^{(j)}}{S_t^{(j)}} \middle| \mathscr{F}_t \right) = \frac{B_{t+1}}{B_t}.$$

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Thus,

$$\mathbb{E}_* \left(\underbrace{\frac{S_{t+1}^{(j)}}{S_t^{(j)}} - 1}_{K_S(t+1)} \middle| \mathscr{F}_t \right) = \underbrace{\frac{B_{t+1}}{B_t} - 1}_{K_B(t+1)}.$$

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• Taking expectations in the previous equation, we get that

$$\mathbb{E}_*(K_S(t+1)) = K_B(t+1), \ 0 \le t \le T-1.$$

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$$\mathbb{Q}(S_t = 100) \neq \mathbb{P}(S_t = 100).$$

$$\mathbb{E}_*(\xi) = \sum_{i=1}^N \xi(\omega_i) \mathbb{Q}(\{\omega_i\}) \neq \sum_{i=1}^N \xi(\omega_i) \mathbb{P}(\{\omega_i\}) = \mathbb{E}(\xi).$$

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- Also conditional expectations will take different values in $\mathbb P$ compared to $\mathbb Q.$
- In the exercise set you will encounter some examples of such situations.

The First Fundamental Theorem of Asset Pricing

Theorem (FFTAP)

Let

$$\mathfrak{M} = \left\{ (\Omega, \mathscr{F}, (\mathscr{F}_t)_{0 \leq t \leq T}, \mathbb{P}), P = (B_t, S_t^{(1)}, \dots, S_t^{(d)})_{0 \leq t \leq T} \right\}$$

be the market described in (1). $\mathfrak M$ is arbitrage free if and only if there is a martingale measure.

Pricing and the FFTAP

Corollary (Pricing Financial Derivatives)

Let M be

$$\mathfrak{M} = \left\{ (\Omega, \mathscr{F}, (\mathscr{F}_t)_{0 \leq t \leq T}, \mathbb{P}), P = (B_t, S_t^{(1)} \dots, S_t^{(d)})_{0 \leq t \leq T} \right\}$$

be the market described in (1) and suppose that there exists a martingale measure \mathbb{Q} . Let $\xi \geq 0$ be the pay-off of a financial derivative with delivery time T>0. If we put

$$\xi_t := \mathbb{E}_* \left(\left. \frac{B_t}{B_T} \xi \right| \mathscr{F}_t \right), \quad 0 \le t \le T,$$

then the extended market

$$\tilde{\mathfrak{M}} = \left\{ (\Omega, \mathscr{F}, (\mathscr{F}_t)_{0 \leq t \leq T}, \mathbb{P}), \tilde{P} = (B_t, S_t^{(1)} \dots, S_t^{(d)}, \xi_t)_{0 \leq t \leq T} \right\},$$

is arbitrage free.

Proof: See the Exercise Set 7.



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Lemma (Lemma 1)

Let $\Theta = (\varphi_t, \theta_t^{(1)}, \dots, \theta_t^{(d)})_{0 \le t \le T}$ be a self-financed strategy. Then

$$\tilde{V}_{t}^{\Theta} = V_{0}^{\Theta} + \sum_{u=1}^{t} \sum_{j=1}^{d} \theta_{u}^{(j)} \Delta \tilde{S}_{u}^{(j)}, \ 1 \leq t \leq T,$$

where $\Delta \tilde{S}_t^{(j)} := \tilde{S}_t^{(j)} - \tilde{S}_{t-1}^{(j)}$. In particular, if Θ is admissible and \mathbb{Q} is a martingale measure, then $(\tilde{V}_t^{\Theta}, \mathscr{F}_t)_{0 \leq t \leq T}$ is a martingale under \mathbb{Q} .

Proof: In the "digital blackboard".

Lemma (Lemma 2)

Let $(X_t, \mathscr{F}_t)_{0 \le t \le T}$ be a martingale. Then

$$\mathbb{E}(X_t) = \mathbb{E}(X_0), \ \forall \ 0 \le t \le T.$$

Proof: In the "digital blackboard".

Lemma (Lemma 3)

If $\xi \geq 0$ is a random variable and $\mathbb{E}(\xi) = 0$, then $\xi \equiv 0$ almost surely.

Proof: In the "digital blackboard".

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1 It has zero initial capital:

$$V_0^{\Theta} = \varphi_0 B_0 + \sum_{i=1}^d \theta_0^{(j)} \cdot S_0^{(j)} = 0.$$

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$$\mathbb{P}\left(\varphi_T B_T + \sum_{j=1}^d \theta_T^{(j)} \cdot S_T^{(j)} > 0\right) > 0.$$

The previous lemmas imply:

• By Lemma 1 the discounted wealth process $(\tilde{V}^{\Theta}_t, \mathscr{F}_t)_{0 \leq t \leq T}$ is \mathbb{Q} -martingale such that

$$\frac{V_T^{\Theta}}{B_T} = \tilde{V}_T^{\Theta} \ge 0, \quad V_0^{\Theta} = 0.$$

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$$\mathbb{E}_*(\tilde{V}_T^{\Theta}) = \mathbb{E}_*(\tilde{V}_0^{\Theta}) = \mathbb{E}_*(\underbrace{\frac{V_0^{\Theta}}{B_0}}) = 0.$$

The previous lemmas imply:

• By Lemma 1 the discounted wealth process $(\tilde{V}^{\Theta}_t, \mathscr{F}_t)_{0 \leq t \leq T}$ is \mathbb{Q} -martingale such that

$$\frac{V_T^{\Theta}}{B_T} = \tilde{V}_T^{\Theta} \ge 0, \quad V_0^{\Theta} = 0.$$

Lemma 2 implies that

$$\mathbb{E}_*(\tilde{V}_T^{\Theta}) = \mathbb{E}_*(\tilde{V}_0^{\Theta}) = \mathbb{E}_*(\underbrace{\frac{V_0^{\Theta}}{B_0}}) = 0.$$

• Lemma 3 now implies that $V_T^\Theta=0$ Q-almost surely. We will use this to show that

$$V_T^{\Theta}(\omega_i) = 0, \quad i = 1, \dots, N.$$

• Recall that if $A \in \mathcal{F}$, then there is $I \subseteq \{1, ..., N\}$ such that

$$A = \cup_{i \in I} \{\omega_i\} \Longrightarrow \mathbb{Q}(A) = \sum_{i \in I} q_i^*, \ \ 0 < q_i^* < 1.$$

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• Since V^{Θ}_{T} is a random variable there is $I \subseteq \{1, \dots, N\}$ such that

$$\mathscr{F}\ni\left\{\omega\in\Omega:V_{T}^{\Theta}(\omega)>0\right\}=\cup_{i\in I}\left\{\omega_{i}\right\}.$$

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• Since V_T^{Θ} is a random variable there is $I \subseteq \{1, \dots, N\}$ such that

$$\mathscr{F} \ni \{\omega \in \Omega : V_T^{\Theta}(\omega) > 0\} = \bigcup_{i \in I} \{\omega_i\}.$$

• Thus,

$$0=\mathbb{Q}(V_T^{\Theta}>0)=\sum_{i\in I}q_i^*>0.$$

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Thus,

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• Therefore $\{\omega \in \Omega : V_T^{\Theta}(\omega) > 0\} = \emptyset$, which contradicts the fact that

$$\mathbb{P}\left(V_T^{\Theta} = \varphi_T B_T + \sum_{i=1}^d \theta_T^{(i)} \cdot S_T^{(i)} > 0\right) > 0.$$

Proof of FFTAP: More lemmas

For the other direction of the proof (which is the hardest conceptually speaking) we need more lemmas.

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Lemma (Lemma 4)

The following sets are vector spaces

$$\begin{split} \mathcal{V} &:= \left\{ V_T^{\Theta} \mid \Theta = (\varphi_t, \theta_t^{(1)}, \dots, \theta_t^{(d)})_{0 \leq t \leq T} \text{ is admissible and } V_0^{\Theta} = 0 \right\}. \\ \mathcal{X} &:= \left\{ (V(\omega_1), V(\omega_2), \dots, V(\omega_N)) \in \mathbb{R}^N \mid V \in \mathcal{V} \right\}. \end{split}$$

Moreover, if no arbitrage is allowed, then $\mathcal{X} \cap \mathcal{Y} = \emptyset$ where

$$\mathcal{Y} = \left\{ \left(\textit{V}(\omega_1), \textit{V}(\omega_2), \dots, \textit{V}(\omega_N) \right) \in \mathbb{R}^N \mid \textit{V r.v.}, \; \textit{V} \geq 0, \, \mathbb{P}(\textit{V} > 0) > 0 \right\}.$$

Proof: See the Exercise Set 7.

Proof of FFTAP: More lemmas

Theorem (The Separating Hyperplane Theorem)

Let $A \subseteq \mathbb{R}^N$ be convex and compact and \mathcal{X} a vector subspace of \mathbb{R}^n . If $A \cap \mathcal{X} = \emptyset$ then there exists $y \in \mathbb{R}^N$ such that:

1 For all $a \in A$

$$\sum_{i=1}^{N} y_i a_i > 0.$$

2 For all $x \in \mathcal{X}$

$$\sum_{i=1}^N y_i x_i = 0.$$

Proof: See Lemma 7.22 in Capinski's book.

Proof (\Rightarrow): We will only check the case T=1. Suppose that there is no arbitrage.

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Recall the notation

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Proof (\Rightarrow): We will only check the case T=1. Suppose that there is no arbitrage.

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By Lemma 4

$$\mathcal{A} := \left\{ (q_1, \dots, q_N) \in \mathcal{Y} \mid 0 \leq q_i \leq 1, \ \sum_{i=1}^N q_i = 1
ight\} \Rightarrow \mathcal{A} \cap \mathcal{X} = \emptyset.$$

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ight\} \Rightarrow \mathcal{A} \cap \mathcal{X} = \emptyset.$$

- Moreover:
 - 1 A is compact: This is obvious because A is closed and bounded.
 - 2 A is convex: Proof in In the "digital blackboard".



- By The Separating Hyperplane Theorem, there exists $y \in \mathbb{R}^N$ such that:
 - **1** For all $(q_1, \ldots, q_N) \in \mathcal{Y}$ with $0 \le q_i \le 1$ and $\sum_{i=1}^N q_i = 1$

$$\sum_{i=1}^N y_i q_i > 0.$$

2 For all $\Theta=(\varphi,\theta^{(1)},\dots,\theta^{(d)})_{0\leq t\leq T}$ portfolios with $V_0^\Theta=0$

$$\sum_{i=1}^{N} y_i V_1^{\Theta}(\omega_i) = 0, \quad \left(\left(V_1^{\Theta}(\omega_1), V_1^{\Theta}(\omega_2), \dots, V_1^{\Theta}(\omega_N) \right) \in \mathcal{X} \right)$$

Define

$$q_i^* = \frac{y_i}{\sum_{i=1}^N y_i}, \ i = 1, \dots, N,$$

and

$$\mathbb{Q}\left(\left\{\omega_{i}\right\}\right)=q_{i}^{*},\ i=1,\ldots,N.$$

Define

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and

$$\mathbb{Q}(\{\omega_i\}) = q_i^*, \quad i = 1, \dots, N.$$

 To conclude the proof we need to show that Q is a martingale measure, i.e.

$$0 < q_i^* < 1$$

and

$$\mathbb{E}_*\left(\frac{S_1^{(j)}}{B_1}\bigg|\mathscr{F}_0 = \{\emptyset, \Omega\}\right) = \frac{S_0^{(j)}}{B_0} = S_0^{(j)}, \ j = 1, \dots, d.$$

• Since for all i = 1, ..., N

$$v=(\underbrace{0,0,\ldots,0}_{i-1},1,0,\ldots,0)\in A,$$

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• We conclude that

$$0 < \sum_{k=1}^{N} y_k v_k = y_i, \ i = 1, \dots, N.$$

• Since for all i = 1, ..., N

$$v = (\underbrace{0, 0, \dots, 0}_{i-1}, 1, 0, \dots, 0) \in A,$$

We conclude that

$$0 < \sum_{k=1}^{N} y_k v_k = y_i, \ i = 1, \dots, N.$$

Which trivially implies that

$$0 < q_i^* = \frac{y_i}{\sum_{i=1}^{N} y_i} < 1.$$

• Recall that by the Separating Hyperplane Theorem, for any portfolio $(\varphi, \theta^{(1)}, \dots, \theta^{(d)}) \in \mathbb{R}^{d+1}$ satisfying that

$$V_0^{\Theta}=0, \ V_1^{\Theta}\geq 0,$$

we have that

$$egin{aligned} \mathbb{E}_*\left(V_1^{\Theta}
ight) &= \sum_{i=1}^N q_i^* V_T^{\Theta}(\omega_i) \ &= rac{1}{\sum_{i=1}^N y_i} \sum_{i=1}^N y_i V_T^{\Theta}(\omega_i) = 0. \end{aligned}$$

• Since B_1 is deterministic we have that from Lemma 1, for these type of portfolios

$$0 = \mathbb{E}_* \left(\frac{V_1^{\Theta}}{B_1} \right) = \mathbb{E}_* \left(V_0^{\Theta} + \sum_{j=1}^d \theta^{(j)} \left[\tilde{S}_1^{(j)} - \tilde{S}_0^{(j)} \right] \right)$$
$$= \mathbb{E}_* \left(\sum_{j=1}^d \theta^{(j)} \left[\tilde{S}_1^{(j)} - \tilde{S}_0^{(j)} \right] \right).$$

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In particular for the portfolio

$$\varphi = -S_0^{(j)}, \ \theta^{(j)} = \begin{cases} 1 & \text{if } j = i, \\ 0 & \text{otherwise}, \end{cases}$$

we have that

$$0 = \mathbb{E}_* \left(\sum_{i=1}^d \theta^{(j)} \left[\tilde{S}_1^{(j)} - \tilde{S}_0^{(j)} \right] \right) = \mathbb{E}_* \left(\left[\tilde{S}_1^{(i)} - \tilde{S}_0^{(i)} \right] \right).$$