

One and Multi-Period Asset Pricing Models

Chapters 2, 3, 7 - Lecture Notes by Markus Brunnermeier
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Today's talk

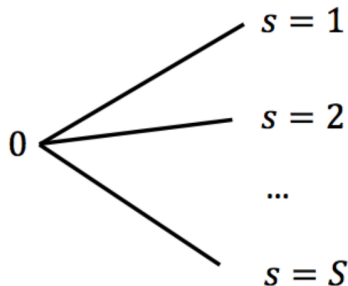
- State-space representation of security structures
- One period setting
- No arbitrage
- State prices and the Fundamental Theorem of Asset Pricing
- Pricing formulas for one period
 - Stochastic discount factors (SDFs)
 - Beta models
- Extension to multiple periods

One-Period Asset Model

Setup

- **Two dates:** present ($t = 0$) and future ($t = 1$)
- World is fixed at $t = 0$ but has S states at $t = 1$
 - e.g. flipping a coin, prices go up or down, $S=2$
- p_j : price of a security j at $t = 0$
- x_s^j : payoff at $t = 1$ for state s
- Payoff vector for asset j

$$\mathbf{x}^j = \begin{bmatrix} x_1^j \\ x_2^j \\ \vdots \\ x_S^j \end{bmatrix} \in \mathbb{R}^S$$



- Market consists of securities $j = 1, \dots, J$
- A *security structure* $\mathbf{X} \in \mathbb{R}^{S \times J}$ is the matrix formed by concatenating payoff vectors

$$\mathbf{X} = [\mathbf{x}^1, \dots, \mathbf{x}^J]$$

- A *portfolio* $\mathbf{h} = [h_1, \dots, h^J] \in \mathbb{R}^J$ denotes amount held for each of the J securities
- The price of the portfolio is $\mathbf{p}^T \mathbf{h}$ and its payoff is $\mathbf{X}\mathbf{h}$
- The *asset span* of \mathbf{X} is the set of achievable payoffs

$$\langle \mathbf{X} \rangle = \{ \mathbf{X}\mathbf{h} : \mathbf{h} \in \mathbb{R}^J \} \quad (1)$$

- We say *the market is complete* if $\langle \mathbf{X} \rangle = \mathbb{R}^S$, or $\text{rank}(\mathbf{X}) = S$

Particular examples

- *Arrow-Debreu securities:* $\mathbf{e}_s = [0, \dots, 1, \dots, 0]^T$, only pays off at state s .
- *Risk-free bonds:* $\mathbf{x}^b = [1, \dots, 1]$, pays off at every state.
- *Calls and puts:* Let $\mathbf{x}^1 = [1, 2, \dots, S]^T$. Then

$$x_s^k = \max(s + 1 - k, 0) \quad \text{and} \quad x_s^k = \max(k - s, 0), \quad k = 2, \dots, S \quad (2)$$

specify payoffs for call and put options on \mathbf{x}^1 , respectively. Then with $\mathbf{X} = [\mathbf{x}^1, \mathbf{x}^2, \dots, \mathbf{x}^S]$

$$\mathbf{X} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 2 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ S & S-1 & \dots & 1 \end{bmatrix} \quad \text{and} \quad \mathbf{X} = \begin{bmatrix} 1 & 1 & 2 & \dots & S-1 \\ 2 & 0 & 1 & \dots & S-2 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ S-1 & 0 & 0 & \dots & 1 \\ S & 0 & 0 & \dots & 0 \end{bmatrix}, \quad (3)$$

so both call and put options can be used to construct complete markets.

- **Notation:**

- $y \geq x \implies y_i \geq x_i \quad \forall i$
- $y > x \implies y_i \geq x_i \quad \forall i$ and $y \neq x$ (at least 1 strict inequality)
- $y \gg x \implies y_i > x_i \quad \forall i$

- **Arbitrage:** $\exists \mathbf{h}$ s.t. $\mathbf{p}^T \mathbf{h} \leq 0$ and $\mathbf{Xh} \geq \mathbf{0}$, with at least 1 strict inequality

- **Strong arbitrage:** $\exists \mathbf{h}$ s.t. $\mathbf{p}^T \mathbf{h} < 0$ and $\mathbf{Xh} \geq \mathbf{0}$

- The converse can be stated as follows:

- **No arbitrage (NA):** $\mathbf{Xh} > \mathbf{0} \implies \mathbf{p}^T \mathbf{h} > 0 \quad \forall \mathbf{h}$
- **No strong arbitrage (NSA):** $\mathbf{Xh} \geq \mathbf{0} \implies \mathbf{p}^T \mathbf{h} \geq 0 \quad \forall \mathbf{h}$

- $\text{NA} \implies \text{NSA}$

- **Law of One Price (LOOP):** Any two portfolios $\mathbf{h}, \mathbf{k} \in \mathbb{R}^J$ with equal payoffs must have equal price

$$\mathbf{X}\mathbf{h} = \mathbf{X}\mathbf{k} \implies \mathbf{p}^T \mathbf{h} = \mathbf{p}^T \mathbf{k}$$

- $\text{NA} \implies \text{NSA} \implies \text{LOOP} \implies$ every portfolio with zero payoff has zero price

- $\text{NSA} \implies \text{LOOP}$:

- Contrapositive: suppose $\exists \mathbf{h}, \mathbf{k}$ s.t. $\mathbf{X}\mathbf{h} = \mathbf{X}\mathbf{k}$, but $\mathbf{p}^T \mathbf{h} \neq \mathbf{p}^T \mathbf{k}$
- Then $\mathbf{X}(\mathbf{h} - \mathbf{k}) = \mathbf{0}$ but $\min(\mathbf{p}^T(\mathbf{h} - \mathbf{k}), \mathbf{p}^T(\mathbf{k} - \mathbf{h})) < 0$
- $\text{!LOOP} \implies \text{!NSA}$ □

- $\text{LOOP} \implies$ every portfolio with zero payoff has zero price:

- Similar to above: let $\mathbf{X}\mathbf{h} = \mathbf{0}$ but $\mathbf{p}^T \mathbf{h} \neq 0$
- Setting $\mathbf{k} = \mathbf{0}$ for LOOP leads to contrapositive □

- A vector of *state prices*: $\mathbf{q} \in \mathbb{R}^S$ s.t.

$$\mathbf{p} = \mathbf{X}^T \mathbf{q} = \sum_{s=1}^S q_s \cdot \mathbf{x}_s, \quad \text{or} \quad p_j = \sum_{s=1}^S q_s \cdot x_s^j \quad \forall j$$

- Implies all assets are discounted *exactly the same way* from payoffs \mathbf{x}^j .
Each state has an associated price.
- When does such a \mathbf{q} exist?

Fundamental Theorem of Asset Pricing (Stiemke's Theorem)

A security structure $\mathbf{X} \in \mathbb{R}^{S \times J}$ provides no arbitrage opportunities iff it yields a positive state price vector.

- Equivalent statement: either $\exists \mathbf{h} \in \mathbb{R}^J$ s.t.

$$\mathbf{p}^T \mathbf{h} \leq 0 \quad \text{and} \quad \mathbf{X} \mathbf{h} \geq \mathbf{0}$$

with at least 1 nonzero, or $\exists \mathbf{q} \in \mathbb{R}^S$ s.t.

$$\mathbf{p} = \mathbf{X}^T \mathbf{q} \quad \text{and} \quad \mathbf{q} \gg \mathbf{0}$$

- An example of a *theorem of the alternative*. Important class of theorems in convex optimization.

Fundamental Theorem of Asset Pricing (Stiemke's Theorem)

- *Proof:* First we restate the alternatives as

$$\text{I. } \exists \mathbf{h} \in \mathbb{R}^n : \mathbf{A}\mathbf{h} > \mathbf{0} \quad (4)$$

$$\text{II. } \exists \mathbf{z} \in \mathbb{R}^m : \mathbf{A}^T \mathbf{z} = \mathbf{0} \quad \text{and} \quad \mathbf{z} \gg \mathbf{0} \quad (5)$$

where

$$\mathbf{A} = \begin{bmatrix} -\mathbf{p}^T \\ \mathbf{X} \end{bmatrix} \in \mathbb{R}^{m \times n} \quad \text{and} \quad \mathbf{z} \propto \begin{bmatrix} 1 \\ \mathbf{q} \end{bmatrix}. \quad (6)$$

- The alternatives are mutually exclusive: (II) implies $\mathbf{h}^T (\mathbf{A}^T \mathbf{z}) = 0$, but (I) implies

$$\mathbf{h}^T \mathbf{A}^T \mathbf{z} = (\mathbf{A}\mathbf{h})^T \mathbf{z} > 0.$$

Fundamental Theorem of Asset Pricing (Stiemke's Theorem)

- Next assume (I) does not hold, which implies that

$$\mathcal{S}_1 = \{\mathbf{u} = \mathbf{A}\mathbf{h} : \mathbf{h} \in \mathbb{R}^n\} \quad \text{and} \quad \mathcal{S}_2 = \{\mathbf{v} \in \mathbb{R}^m : \mathbf{v} > \mathbf{0}\} \quad (7)$$

are disjoint. By hyperplane separation there is a unit vector $\mathbf{z} \in \mathbb{R}^m$ s.t.

$$\mathbf{z}^T \mathbf{v} > 0, \quad \forall \mathbf{v} \in \mathcal{S}_2 \quad \text{and} \quad \mathbf{z}^T \mathbf{A}\mathbf{h} = 0, \quad \forall \mathbf{h} \in \mathbb{R}^n. \quad (8)$$

- Then $\mathbf{A}^T \mathbf{z} = \mathbf{0}$, otherwise

$$\|\mathbf{A}^T \mathbf{z}\|^2 = \mathbf{z}^T \mathbf{A} \mathbf{A}^T \mathbf{z} > 0 \quad (9)$$

but this contradicts the hyperplane condition for \mathcal{S}_1 (set $\mathbf{h} = \mathbf{A}^T \mathbf{z}$).

- Finally show $\mathbf{z} \gg \mathbf{0}$: All the basis vectors of \mathbb{R}^m are in \mathcal{S}_2 .

The hyperplane condition then implies all coordinates of \mathbf{z} must be positive. □

State Prices in Incomplete Markets

- In general there can be multiple \mathbf{q} s, e.g. if $J < S$.
- However there is a unique \mathbf{q}^* in $\langle \mathbf{X} \rangle$, known as the *pricing kernel*. For instance,

$$\mathbf{X}^T(\mathbf{X}\mathbf{u}) = \mathbf{p} \quad (10)$$

provides a unique solution for $\mathbf{q}^* = \mathbf{X}\mathbf{u}$ as long as the \mathbf{x}^j s are linearly independent. Furthermore, for any valid \mathbf{q} ,

$$\mathbf{q}^* = \text{proj}(\mathbf{q} \mid \langle \mathbf{X} \rangle). \quad (11)$$

- Finally, there exists a unique $\mathbf{q} \gg 0$ iff markets are complete and there is no arbitrage.
 - \Longleftarrow Just solve $\mathbf{X}^T \mathbf{q} = \mathbf{p}$.
 - \Longrightarrow If markets are not complete but there is no arbitrage, then $\exists \mathbf{v} \neq \mathbf{0}$ s.t. $\mathbf{X}^T \mathbf{v} = \mathbf{0}$. Moreover we can find $\mathbf{q} \gg \mathbf{0}$ and some $\alpha \in \mathbb{R}$ s.t.

$$\mathbf{q} + \alpha \mathbf{v} \gg \mathbf{0}.$$

So there are infinitely many solutions to $\mathbf{X}^T \mathbf{q} = \mathbf{p}$. Contrapositive. □

One-Period Asset Pricing: Stochastic Discount Factor (SDF)

- Discounted price if no arbitrage: $p_j = \mathbf{q}^T \mathbf{x}^j$
- Introduce state probabilities π_s and **stochastic discount factor (SDF)** $m_s = \frac{q_s}{\pi_s}$ (state price per unit probability)

$$p_j = \sum_{s=1}^S \pi_s \frac{q_s}{\pi_s} x_s^j = \sum_s \pi_s m_s x_s^j = \mathbb{E}[\mathbf{m} \odot \mathbf{x}^j] \quad (12)$$

- Consider a risk-free bond with $\mathbf{x}^b = \mathbf{1}$. Since $\mathbb{E}[\mathbf{m} \odot \mathbf{x}^j] = \mathbb{E}[\mathbf{m}] \mathbb{E}[\mathbf{x}^j] + \text{Cov}(\mathbf{m}, \mathbf{x}^j)$,

$$p_b = \mathbb{E}[\mathbf{m}] = 1/R^f \implies p_j = \frac{\mathbb{E}[\mathbf{x}^j]}{R^f} + \text{Cov}[\mathbf{m}, \mathbf{x}^j] \quad (13)$$

and for any asset j .

- Since prices tend to grow over time, typically $\text{Cov}[\mathbf{m}, \mathbf{x}^j] < 0$.

One-Period Asset Pricing: Stochastic Discount Factor (SDF)

- Define the return of asset j as $\mathbf{R}^j \doteq \mathbf{x}^j / p_j$, then $\mathbb{E}[\mathbf{m} \odot \mathbf{R}^j] = 1$.
- Moreover, since $R^f = 1/\mathbb{E}[\mathbf{m}]$,

$$\mathbb{E}[\mathbf{m} \odot (\mathbf{R}^j - R^f)] = \mathbb{E}[\mathbf{m}] \left(\mathbb{E}[\mathbf{R}^j] - R^f \right) + \text{Cov}(\mathbf{m}, \mathbf{R}^j) = 0 \quad (14)$$

- Or

$$\mathbb{E}[\mathbf{R}^j] - R^f = -\frac{\text{Cov}(\mathbf{m}, \mathbf{R}^j)}{\mathbb{E}[\mathbf{m}]} \quad (15)$$

- The excess return for a generic asset j is determined solely by its covariance with the SDF. Investors are only compensated for holding *systematic risk*.

- Consider the unique SDF $\mathbf{m}^* = [q_1^*/\pi_1, \dots, q_S^*/\pi_S]^T$
- Treat \mathbf{m}^* as a payoff vector with price $p_{m^*} > 0$. We can define $\mathbf{R}^* = \mathbf{m}^*/p_{m^*}$ and write

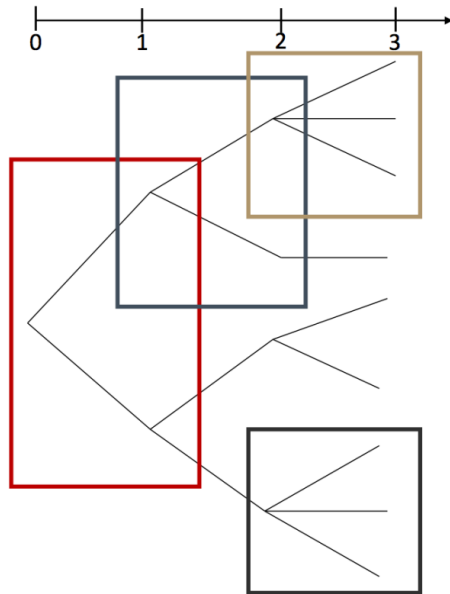
$$\mathbb{E}[\mathbf{R}^j] - R^f = -\frac{\text{Cov}(\mathbf{R}^*, \mathbf{R}^j)}{\mathbb{E}[\mathbf{R}^*]} = -\underbrace{\frac{\text{Cov}(\mathbf{R}^*, \mathbf{R}^j)}{\text{Var}(\mathbf{R}^*)}}_{\beta_j} \cdot \frac{\text{Var}(\mathbf{R}^*)}{\mathbb{E}[\mathbf{R}^*]} \quad (16)$$

- By plugging in $\mathbb{R}^* \rightarrow \mathbb{R}^j$ to (16), we get $\mathbb{E}[\mathbf{R}^*] - R^f = -\text{Var}(\mathbf{R}^*)/\mathbb{E}[\mathbf{R}^*]$, and

$$\mathbb{E}[\mathbf{R}^j] - R^f = \beta_j \left(\mathbb{E}[\mathbf{R}^*] - R^f \right). \quad (17)$$

- It remains to show that \mathbf{m}^* corresponds to the payoff of the market portfolio.

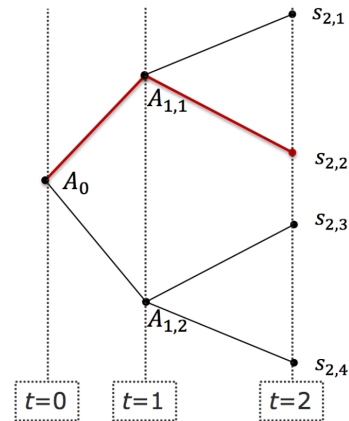
Multi-Period Asset Model



Dynamic Trading and Model Completeness

- T periods. Look at each branch as a one-period setting
- Even if with limited assets, we can trade at each time-period to achieve completeness
- *Short-lived assets* only pays out at the period after purchase. Need T assets for completeness.
First replicate an asset bought at $t = 0$ and pays \$1 at $s_{2,2}$
 - Arrow security: costs \$1, pays \$1 for a specific next state
 - $q_{2,2}$: state price for $s_{2,2}$
 - $\implies q_{2,2}$ Arrow securities are needed when $A_{1,1}$ occurs
 - To get \$1 at $A_{1,1}$, need $q_{1,1}$ Arrow securities at $t = 0$
 - \implies Buy $q_{1,1}q_{2,2}$ Arrow securities at $t = 0$, use payoff to purchase $q_{2,2}$ Arrow securities at $t = 1$

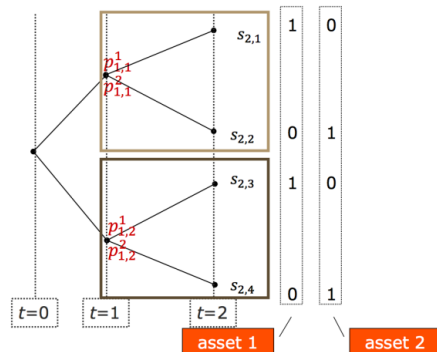
Repeat for other states to establish completeness



Dynamic Trading and Model Completeness

- *Long-lived assets* pay off over many time periods.
Can be traded any time.
- Example: Two assets paying off at $t = 2$.
 - Trading strategy $[j, A_{t,i}]$: the cash flow of asset j purchased in event $A_{t,i}$ and sold one period later.
 - $p_{t,i}^j$: price of asset j at time and state t, i
 - Enumerating strategies creates payoff matrix

Strategy	$[1, A_0]$	$[2, A_0]$	$[1, A_{1,1}]$	$[2, A_{1,1}]$	$[1, A_{1,2}]$	$[2, A_{1,2}]$
Event A_0	$-p_0^1$	$-p_0^2$	0	0	0	0
Event $A_{1,1}$	$p_{1,1}^1$	$p_{1,1}^2$	$-p_{1,1}^1$	$-p_{1,1}^2$	0	0
Event $A_{1,2}$	$p_{1,2}^1$	$p_{1,2}^2$	0	0	$-p_{1,2}^1$	$-p_{1,2}^2$
State $s_{2,1}$	0	0	1	0	0	0
State $s_{2,2}$	0	0	0	1	0	0
State $s_{2,3}$	0	0	0	0	1	0
State $s_{2,4}$	0	0	0	0	0	1

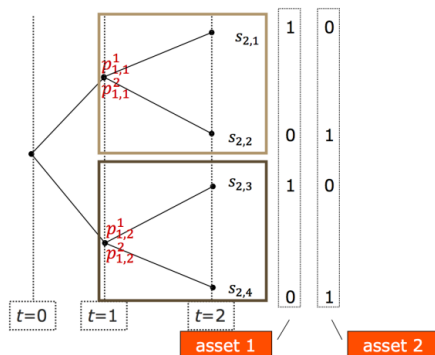


Dynamic Trading and Model Completeness

- Example contd.

Strategy	$[1, A_0]$	$[2, A_0]$	$[1, A_{1,1}]$	$[2, A_{1,1}]$	$[1, A_{1,2}]$	$[2, A_{1,2}]$
Event A_0	$-p_0^1$	$-p_0^2$	0	0	0	0
Event $A_{1,1}$	$p_{1,1}^1$	$p_{1,1}^2$	$-p_{1,1}^1$	$-p_{1,1}^2$	0	0
Event $A_{1,2}$	$p_{1,2}^1$	$p_{1,2}^2$	0	0	$-p_{1,2}^1$	$-p_{1,2}^2$
State $s_{2,1}$	0	0	1	0	0	0
State $s_{2,2}$	0	0	0	1	0	0
State $s_{2,3}$	0	0	0	0	1	0
State $s_{2,4}$	0	0	0	0	0	1

- The payoff matrix has full rank iff the highlighted submatrix has full rank
- Generically, the markets are *dynamically complete*
- Generically the *branching number* of the tree determines assets needed for completeness
 - e.g. binomial tree \implies Black-Scholes, 2 assets



Some changes to notation:

- For an asset, replace $\mathbf{x} \rightarrow p_{t+1} + x_{t+1}$ and $\mathbf{m} \rightarrow m_{t+1}$
 - need to keep track of the prices p_t at each time
 - x_t represents deviation from price
- Both $\{x_t\}_{t=0}^T$ and $\{m_t\}_{t=0}^T$ are *random processes*
 - x_t and m_t are random variables over some state space Ω (e.g. corresponding directly to stock price fluctuations)
- Each x_t and m_t measurable w.r.t. algebra \mathcal{F}_t of the state space Ω
- Furthermore $\{\mathcal{F}_t\}_{t=0}^T$ is a filtration: $\mathcal{F}_u \subseteq \mathcal{F}_v$ if $u \leq v$.
 - Uncertainty increases over time; cannot know the future

Multi-period Stochastic Discount Factors

- One-period pricing becomes

$$p_t = \mathbb{E}_t [m_{t+1} \cdot (p_{t+1} + x_{t+1})] \quad (18)$$

- Multi-period SDF: $M_{t+1} \doteq m_1 \cdot m_2 \cdot \dots \cdot m_{t+1}$
- Multiplying both sides by M_t :

$$M_t p_t = \mathbb{E}_t [M_{t+1} (p_{t+1} + x_{t+1})] \quad (19)$$

- Then for $t = 0$ and $t = 1$,

$$p_0 = \mathbb{E}_0 [M_1 (p_1 + x_1)] \quad \text{and} \quad M_1 p_1 = \mathbb{E}_1 [M_2 (p_2 + x_2)]$$

- Therefore

$$p_0 = \mathbb{E}_0 [\mathbb{E}_1 [M_2 (p_2 + x_2)] + M_1 x_1] = \mathbb{E}_0 [\mathbb{E}_1 [M_2 p_2 + M_2 x_2 + M_1 x_1]] \quad (20)$$

Multi-period Stochastic Discount Factors

- Law of Iterated Expectations: $\mathbb{E}_u[\mathbb{E}_v[X]] = \mathbb{E}_u[X]$ for all $\mathcal{F}_u \subseteq \mathcal{F}_v$
- Therefore

$$p_0 = \mathbb{E}_0[\mathbb{E}_1[M_2 p_2 + M_2 x_2 + M_1 x_1]] = \mathbb{E}_0[M_2 p_2 + M_2 x_2 + M_1 x_1] \quad (21)$$

- Iterating k steps yields

$$p_0 = \mathbb{E}_0[M_k p_k] + \sum_{t=1}^k \mathbb{E}_0[M_t x_t] \quad (22)$$

- Assuming that $\lim_{k \rightarrow \infty} \mathbb{E}_0[M_k p_k] = 0$, we have

$$p_0 = \sum_{t=1}^{\infty} \mathbb{E}_0[M_t x_t]$$

Some Ideas for Next Steps

- Extension to continuous time / state space
- Options pricing
- Modern derivation of portfolio theory / Black-Scholes
- Clarification on the multi-period setting
 - Some of the notation / objects in the multi-period setting are not clearly explained
 - Seems to be a widespread problem in financial economics literature
- Multi-period SDF: crucial for pricing and represents systemic risk
 - Relationship to market variables?
 - Empirical models: Fama-French, factor models, PCA / ML models

Thanks!