

MFE 230Q [Spring 2021]

Introduction to Stochastic Calculus

GSI Session 2



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Review

- Risk-Neutral Pricing
- Basic Probability Theory
- The Multi-Period Security market
- **Sample Problems!**

Risk-Neutral Pricing

A random variable (r.v.) is a function $\tilde{X} : \Omega \rightarrow \mathbb{R}$.

In this part of course, discrete finite sample spaces ($M < \infty$).

The expectation of a r.v. is $E_{\mathbb{P}}[\tilde{X}] = \sum_i \tilde{X}(\omega_i) \mathbb{P}(\omega_i)$.

Could have two probability measures defined over same sample space, \mathbb{P} and \mathbb{Q} .

- \mathbb{P} and \mathbb{Q} are *equivalent* if $\mathbb{P}(A) = 0 \Leftrightarrow \mathbb{Q}(A) = 0$.

- \mathbb{Q} *absolutely continuous* w.r.t. \mathbb{P} , $\mathbb{Q} \ll \mathbb{P}$, if $\mathbb{P}(A) = 0 \Rightarrow \mathbb{Q}(A) = 0$.

If $\mathbb{Q} \ll \mathbb{P}$, *Radon-Nikodym derivative* (or *likelihood ratio*) of \mathbb{Q} w.r.t. \mathbb{P} is

$$L(A) = \frac{d\mathbb{Q}}{d\mathbb{P}} \stackrel{\text{def}}{=} \frac{\mathbb{Q}(A)}{\mathbb{P}(A)}, \quad \mathbb{P}(A) > 0.$$

Follows immediately that for any r.v., \tilde{X} , $E_{\mathbb{P}}[L\tilde{X}] = E_{\mathbb{Q}}[\tilde{X}]$.

Pricing Equations

$$\begin{aligned} V^0 &= \mathbf{V}^1 \psi^{\text{SPV/AD}} \\ &= \frac{1}{R} E_Q[\tilde{V}^1], \quad \rightarrow \text{RN} \\ &= \frac{1}{R} E_P[L\tilde{V}^1], \quad q_i = \frac{\psi_i}{\hat{\psi}}, \quad R = \frac{1}{\hat{\psi}}, \\ \underline{LR} \\ \underline{SDF} &= E_P \left[\frac{M_1}{M_0} \tilde{V}^1 \right], \quad L_i = \frac{q_i}{p_i}, \\ &\quad \underline{(M_1)_i = \frac{L_i}{R} = \frac{\psi_i}{p_i}, \quad M_0 = 1.} \end{aligned}$$

Basic Probability

- Probability Space: $(\Omega, \mathcal{F}, \mathbb{P})$
- Sample space: Ω , set of all possible outcomes
- A set of events: \mathcal{F} , where each event is a set containing zero or more outcomes
- The assignment of probability to the events:
 $P : \mathcal{F} \rightarrow [0, 1]$

Basic Probability

$$\Omega = \{a, b, c, d\} \quad \emptyset \neq \{\emptyset\}$$

A σ -algebra \mathcal{F} is a set of subsets of Ω such that

1. $\emptyset \in \mathcal{F}$ ✓
2. $B \in \mathcal{F} \implies B^c \in \mathcal{F}$ ✓
3. $B_1, B_2, \dots \in \mathcal{F} \implies \bigcup_{n=1}^{\infty} B_n \in \mathcal{F}$ ✓

$$\mathcal{F} = \left\{ \emptyset, \{a, b\}, \{c, d\}, \{a, b, c, d\} \right\}$$

Properties:

1. $\bigcap_{n=1}^{\infty} B_n \in \mathcal{F}$
2. $A, B \in \mathcal{F} \implies A \setminus B \in \mathcal{F}$
3. $\{\emptyset, \Omega\} \subset \mathcal{F} \subset 2^{\Omega}$.

Basic Probability

$$\Omega = \{a, b, c\}$$

$$\{\{a\}, \{b\}, \{c\}\}$$

Let $\Omega = \{\omega_1, \dots, \omega_M\}$. Then $\{B_1, \dots, B_m\}$ is a partition of Ω iff:

1. $\cup_{i=1}^m B_i = \Omega$ ✓
2. $B_i \cap B_j = \emptyset, i \neq j.$

$$\{\{a, b\}, \{c\}\}$$

Note: Each of the B_i 's are sets!

Let \mathcal{B} be a non-empty collection of subsets of Ω . Note: Each element of \mathcal{B} is a set. Denote $\sigma(\mathcal{B})$ as the smallest sigma-algebra containing all the elements of \mathcal{B} .

Here, smallest means that any sigma-algebra containing the elements of \mathcal{B} would have to contain all the elements of $\sigma(\mathcal{B})$ as well.

generated σ-algebra

Basic Probability

$$\Omega = \{a, b, c, d\} \quad \mathcal{B} = \{\{a\}, \{b\}, \{c\}, \{d\}\}$$

We will consider sigma-algebras generated by partitions of Ω .

Let \mathcal{B} be a partition of Ω , then $\sigma(\mathcal{B})$ is constructed by taking all possible unions and complements of the elements in \mathcal{B} .

Note: $\sigma(\{\{\omega_1\}, \dots, \{\omega_M\}\}) = 2^{\Omega}$.

- Examples:

$$\sigma(\mathcal{B}) = \{\emptyset, \Omega, \{a\}, \{a\}^c, \{a, b\}, \{a, b\}^c, \{a, c\}, \{a, c\}^c, \dots\}$$

$$\Omega = \{\omega_1, \omega_2, \omega_3\}$$

$$\mathcal{F}_0 = \sigma(\{\Omega\}) = \{\emptyset, \Omega\}$$

$$\mathcal{F}_1 = \sigma(\{\omega_1, \omega_2\}, \{\omega_3\}) = \{\emptyset, \{\omega_1, \omega_2\}, \{\omega_3\}, \Omega\}$$

$$\mathcal{F}_2 = \sigma(\{\omega_1\}, \{\omega_2\}, \{\omega_3\}) = 2^{\Omega}$$

Basic Probability

- Sigma-algebra:

Interpretation: We can view a σ -algebra \mathcal{F} as formalizing the idea of information. More precisely: A σ -algebra \mathcal{F} is a collection of events, and if we assume that we have access to the information contained in \mathcal{F} , this means that for every $A \in \mathcal{F}$ we know exactly if A has occurred or not.

If we have another σ -algebra \mathcal{G} with $\mathcal{G} \subseteq \mathcal{F}$ then we interpret this as " \mathcal{G} contains less information than \mathcal{F} ".

Basic Probability

• Measurability:

(1) A set $A \subset \Omega$ is said to be \mathcal{F} - measurable, if $A \in \mathcal{F}$.

Example:

Assume $\Omega = \{\omega_1, \omega_2, \omega_3\}$, $\mathcal{F} = \sigma(\{\omega_1, \omega_2\}, \{\omega_3\})$

Then: $\{\omega_3\}$ is \mathcal{F} - measurable, $\{\omega_1\}, \{\omega_2, \omega_3\}$ are not.

(2) For R. V. \hat{X} and finite sample space Ω , \hat{X} is \mathcal{F} - measurable if \hat{X} is constant for all elements in B_i , for all $i = 1, \dots, K$, where $\mathcal{F} = \sigma(\{B_1, \dots, B_K\})$.

Examples:

Assume $\Omega = \{\omega_1, \omega_2, \omega_3\}$, $\mathcal{F} = \sigma(\{\omega_1, \omega_2\}, \{\omega_3\})$

Then $\hat{X}(\omega_i) = i$ is not \mathcal{F} - measurable.

$\hat{X}(\omega_i) = 1_{i \leq 2}$ is \mathcal{F} - measurable.

$$\left\{ \begin{array}{l} \omega \in \Omega | \hat{X}(\omega) \\ || \\ (-i) \end{array} \right.$$

(3) Similarly, for infinite sample space Ω , \hat{X} is \mathcal{F} - measurable if $\hat{X}^{-1} \in \mathcal{F}$ for every interval, $I \subset \mathcal{R}$

$$\hat{X}(w_1) = 1$$
$$\hat{X}(w_2) = 2$$

$$B_1 = \{w_1, w_2\}$$

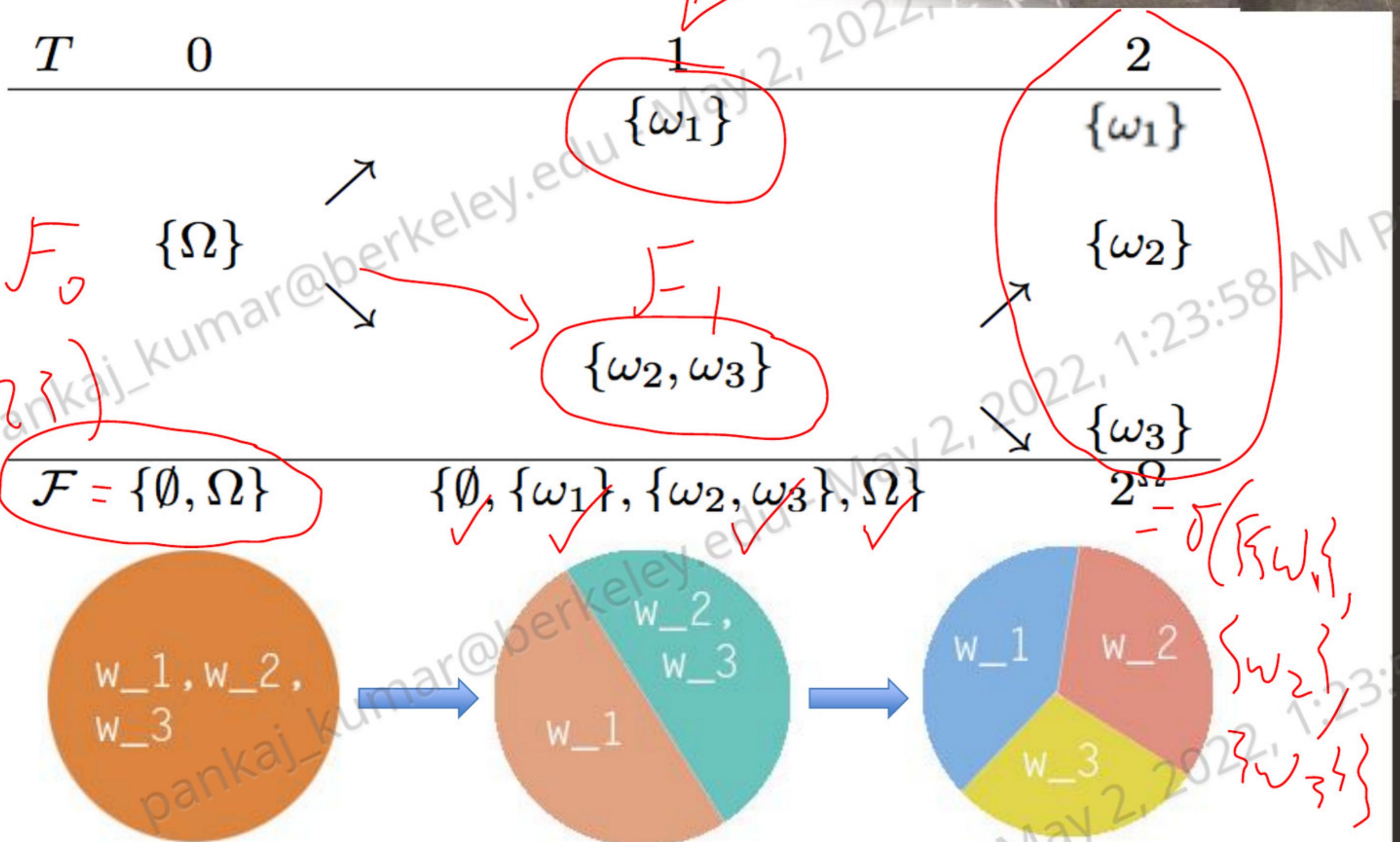
$$B_2 = \{w_3\}$$

$$\hat{X}(w_1) = 1$$
$$\hat{X}(w_2) = 1$$

$$\hat{X}(w_3) = 0$$

Basic Probability

Ω



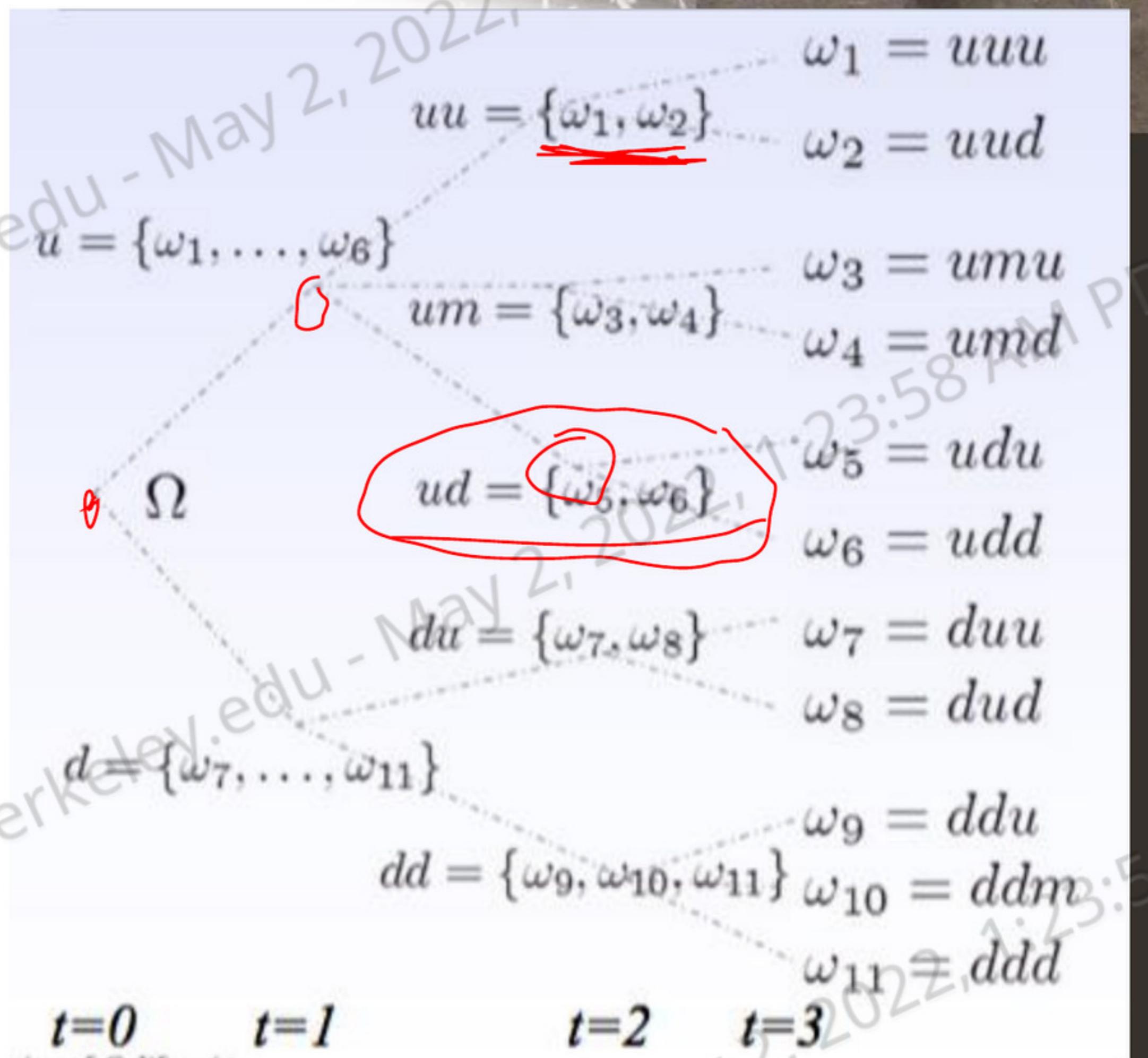
The Multi-Period Market

- Filtration

- Increasing sequence of σ -algebras: $\mathcal{F}_0 \subset \mathcal{F}_1 \subset \dots$
- Information revelation
- Think of it as a tree (if it helps)

- Adapted processes

- X_t is adapted to filtration $\underline{\mathcal{F}} = \{\mathcal{F}_t\}_{t \geq 0}$ if for all t , the process X_t is \mathcal{F}_t -measurable



The Multi-Period Market

- Adapted Processes:

| Embedding | | | | | |
|-----------|--|-------|--|---------------------|--|
| | | | | | |
| | | | | | |
| | | | | $\omega_1 = uuu$ | |
| | | | | $\omega_2 = uud$ | |
| | | | | $\omega_3 = umu$ | |
| | | | | $\omega_4 = umd$ | |
| | | | | $\omega_5 = udu$ | |
| | | | | $\omega_6 = udd$ | |
| | | | | $\omega_7 = duu$ | |
| | | | | $\omega_8 = dud$ | |
| | | | | $\omega_9 = ddu$ | |
| | | | | $\omega_{10} = ddm$ | |
| | | | | $\omega_{11} = ddd$ | |
| $t=0$ | | $t=1$ | | $t=2$ | |
| | | | | $t=3$ | |

Which of the following processes are *adapted to the filtration* in Example 7 from Lecture 3?

$$f_t : \Omega \rightarrow \mathbb{R}$$

1. $f(t, \omega) = i + t$
2. $f(t, \omega) = t$
3. $f(t, \omega) = (t^2 - t) \mathbf{1}_{\{i \geq 5\}}$
4. $f(t, \omega) = (t^2 - t) \mathbf{1}_{\{i \geq 6\}}$

$$\mathcal{F}_0 = \sigma(\{\emptyset\}) = \{\emptyset, \Omega\}$$

$$\mathcal{F}_1 = \sigma(\{\emptyset, \{\omega\}\}) = \{\emptyset, \Omega, \{\omega\}, \{\bar{\omega}\}\}$$

$$\mathcal{F}_2 = \sigma(\{\emptyset, \{\omega\}, \{\omega_1, \omega_2\}, \{\omega_1, \bar{\omega}_2\}, \{\bar{\omega}_1, \omega_2\}, \{\bar{\omega}_1, \bar{\omega}_2\}\}) = \dots$$

$$\mathcal{F}_3 = 2^{\Omega}$$

1. $t=0$, $f(0, \omega_1) = 1$, $f(0, \omega_2) = 2$ \times

2. $t=0$, $f(0, \omega_1) = 0$, $f(0, \omega_2) = 0$ \dots
 $t=1$, \dots

3. $t=0$, $f(0, w_1) = 0 \quad \dots \quad \checkmark$

$t=1$, $f(1, w_1) = 0 \quad \dots \quad \checkmark$

$t=2$, $f(2, w_1) = 0, f(2, w_2) = 0 \quad \checkmark$

$f(2, w_3) = 0 \quad f(2, w_4) = 0 \quad \checkmark$

$f(2, w_5) = 2 \quad f(2, w_6) = 2 \quad \checkmark$

⋮

4. $f(2, w_5) = 0 \neq f(2, w_6) = 2 \quad \checkmark \times$

The Multi-Period Market

Given a probability space $(\Omega, \mathcal{F}, \mathbb{P})$, a filtration, $\underline{\mathcal{F}} = (\mathcal{F}_0, \dots, \mathcal{F}_T)$, with $\mathcal{F}_0 = \{\emptyset, \Omega\}$, $\mathcal{F}_T = \sigma(\Omega)$, and a market with N traded assets:

- **Security level:**

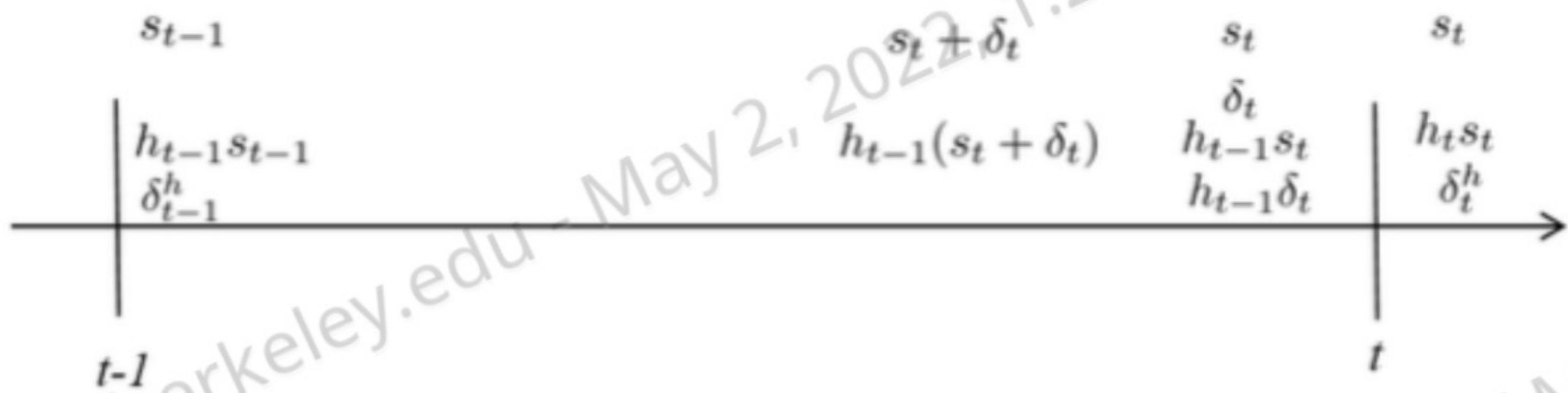
- *adapted* Dividend process: $\delta_t \in \mathbb{R}^N$,
- *adapted* (ex Dividend) Price process: $\mathbf{s}_t \in \mathbb{R}^N$,
- *adapted* Cum dividend process: $\mathbf{s}_t + \delta_t \in \mathbb{R}^N$,
dividend-price pair (δ, \mathbf{s})

- **Portfolio level:**

- *adapted* Trading strategy: $\mathbf{h}_t \in \mathbb{R}^N$, $\mathbf{h}_{-1} = \mathbf{0}$.
- *adapted* Payoff process (gain process):
 - **Case 1:** dividends: $\delta_t^{\mathbf{h}} = \mathbf{h}'_{t-1}(\mathbf{s}_t + \delta_t) - \mathbf{h}'_t \mathbf{s}_t$.
 - **Case 2:** No dividends: $\delta_t^{\mathbf{h}} = (\mathbf{h}_{t-1} - \mathbf{h}_t)' \mathbf{s}_t = -(\Delta_{-} \mathbf{h}_t)' \mathbf{s}_t$.
- *adapted* Value $V_t^{\mathbf{h}} = \mathbf{h}'_t \mathbf{s}_t$.
- *adapted* Consumption process: $c_t \in \mathbb{R}$,
 - Self-financing trading strategy: $\delta_t^{\mathbf{h}} = c_t$ for $t = 1, \dots, T$.

The Multi-Period Market

- Asset prices
- Asset dividends
- Portfolio value
- Net portfolio pay



• Several Insights:

- Total payout at time t = returns from asset holdings – change in portfolio value + dividend generated.

$$\underbrace{\delta_t^h}_{\text{Total time } t \text{ payout}} = \underbrace{\mathbf{h}'_{t-1}(\Delta - \mathbf{s}_t)}_{\text{Change in value from asset returns}} - \underbrace{\Delta - V_t}_{\text{Change in value of portfolio holdings}} + \underbrace{\mathbf{h}_{t-1}\delta_t}_{\text{Dividends}}$$

The Multi-Period Market

- **Several Insights:**

- Change in portfolio value = *returns from asset holdings + change from rebalancing.*

$$\Delta_{-}V_t = \mathbf{h}'_{t-1} \Delta_{-}\mathbf{s}_t + \Delta_{-}\mathbf{h}'_t (\mathbf{s}_{t-1} + \Delta_{-}\mathbf{s}_t).$$

- For self-financed portfolio with no dividend and no consumption, *change in portfolio value = returns from asset holdings.*

$$\underbrace{\Delta_{-}V_t}_{\substack{\text{Change in value} \\ \text{of portfolio holdings.}}} = \underbrace{\mathbf{h}'_{t-1}(\Delta_{-}\mathbf{s}_t)}_{\substack{\text{Change in value} \\ \text{from asset returns}}}$$

Sample Problem I

Consider a simple one-period binomial model. There are two traded assets, a bond with price $B(t)$ and a stock with price $S(t)$. Prices in the possible states of the world are given by.

$$\begin{array}{ccc} \text{State} & & \\ \bar{U} = U & \xrightarrow{\quad} & \bar{D} = L \\ B(0) = 100, S(0) = 100 & & B(T) = 105, S_{\uparrow}(T) = 110 \\ & \searrow & \\ & & B(T) = 105, S_{\downarrow}(T) = 90 \end{array}$$

A couple of quick questions to warm up:

- If $T = 1$ and interest is compounded once per period, what is the risk free rate in this economy?
- Is the market complete?

$$R_f = \frac{105}{100} = 1.05$$

$$D = \begin{bmatrix} & 105 & 105 \\ 110 & & 90 \end{bmatrix}$$

#asset States

\rightarrow

$$\text{rank}(D) = 2$$

$$S = D \begin{pmatrix} \psi_1 \\ \psi_2 \end{pmatrix}$$
$$\begin{pmatrix} \psi_1 \\ \psi_2 \end{pmatrix} = \begin{pmatrix} 5/7 \\ -5/21 \end{pmatrix}$$

Sample Problem I

$$B(0) = 100, S(0) = 100$$

$$B(T) = 105, S_{\uparrow}(T) = 110$$

2100

$$B(T) = 105, S_{\downarrow}(T) = 90$$

0

We wish to find the price of the derivative with the payoff of $\max\{S(T)^2 - K, 0\}$ where $K = 10,000$. This derivative is called a power call. What tools have we learned so far to price this option?

- Replication
- Arrow–Debreu Securities
- Martingale/risk-neutral measure

Rep.

$$\underline{h_S} / 10 + \underline{n_B} / 105 = 2100$$

$$\underline{h_S} 90 + \underline{n_B} 105 = 0$$

\Rightarrow

$$h_S = 105 \quad n_B = -90$$

$$\angle DOP \Rightarrow 105 \times 100 - 90 \times 100 = 1500$$

AD/SPV

$$C = \begin{pmatrix} 2100 & 0 \end{pmatrix} \begin{pmatrix} 5/7 \\ 5/21 \end{pmatrix} = 1500$$

Fmr / RN

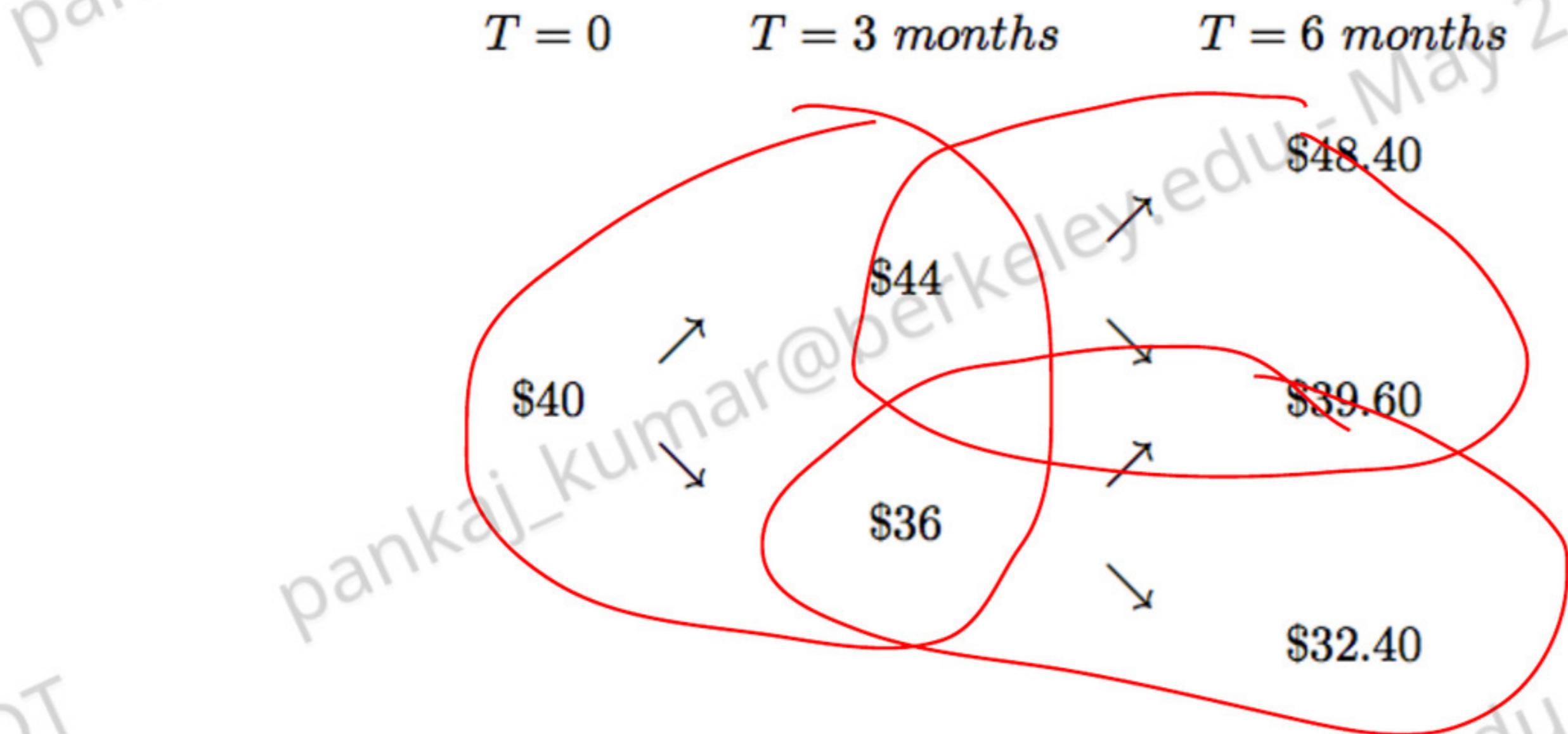
$$g_i = \frac{\psi_i}{\psi} \Rightarrow g_1 = \frac{3}{4}, \quad g_2 = \frac{1}{4}$$

$$C = \frac{1}{12} E_Q(V) = \frac{1}{1.05} \left(\frac{3}{4} \times 2100 + \frac{1}{4} \times 0 \right) = 1500$$

Sample Problem II

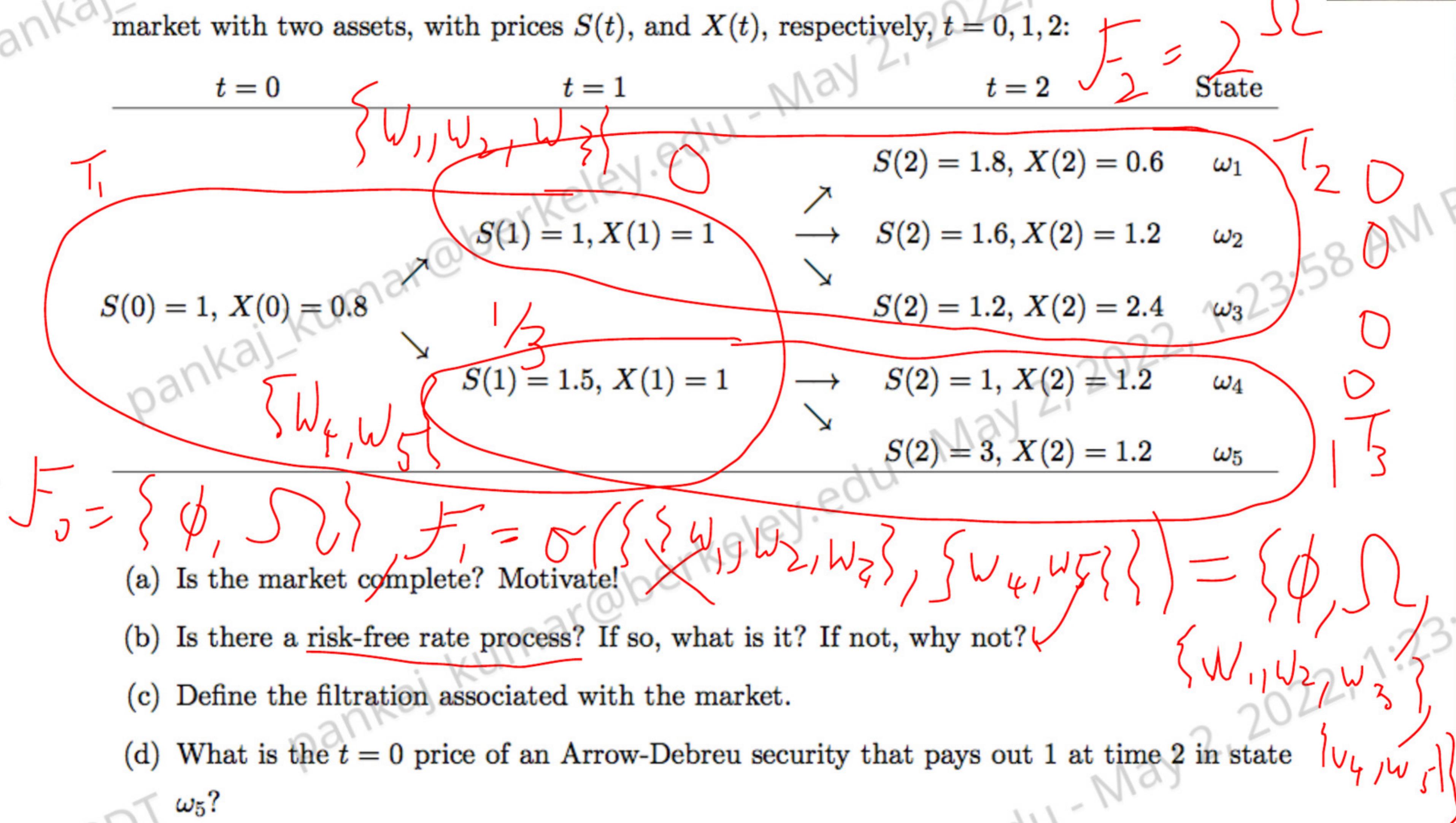
Current stock price of MFE2014Class Inc. is \$40, over the next two 3-month periods it's expected to go up by 10% or down by 10%. The risk free rate is 12% per year with continuous compounding.

- Compute the price of a 6-month European put option with a strike price of \$42?
- Compute the price of a 6-month American put option with a strike price of \$42?



Sample Problem III

Multi-period model: Consider the following tree-representation of a two-period (three-date) market with two assets, with prices $S(t)$, and $X(t)$, respectively, $t = 0, 1, 2$:



$$T_2 \quad D = \begin{bmatrix} 1.8 & 1.6 \\ 0.6 & 1.2 \end{bmatrix} \quad 2.4 \quad \times$$

$$T_3 \quad D = \begin{bmatrix} 1 & 3 \\ 1.2 & 1.2 \end{bmatrix} \quad \checkmark$$

$$T_1 \quad D = \begin{bmatrix} 1 & & 1.5 \\ & 1 & \end{bmatrix} \quad \checkmark$$

$$T_1 \quad R_F = \cancel{0.8} \quad \checkmark \quad T_2 \quad (\alpha \beta) \begin{pmatrix} 1.8 & 1.6 & 1.2 \\ 0.6 & 1.2 & 2.4 \end{pmatrix}$$

$$\cancel{T_3} \quad R_F = \cancel{1.2} \quad \checkmark \quad = (r \quad \checkmark \quad r)$$

$$\alpha = \frac{r}{2} \quad \beta = \frac{r}{6}$$

$$\frac{r}{2}x_1 + \frac{r}{6}x_1 = \frac{2r}{3}$$
$$R_f = \frac{r}{\frac{2r}{3}} = \frac{3}{2} = 1.5$$

$$(0) \quad \begin{pmatrix} & 0.5 \\ 1 & \end{pmatrix} = \frac{1}{3}$$

$$(0) \quad \begin{pmatrix} & 2/5 \\ -3/5 & \end{pmatrix} = \frac{-2}{15}$$