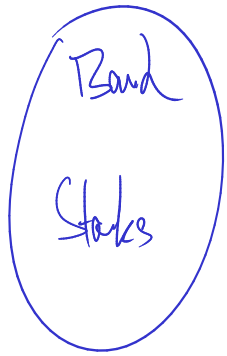


Last time: No Arb (assumption)



Market

NTA (option)

AFP: V_0 \rightarrow if when allowed to trade the
NTA at price V_0

the extended market remains arbitrage-free.

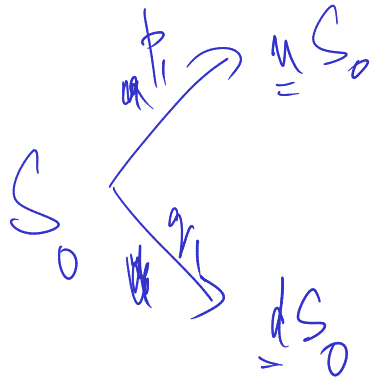
Replication!

If the payoff of the NTA can be
replicated using only tradeable assets

then $X_{0N} =$ initial wealth of the rep port
 $= V_{0N} = \text{AFP}.$

u

last time t_0 Binomial model: Multi period version



Formula: Payoff = V_N

$$\underline{AFP} = \frac{1}{D_n} \overset{2}{E}_n(\underline{D}_n \underline{V}_n)$$

4.4. Conditional expectation.

Definition 4.28. Let X be a random variable, and $n \leq N$. We define $\mathbf{E}(X | \mathcal{F}_n) = \mathbf{E}_n X$ to be the *random variable* given by

$$\mathbf{E}_n X(\omega) = \sum_{x_i \in \text{Range}(X)} x_i \mathbf{P}(X = x_i | \Pi_n(\omega))$$

where

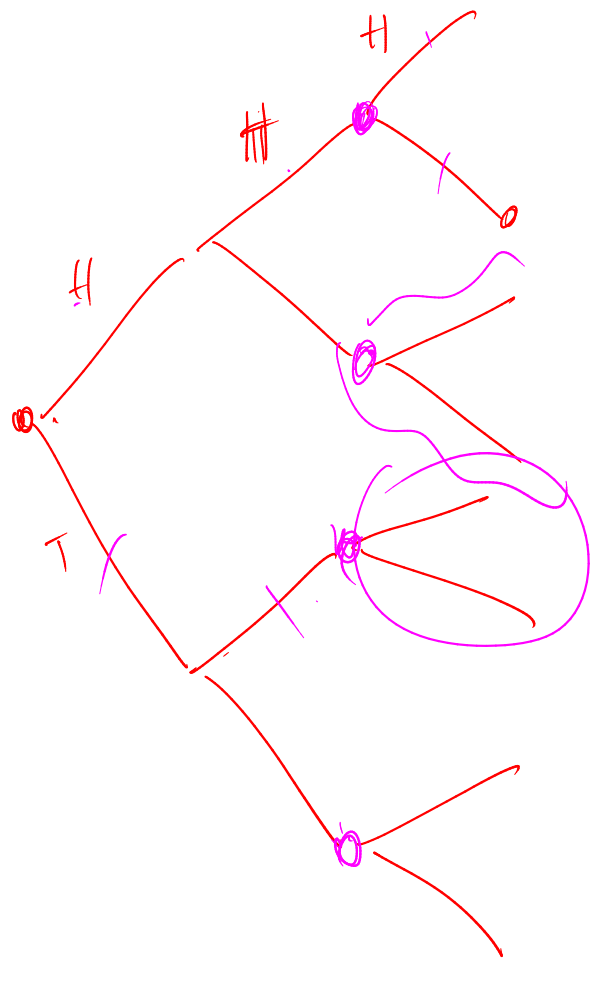
$$\Pi_n(\omega) = \{\omega' \in \Omega \mid \omega'_1 = \omega_1, \dots, \omega'_n = \omega_n\}$$

Remark 4.29. The above formula does not generalize well to infinite probability spaces. We will develop certain properties of \mathbf{E}_n , and then only use those properties going forward.

Example 4.30. If we represent Ω as a tree, $\mathbf{E}_n X$ can be computed by averaging over leaves.

Remark 4.31. $\mathbf{E}_n X$ is the “best approximation” of X given only the first n coin tosses.

$\mathbf{E}_n X \rightarrow$ “Best approximation of X given info up to time n ”



- 1.
- 2.
- 3.
- 4.
- 5.
- 6.
- 7.
- 8.

$E_2 X$

$\left. \begin{array}{l} 1 \\ 2 \end{array} \right\} \rightarrow p_1 \cdot 1 + q_1 \cdot 2$

$\rightarrow p_1 \cdot 3 + q_1 \cdot 4$

$\rightarrow p_1 \cdot 3 + q_1 \cdot 4$

(3 coin tosses)

Proposition 4.32. The conditional expectation $\mathbf{E}_n X$ defined by the above formula satisfies the following two properties:

(1) $\mathbf{E}_n X$ is an \mathcal{F}_n -measurable random variable.

(2) For every $A \in \mathcal{F}_n$, $\sum_{\omega \in A} \mathbf{E}_n X(\omega) p(\omega) = \sum_{\omega \in A} X(\omega) p(\omega)$.

($\mathbf{E}_n X$ dep only on first n coin tosses.)

Remark 4.33. This property is used to define conditional expectations in the continuous time setting. It turns out that there is exactly one random variable that satisfies both the above properties; and thus we define $\mathbf{E}_n X$ to be the unique random variable which satisfies both the above properties.

Remark 4.34. Note, choosing $A = \Omega$, we see $\mathbf{E}(\mathbf{E}_n X) = \mathbf{E}X$.

On A : Avg of $X = \frac{1}{P(A)} \sum_{\omega \in A} X(\omega) p(\omega)$

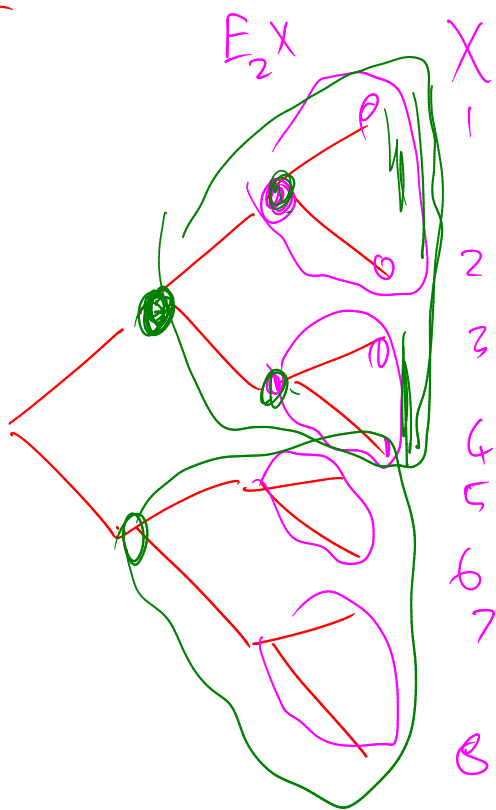
→ Avg of $\mathbf{E}_n X$ (on A) = $\frac{1}{P(A)} \sum_{\omega \in A} \mathbf{E}_n X(\omega) p(\omega)$

← equal.

Proposition 4.35. (1) If X, Y are two random variables and $\alpha \in \mathbb{R}$, then $\underline{E}_n(X + \alpha Y) = \underline{E}_n X + \alpha \underline{E}_n Y$.

(2) (Tower property) If $m \leq n$, then $\underline{E}_m(\underline{E}_n X) = \underline{E}_m X$.

(3) If X is \mathcal{F}_n measurable, and Y is any random variable, then $\underline{E}_n(XY) = X \underline{E}_n Y$.

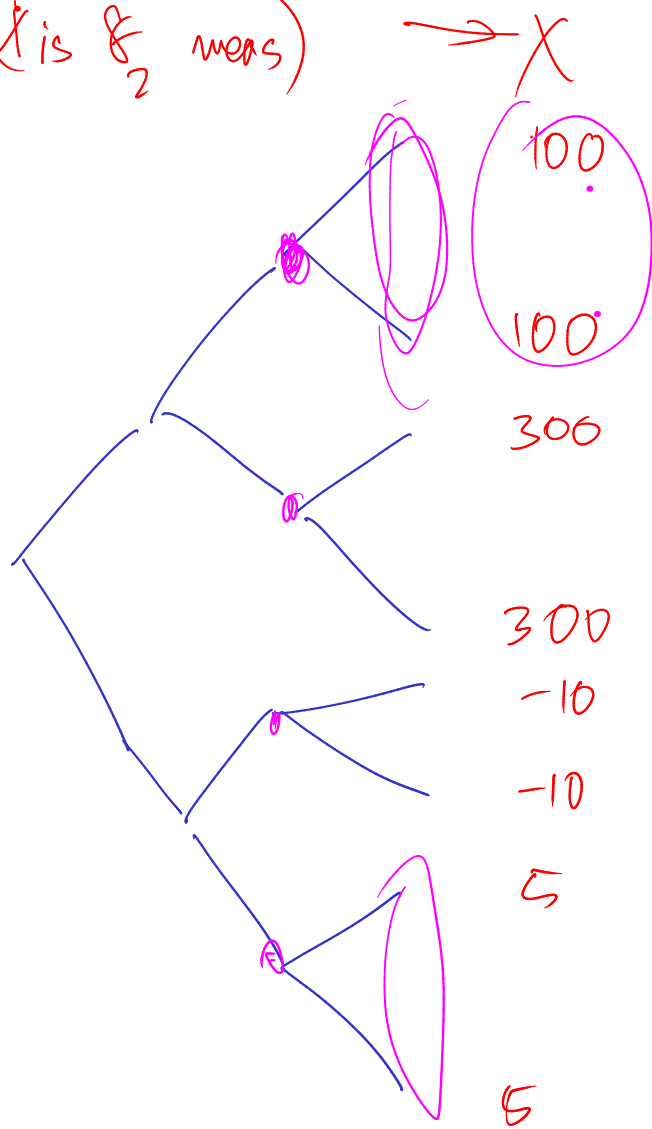


$E_2 X$

$E_1 X$

h

(X is ξ_2 meas)



Y

1 }
2 }
3 }
4 }
5 }
6 }
7 }
8 }

$E_2(XY)$

$$100(1p_1 + 2q_1)$$

Proposition 4.36. (1) If X is measurable with respect to \mathcal{F}_n , then $E_n X = X$.

(2) If X is independent of \mathcal{F}_n then $E_n X = EX$.

Remark 4.37. We say X is independent of \mathcal{F}_n if for every $A \in \mathcal{F}_n$ and $B \subseteq \mathbb{R}$, the events A and $\{X \in B\}$ are independent.

Example 4.38. If X only depends on the $(n+1)^{\text{th}}$, $(n+2)^{\text{th}}$, \dots , n^{th} coin tosses and *not* the 1^{st} , 2^{nd} , \dots , n^{th} coin tosses, then X is independent of \mathcal{F}_n .

Notation: $\{X \in B\} = \{\omega \mid X(\omega) \in B\}$.

Proposition 4.39 (Independence lemma). If X is independent of \mathcal{F}_n and Y is \mathcal{F}_n -measurable, and $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ is a function then

$$\mathbb{E}_n f(X, Y) = \sum_{i=1}^m f(x_i, Y) P(X = x_i), \quad \text{where } \{x_1, \dots, x_m\} = X(\Omega).$$

Range(X).

Say $Y = y$ (some const)

$$\mathbb{E}_n f(X, \underline{y}) = \mathbb{E} f(X, \underline{y}) = \sum_{x_i \in \text{Range}(X)} f(x_i, \underline{y}) P(X = x_i)$$

Average the ind RV treating the other RV as a const.

4.5. Martingales.

Definition 4.40. A stochastic process is a collection of random variables X_0, X_1, \dots, X_N .

Example 4.41. Typically X_n is the wealth of an investor at time n , or S_n is the price of a stock at time n .

Definition 4.42. A stochastic process is adapted if X_n is \mathcal{F}_n -measurable for all n . (Non-anticipating.)

Remark 4.43. Requiring processes to be adapted is fundamental to Finance. Intuitively, being adapted forbids you from trading today based on tomorrow's stock price. All processes we consider (prices, wealth, trading strategies) will be adapted.

Example 4.44 (Money market). Let $Y_0 = Y_0(\omega) = \underline{a} \in \mathbb{R}$. Define $\underline{Y_{n+1}} = (1+r)\underline{Y_n}$. (Here \underline{r} is the interest rate.)

Example 4.45 (Stock price). Let $S_0 \in \mathbb{R}$. Define $\underline{S_{n+1}}(\omega) = \begin{cases} \underline{uS_n}(\omega) & \underline{\omega_{n+1} = 1}, \\ \underline{dS_n}(\omega) & \underline{\omega_{n+1} = -1}. \end{cases}$

$\hookrightarrow S_n$ is a stock process
 S_n is an ADAPTED process

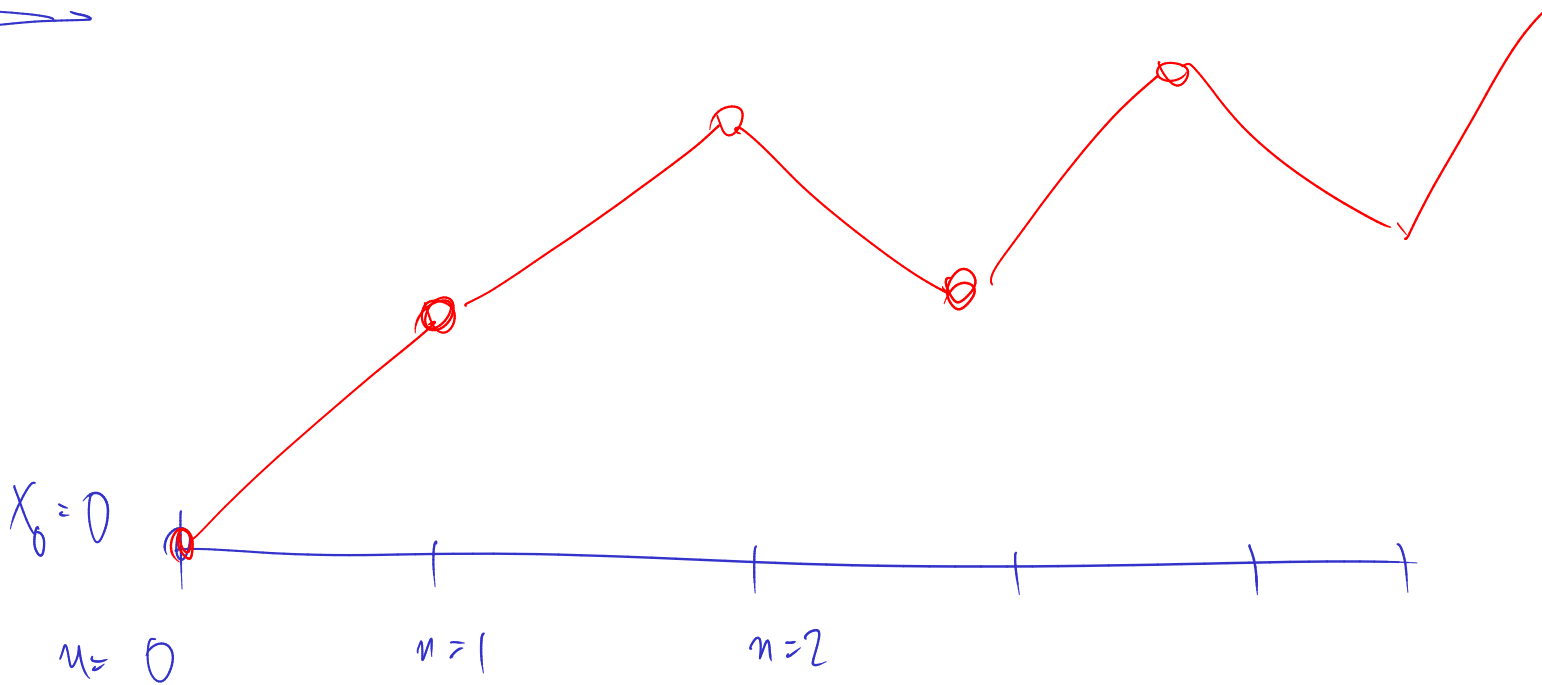
$Y_n = (1+r)^n a.$
(adapted process)

Definition 4.46. We say an adapted process \underline{M}_n is a martingale if $\underline{E}_n \underline{M}_{n+1} = \underline{M}_n$. (Recall $\underline{E}_n Y = \underline{E}(Y | \mathcal{F}_n)$.)

Remark 4.47. Intuition: A martingale is a “fair game”.

Example 4.48 (Unbiased random walk). If ξ_1, \dots, ξ_N are i.i.d. and mean zero, then $\underline{X}_n = \sum_{k=1}^n \xi_k$ is a martingale.

$E \xi_n = 0 \forall n$



Check X is a mg: Note $X_{n+1} = X_n + \xi_{n+1}$

$$\Rightarrow \underline{E}_n (X_{n+1}) = \underline{E}_n (X_n + \xi_{n+1})$$

$$= E_n X_n + E_n \zeta_{n+1}$$

X_n is \mathcal{F}_n meas ζ_{n+1} is ind of \mathcal{F}_n .

$$= X_n + \cancel{E_n \zeta_{n+1}}$$

\downarrow
0

$$\Rightarrow E_n X_{n+1} = X_n$$

$\Rightarrow X$ is a mg!

Remark 4.49. If M is a martingale, then for every $\underline{m} \leq \underline{n}$, we must have $\underline{E}_m M_n = \underline{M}_m$.

Remark 4.50. If \underline{M} is a martingale then $\underline{E} M_n = \underline{E} M_0 = \underline{M}_0$.

$$E_n M_{n+2} \stackrel{\text{tower}}{=} E_n E_{n+1} M_{n+2} \stackrel{\text{Mg}}{=} E_n M_{n+1} \stackrel{\text{Mg}}{=} M_n$$

→ Note

$$\underline{E} M_n = E_0 M_n \stackrel{4.49}{=} M_0$$

4.6. Change of measure.

- Gambling in a Casino: If it's a martingale, then on average you won't make or lose money.
- Stock market: Bank always pays interest! Not looking for a "break even" strategy.
- Mathematical tool that helps us price securities: Find a Risk Neutral Measure.
 - ▷ Discounted stock price is (usually) not a martingale.
 - ▷ Invent a "risk neutral measure" which the discounted stock price is a martingale.
 - ▷ Securities can be priced by taking a conditional expectation *with respect to the risk neutral measure*. (That's the meaning of \tilde{E}_n in Proposition 4.1.)

$$\text{AFP at time } n = \frac{1}{\underline{\underline{D_n}}} \left(\tilde{E}_n (D_n V_D) \right)$$

Cond exp w.r.t the "Risk Neutral Measure"

Definition 4.51. Let $D_n = (1 + r)^{-n}$ be the discount factor. (So D_n \$ in the bank at time 0 becomes 1 \$ in the bank at time n .)

- Invent a new probability mass function \tilde{p} .
- Use a tilde to distinguish between the new, invented, probability measure and the old one.
 - ▷ \tilde{P} the probability measure obtained from the PMF \tilde{p} (i.e. $\tilde{P}(A) = \sum_{\omega \in A} \tilde{p}(\omega)$).
 - ▷ \tilde{E}, \tilde{E}_n conditional expectation with respect to \tilde{P} (the new “risk neutral” coin)

Definition 4.52. We say P and \tilde{P} are equivalent if for every $A \in \mathcal{F}_N$, $P(A) = 0$ if and only if $\tilde{P}(A) = 0$.

Definition 4.53. A risk neutral measure is an equivalent measure \tilde{P} under which $D_n S_n$ is a martingale. (I.e. $\tilde{E}_n(D_{n+1} S_{n+1}) = D_n S_n$ for all n)

Remark 4.54. If there are more than one risky assets, S^1, \dots, S^k , then we require $D_n S_n^1, \dots, D_n S_n^k$ to all be martingales under the risk neutral measure \tilde{P} .

Remark 4.55. Proposition 4.1 says that any security with payoff V_N at time N has arbitrage free price $V_n = \frac{1}{D_n} \tilde{E}_n(D_N V_N)$ at time n . (Called the risk neutral pricing formula.)

hde

Proposition 4.56. Let $\tilde{\mathbf{P}}$ be an equivalent measure under which the coins are i.i.d. and land heads with probability \tilde{p}_1 and tails with probability $\tilde{q}_1 = 1 - \tilde{p}_1$.

(1) Under $\tilde{\mathbf{P}}$, we have $\tilde{\mathbf{E}}_n(D_{n+1}S_{n+1}) = \frac{\tilde{p}_1 u + \tilde{q}_1 d}{1+r} D_n S_n$. ~~A~~

(2) $\tilde{\mathbf{P}}$ is the risk neutral measure if and only if $\tilde{p}_1 u + \tilde{q}_1 d = 1+r$. (Explicitly $\tilde{p}_1 = \frac{1+r-d}{u-d}$, and $\tilde{q}_1 = \frac{u-(1+r)}{u-d}$.)

① $\tilde{\mathbf{E}}_n(D_{n+1}S_{n+1}) = (1+r)^{-(n+1)} \tilde{\mathbf{E}}_n S_{n+1}$

$$\implies (1+r)^{-(n+1)} \tilde{\mathbf{E}}_n(S_n X_{n+1})$$

where $X_{n+1} = \begin{cases} u & \omega_{n+1} = 1 \\ d & \omega_{n+1} = -1 \end{cases}$

$$= (1+r)^{-(n+1)} \tilde{\mathbf{E}}_n(S_n X_{n+1})$$

(S_n is \mathcal{F}_n meas, X_{n+1} is ind)

$$= (1+r)^{-(n+1)} S_n \tilde{E} X_{n+1}$$

$$= (1+r)^{-n+1} S_n (\tilde{r}_1^n + \tilde{r}_1 d)$$

$$= \left(\frac{\tilde{r}_1^n + \tilde{r}_1 d}{1+r} \right) D_n S_n //$$

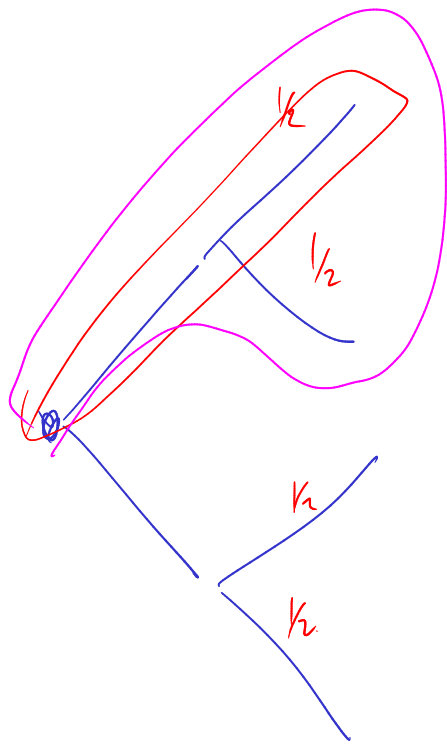
Theorem 4.57. Let X_n represent the wealth of a portfolio at time n . The portfolio is self-financing portfolio if and only if the discounted wealth $D_n X_n$ is a martingale under the risk neutral measure \tilde{P} .

Remark 4.58. Recall a portfolio is self financing if $X_{n+1} = \Delta_n S_{n+1} + (1+r)(X_n - \Delta_n S_n)$ for some adapted process Δ_n .

- (1) That is, self-financing portfolios use only tradable assets when trading, and don't look into the future.
- (2) All replication has to be done using self-financing portfolios.

Check: Self fin $\Rightarrow D_n X_n$ is a \tilde{P} mg.

$$\begin{aligned}
 \tilde{E}_n(D_{n+1} X_{n+1}) &= \tilde{E}_n(D_{n+1} (\Delta_n S_{n+1} + (1+r)(X_n - \Delta_n S_n))) \\
 &= \Delta_n \underbrace{\tilde{E}_n(D_{n+1} S_{n+1})}_{D_n S_n} + D_n (X_n - \Delta_n S_n) \\
 &= D_n X_n \quad \text{Q.E.D.}
 \end{aligned}$$



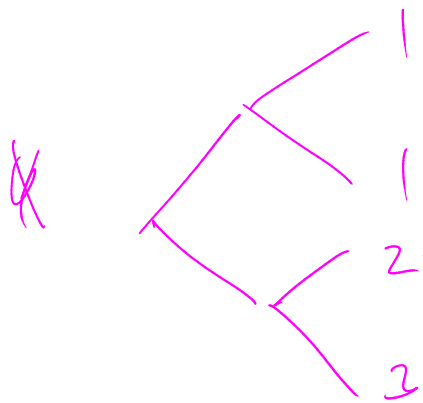
x
 1
 2
 3
 4

$\frac{3}{2}$
 $\frac{3}{2}$
 $\frac{7}{2}$

$$A = \{(1, 1)\} \in \mathcal{R}_2 \not\subseteq \mathcal{R}_1$$

$$B = \{(1, 1), (1, -1)\} \in \mathcal{R}_1$$

$$\tau(x) = \{ \{x \in B\} \mid B \subseteq \mathcal{R} \}$$



$$\tau(x) = \{ \{ (1, 1), (1, -1) \}, \{ (1, 1) \}, \{ (-1, -1) \}, \{ (-1, 1), (-1, -1) \}, \emptyset, \Omega \}$$

