

①  $\exists$  no arb if and only if  $\exists \mathbb{Q}$  s.t.

$$C_0 = \frac{1}{1+r} \mathbb{E}^{\mathbb{Q}} [C_1] \quad \text{risk-neutral prob.}$$

for any traded asset  $C$ .

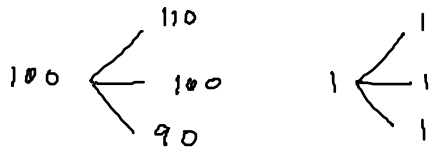
①  $\Leftrightarrow A_d < A_0(1+r) < A_u$

$\Leftrightarrow 0 < q < 1$

$$A_0 \begin{cases} A_u \\ A_d \end{cases} \begin{matrix} \alpha \\ \beta \end{matrix} + 1 \begin{cases} 1+r \\ 1+r \end{cases} = \underbrace{A_0 \alpha + \beta}_{C_0} \begin{cases} A_u \alpha + (1+r)\beta = C_u \\ A_d \alpha + (1+r)\beta = C_d \end{cases}$$

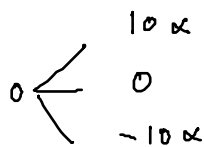
## Incomplete Markets

(more outcomes than traded assets)



① direct approach --  $(\alpha, \beta)$

$$V_0 = 0 \Rightarrow \beta = -100\alpha$$



for any arb  $\mathbb{P}(V_1 \geq 0) = 1$   
 $\Rightarrow \alpha = 0$

then  $\mathbb{P}(V_1 > 0) = 0$

but an arb also has  $\mathbb{P}(V_1 > 0) > 0$

$\therefore$  there are no arb in this market!

②  $A_0 = \frac{1}{1+r} \mathbb{E}^Q[A_1]$

$$100 = 1 \left( 110 q_u + 100 q_m + 90 q_d \right)$$

$1 - q_u - q_d$

seek:  $q_u + q_m + q_d = 1$ ,  $q_u > 0$ ,  $q_m > 0$ ,  $q_d > 0$

$$\Rightarrow 0 = 10 q_u - 10 q_d \Rightarrow q_u = q_d = q$$

$$q_u > 0 + q_m > 0 \Rightarrow q > 0$$

$$q_d > 0 \Rightarrow 1 - q_u - q_d > 0 \Rightarrow 1 - 2q > 0$$

$$\Rightarrow q < \frac{1}{2}$$

$\forall 0 < q < \frac{1}{2} \exists$  no arb

How to value ?  $\begin{matrix} 0 \\ 10 \\ 0 \end{matrix}$

$$B_0 = \frac{1}{1+r} \mathbb{E}^Q [B_1] \quad \text{but } Q \text{ is not unique!}$$

$$= \frac{1}{1+r} (6 \cdot q + 10 \cdot (1-2q) + 0 \cdot q)$$

$$= 10(1-2q) = B_0^* \Rightarrow q = (1 - B_0^*/10) \frac{1}{2}$$

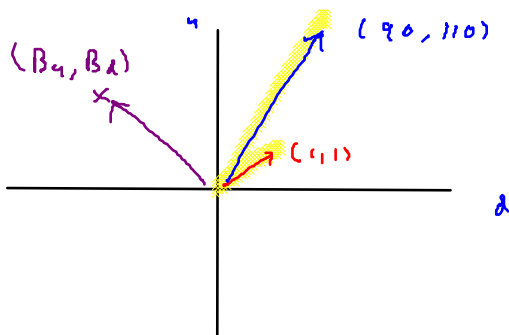
$$0 < B_0 < 10$$

any such  $B_0$  is a viable no arb. price!

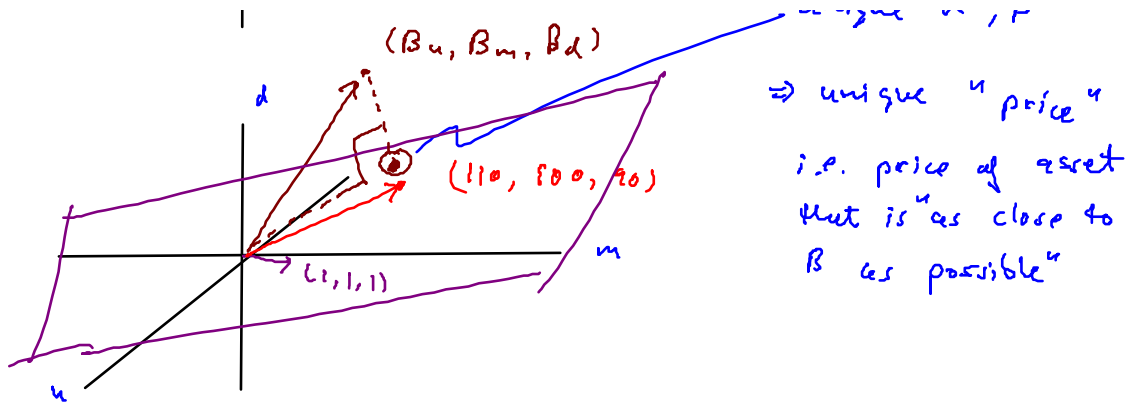
try replication:

$$100 \begin{matrix} \swarrow 110 \\ \searrow 90 \end{matrix} \begin{matrix} \swarrow \\ \searrow \end{matrix} \begin{matrix} \alpha \\ \beta \end{matrix} = \begin{cases} 110\alpha + \beta = B_u \\ 100\alpha + \beta = B_m \\ 90\alpha + \beta = B_d \end{cases}$$

$$100 \begin{matrix} \swarrow 110 \\ \searrow 90 \end{matrix} \begin{matrix} \swarrow 1 \\ \searrow 1 \end{matrix} = \begin{cases} 110\alpha + \beta = B_u \\ 90\alpha + \beta = B_d \end{cases}$$



unique  $\alpha, \beta$



$$\hat{B}_1 = \arg \min_{\hat{B}_1} \mathbb{E}^{\mathbb{P}} \left[ (\hat{B}_1 - B_1)^2 \right]$$

$\hookrightarrow \alpha A_1 + \beta$

Value of  $B = \alpha A_0 + \beta \leftarrow$  is it arbitrage free?

Such value is called the min var hedge price.

$$B_0 \begin{cases} 20 \\ 10 \\ 0 \end{cases}$$

$$B_0 = 1(20 - q) + (1 - 2q)10 + q \cdot 0$$

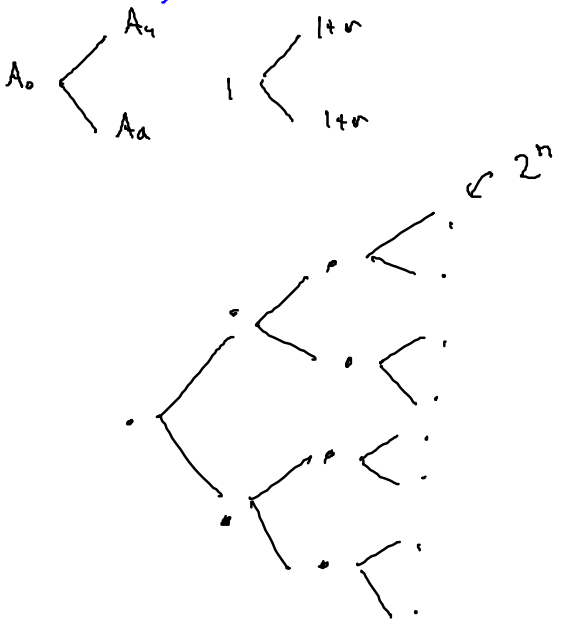
$$= 10$$

what if  $B_0 = 11 \dots$  find the arb.

$$\begin{cases} 0 \\ 10 \\ 20 \end{cases}$$

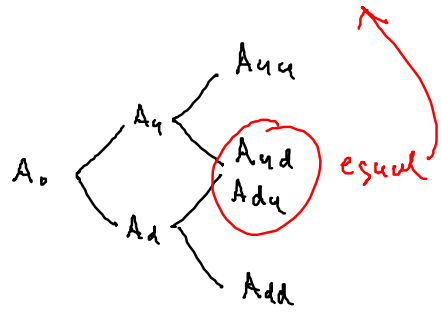


now to write in terms of observables?

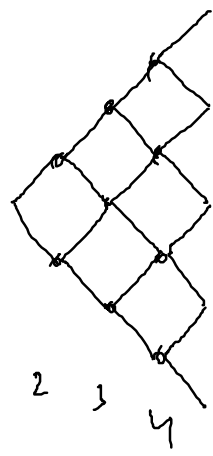


$O(2^n)$

recombining trees:

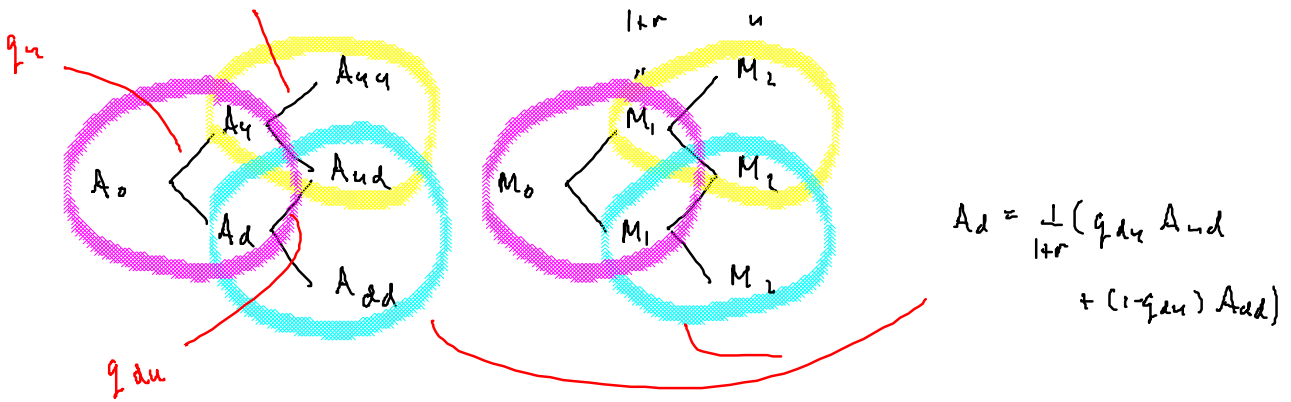


$O(n^2)$



Qua

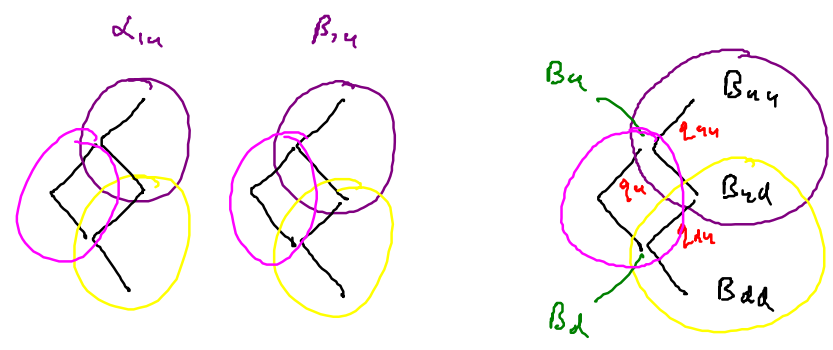
$(1+n)^2$



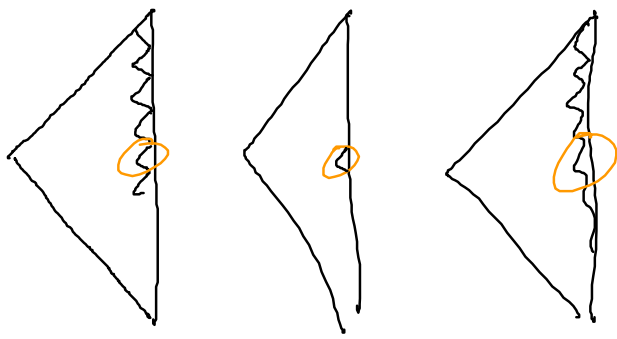
$$\begin{aligned}
 (\alpha_0, \beta_0) &= (\alpha_{1u}, \beta_{1u}) & \alpha_{1u} A_u + \beta_{1u} M_u &= \alpha_0 A_u + \beta_0 M_u \\
 &= (\alpha_{1d}, \beta_{1d}) & \alpha_{1d} A_d + \beta_{1d} M_d &= \alpha_0 A_d + \beta_0 M_d \\
 &= \alpha_1 A_1 + \beta_1 M_1 & & \text{Self-financing conditions} \\
 &= \alpha_0 A_1 + \beta_0 M_1 & & \\
 \Rightarrow (\alpha_1 - \alpha_0) A_1 + (\beta_1 - \beta_0) M_1 &= 0
 \end{aligned}$$

$\exists \alpha \in \mathbb{R}$  s.t.  $A_t = \frac{1}{1+r} \mathbb{E}^\alpha [A_{t+1}]$  iff there is no arb.

so find  $q_u, q_{ud}, q_{du}$  and check that  $q_u \in (0,1), q_{ud} \in (0,1), q_{du} \in (0,1)$ .



$$\beta_u = \frac{1}{1+r} (q_{uu} \beta_{uu} + (1-q_{uu}) \beta_{ud})$$



$A_{t_0}, A_{t_1}, A_{t_2}, \dots, A_{t_n} \leftarrow$  observed

$$r_{t_n} = \frac{A_{t_n} - A_{t_{n-1}}}{A_{t_{n-1}}}$$

$$\rightarrow r_{t_n} = \ln \left( \underbrace{A_{t_n} / A_{t_{n-1}}}_{\uparrow} \right) \leftarrow$$

$$\ln(\alpha - 1 + 1) = \ln(1 - \underline{(1-\alpha)}) \sim \alpha - 1 + \dots$$

$$= \frac{A_{t_n} - 1}{A_{t_{n-1}}}$$

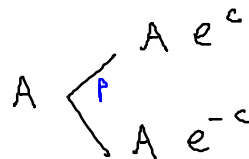
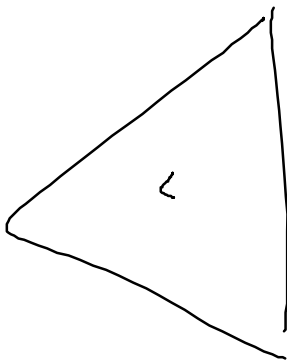
$$= \frac{A_{t_n} - A_{t_{n-1}}}{A_{t_{n-1}}}$$

$$\ln(1+x) \sim x$$

convexity correction

$$\mathbb{E}^{\mathbb{P}}[r] = \left( \mu^* - \frac{1}{2}(\sigma^*)^2 \right) \Delta t = \frac{1}{N} \sum_{n=1}^N r_{t_n}$$

$$\mathbb{V}^{\mathbb{P}}[r] = (\sigma^*)^2 \Delta t = \frac{1}{N} \sum_{n=1}^N (r_{t_n} - \mathbb{E}^{\mathbb{P}}[r])^2$$



$$\hookrightarrow A_n = A_{n-1} e^{c x_n}$$

$x_1, x_2, \dots$  iid Bernoulli r.v.

$$\mathbb{P}(x_1 = +1) = p$$

$$\mathbb{P}(x_1 = -1) = 1-p$$

$$r = \ln(A_n / A_{n-1}) = c x_n$$

$$\mathbb{E}^{\mathbb{P}}[r] = \left( \mu^* - \frac{1}{2}(\sigma^*)^2 \right) \Delta t = c \mathbb{E}^{\mathbb{P}}[x_n] = c (1 \cdot p + (-1) \cdot (1-p))$$

$$= c (2p-1)$$



$$\Rightarrow p = \frac{1}{2} \left[ 1 + \frac{\mu^* - \frac{1}{2}(\sigma^*)^2}{c} \Delta t \right] \quad \leftarrow$$

$$\begin{aligned} \mathbb{V}^{\mathbb{P}}[r] &= (\sigma^*)^2 \Delta t = c^2 \mathbb{V}^{\mathbb{P}}[x_n] = c^2 ( \mathbb{E}^{\mathbb{P}}[x_n^2] - (\mathbb{E}^{\mathbb{P}}[x_n])^2 ) \\ &= c^2 (1 - (2p-1)^2) = c^2 \left( 1 - \frac{(\mu^* - \frac{1}{2}(\sigma^*)^2)^2 \Delta t^2}{c^2} \right) \end{aligned}$$

$$\Rightarrow c^2 = (\sigma^*)^2 \Delta t + (\mu^* - \frac{1}{2}(\sigma^*)^2)^2 \Delta t^2$$

$$\Rightarrow c = \sigma^* \sqrt{\Delta t} \sqrt{1 + \left( \frac{\mu^* - \frac{1}{2}(\sigma^*)^2}{\sigma^*} \right)^2 \Delta t} \quad \leftarrow$$

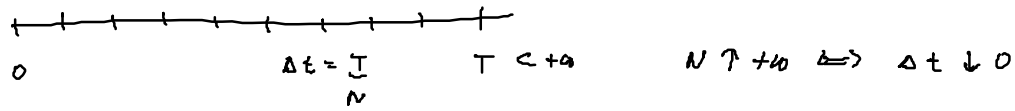
$$\sim \sigma^* \sqrt{\Delta t} + o(\sqrt{\Delta t}) \quad \lim_{\Delta t \downarrow 0} \frac{o(\sqrt{\Delta t})}{\sqrt{\Delta t}} = 0$$

$\Rightarrow$

$$p = \frac{1}{2} \left( 1 + \frac{\mu^* - \frac{1}{2}(\sigma^*)^2}{\sigma^*} \sqrt{\Delta t} \right) + o(\sqrt{\Delta t})$$

$$c = \sigma^* \sqrt{\Delta t} + o(\sqrt{\Delta t})$$

$$A \begin{cases} e^{\sigma^* \sqrt{\Delta t}} A \\ e^{-\sigma^* \sqrt{\Delta t}} A \end{cases}$$



$$A \begin{cases} A e^{\sigma^* \sqrt{\Delta t}} \\ A e^{-\sigma^* \sqrt{\Delta t}} \end{cases}$$

$$A_T = A_0 \exp \left\{ \underbrace{\sigma^* \sqrt{\Delta t} (x_1 + x_2 + \dots + x_N)}_{X_T} \right\}$$

$$X_T \xrightarrow[N \rightarrow +\infty]{d} \mathcal{N}(-; \cdot) \quad \text{by CLT}$$

$$\mathbb{E}^P[X_T] = \sigma^* \sqrt{\Delta t} \cdot N \cdot \mathbb{E}^P[x_1]$$

$$= \sigma^* \sqrt{\Delta t} \cdot N \cdot (2\rho - 1)$$

$$= \sigma^* \sqrt{\Delta t} \cdot N \cdot \left( \frac{\mu^* - \frac{1}{2}(\sigma^*)^2}{\sigma^*} \sqrt{\Delta t} + o(\sqrt{\Delta t}) \right)$$

$\hookrightarrow \sqrt{T/N}$

$$\xrightarrow{\Delta t \downarrow 0} (\mu^* - \frac{1}{2}(\sigma^*)^2) T$$

$$\mathbb{V}^P[X_T] = (\sigma^*)^2 \Delta t \cdot N \cdot \mathbb{V}^P[x_1]$$

$$= (\sigma^*)^2 \Delta t \cdot N \cdot (1 - (2\rho - 1)^2)$$

$$\hookrightarrow \sqrt{T/N} \quad \hookrightarrow \left( \frac{\mu^* - \frac{1}{2}(\sigma^*)^2}{\sigma^*} \right)^2 \Delta t + o(\Delta t)$$

$$\xrightarrow{\Delta t \downarrow 0} (\sigma^*)^2 T$$

$$X_T \stackrel{d}{\underset{\mathbb{P}}{\rightarrow}} \mathcal{N}\left(\left(\mu^* - \frac{1}{2}(\sigma^*)^2\right)T; (\sigma^*)^2 T\right)$$

$$A_T \stackrel{d}{=} A_0 e^{X_T} \quad \text{is a log-normal dist.}$$

$$\stackrel{d}{=} A_0 e^{\left(\mu^* - \frac{1}{2}(\sigma^*)^2\right)T + \sigma^* \sqrt{T} z}$$

$$z \underset{\mathbb{P}}{\sim} \mathcal{N}(0; 1)$$

$$\mathbb{E}^{\mathbb{P}}[A_T] = A_0 \mathbb{E}^{\mathbb{P}}[e^{X_T}]$$

$$= A_0 e^{\left(\mu^* - \frac{1}{2}(\sigma^*)^2\right)T} \mathbb{E}^{\mathbb{P}}[e^{\sigma^* \sqrt{T} z}]$$

reminder mgf of normal  $\mathbb{E}[e^{uz}] = e^{\frac{1}{2}u^2}$   $z \sim \mathcal{N}(0,1)$

$$\Rightarrow \mathbb{E}^{\mathbb{P}}[A_T] = A_0 e^{\mu^* T}$$

$\mu^*$  is the "drift" of the asset

$$A \begin{cases} A e^{\sigma^* \sqrt{\Delta t}} \\ A e^{-\sigma^* \sqrt{\Delta t}} \end{cases} \quad 1 \begin{cases} e^{r \Delta t} \\ e^{r \Delta t} \end{cases} \quad \triangleleft$$

$$A = e^{-r \Delta t} ( q A e^{\sigma^* \sqrt{\Delta t}} + (1-q) A e^{-\sigma^* \sqrt{\Delta t}} )$$

$$\Rightarrow q = \frac{e^{r \Delta t} - e^{-\sigma^* \sqrt{\Delta t}}}{e^{\sigma^* \sqrt{\Delta t}} - e^{-\sigma^* \sqrt{\Delta t}}}$$

$$e^x \sim 1 + x + \frac{1}{2} x^2 + \dots$$

$$= \frac{(1 + r \Delta t) - (1 - \sigma^* \sqrt{\Delta t} + \frac{1}{2} (\sigma^*)^2 \Delta t + \dots)}{(1 + \sigma^* \sqrt{\Delta t} + \frac{1}{2} (\sigma^*)^2 \Delta t + \dots) - (1 - \sigma^* \sqrt{\Delta t} + \frac{1}{2} (\sigma^*)^2 \Delta t)}$$

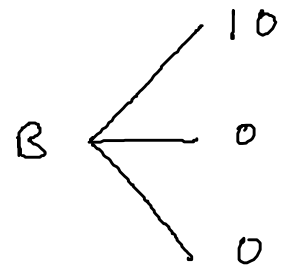
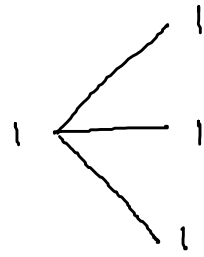
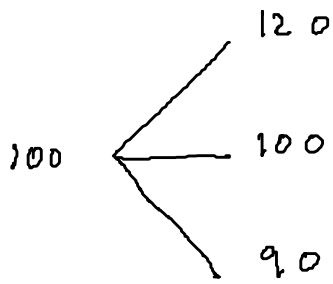
$$= \frac{\sigma^* \sqrt{\Delta t} + (r - \frac{1}{2} (\sigma^*)^2) \Delta t}{2 \sigma^* \sqrt{\Delta t}} + \dots$$

$$q = \frac{1}{2} \left( 1 + \frac{r - \frac{1}{2} (\sigma^*)^2}{\sigma^*} \sqrt{\Delta t} \right) + \dots$$

$$p = \frac{1}{2} \left( 1 + \frac{\frac{1}{2} (\sigma^*)^2 - r}{\sigma^*} \sqrt{\Delta t} \right) + \dots$$

$$A_T \stackrel{d}{=} A_0 e^{X_T}$$

$$X_T \underset{Q}{\sim} \mathcal{N} \left( (r - \frac{1}{2} (\sigma^*)^2) T ; (\sigma^*)^2 T \right)$$



Find no and range for B.