

FTAP: no arbitrage  $\Leftrightarrow \exists \mathbb{Q} \sim \mathbb{P}$  s.t.  
 $\forall$  traded assets  $X$ , we have

$$\tilde{X}_t = E^{\mathbb{Q}} [ \tilde{X}_T | \mathcal{F}_t ]$$

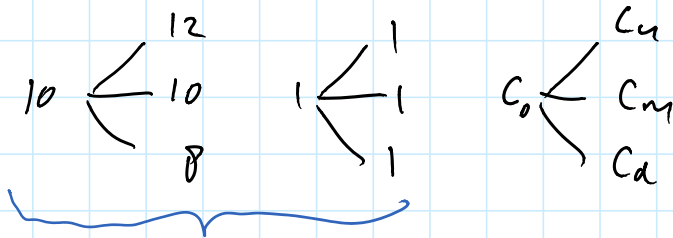
$$\tilde{X}_t = \frac{X_t}{B_t}, \text{ where } B_t \text{ is a}$$

numeraire asset

$$( B_t > 0 \text{ a.s.} )$$

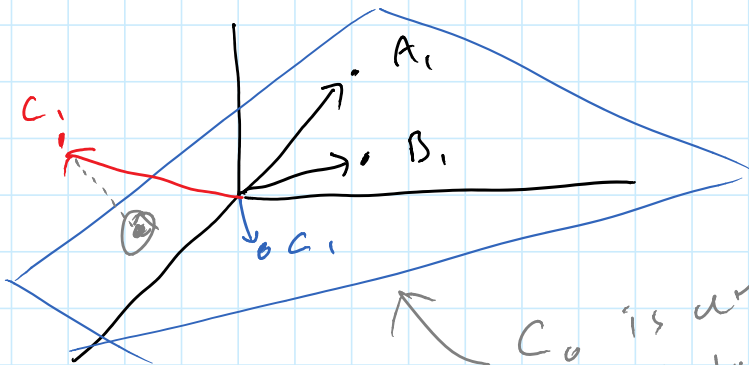
$\mathbb{Q}$  is not necessarily unique!

e.g.



$\mathbb{Q}$  is not unique  $\Rightarrow$

$C_0$  is not necessarily unique



$C_0$  is unique if it lands in  $\text{span}(A_1, B_1)$

$$\min_{(\alpha, \beta)} E^{\mathbb{P}} [ (\alpha A_1 + \beta B_1 - C_1)^2 ]$$

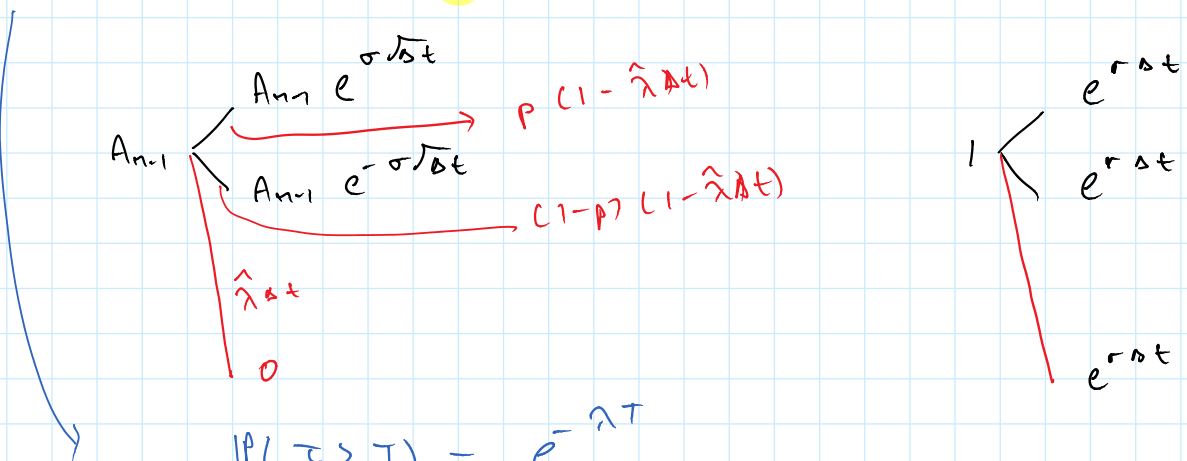
local risk minimizing hedge.

$$A_n = A_{n-1} e^{\sigma \sqrt{\Delta t} x_n}, \quad x_1, x_2, \dots \text{ iid Bernoulli } (\pm 1)$$

$$P(x_1 = +1) = \frac{1}{2} \left( 1 + \frac{r - \frac{1}{2} \sigma^2}{\sigma} \sqrt{\Delta t} \right) = p$$

time of default  $\tau \sim \text{exp}$ , rate  $\hat{\lambda}$

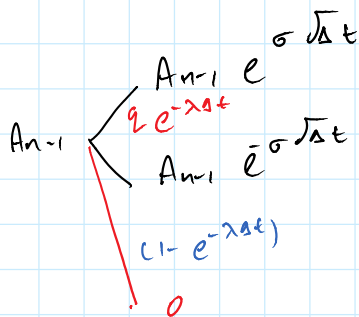
$$P(\tau \in (t, t + \Delta t] | \tau > t) = (1 - e^{-\hat{\lambda} \Delta t}) \sim \hat{\lambda} \Delta t$$



$$P(\tau > T) = e^{-\lambda T}$$

$$P(\tau \in (t, t + \Delta t]) = e^{-\lambda t} - e^{-\lambda(t + \Delta t)}$$

$$= e^{-\lambda t} (1 - e^{-\lambda \Delta t})$$



$\lambda$  is the  $\mathbb{Q}$ -intensity of default

remember when  $\lambda = 0$ ,

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

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

$$\frac{A_{n-1}}{B_{n-1}} = \mathbb{E}^{\mathbb{Q}} \left[ \frac{A_n}{B_n} \right] \quad \text{mtg cond.}$$

$\lambda \Delta t$

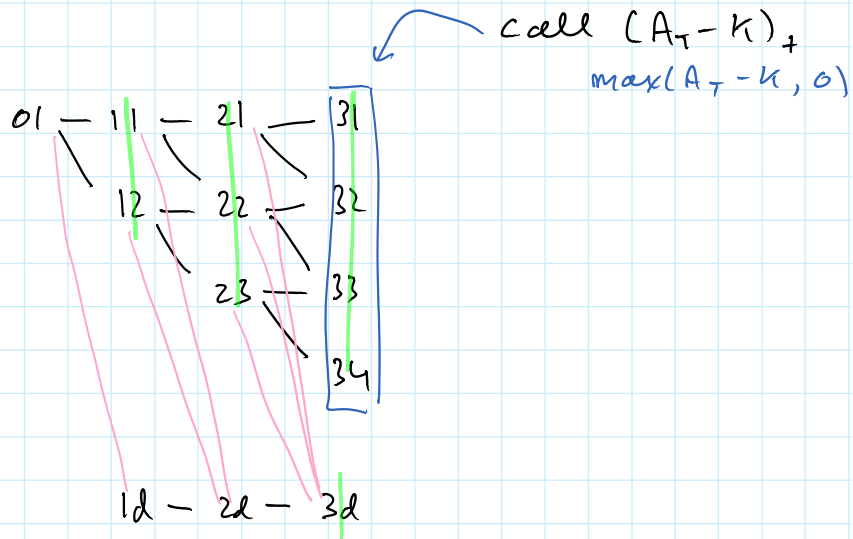
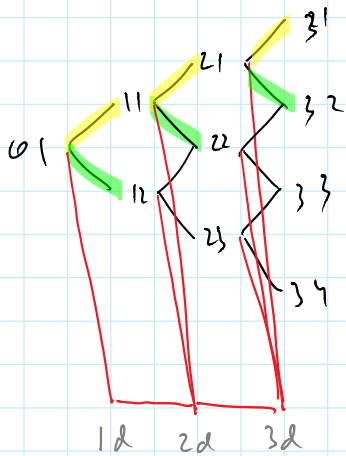
$$\frac{A_{n-1}}{B_{n-1}} = \mathbb{E} \left[ \frac{H_n}{B_n} \right] \quad \text{mtg cond.}$$

$$1 = \overset{\uparrow e^{-\lambda \Delta t}}{q} (1 - \lambda \Delta t) \cdot e^{\sigma \sqrt{\Delta t} - r \Delta t} + (1 - q) \overset{\uparrow e^{-\lambda \Delta t}}{(1 - q)} (1 - \lambda \Delta t) e^{-\sigma \sqrt{\Delta t} - r \Delta t} + o(\Delta t)$$

$$\Rightarrow q = \frac{e^{(r+\lambda)\Delta t} - e^{-\sigma\sqrt{\Delta t}}}{e^{+\sigma\sqrt{\Delta t}} - e^{-\sigma\sqrt{\Delta t}}} \approx \frac{1}{2} \left( 1 + \frac{r+\lambda - \frac{1}{2}\sigma^2}{\sigma} \sqrt{\Delta t} \right) + o(\sqrt{\Delta t})$$

$$\stackrel{\mathcal{Q}}{\mathbb{E}} [A_T | \tau > T] = A_0 e^{(r+\lambda)T}$$

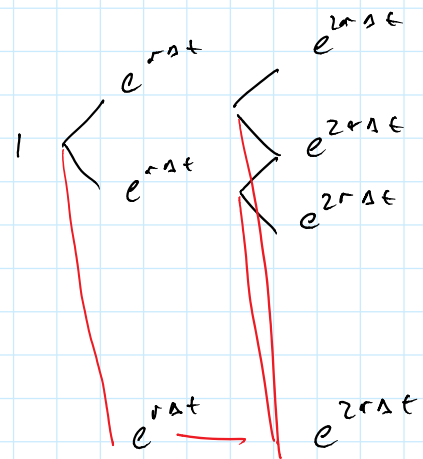
$$\stackrel{\mathcal{Q}}{\mathbb{E}} [A_T] = A_0 e^{rT}$$



$$\frac{C_{n-1}}{B_{n-1}} = E^Q \left[ \frac{C_n}{B_n} \right]$$

$$C_{n-1,j} = q \frac{e^{-\lambda \Delta t}}{e^{r \Delta t}} C_{n,j} + (1-q) \frac{e^{-\lambda \Delta t}}{e^{r \Delta t}} C_{n,j+1} + (1 - e^{-r \Delta t}) \cdot C_{nd}$$

$$C_{n-1,d} = C_{n,d} e^{-r \Delta t}$$



$$\frac{C_0}{B_0} = E^Q \left[ \frac{C_T}{B_T} \right] \sim \frac{1}{N} \sum_{n=1}^N \frac{(A_T^{(n)} - K)_+}{e^{rT}}$$

$$= k - (N - k) = 2k - N$$

$$A_T = A_0 e^{\sigma \sqrt{\Delta t} \sum_{n=1}^N z_n} \quad \mathbb{1}_{T > T}$$

$$\binom{N}{k} q^k (1-q)^{N-k}$$

$$\sigma \sqrt{kt} \sum_{n=1}^N x_n \xrightarrow[N \rightarrow +\infty]{d} \mathcal{N} \left( (r + \lambda - \frac{1}{2}\sigma^2)T; \sigma^2 T \right)$$

$$A_T \xrightarrow{d} A_0 e^{(r + \lambda - \frac{1}{2}\sigma^2)T + \sigma \sqrt{T} Z} \mathbb{1}_{\tau > T}$$

$$Z \sim \mathcal{N}(0, 1)$$

$$\tau \sim \text{exp, intensity } \lambda.$$

$$\frac{C_0}{B_0} = \mathbb{E}^Q \left[ \frac{C_T}{B_T} \right]$$

$$= \mathbb{E}^Q \left[ \frac{(A_T - K)_+}{B_T} \right]$$

$$= \mathbb{E}^Q \left[ \frac{A_T}{B_T} \mathbb{1}_{A_T > K} - \frac{K}{B_T} \mathbb{1}_{A_T > K} \right]$$

$$= \mathbb{E}^Q \left[ \frac{A_T}{B_T} \mathbb{1}_{A_T > K} \right] - \frac{K}{B_T} \mathbb{Q}(A_T > K)$$

$$\frac{F_0}{B_0} = \mathbb{E}^Q \left[ \frac{F_T}{B_T} \right]$$

$\lambda \neq 0$ : take  $A$  as the numeraire!

$$\frac{F_0}{A_0} = \mathbb{E}^{Q^A} \left[ \frac{F_T}{A_T} \right] = \mathbb{E}^{Q^A} \left[ \mathbb{1}_{A_T > K} \right]$$

$$= \mathbb{Q}^A(A_T > K)$$

$$C_0 = F_0 - K e^{-rT} \mathbb{Q}(A_T > K)$$

$$C_0 = A_0 \mathbb{Q}^A(A_T > K) - K e^{-rt} \mathbb{Q}(A_T > K)$$

$$A_{T-1} \begin{cases} q^a A_{T-1} e^{\sigma\sqrt{\Delta t}} \\ A_{T-1} e^{-\sigma\sqrt{\Delta t}} \end{cases} \quad 1 \begin{cases} e^{r\Delta t} \\ e^{-r\Delta t} \end{cases}$$

mtg. cond is  $\frac{1}{A_{T-1}} = q^a \frac{e^{r\Delta t}}{A_{T-1} e^{\sigma\sqrt{\Delta t}}} + (1-q^a) \frac{e^{-r\Delta t}}{A_{T-1} e^{-\sigma\sqrt{\Delta t}}}$

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

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

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

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

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

$$q^a = \frac{1}{2} \left[ 1 + \frac{r + \frac{1}{2}\sigma^2}{\sigma} \sqrt{\Delta t} \right] + \dots$$

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

so  $\mathbb{Q} \rightarrow \mathbb{Q} \quad r \rightarrow r + \sigma^2$

$$\therefore A_T \xrightarrow[\mathbb{Q}^a]{d} A_0 e^{(r + \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}Z}$$

$$Z \sim_{\mathbb{Q}^a} \mathcal{N}(0,1)$$

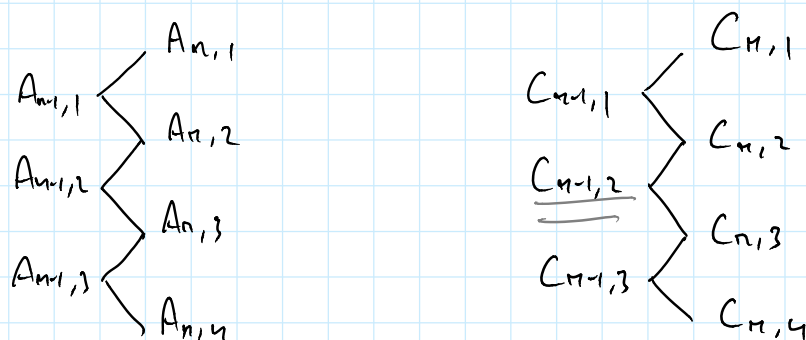
$$C_0 = A_0 \Phi(d_+) - K e^{-rT} \Phi(d_-)$$

$$d_{\pm} = \frac{\ln(A_0/K) + (r \pm \frac{1}{2} \sigma^2)T}{\sigma \sqrt{T}}, \quad \Phi(\cdot) \text{ is } \text{std. norm cdf.}$$

American option lets you set the maturity up to a max date

$s$  - stopping time bounded by  $T$   
 at  $s$   $q = (A_s - K)_+$

$$\frac{C_0}{B_0} = \mathbb{E}^Q \left[ \frac{(A_s - K)_+}{B_s} \right]$$



holding value

$$\frac{C_{n-1,k}^H}{B_{n-1,k}} = q \frac{C_{n,k}}{B_{n,k}} + (1-q) \frac{C_{n,k+1}}{B_{n,k+1}}$$

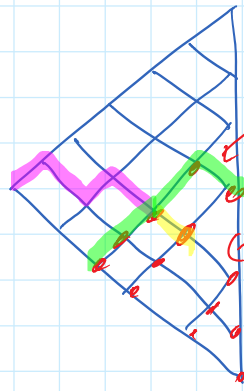
$$\Rightarrow C_{n-1,k}^H = e^{-r\Delta t} (q C_{n,k} + (1-q) C_{n,k+1})$$

intrinsic value

$$C_{n-1,k}^I = (A_{n-1,k} - K)_+ , (K - A_{n-1,k})_+$$

$$C_{n-1,k} = \max(C_{n-1,k}^I, C_{n-1,k}^H)$$

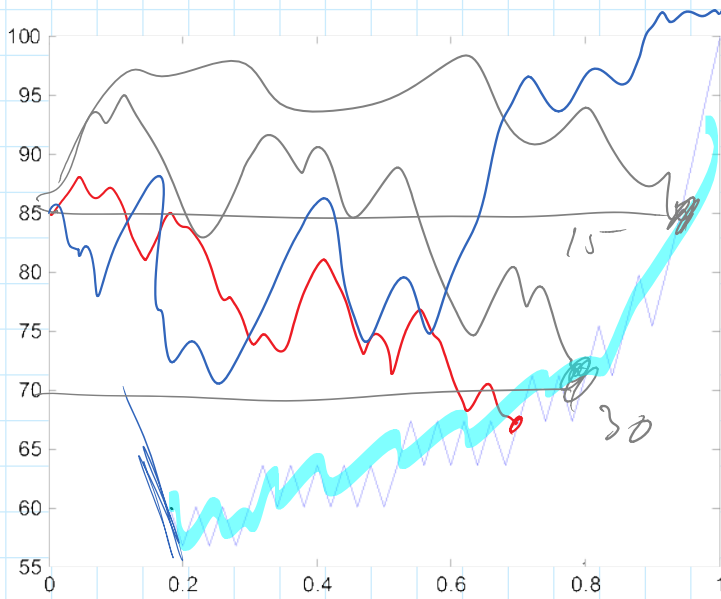




optimal exercise  
Sd.

optimal to  
exercise

$$C_{n-1, k}^I > C_{n-1, k}^n$$



Sun 29, 2013  
mid night

$$(K - A_s)^+$$

protected put gets 0  
upon default.



$$\mu = 10\%$$

$$\hat{\lambda} = 1.5\%$$