

$$A_t = A_u x + A_d (1-x)$$

$$B_t = B_u x + B_d (1-x)$$

$x$  is Bernoulli  
 $IP(x=+1) = p$

an arbitrage is trading strategy  $(\alpha_t, \beta_t)$  s.t.  
 $(\alpha_0, \beta_0)$

$$V_0 = 0$$

$$IP(V_1 \geq 0) = 1$$

$$IP(V_1 > 0) > 0$$

no arbitrage  $\Leftrightarrow \exists Q \sim IP$  s.t.

$$\frac{A_0}{B_0} = E^Q \left[ \frac{A_1}{B_1} \right]$$

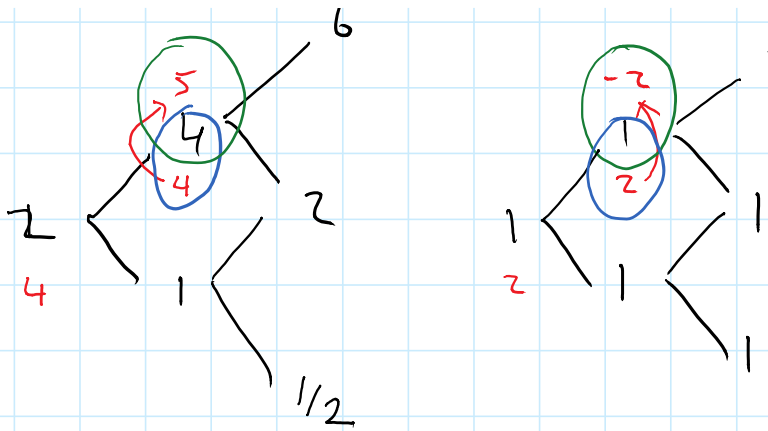
associated with the numeraire  $B$ .

$$\tilde{A}_0 = E^Q [\tilde{A}_1]$$

self-financing strategy  $(\alpha_t, \beta_t) \in \mathcal{F}_t$   
(adapted to the filtration generated by asset prices)

and  $\alpha_t A_{t+1} + \beta_t B_{t+1} = \alpha_{t+1} A_{t+1} + \beta_{t+1} B_{t+1}$





$$(\alpha_0, \beta_0) = (4, 2), \quad V_0 = 10$$

$$(\alpha_1^u, \beta_1^u) = (5, 10)$$

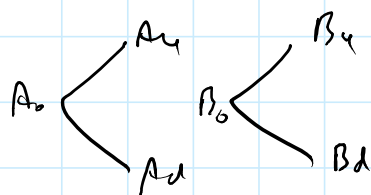
an arbitrage strategy is a self-financing strategy  $(\alpha_t, \beta_t)$

s.t.  $V_0 = 0$

$\exists t$  a)  $IP(V_t \geq 0) = 1$

b)  $IP(V_t > 0) > 0$ .

Replication of claims: trading assets to produce other assets



$$C_u = \alpha A_u + \beta B_u$$

$$C_d = \alpha A_d + \beta B_d$$

$$\alpha A_0 + \beta B_0$$

incomplete market: not all claims are replicable,

( number of traded assets  $\geq$   
 number of branches (each step)  
 $\Leftrightarrow$  complete )

$\Leftrightarrow Q$  is not unique

## Fundamental Thm of Finance

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

(martingale) 
$$\tilde{A}_s = \mathbb{E}^{\mathbb{Q}}[\tilde{A}_t | \mathcal{F}_s] \quad (s < t)$$

where 
$$\tilde{A}_t = \frac{A_t}{B_t}$$

$\hookrightarrow$  numeraire asset.

markets are complete  $\Leftrightarrow \mathbb{Q}$  from above is unique

CRR:  $A_{n+1} = A_n e^{\sigma \sqrt{\Delta t} z_n}$   
 $\{-1, +1\} \Rightarrow z_1, z_2, \dots$  iid Bernoulli  
 $P(z_1 = +1) = p = \frac{1}{2} \left( 1 + \frac{\gamma}{\sigma} \sqrt{\Delta t} \right)$   
 $\Delta t = T/N$

$$E^P \left[ \ln(A_T/A_0) \right] = \gamma \cdot T$$

$$V^P \left[ \ln(A_T/A_0) \right] \xrightarrow{N \rightarrow \infty} \sigma^2 \cdot T$$

$$E^P \left[ e^{iu X} \right] \xrightarrow{N \rightarrow \infty} e^{iu \gamma T - \frac{1}{2} \sigma^2 T u^2}$$

$\Rightarrow X$  is a normal r.v.  
 ↑ mean      ↑ variance

$$E \left[ e^{iu z} \right] \stackrel{\sim N(0,1)}{=} e^{-\frac{1}{2} u^2}$$

$$Y = a + b z \sim N(a, b^2)$$

$$E \left[ e^{iu Y} \right] = E \left[ e^{i(a + b z)u} \right]$$

$$= e^{iau} E \left[ e^{ib u z} \right]$$

$$= e^{iau} e^{-\frac{1}{2} b^2 u^2}$$

$$E \left[ e^{iu z} \right] = \int_{-\infty}^{\infty} e^{iu z} \left( \frac{e^{-\frac{1}{2} z^2}}{\sqrt{2\pi}} \right) dz$$

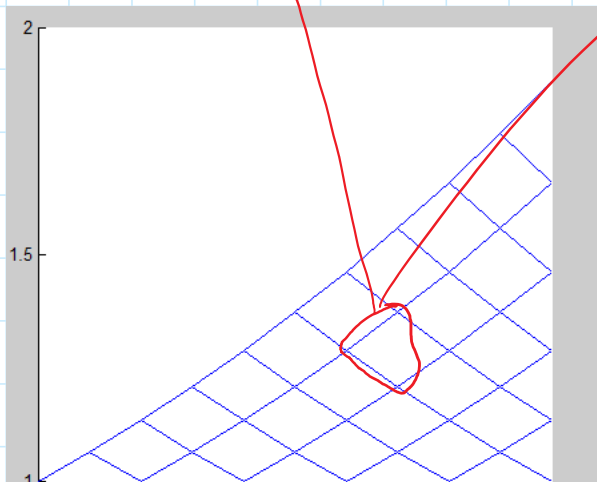
$$= \int_{-\infty}^{\infty} e^{iu z - \frac{1}{2} z^2} dz$$

$$\begin{aligned}
 &= \int_{-\infty}^{\infty} e^{i u z - \frac{1}{2} z^2} \frac{dz}{\sqrt{2\pi}} \\
 &= \int_{-\infty}^{\infty} e^{-\frac{1}{2}[(z - i u)^2 - (i u)^2]} \frac{dz}{\sqrt{2\pi}} \\
 &= e^{-\frac{1}{2} u^2} \int_{-\infty}^{\infty} e^{-\frac{1}{2} (z - i u)^2} \frac{dz}{\sqrt{2\pi}} \\
 &= e^{-\frac{1}{2} u^2} \int_{-\infty}^{\infty} e^{-\frac{1}{2} z^2} \frac{dz}{\sqrt{2\pi}} \quad \rightarrow 1
 \end{aligned}$$

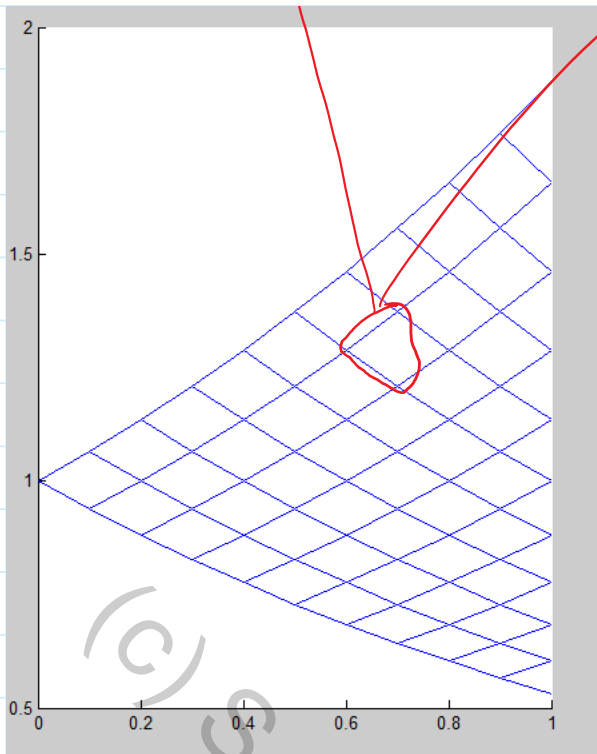
$$\begin{aligned}
 B_{n+1} &= e^{r \Delta t} B_n, & A_{n+1} &= e^{\sigma \sqrt{\Delta t} z_n} A_n \\
 B_0 &= 1 & &
 \end{aligned}$$

$\text{IP}(z_1 = \pm 1) = \frac{1}{2} \left( 1 \pm \frac{\sigma}{r} \sqrt{\Delta t} \right)$

$$A_{n,j} \begin{cases} e^{\sigma \sqrt{\Delta t}} A_{n,j} \\ e^{-\sigma \sqrt{\Delta t}} A_{n,j} \end{cases} \quad e^{r n \Delta t} \begin{cases} e^{r(n+1) \Delta t} \\ e^{r(n+1) \Delta t} \end{cases}$$



$$\tilde{A} \begin{cases} e^{\sigma \sqrt{\Delta t} - r(n+1) \Delta t} A_{n,j} \\ e^{-\sigma \sqrt{\Delta t} - r(n+1) \Delta t} A_{n,j} \end{cases}$$



$\tilde{A}$

$$e^{-r\Delta t} A_{n,i,j} = q_{n,i,j} e^{\sigma\sqrt{\Delta t} - r(n+1)\Delta t} A_{n,i,j} + (1 - q_{n,i,j}) e^{-\sigma\sqrt{\Delta t} - r(n+1)\Delta t} A_{n,i,j}$$

$$e^{-r\Delta t} A_{n,i,j} = q_{n,i,j} e^{\sigma\sqrt{\Delta t} - r(n+1)\Delta t} A_{n,i,j} + (1 - q_{n,i,j}) e^{-\sigma\sqrt{\Delta t} - r(n+1)\Delta t} A_{n,i,j}$$

$$e^{r\Delta t} = q_{n,i,j} e^{\sigma\sqrt{\Delta t}} + (1 - q_{n,i,j}) e^{-\sigma\sqrt{\Delta t}} \quad (\times e^{r(n+1)\Delta t})$$

$$\Rightarrow e^{r\Delta t} = q_{n,i,j} e^{\sigma\sqrt{\Delta t}} + (1 - q_{n,i,j}) e^{-\sigma\sqrt{\Delta t}}$$

$$\Rightarrow q_{n,i,j} = q = \frac{e^{r\Delta t} - e^{-\sigma\sqrt{\Delta t}}}{e^{\sigma\sqrt{\Delta t}} - e^{-\sigma\sqrt{\Delta t}}}$$

risk-neutral probabilities

recall  $e^y = 1 + y + \frac{1}{2}y^2 + o(y^2)$

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

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

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

$$= 2\sigma\sqrt{\Delta t} + o(\Delta t)$$

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

r & call

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

$$\log \frac{A_T}{A_0} \xrightarrow[N \rightarrow \infty]{IP} \mathcal{N}(\sigma T; \sigma^2 T)$$

$$\log \frac{A_T}{A_0} \xrightarrow[N \rightarrow \infty]{Q} \mathcal{N}((r - \frac{1}{2}\sigma^2)T; \sigma^2 T)$$

$$E^{IP}[A_T] = E^{IP}[A_0 e^X]$$

$$= E^{IP}[A_0 e^{\sigma T + \sigma\sqrt{T}Z}] \quad \sim \mathcal{N}(0,1)$$

$$(E^{IP}[e^{uZ}] = e^{\frac{1}{2}u^2})$$

$$= A_0 e^{\sigma T} E^{IP}[e^{\sigma\sqrt{T}Z}]$$

$$= A_0 e^{\sigma T} e^{\frac{1}{2}\sigma^2 T}$$

$$= A_0 e^{(\sigma + \frac{1}{2}\sigma^2)T} = A_0 e^{\mu T}$$

$$(\sigma = \mu - \frac{1}{2}\sigma^2)$$

$$E^Q[A_T] = A_0 e^{rT}$$

$$A_0 = \underline{A_0} = E^Q[A_T] = E^Q[\underline{A_T}]$$

$$A_0 = \frac{A_0}{B_0} = \mathbb{E}^Q \left[ \frac{A_T}{B_T} \right] = \mathbb{E}^Q \left[ \frac{A_T}{e^{rt}} \right] \Bigg| \mathcal{F}_0$$

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$$A_T \stackrel{\Delta}{=} A_0 e^x, \quad x \stackrel{\Delta}{\sim} N\left(\left(r - \frac{1}{2}\sigma^2\right)T; \sigma^2 T\right)$$

value a call option pays  
 $(A_T - K)_+$  @ T

$$\frac{C_0}{B_0} = E^Q \left[ \frac{C_T}{B_T} \mid \mathcal{F}_0 \right]$$

$$\begin{aligned} \Rightarrow C_0 &= e^{-rT} E^Q \left[ (A_T - K)_+ \right] \\ &= e^{-rT} E^Q \left[ (A_0 e^x - K)_+ \right] \\ &= e^{-rT} E^Q \left[ (A_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}z} - K)_+ \right] \end{aligned}$$

$$E = \int_{-\infty}^{\infty} (A_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}z} - K)_+ \frac{e^{-\frac{1}{2}z^2}}{\sqrt{2\pi}} dz$$

$$= \int_{z^*}^{\infty} (A_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}z} - K) \frac{e^{-\frac{1}{2}z^2}}{\sqrt{2\pi}} dz$$

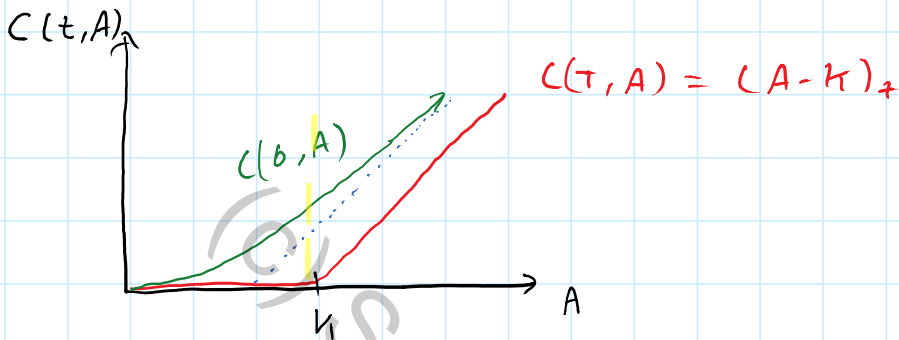
$$A_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}z^*} - K = 0$$

= ...

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

$$d_{\pm} = \frac{\ln(A_0/K) + (r \pm \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}$$

Black-Scholes pricing formula



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

asset or nothing  $N$  digital  $D$

$$(A_T - K)_+ = A_T \mathbb{1}_{A_T > K} - K \mathbb{1}_{A_T > K}$$

$$\mathbb{1}_{\{D\}} = \begin{cases} 1 & \text{if } D \text{ holds} \\ 0 & \text{otherwise} \end{cases}$$

$$\frac{D_0}{B_0} = E^Q \left[ \frac{D_T}{B_T} \right] \Rightarrow D_0 = e^{-rT} E^Q [K \mathbb{1}_{A_T > K}]$$

$$\Rightarrow D_0 = K e^{-rT} Q^B(A_T > K)$$

$$\frac{N_0}{B_0} = E^Q \left[ \frac{N_T}{B_T} \right]$$

$$\frac{N_0}{A_0} = E^Q \left[ \frac{N_T}{A_T} \right] \Rightarrow N_0 = A_0 E^Q \left[ \frac{A_T \mathbb{1}_{A_T > K}}{A_T} \right]$$

$$\Rightarrow N_0 = A_0 Q^A(A_T > K)$$

$$C_0 = A_0 Q^A(A_T > K) - K e^{-rT} Q^B(A_T > K)$$

$$Q^B(A_T > K) = Q^B \left( A_0 e^{(r - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}z} > K \right)$$

$$= Q^B \left( z > \frac{\ln(K/A_0) - (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}} \right)$$

$$= \Phi \left( \frac{\ln(A_0/K) + (r - \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}} \right)$$



$$\frac{e^{r\Delta t}}{A_n} \begin{cases} q^A \frac{e^{r(n+1)\Delta t - \sigma\sqrt{\Delta t}}}{A_n} \\ (1-q^A) \frac{e^{r(n+1)\Delta t + \sigma\sqrt{\Delta t}}}{A_n} \end{cases}$$

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

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

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

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

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

$$q = \frac{1}{2} \left( 1 + \frac{\sigma}{\sigma} \sqrt{\Delta t} \right)$$

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

$$\cdot \left( r + \frac{1}{2}\sigma^2 \right) T + \sigma\sqrt{T} z^A$$

$$A_T = A_0 e^{(r + \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}Z^A}$$

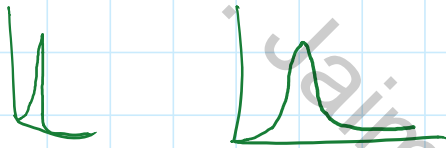
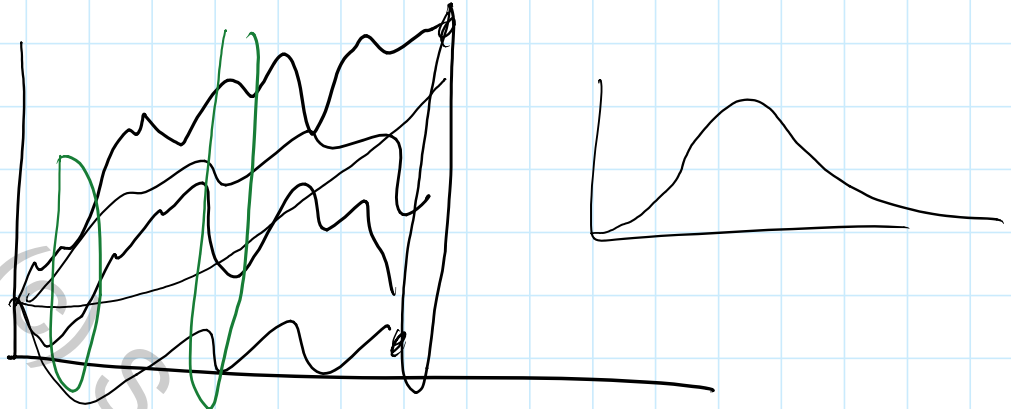
$Z^A \stackrel{Q^A}{\sim} N(0,1)$

$$\begin{aligned} \Rightarrow Q^A(A_T > K) &= \Phi\left(\frac{\log(A_0/K) + (r + \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}\right) \\ &= \Phi\left(\frac{\log(A_0/K) + (r + \frac{1}{2}\sigma^2)T}{\sigma\sqrt{T}}\right) \end{aligned}$$

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$$A_t = A_0 e^{(r - \frac{1}{2}\sigma^2)t + \sigma\sqrt{t}Z}$$

$Z \sim N(0,1)$

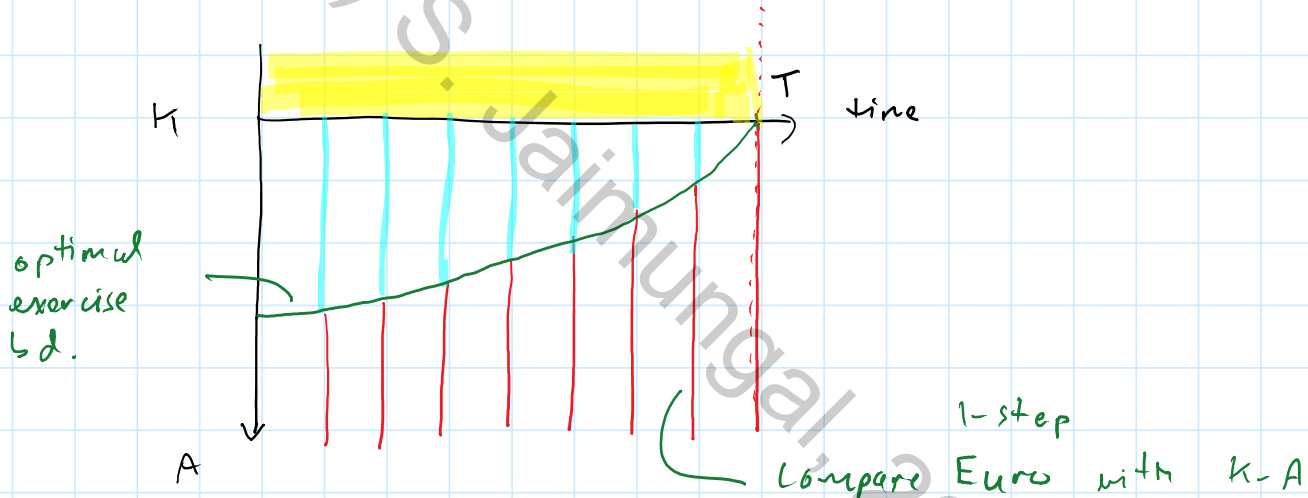


# American Options

American put option: pays  $(K - A_\tau)_+ @ \tau$   
 $\tau \in (0, T]$

$$\frac{P_0}{B_0} = \sup_{\tau \in \mathcal{T}} \mathbb{E}^Q \left[ \frac{(K - A_\tau)_+}{B_\tau} \right]$$

$\mathcal{T}$  set of  $\mathcal{F}$ -stopping times.

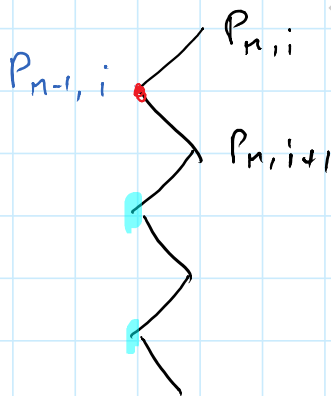


hold value

$$P_{n-1,i}^H = \mathbb{E}^Q [ P_n | \mathcal{F}_{n-1,i} ]$$

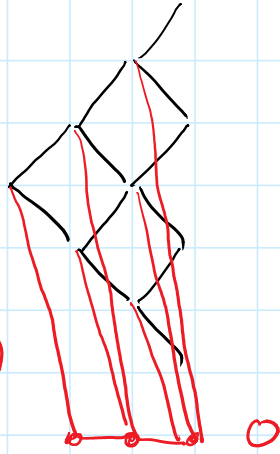
exercise value

$$P_{n-1,i}^E = (K - A_{n-1,i})_+$$



$$P_{n-1,i} = \max( P_{n-1,i}^H, P_{n-1,i}^E )$$

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



$$A_{n+1} = e^{\sigma \sqrt{\Delta t} x_n} A_n$$

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