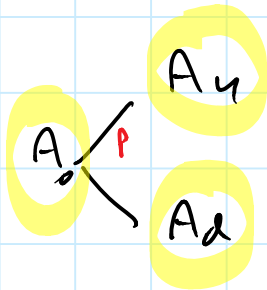


# One-Period Binomial

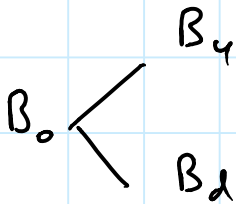
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$p \in (0, 1)$



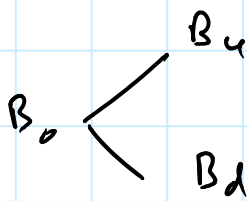
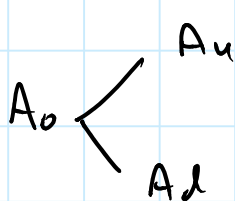
$$A_0 = \frac{p A_u + (1-p) A_d}{1+r}$$



$$A_1 = A_u x_1 + A_d (1-x_1)$$

$$x_1 = \begin{matrix} 1, 0 \\ p, (1-p) \end{matrix}$$

$$B_1 = B_u x_1 + B_d (1-x_1)$$



$$\left( \frac{A_u}{B_u} > \frac{A_d}{B_d} \right)$$

$$B_0 \neq 0$$

$$B_u \neq 0$$

$$B_d \neq 0$$

$\alpha$

$\beta$

$$V_0 = \alpha A_0 + \beta B_0$$

$$V_1 = \alpha A_1 + \beta B_1$$

an arbitrage is a strategy s.t.

an arbitrage is a strategy s.t.

i)  $V_0 = 0$

ii)  $\exists t$  s.t. a)  $\mathbb{P}(V_t \geq 0) = 1$

b)  $\mathbb{P}(V_t > 0) > 0$

---

$$V_0 = 0 \Rightarrow \alpha A_0 + \beta B_0 = 0$$

$$\Rightarrow \beta = -\alpha \frac{A_0}{B_0}$$

$$0 \begin{cases} \alpha A_u + \beta B_u = \alpha \left( A_u - \frac{A_0}{B_0} B_u \right) \\ \alpha A_d + \beta B_d = \alpha \left( A_d - \frac{A_0}{B_0} B_d \right) \end{cases}$$

$$\left. \begin{array}{l} A_u - \frac{A_0}{B_0} B_u > 0 \\ \alpha A_d - \frac{A_0}{B_0} B_d < 0 \end{array} \right\} \begin{array}{l} \text{all other cases} \\ \text{lead to arb.} \end{array}$$

$$\frac{A_d}{B_d} < \frac{A_0}{B_0} < \frac{A_u}{B_u}$$

 no arbitrage condition



$\exists$  some  $q \in (0, 1)$  s.t.

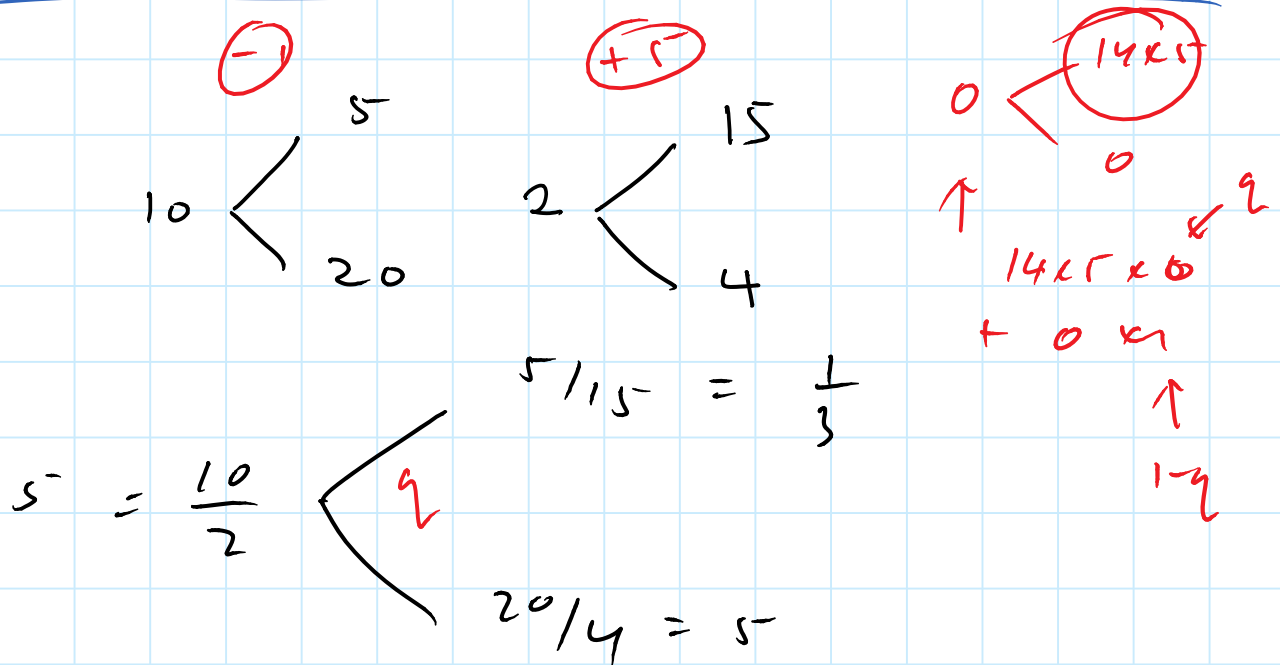
$$\frac{A_0}{B_0} = q \frac{A_u}{B_u} + (1-q) \frac{A_d}{B_d}$$

↔ no arbitrage!

no arbitrage iff  $\exists q \in (0,1)$  s.t.

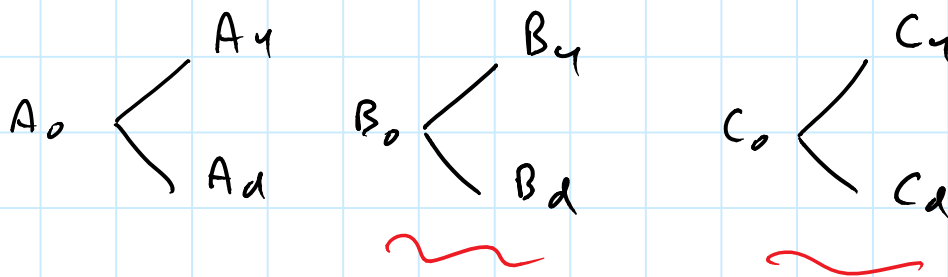
$$\tilde{A}_0 = E^Q[\tilde{A}_1]$$

where  $\tilde{A}_t = \frac{A_t}{B_t}$  (relative price)



# A Third Asset

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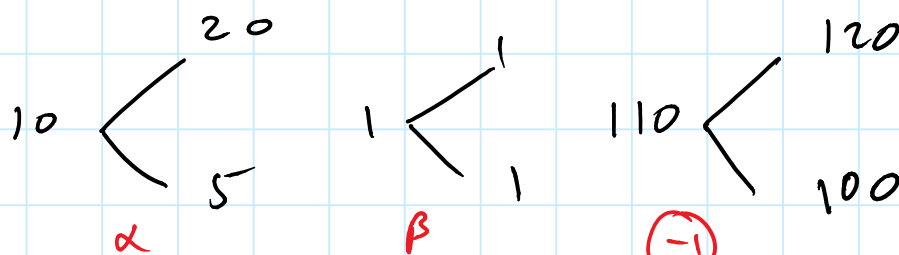


$$\tilde{A}_0^B = \mathbb{E}^{\mathbb{Q}^{BA}} [\tilde{A}_1^B] \rightarrow q^{BA} \in (0,1)$$

$$\tilde{C}_0^B = \mathbb{E}^{\mathbb{Q}^{BC}} [\tilde{C}_1^B] \rightarrow q^{BC} \in (0,1)$$

suppose we know  $q^{BA}$ , then compute

$$\frac{\hat{C}_0}{B_0} \stackrel{?}{=} \frac{C_u}{B_u} - q^{BA} + \frac{C_d}{B_d} (1 - q^{BA}) > \frac{C_0}{B_0}$$



$$\Rightarrow \begin{cases} 10 = 20 q^{BA} + 5(1 - q^{BA}) \\ 110 = 120 q^{BC} + 100(1 - q^{BC}) \end{cases} \Rightarrow q^{BC} = \frac{10}{20} = \frac{1}{2}$$

$$\hat{C}_0 = 120 \times \frac{1}{3} + 100 \times \frac{2}{3} = 40 + 66\frac{2}{3} < 110$$

$$\vec{C}_0 = 120 \frac{1}{3} + 100 \times \frac{2}{3} = 40 + 66\frac{2}{3} < 110$$

$$V_0 = 10\alpha + \beta - 110 = 0$$

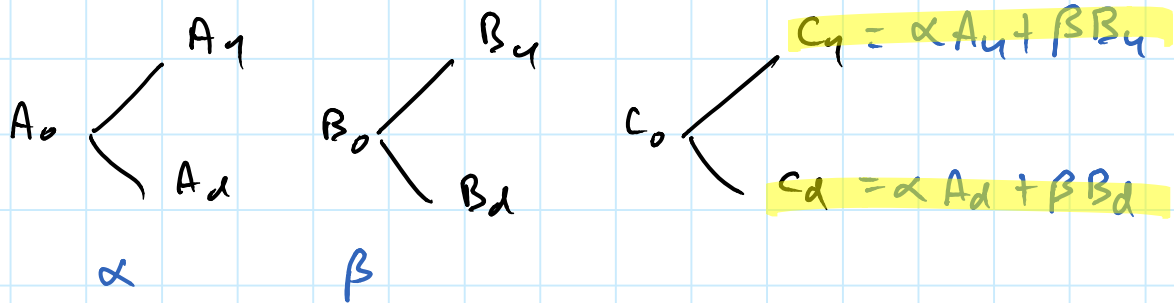
$$\Rightarrow \beta = 110 - 10\alpha = 90$$

$$0 \left\{ \begin{array}{l} 20\alpha + \beta - 120 = 10\alpha - 10 = 10 \\ 5\alpha + \beta - 100 = -5\alpha + 10 = 0 \\ \alpha = 2 \end{array} \right.$$

# Replication

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choose  $(\alpha, \beta)$  s.t.  $\alpha A_u + \beta B_u = C_u$  p.a.s.

$$\Rightarrow C_0 = \alpha A_0 + \beta B_0!$$

otherwise  $\exists$  an arb!

you will find that

$$\frac{C_0}{B_0} = q \frac{C_u}{B_u} + (1-q) \frac{C_d}{B_d}$$

$$\text{where } q = \frac{\frac{A_0}{B_0} - \frac{A_d}{B_d}}{\frac{A_u}{B_u} - \frac{A_d}{B_d}}$$

moreover  $q$  satisfies

$$\frac{A_0}{B_0} = q \frac{A_u}{B_u} + (1-q) \frac{A_d}{B_d}$$

no arb  $\Leftrightarrow \exists \mathbb{Q} \sim \mathbb{P}$  s.t. For all traded assets  $X$ , we have

... numeraires, ...

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

where  $\tilde{X}_t = \frac{X_t}{B_t}$  and

$P(B_t > 0) = 1$  (B is called a numeraire asset)

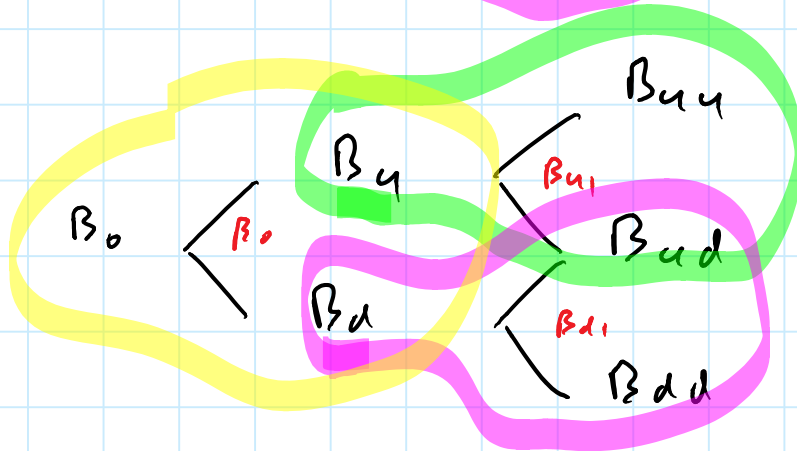
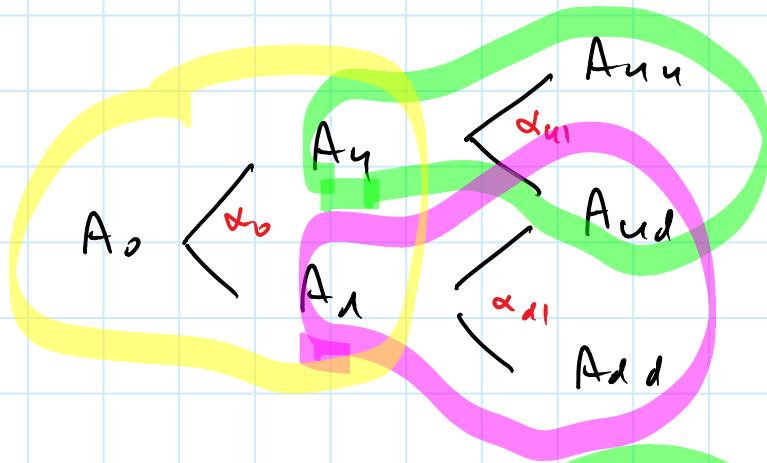
# Multi-Period Binomial

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

$x_1, x_2, \dots$  iid Bernoulli  
 $\pm 1$  with  $\frac{p}{1-p}$



an arbitrage is  
 strategy that  
 is adapted to  
 the filtration  
 generated by  
 asset prices  
 ( $x_t \in \mathcal{F}_t$ )

s.t.

i)  $V_0 = 0$

ii)  $\exists t$  s.t.

1)  $\mathbb{P}(V_t \geq 0) = 1$

2)  $\mathbb{P}(V_t > 0) > 0$

martingale measure

no arb  $\Leftrightarrow \exists \mathbb{Q} \sim \mathbb{P}$  s.t.  $\forall t < s$   
 and all traded assets  $X$ , we have

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

i.e. relative prices  
 are martingale!





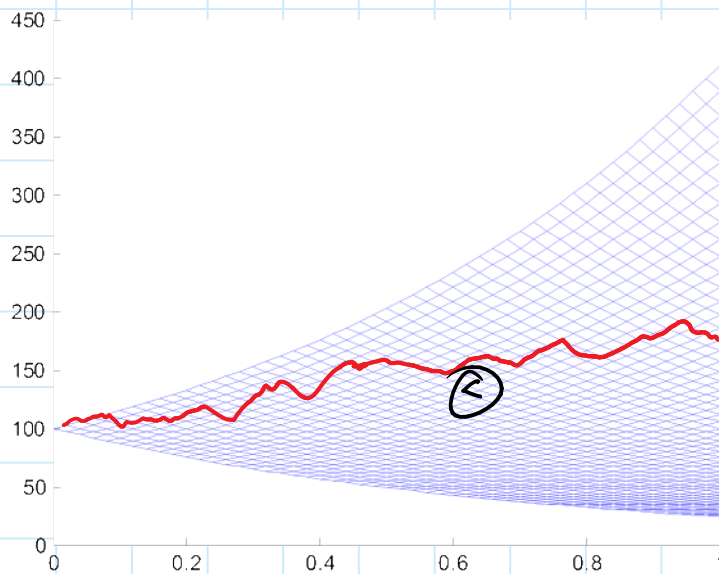
if  $B$  is the bank account, then  $\mathbb{Q}$  is called the risk-neutral measure

# Matching Drift & Volatility

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



$$\ln(A_T/A_0) = \sigma \sqrt{\Delta t} \sum_{n=1}^N x_n$$

$\hookrightarrow N \Delta t = T$

$$\begin{aligned} \mathbb{E} \left[ \ln \left( \frac{A_T}{A_0} \right) \right] &= \sigma \sqrt{\Delta t} \sum_{n=1}^N \mathbb{E}[x_n] \\ &= \sigma \sqrt{\Delta t} N \cdot (2p-1) = (\mu - \frac{1}{2}\sigma^2) T \end{aligned}$$

$$\Rightarrow 2p-1 = \frac{\mu - \frac{1}{2}\sigma^2}{\sigma} \cdot \sqrt{\Delta t}$$

$$p = \frac{1}{2} \left[ 1 + \frac{\mu - \frac{1}{2}\sigma^2}{\sigma} \sqrt{\Delta t} \right]$$

$$\mathbb{V} \left[ \ln \left( \frac{A_T}{A_0} \right) \right] = \sigma^2 \Delta t \cdot N \cdot \mathbb{V} [x_i]$$

$$\begin{aligned} \mathbb{V} [x_i] &= \mathbb{E} [x_i^2] - (\mathbb{E} [x_i])^2 \\ &= 1 - \left( \left( \frac{\mu - \frac{1}{2}\sigma^2}{\sigma} \right) \sqrt{\Delta t} \right)^2 \end{aligned}$$

$$\Rightarrow \mathbb{V} \left[ \ln \frac{A_T}{A_0} \right] = \sigma^2 T \cdot \left( 1 - \left( \frac{\mu - \frac{1}{2}\sigma^2}{\sigma} \right)^2 \Delta t \right)$$

$$B_t = e^{rt}, \quad B_n = B_{n-1} e^{r \Delta t}$$

$$\frac{A_t}{B_t} = \mathbb{E}^Q \left[ \frac{A_s}{B_s} \mid \mathcal{F}_t \right]$$

$$A_{n-1} \begin{cases} A_{n-1} e^{\sigma \sqrt{\Delta t}} \\ A_{n-1} e^{-\sigma \sqrt{\Delta t}} \end{cases}$$

$$B_{n-1} \begin{cases} B_{n-1} e^{r \Delta t} \\ B_{n-1} e^{r \Delta t} \end{cases}$$

$$\frac{A_{n-1}}{B_{n-1}} = \frac{A_{n-1}}{B_{n-1}} e^{\sigma \sqrt{\Delta t} - r \Delta t} q + \frac{A_{n-1}}{B_{n-1}} e^{-\sigma \sqrt{\Delta t} - r \Delta t} (1-q)$$

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

$$e^x \sim 1 + x + \frac{1}{2}x^2 + o(x^2)$$

$$e^x \sim 1 + x + \frac{1}{2}x^2 + o(x^2)$$

$$\begin{aligned} \text{num} &= (1 + r \Delta t) - (1 - \sigma \sqrt{\Delta t} + \frac{1}{2} \sigma^2 \Delta t) + \dots \\ &= \sigma \sqrt{\Delta t} + (r - \frac{1}{2} \sigma^2) \Delta t + \dots \end{aligned}$$

$$\begin{aligned} \text{denom} &= (1 + \sigma \sqrt{\Delta t} + \frac{1}{2} \sigma^2 \Delta t) - (1 - \sigma \sqrt{\Delta t} + \frac{1}{2} \sigma^2 \Delta t) + \dots \\ &= 2 \sigma \sqrt{\Delta t} + \dots \end{aligned}$$

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

$$\mathbb{E}^Q \left[ \ln \frac{A_T}{A_0} \right] = (r - \frac{1}{2} \sigma^2) T$$

$$\mathbb{V}^Q \left[ \ln \frac{A_T}{A_0} \right] = \sigma^2 T + \int \Delta t$$

# Limiting Distribution

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$$\ln\left(\frac{A_T}{A_0}\right) = \sigma \sqrt{\Delta t} \sum_{n=1}^N x_n \xrightarrow[N \rightarrow \infty]{d} \mathcal{N}(\cdot; \cdot)$$

$$\text{mean} = \left(\mu - \frac{1}{2}\sigma^2\right)T$$

$$\text{var} = \sigma^2 T$$

$$\Leftrightarrow A_T \stackrel{d}{=} A_0 e^{(\mu - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T}Z}$$

$$Z \stackrel{P}{\sim} \mathcal{N}(0; 1)$$

$$\mathbb{E}[A_T] = A_0 e^{(\mu - \frac{1}{2}\sigma^2)T}$$

$$\mathbb{E}\left[e^{\sigma\sqrt{T}Z}\right]$$

$$\hookrightarrow e^{\frac{1}{2}\sigma^2 T}$$

# Existence of Martingale Measure $\Rightarrow$ No Arbitrage

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$A_t^{(1)}, A_t^{(2)}, \dots, A_t^{(n)}$  non-negative IP-a.s.  
 $B_t$  strictly positive IP-a.s.  
 $t = \{0, 1\}$

suppose  $\exists \mathbb{Q} \sim \mathbb{P}$  s.t.

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

take a portfolio  $\beta$  of  $B$   
 $\alpha^{(k)}$  of  $A^{(k)}$

s.t.  $\mathbb{P}(V_1 \geq 0) \geq 1$   
 $\mathbb{P}(V_1 > 0) > 0$

note:  $\mathbb{E}^{\mathbb{Q}}[V_1] > 0$

$$\begin{aligned}
 V_0 &= \beta B_0 + \alpha \cdot A_0 \\
 &= \beta B_0 + \sum_{k=1}^n \alpha^{(k)} \frac{A_0^{(k)}}{B_0} \\
 &= \beta B_0 \mathbb{E}^{\mathbb{Q}} \left[ \frac{B_1}{B_0} \right] + \sum_{k=1}^n \alpha^{(k)} \frac{A_1^{(k)}}{B_1}
 \end{aligned}$$

$$= \beta_0 \mathbb{E} \left[ \frac{\beta_1 \beta + \sum_{k=1}^n \alpha^{(k)} \frac{A_k^{(k)}}{\beta_1}}{\beta_1} \right]$$

$$= \beta_0 \mathbb{E}^{\mathcal{Q}} \left[ \frac{\beta \beta_1 + \sum_{k=1}^n \alpha^{(k)} A_k^{(k)}}{\beta_1} \right]$$

$$= \beta_0 \mathbb{E}^{\mathcal{Q}} \left[ \frac{V_1}{\beta_1} \right] > 0$$

$$\text{let } X = \sigma \sqrt{\Delta t} \sum_{n=1}^N x_n, \quad \Delta t = \frac{T}{N}$$

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

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

$$\text{let } \hat{\mu} \triangleq (\mu - \frac{1}{2}\sigma^2)$$

Want to show that

$$X \xrightarrow[N \rightarrow +\infty]{d} \mathcal{N}(\hat{\mu}T, \sigma^2 T)$$

Do this via characteristic fn.

$$\begin{aligned} \mathbb{E}[e^{iuX}] &= \mathbb{E}\left[ e^{iu\sigma\sqrt{\Delta t} \sum_{n=1}^N x_n} \right] \\ &= \left( \mathbb{E}\left[ e^{iu\sigma\sqrt{\Delta t} x_1} \right] \right)^N \end{aligned}$$

now,  $\mathbb{E}\left[ e^{iu\sigma\sqrt{\Delta t} x_1} \right]$

$$= \mathbb{E}\left[ 1 + iu\sigma\sqrt{\Delta t} x_1 - \frac{1}{2}u^2\sigma^2\Delta t x_1^2 + o(\Delta t) \right]$$

$$= 1 + iu\sigma\sqrt{\Delta t} \mathbb{E}[x_1] - \frac{1}{2}u^2\sigma^2\Delta t \mathbb{E}[x_1^2] + o(\Delta t)$$

$$\text{next, } \mathbb{E}[x_1] = 2p - 1 = \frac{\hat{\mu}}{\sigma} \sqrt{\Delta t}$$

$$\text{and } \mathbb{E}[x_1^2] = 1$$



$$\Rightarrow \mathbb{E} \left[ e^{i u \sigma \sqrt{\Delta t} x_t} \right]$$

$$= 1 + i u \hat{\mu} \Delta t - \frac{1}{2} u^2 \sigma^2 \Delta t + o(\Delta t)$$

So then,

$$\mathbb{E} \left[ e^{i u X} \right] = \left( 1 + (i \hat{\mu} u - \frac{1}{2} \sigma^2 u^2) \Delta t + o(\Delta t) \right)^{\frac{T}{\Delta t}}$$

↑  
T/n

$$\xrightarrow{N \rightarrow +\infty} e^{(i \hat{\mu} u - \frac{1}{2} \sigma^2 u^2) T}$$

(since  $(1 + \frac{a}{n})^n \xrightarrow{n \rightarrow +\infty} e^a$ )

$$\therefore X \xrightarrow[N \rightarrow +\infty]{d} \mathcal{N}(\hat{\mu} T; \sigma^2 T)$$