

YING KIT HUI 20933849 Graduate Level

Q1a.

To sum up, we get

$$C_0 = 2 = q^* (12u - 12) \quad (1)$$

$$\tilde{C}_0 = 1.5 = q^* (12u - 13.5) \quad (2)$$

$$\frac{(1)}{(2)} \Rightarrow \frac{2}{1.5} = \frac{12u - 12}{12u - 13.5}$$

$$\Rightarrow \frac{4}{3} (12u - 13.5) = 12u - 12$$

$$\Rightarrow 4u = \frac{4}{3} 13.5 - 12 = 6$$

$$\Rightarrow u = \frac{3}{2}$$

$$\text{Thus, by (1), } 2 = q^* (12 \cdot \frac{3}{2} - 12)$$

$$\Rightarrow q^* = \frac{1}{3}$$

$$\text{Also } q^* = \frac{1-d}{u-d} = \frac{1-d}{\frac{3}{2}-d} = \frac{1}{3}$$

$$\Rightarrow 3(1-d) = \frac{3}{2} - d$$

$$\Rightarrow d = \frac{3}{4}$$

Then, binomial model for P_0 is

$$P_0 \begin{cases} \max(15 - 12u, 0) = \max(15 - 12 \cdot \frac{3}{2}, 0) = \max(15 - 18, 0) = 0 \\ \max(15 - 12d, 0) = \max(15 - 12 \cdot \frac{3}{4}, 0) = \max(15 - 9, 0) = 6 \end{cases}$$

By risk-neutral valuation,

$$P_0 = e^{-rT} (q^* \cdot 0 + (1-q^*) \cdot 6)$$

$$= (1 - \frac{1}{3}) \cdot 6 = \frac{2}{3} \cdot 6 = 4.$$

Hence, the fair value of a European put option with a strike price of \$15 is \$4.

Hibon

Q16)

Let Π_t denote our portfolio at time t .

Let η_t and δ_t denote the amount of bond and stock that we hold at time t .

To hedge a short position in the put option specified in (a) amounts to find η_0, δ_0 s.t.

the portfolio $\Pi_0 = \underbrace{-P_0}_{\text{short put}} + \underbrace{\eta_0}_{\text{bond}} + \underbrace{\delta_0 S_0}_{\text{stock}}$

has the value $\Pi_T = 0$ at expiring time T .

$$\text{So we want } \Pi_T = -P_T + \eta_0 e^{rT} + \delta_0 S_T = 0$$

$$\Leftrightarrow \eta_0 e^{rT} + \delta_0 S_T = P_T$$

That is, we want to replicate the payoff P_T , this amounts to solve the following (based on our binomial model)

$$(b) \begin{bmatrix} 1 & uS_0 \\ 1 & dS_0 \end{bmatrix} \begin{bmatrix} \eta_0 \\ \delta_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 6 \end{bmatrix}$$

$$\text{noting that } e^{rT} = e^0 = 1.$$

$$(c) \Rightarrow \eta_0 + \frac{3}{2} \cdot 12 \cdot \delta_0 = 0$$

$$\eta_0 + \frac{3}{4} \cdot 12 \cdot \delta_0 = 6$$

$$\Rightarrow 12 \left(\frac{3}{2} - \frac{3}{4} \right) \delta_0 = -6$$

$$\delta_0 = -\frac{2}{3}$$

$$\Rightarrow \eta_0 = -\frac{3}{2} \cdot 12 \cdot \delta_0 = -\frac{3}{2} \cdot 12 \cdot \left(-\frac{2}{3}\right) = 12$$

Thus, $\delta_0 = -\frac{2}{3}$ unit of the underlying is required at $t=0$ to hedge. That is, we need to short $\frac{2}{3}$ unit of the stock and long 12 dollars of bond to hedge. *History*

Q1c

We use the notation $E^P(\cdot)$ and $E^Q(\cdot)$ for expectation under actual probability p and under risk neutral expectation q^* respectively.

Then, using actual probability p , the expected option payoff for the European put in (a) is

$$E^P(P_T) = p \cdot 0 + (1-p) \cdot 6 = 6(1-p)$$

Pricing this put option at this expected payoff value will cause arbitrage if $p \neq q^*$.

Now we demonstrate how to construct arbitrage assuming $p \neq q^*$ and suppose P_0 is given by $E^P(P_T)$

Case 1: $p > q^*$, equivalently $1-p < 1-q^*$.

Then

$$P_0 = E^P(P_T) = (1-p)6 < (1-q^*)6 = E^Q(P_T) = 4$$

So, intuitively, this put is underpriced.

Construct portfolio by

$$\pi_0 = \underbrace{P_0}_{< 4} - \underbrace{(\eta_0 + \delta_0 S_0)}_{= 4} < 0$$

(note that $\eta_0 + \delta_0 S_0 = 1.2 + \frac{2}{3} \cdot 1.2 = 4$.)

$$\text{and } \pi_T = P_T - (\eta_0 + \delta_0 S_T) \\ = P_T - P_T = 0$$

Since η_0, δ_0 are chosen s.t.

$$\eta_0 + \delta_0 S_T = P_T \text{ under each}$$

scenario in binomial model).

This is an arbitrage since $\pi_0 < 0$, $\pi_T = 0$.

Arbitrage

Case: $p < \frac{1}{2}$, equivalently $1-p > 1-\frac{1}{2}$.

$$S_0, P_0 = \mathbb{E}^P(P_T) = 6(1-p) > 6(1-\frac{1}{2}) = 4 = \mathbb{E}^Q(P_T)$$

So, intuitively, this put is overpriced.

Construct portfolio by

$$\pi_0 = \underbrace{-P_0}_{< -4} + \underbrace{(y_0 + \delta_0 S_0)}_{= 4} < 0$$

$$\text{Then } \pi_T = -P_T + (y_0 + \delta_0 S_T) \\ = -P_T + P_T = 0$$

So $\pi_0 < 0$, $\pi_T = 0$, this is an arbitrage.

Hilary

$$a) \quad E[S(t+\Delta t) | S(t)] \\ = S(t)u \cdot q + S(t)d \cdot (1-q)$$

$$\text{Note } uq + d(1-q) \\ = e^{\sigma \sqrt{\Delta t}} \left(\frac{e^{r\Delta t} - e^{-\sigma \sqrt{\Delta t}}}{e^{\sigma \sqrt{\Delta t}} - e^{-\sigma \sqrt{\Delta t}}} \right) + e^{-\sigma \sqrt{\Delta t}} \left(\frac{e^{\sigma \sqrt{\Delta t}} - e^{r\Delta t}}{e^{\sigma \sqrt{\Delta t}} - e^{-\sigma \sqrt{\Delta t}}} \right) \\ = \frac{e^{\sigma \sqrt{\Delta t} + r\Delta t} - e^{-\sigma \sqrt{\Delta t} + r\Delta t}}{e^{\sigma \sqrt{\Delta t}} - e^{-\sigma \sqrt{\Delta t}}} \\ = e^{r\Delta t}$$

$$\text{Hence } E[S(t+\Delta t) | S(t)] \\ = S(t)u \cdot q + S(t)d \cdot (1-q) \\ = S(t)e^{r\Delta t}$$

$$b) \quad \text{Var}[S(t+\Delta t) | S(t)] \\ = E[S(t+\Delta t)^2 | S(t)] - E[S(t+\Delta t) | S(t)]^2$$

$$\text{Note } E[S(t+\Delta t)^2 | S(t)]$$

$$= S(t)^2 u^2 q + S(t)^2 d^2 (1-q)$$

$$\text{and } u^2 q + d^2 (1-q)$$

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

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

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

$$\text{using } a^2 - b^2 = (a+b)(a-b)$$

$$\text{Hence, } \text{Var}[S(t+\Delta t)^2 | S(t)] \\ = S(t)^2 (-1 + e^{r\Delta t} (e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t})) \\ - S(t)^2 e^{2r\Delta t}$$

$$\text{note } e^{r\Delta t} = 1 + r\Delta t + \frac{(r\Delta t)^2}{2} + \dots \\ = 1 + r\Delta t + O(\Delta t^2)$$

$$\text{similarly } e^{\sigma^2 \Delta t} = 1 + \sigma^2 \Delta t + \frac{(\sigma^2 \Delta t)^2}{2} + \dots \\ \text{by Taylor expansion} \\ e^{-\sigma^2 \Delta t} = 1 - \sigma^2 \Delta t + \frac{(\sigma^2 \Delta t)^2}{2} - \frac{(\sigma^2 \Delta t)^3}{6} + O(\Delta t^4) \\ \Rightarrow e^{r\Delta t} (e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t}) \\ = (1 + r\Delta t + O(\Delta t^2)) (e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t}) \\ = e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t} + (r\Delta t + O(\Delta t^2)) (e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t}) \\ = e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t} + (r\Delta t + O(\Delta t^2)) (2 + O(\Delta t^2)) \\ = e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t} + 2r\Delta t + O(\Delta t^2)$$

$$\text{So, } \text{Var}[S(t+\Delta t)^2 | S(t)] = S(t)^2 (-1 + e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t} + 2r\Delta t + O(\Delta t^2)) - S(t)^2 e^{2r\Delta t} \\ \text{(by (*))} \\ = S(t)^2 (-1 + 2 + \sigma^2 \Delta t + r\Delta t + O(\Delta t^2)) \\ - S(t)^2 e^{2r\Delta t} \\ = S(t)^2 (1 + r\Delta t + \sigma^2 \Delta t + O(\Delta t^2)) \\ - S(t)^2 (1 + 2r\Delta t + O(\Delta t^2)) \\ \text{(Taylor expansion for } e^{2r\Delta t} \text{)} \\ = S(t)^2 (\sigma^2 \Delta t + O(\Delta t^2)) = S(t)^2 [\sigma^2 \Delta t + O(\Delta t^2)]$$

$$\text{noting that } e^{\sigma^2 \Delta t} + e^{-\sigma^2 \Delta t} = 1 + \sigma^2 \Delta t + \frac{\sigma^4 \Delta t^2}{2} + \frac{\sigma^6 \Delta t^3}{6} + O(\Delta t^4) \\ + 1 - \sigma^2 \Delta t + \frac{\sigma^4 \Delta t^2}{2} - \frac{\sigma^6 \Delta t^3}{6} + O(\Delta t^4)$$

$$(*) = 2 + \sigma^2 \Delta t + O(\Delta t^2)$$

$$\text{and } O(\Delta t^2) = O(\Delta t)^2$$

Hilary

Q3a Construct a portfolio by

$$\begin{aligned}\Pi_0 &= e^{-rT}(\bar{K} - K) - (\bar{P}_0 - P_0) \\ &= \underbrace{e^{-rT}(\bar{K} - K)}_{\text{hold bond}} - \underbrace{\bar{P}_0}_{\text{short put}} + \underbrace{P_0}_{\text{long put}}\end{aligned}$$

$$\text{so } \Pi_t = e^{-(r-\epsilon)t}(\bar{K} - K) - \bar{P}_t + P_t \quad \text{for } 0 \leq t \leq T.$$

$$\text{Note } \Pi_T = \bar{K} - K - \bar{P}_T + P_T.$$

Let $S(T)$ be the price of underlying at expiry T .

If $S(T) \geq \bar{K}$, then

$$\bar{P}_T = \max(\bar{K} - S(T), 0) = 0,$$

$$P_T = \max(K - S(T), 0) = 0$$

$$\text{and so } \Pi_T = \bar{K} - K > 0.$$

If $K \leq S(T) < \bar{K}$, then

$$\bar{P}_T = \bar{K} - S(T),$$

$$P_T = 0$$

$$\begin{aligned}\text{and so } \Pi_T &= \bar{K} - K - (\bar{K} - S(T)) + 0 \\ &= S(T) - K \geq 0.\end{aligned}$$

If $S(T) < K$, then

$$\bar{P}_T = \bar{K} - S(T),$$

$$P_T = K - S(T)$$

$$\begin{aligned}\text{and so } \Pi_T &= \bar{K} - K - (\bar{K} - S(T)) + K - S(T) \\ &= 0.\end{aligned}$$

Thus, we see that in any case,
 $\Pi_T \geq 0$.

Thus, no arbitrage implies that

$$\Pi_t \geq 0, \quad \text{for all } 0 \leq t \leq T. \quad \text{at time } t$$

Otherwise, if $\Pi_t < 0$ for some $0 \leq t < T$, then \forall we can buy this portfolio (obtaining cash $-\Pi_t$) and short this portfolio at time T , obtaining $\Pi_T \geq 0$, this is an arbitrage.

$\pi_t \geq 0$ for $0 \leq t \leq T$ implies

$$\pi_t = e^{-r(T-t)} (\bar{K} - K) - (\bar{P}_t - P_t) \geq 0$$

$$\Rightarrow \bar{P}_t - P_t \leq e^{-r(T-t)} (\bar{K} - K).$$

Q3b) We use backwards induction. Employ S_j^N notation for undelaying as in lecture 4.

Base case:

$$\bar{P}_j^N - P_j^N = \max(\bar{K} - S_j^N, 0) - \max(K - S_j^N, 0)$$

$$= \begin{cases} \bar{K} - K & \text{if } S_j^N < K \\ \bar{K} - S_j^N & \text{if } K \leq S_j^N < \bar{K} \\ 0 & \text{if } K \leq S_j^N \end{cases}$$

$$\leq \bar{K} - K = e^{-r(N-N)\Delta t} (\bar{K} - K)$$

noting that if $K \leq S_j^N < \bar{K}$, then

$$\bar{K} - S_j^N \leq \bar{K} - K.$$

(which is equivalent to $S_j^N \geq K$.)

Inductive case: Let n s.t. $0 \leq n < N$, $n \in \mathbb{Z}$ and $0 \leq j \leq n$.

Assume we have

$$\bar{P}_j^{n+1} - P_j^{n+1} \leq e^{-r(N-(n+1))\Delta t} (\bar{K} - K)$$

for $0 \leq j \leq n+1$.

Let $q^* = \frac{e^{r\Delta t}d}{u-d}$. (We use risk-neutral valuation.)

Then for $0 \leq j \leq n$,

$$\bar{P}_j^n - P_j^n = e^{-r\Delta t} (q^* \bar{P}_{j+1}^{n+1} + (1-q^*) \bar{P}_j^{n+1} - q^* P_{j+1}^{n+1} - (1-q^*) P_j^{n+1})$$

$$= e^{-r\Delta t} (q^* (\bar{P}_{j+1}^{n+1} - P_{j+1}^{n+1}) + (1-q^*) (\bar{P}_j^{n+1} - P_j^{n+1}))$$

(inductive hypothesis)

$$\leq e^{-r\Delta t} (q^* e^{-r(N-n-1)\Delta t} (\bar{K} - K) + (1-q^*) e^{-r(N-n-1)\Delta t} (\bar{K} - K))$$

$$= e^{-r(N-n)\Delta t} (\bar{K} - K).$$

This finishes the induction.

3b) To be precise, the statement that we are inducting is
(denoted by $P(n)$, for $0 \leq n \leq N$)

$$P(n): \text{for any } 0 \leq j \leq n, \quad \bar{p}_j^n - p_j^n \leq e^{-\frac{n(N-n)}{K}} (\bar{K} - K)$$

(49)

$$S_j^n \begin{cases} S_{j+1}^{n+1} = u S_j^n \\ S_{j+1}^{n+1} = d S_j^n \end{cases}$$

Note that $S_0^N = d^N S_0 = \left(\frac{d}{u}\right)^N S_0$

$$= S_0 e^{-\frac{rN\Delta t}{\sqrt{N}}}$$

$$= S_0 e^{-\sigma \sqrt{N\Delta t}}$$

noting that $\Delta t = \frac{T}{N}$,

Thus, $\Delta t \rightarrow 0$ iff $N \rightarrow \infty$.

Thus we see that $S_0^N = S_0 e^{-\sigma \sqrt{N\Delta t}} \rightarrow 0$ as $N \rightarrow \infty$

Thus $p_0^N = \max(K - S_0^N, 0) \rightarrow K$ as $N \rightarrow \infty$.

Since $S_0^N \rightarrow 0$ and $K \geq 0$.

b) Recall that $\frac{u}{d} = \frac{e^{r\Delta t} - d}{u - d} = \frac{e^{r\Delta t} - e^{-r\Delta t}}{e^{r\Delta t} - e^{-r\Delta t}}$. Assume $\sigma > 0$.

Then, we see $p \leq \frac{u}{d} \leq 1$

iff $0 \leq \frac{e^{r\Delta t} - e^{-r\Delta t}}{e^{r\Delta t} - e^{-r\Delta t}} \leq 1$

(*) iff $(e^{-\sigma\sqrt{\Delta t}} \leq e^{r\Delta t} \text{ and } e^{r\Delta t} \leq e^{\sigma\sqrt{\Delta t}})$

Note $\lim_{\Delta t \rightarrow 0} (r + \sigma\sqrt{\Delta t}) = r$

and $\lim_{\Delta t \rightarrow 0} (r - \sigma\sqrt{\Delta t}) = r$.

Hence if Δt is sufficiently small and $\sigma > 0$, then we have

$r + \sigma\sqrt{\Delta t} > 0$ and $r - \sigma\sqrt{\Delta t} > 0$,

and so $\sqrt{\Delta t} (r + \sigma\sqrt{\Delta t}) > 0$ and $\sqrt{\Delta t} (r - \sigma\sqrt{\Delta t}) > 0$.

(f) Thus, under the assumption that Δt is sufficiently small and $\sigma > 0$,

$$\frac{e^{rat}}{e^{-\sigma\Delta t}} = e^{rat + \sigma\Delta t} = e^{\Delta t(\sigma + r/\Delta t)} > 1 \quad \text{since } \Delta t(\sigma + r/\Delta t) > 0$$

$$\frac{e^{\sigma\Delta t}}{e^{rat}} = e^{\sigma\Delta t - rat} = e^{\Delta t(\sigma - r/\Delta t)} > 1 \quad \text{since } \Delta t(\sigma - r/\Delta t) > 0$$

so we have $e^{rat} > e^{-\sigma\Delta t}$ and $e^{\sigma\Delta t} > e^{rat}$.

Thus, by (*), we have $0 \leq q^* \leq 1$, if Δt is sufficiently small and $\sigma > 0$.

$$q^* S_{j+1}^{n+1} + (1 - q^*) S_j^{n+1}$$

$$= \frac{e^{rat} - e^{-\sigma\Delta t}}{e^{\sigma\Delta t} - e^{-\sigma\Delta t}} S_j^n e^{\sigma\Delta t} + \frac{e^{\sigma\Delta t} - e^{rat}}{e^{\sigma\Delta t} - e^{-\sigma\Delta t}} S_j^n e^{-\sigma\Delta t}$$

$$= S_j^n \left(\frac{e^{\sigma\Delta t + rat} - 1}{e^{\sigma\Delta t} - e^{-\sigma\Delta t}} + \frac{1 - e^{rat - \sigma\Delta t}}{e^{\sigma\Delta t} - e^{-\sigma\Delta t}} \right)$$

$$= S_j^n \frac{e^{\sigma\Delta t + rat} - e^{rat - \sigma\Delta t}}{e^{\sigma\Delta t} - e^{-\sigma\Delta t}}$$

$$= e^{rat} S_j^n$$

$$\Rightarrow e^{-rat} (q^* S_{j+1}^{n+1} + (1 - q^*) S_j^{n+1})$$

$$= S_j^n$$

Hilary

(4c) We use backwards induction.

Base case: Note that since $S_0^0 \geq 0$, and that $u, d > 0$, we have $S_j^n \geq 0$ for any $0 \leq j \leq n$, $0 \leq n \leq N$. Thus, for any $0 \leq j \leq N$,

$$0 \leq P_j^N = \max(K - S_j^N, 0) \leq K - S_j^N \leq K.$$

Note that we assume Δt is sufficiently small and $\sigma > 0$, so that $0 \leq \tilde{q}^* \leq 1$. We can use risk-neutral valuation for P_j^n , $\forall 0 \leq j \leq n$, $0 \leq n \leq N$.

Inductive case:

Let n s.t. $0 \leq n < N$ and $n \in \mathbb{Z}$.

Assume we have

$$0 \leq P_j^{n+1} \leq K \quad \text{for } 0 \leq j \leq n+1.$$

$$\text{Let } 0 \leq j \leq n, \text{ Then } P_j^n = e^{-r\Delta t} (P_{j+1}^{n+1} \tilde{q}^* + P_j^{n+1} (1 - \tilde{q}^*))$$

$$\begin{aligned} \text{(inductive hypothesis)} \quad &\leq e^{-r\Delta t} (K \tilde{q}^* + K (1 - \tilde{q}^*)) \\ &= e^{-r\Delta t} K \leq K \end{aligned}$$

$$\begin{aligned} \text{also } P_j^n &= e^{-r\Delta t} (P_{j+1}^{n+1} \tilde{q}^* + P_j^{n+1} (1 - \tilde{q}^*)) \\ \text{(inductive hypothesis)} \quad &\geq e^{-r\Delta t} (0 + 0) = 0. \end{aligned}$$

This is true for any $0 \leq j \leq n$.
This finishes induction.

To be precise, the statement that we are inducting is this (denoted by $P(n)$, where $0 \leq n \leq N$)

$$P(n): \text{ for any } 0 \leq j \leq n, \quad 0 \leq P_j^n \leq K.$$

Hilary

Q5) The statement we want to induct is, where $N \in \mathbb{Z}^+$,

$P(N)$: In a N -period model,

$$V_0 = e^{-rN\Delta t} \sum_{k=0}^N \binom{N}{k} (q^*)^k (1-q^*)^{N-k} V_k^N$$

Base Case: Want to show $P(1)$ holds.

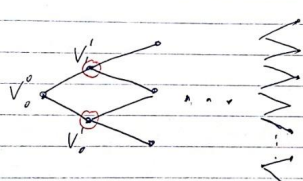
Recall in class that for a 1-period model,

$$V_0 = e^{-r\Delta t} (q^* V_1' + (1-q^*) V_0')$$

$$= e^{-r\Delta t} \sum_{k=0}^1 \binom{1}{k} (q^*)^k (1-q^*)^{1-k} V_k'$$

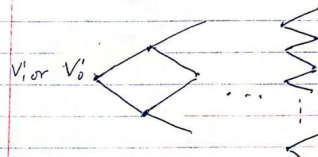
Inductive step: Assume $P(N)$ holds, where $N \in \mathbb{Z}^+$.

Want to show $P(N+1)$ holds.



a $(N+1)$ -period model

If we start from V_1' or V_0' and treat them as the starting node of the binomial lattice, then we obtain a N -period model



a N -period model
(*)

Hilary

(25) Hence, by $P(N)$, we have

$$V_i^1 = e^{-N\alpha t} \sum_{k=1}^{N+1} \binom{N}{k-1} \left(\frac{\alpha}{\beta}\right)^{k-1} (1-\frac{\alpha}{\beta})^{N+1-k} V_k^{N+1}$$

(remark: the end nodes of $(**)$ for the case V_i^1 will be $V_{N+1}^{N+1}, V_N^{N+1}, \dots, V_1^{N+1}$,

The summation formula looks a little bit different since the starting node is V_i^1 .)

$$\text{Similarly, } V_0^1 = e^{-N\alpha t} \sum_{k=0}^N \binom{N}{k} \left(\frac{\alpha}{\beta}\right)^k (1-\frac{\alpha}{\beta})^{N-k} V_k^{N+1}$$

$$\text{Thus, } V_0^0 = e^{-\alpha t} \left(\frac{\alpha}{\beta} V_i^1 + (1-\frac{\alpha}{\beta}) V_0^1 \right)$$

$$= e^{-(N+1)\alpha t} \left(\frac{\alpha}{\beta} \sum_{k=1}^{N+1} \binom{N}{k-1} \left(\frac{\alpha}{\beta}\right)^{k-1} (1-\frac{\alpha}{\beta})^{N+1-k} V_k^{N+1} \right.$$

You should expand at the last layer, i.e. establish relationship between V_i^N and V_i^{N+1} . In fact, the problem is unclear induction set up, rather than anything else.

$$+ (1-\frac{\alpha}{\beta}) \sum_{k=0}^N \binom{N}{k} \left(\frac{\alpha}{\beta}\right)^k (1-\frac{\alpha}{\beta})^{N-k} V_k^{N+1})$$

$$= e^{-(N+1)\alpha t} \sum_{k=0}^{N+1} \binom{N+1}{k} \left(\frac{\alpha}{\beta}\right)^k (1-\frac{\alpha}{\beta})^{N+1-k} V_k^{N+1}$$

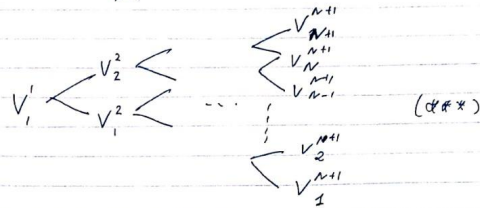
This finishes induction.

History

25) Further explanation on

$$V_1' = e^{-N\Delta t} \sum_{k=1}^{N+1} \binom{N}{k-1} \left(\frac{q}{p}\right)^{k-1} (1-\frac{q}{p})^{N+1-k} V_k^{N+1}$$

note

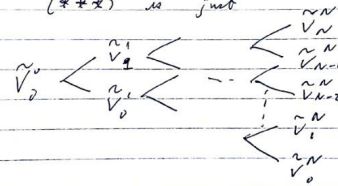


if we rename $V_1', V_2^2, V_1^2, \dots$ by

$$V_1' = V_0^0$$

$$V_j^n = V_{j-1}^{n-1} \text{ for } 1 \leq j \leq n, 1 \leq n \leq N+1,$$

then $(***)$ is just



$$\text{so that } V_1' = V_0^0 = e^{-N\Delta t} \sum_{k=0}^N \binom{N}{k} \left(\frac{q}{p}\right)^k (1-\frac{q}{p})^{N-k} V_k^N$$

$$= e^{-N\Delta t} \sum_{k=1}^{N+1} \binom{N}{k-1} \left(\frac{q}{p}\right)^{k-1} (1-\frac{q}{p})^{N+1-k} V_{k-1}^N$$

$$= e^{-N\Delta t} \sum_{k=1}^{N+1} \binom{N}{k-1} \left(\frac{q}{p}\right)^{k-1} (1-\frac{q}{p})^{N+1-k} V_k^{N+1}$$

Library

Q6. Consider $G(z, t) : \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by

$$G(z, t) = \frac{1}{2} z^2 - \frac{t}{2}.$$

Then $\frac{\partial G}{\partial z} = z$ $\frac{\partial G}{\partial t} = -\frac{1}{2}$

$$\frac{\partial^2 G}{\partial z^2} = 1$$

By Itô's Lemma,

$$dG(z(t), t) = \frac{\partial G}{\partial z}(z(t), t) \cdot dz(t) + \frac{\partial G}{\partial t}(z(t), t) dt$$

$$+ \frac{1}{2} \frac{\partial^2 G}{\partial z^2}(z(t), t) (dz(t))^2$$

(note $(dz(t))^2 = dt$)

$$= z(t) dz(t) - \frac{1}{2} dt + \frac{1}{2} dt$$

$$= z(t) dz(t)$$

Hence $G(z(T), T) - G(z(0), 0) = \int_0^T z(t) dz(t)$

$$\Rightarrow \int_0^T z(t) dz(t) = G(z(T), T)$$

$$= \frac{1}{2} z(T)^2 - \frac{T}{2}$$

Since $G(z(0), 0) = \frac{1}{2} \cdot 0^2 - \frac{0}{2} = 0$.

Hint

