STA 2503/MMF 1928 Project 1 - American Options

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Question 1

Proof. We start the proof by considering

$$X^{(N)} = \log(\frac{S_T}{S_0}) = \sum_{n=1}^{N} (r\Delta t + \sigma \sqrt{\Delta t} \epsilon_n)$$

The m.g.f of $X^{(N)}$ is:

$$\mathbb{E}^{\mathbb{P}}[e^{uX^{(N)}}] = \mathbb{E}^{\mathbb{P}}[e^{\sum_{n=1}^{N} (ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n)}]$$

$$= \mathbb{E}^{\mathbb{P}}[\prod_{n=1}^{N} e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n}]$$

$$= (\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}])^N \quad \text{as } \epsilon_n \text{ is i.i.d}$$

We then investigate the inner term $\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + uo\sqrt{\Delta t}\epsilon_1}]$:

$$\begin{split} \mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}] &= e^{ur\Delta t + u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 + \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \\ &\quad + e^{ur\Delta t - u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 - \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \\ &= [1 + ur\Delta t + u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 + \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \\ &\quad + [1 + ur\Delta t - u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 - \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \end{split}$$

Note that we collect all the terms containing Δt with order 2 or higher to be $o(\Delta t)$ because as $N \to \infty$ and $\Delta t \to 0$, $o(\Delta t)$ will converge to 0 faster than Δt or Δt with lower orders.

We then continue the algebra manipulation:

$$\begin{split} \mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n}] &= [1 + ur\Delta t + u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 + \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \\ &\quad + [1 + ur\Delta t - u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 - \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t}) \\ &= \frac{1}{2}[1 + \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t} + ur\Delta t + u\sigma\sqrt{\Delta t} \\ &\quad + u((\mu - r) - \frac{1}{2}\sigma^2)\Delta t + \frac{1}{2}u^2\sigma^2\Delta t \\ &\quad + 1 - \frac{(\mu - r) - \frac{1}{2}\sigma^2}{\sigma}\sqrt{\Delta t} + ur\Delta t - u\sigma\sqrt{\Delta t} \\ &\quad + u((\mu - r) - \frac{1}{2}\sigma^2)\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \\ &= 1 + ur\Delta t + u((\mu - r) - \frac{1}{2}\sigma^2)\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t) \\ &= 1 + (ur + u\mu - ur - \frac{1}{2}u\sigma^2 + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t) \\ &= 1 + (u(\mu - \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t) \\ &= e^{(u(\mu - \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t} + o(\Delta t) \end{split}$$

As a result:

$$\begin{split} \mathbb{E}^{\mathbb{P}}[e^{uX^{(N)}}] &= (e^{(u(\mu - \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t} + o(\Delta t))^N \\ &= e^{(u(\mu - \frac{1}{2}\sigma^2)T + \frac{1}{2}u^2\sigma^2T)} \quad \text{as } N \to \infty \text{ and } \Delta t = \frac{T}{N} \end{split}$$

We notice that the m.g.f of the $X^{(N)}$ is equal to the m.g.f of a random variable Y which follows the normal distribution with mean to be $(\mu - \frac{1}{2}\sigma^2)T$ and variance to be σ^2T .

Thus, we prove that:

$$X^{(N)} \xrightarrow[N \to \infty]{d} (\mu - \frac{1}{2}\sigma^2)T + \sigma^2 TZ$$

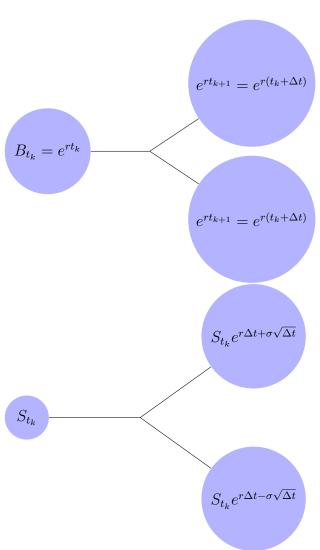
where

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

Question 2

Part $1-\mathbb{Q}$

We firstly find the martingale measure $\mathbb{Q}(\epsilon_k = \pm 1)$ where $\mathbb{Q}(\epsilon_k = 1) = q_k$ and $\mathbb{Q}(\epsilon_k = -1) = 1 - q_k$. Consider time t_k and t_{k+1} , we have the following for B and S.



We construct the following Q-martingale:

$$\begin{split} \frac{S_{t_k}}{e^{rt_k}} &= q_k \frac{S_{t_k} e^{r\Delta t + \sigma\sqrt{\Delta t}}}{e^{r(t_k + \Delta t)}} + (1 - q_k) \frac{S_{t_k} e^{r\Delta t - \sigma\sqrt{\Delta t}}}{e^{r(t_k + \Delta t)}} \\ \Rightarrow 1 &= q_k \frac{e^{r\Delta t + \sigma\sqrt{\Delta t}}}{e^{r\Delta t}} + (1 - q_k) \frac{e^{r\Delta t - \sigma\sqrt{\Delta t}}}{e^{r\Delta t}} \\ \Rightarrow q_k (e^{r\Delta t + \sigma\sqrt{\Delta t}} - e^{r\Delta t - \sigma\sqrt{\Delta t}}) &= e^{r\Delta t} - e^{r\Delta t - \sigma\sqrt{\Delta t}} \\ \Rightarrow q_k &= q = \frac{e^{r\Delta t} - e^{r\Delta t - \sigma\sqrt{\Delta t}}}{e^{r\Delta t + \sigma\sqrt{\Delta t}} - e^{r\Delta t - \sigma\sqrt{\Delta t}}} \qquad \text{As } q_k \text{ does not depend on } k \\ \Rightarrow q &= \frac{1 - e^{-\sigma\sqrt{\Delta t}}}{e^{\sigma\sqrt{\Delta t}} - e^{-\sigma\sqrt{\Delta t}}} \end{split}$$

We then Taylor-expand the exponential terms and collect all all the terms containing Δt with order 2 or higher to be $o(\Delta t)$:

$$\begin{split} q &= \frac{e^{r\Delta t} - e^{r\Delta t - \sigma\sqrt{\Delta t}}}{e^{r\Delta t + \sigma\sqrt{\Delta t}} - e^{r\Delta t - \sigma\sqrt{\Delta t}}} \\ &= \frac{1 + r\Delta t - (1 + r\Delta t - \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t) + o(\Delta t)}{1 + r\Delta t + \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t - (1 + r\Delta t - \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t) + o(\Delta t)} \\ &= \frac{\sigma\sqrt{\Delta t} - \frac{1}{2}\sigma^2\Delta t + o(\Delta t)}{2\sigma\sqrt{\Delta t} + o(\Delta t)} \\ &= \frac{1}{2} - \frac{1}{4}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t}) \\ &= \frac{1}{2}(1 - \frac{1}{2}\sigma\sqrt{\Delta t}) + o(\sqrt{\Delta t}) \end{split}$$

Now let us find the \mathbb{Q} distribution of S_T as $N \to \infty$. We first define $X_T = log(\frac{S_T}{S_0})$, then we have:

$$X_T = \log(\frac{S_T}{S_0}) = \log(e^{\sum_{n=1}^N r\Delta t + \sigma\sqrt{\Delta t}\epsilon_n}) = \sum_{n=1}^N r\Delta t + \sigma\sqrt{\Delta t}\epsilon_n$$

Consider the m.g.f of X_T :

$$\mathbb{E}^{\mathbb{P}}[e^{uX_T}] = \mathbb{E}^{\mathbb{P}}[e^{\sum_{n=1}^{N}(ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n)}]$$

$$= \mathbb{E}^{\mathbb{P}}[\prod_{n=1}^{N}e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n}]$$

$$= (\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}])^N \quad \text{as } \epsilon_n \text{ is i.i.d}$$

We then investigate the inner term $\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}]$:

$$\begin{split} \mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}] &= e^{ur\Delta t + u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 - \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &\quad + e^{ur\Delta t - u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 + \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &= [1 + ur\Delta t + u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 - \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &\quad + [1 + ur\Delta t - u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 + \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &= \frac{1}{2}[1 - \frac{1}{2}\sigma\sqrt{\Delta t} + ur\Delta t + u\sigma\sqrt{\Delta t} - \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t \\ &\quad + 1 + \frac{1}{2}\sigma\sqrt{\Delta t} + ur\Delta t - u\sigma\sqrt{\Delta t} - \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \\ &= 1 + ur\Delta t - \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t) \\ &= 1 + (u(r - \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t) \\ &= e^{(u(r - \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t)} \end{split}$$

Then we have the m.g.f of X_T to be:

$$\mathbb{E}^{\mathbb{P}}[e^{uX_T}] = (\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}])^N$$

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

$$= e^{(u(r - \frac{1}{2}\sigma^2)T + \frac{1}{2}u^2\sigma^2T)} \quad \text{as } N \to \infty \text{ and } \Delta t = \frac{T}{N}$$

We notice that the m.g.f of the X_T is equal to the m.g.f of a random variable Y which follows the normal distribution with mean to be $(r - \frac{1}{2}\sigma^2)T$ and variance to be σ^2T . Thus,

$$S_T \stackrel{\mathbb{Q}}{\sim} lognormal((r - \frac{1}{2}\sigma^2)T + \log(S_0), \sigma^2T)$$

Part 2 $- \mathbb{Q}^S$

Similarly, we want to find the martingale measure $\mathbb{Q}^S(\epsilon_k = \pm 1)$ where $\mathbb{Q}^S(\epsilon_k = 1) = h_k$ and $\mathbb{Q}^S(\epsilon_k = -1) = 1 - h_k$. Consider time t_k and t_{k+1} , this time we use S as the numeraire.

We construct the following \mathbb{Q}^S -martingale:

$$\begin{split} \frac{e^{rt_k}}{S_{t_k}} &= h_k \frac{e^{r(t_k + \Delta t)}}{S_{t_k} e^{r\Delta t + \sigma \sqrt{\Delta t}}} + (1 - h_k) \frac{e^{r(t_k + \Delta t)}}{S_{t_k} e^{r\Delta t - \sigma \sqrt{\Delta t}}} \\ \Rightarrow 1 &= h_k \frac{e^{r\Delta t}}{e^{r\Delta t + \sigma \sqrt{\Delta t}}} + (1 - h_k) \frac{e^{r\Delta t}}{e^{r\Delta t - \sigma \sqrt{\Delta t}}} \\ \Rightarrow e^{2r\Delta t} &= h_k e^{2r\Delta t - \sigma \sqrt{\Delta t}} + (1 - h_k) e^{2r\Delta t + \sigma \sqrt{\Delta t}} \\ \Rightarrow e^{2r\Delta t} &= e^{2r\Delta t} - e^{2r\Delta t + \sigma \sqrt{\Delta t}} = h_k (e^{2r\Delta t - \sigma \sqrt{\Delta t}} - e^{2r\Delta t + \sigma \sqrt{\Delta t}}) \\ \Rightarrow h_k &= h = \frac{e^{2r\Delta t} - e^{2r\Delta t + \sigma \sqrt{\Delta t}}}{e^{2r\Delta t - \sigma \sqrt{\Delta t}} - e^{2r\Delta t + \sigma \sqrt{\Delta t}}} \qquad \text{As } h_k \text{ does not depend on } k \\ \Rightarrow h &= \frac{1 - e^{\sigma \sqrt{\Delta t}}}{e^{-\sigma \sqrt{\Delta t}} - e^{\sigma \sqrt{\Delta t}}} \end{split}$$

We then Taylor-expand the exponential terms and collect all all the terms containing Δt with order 2 or higher to be $o(\Delta t)$:

$$\begin{split} h &= \frac{e^{2r\Delta t} - e^{2r\Delta t + \sigma\sqrt{\Delta t}}}{e^{2r\Delta t - \sigma\sqrt{\Delta t}} - e^{2r\Delta t + \sigma\sqrt{\Delta t}}} \\ &= \frac{1 + 2r\Delta t - (1 + 2r\Delta t + \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t) + o(\Delta t)}{1 + 2r\Delta t - \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t - (1 + 2r\Delta t + \sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t) + o(\Delta t)} \\ &= \frac{-\sigma\sqrt{\Delta t} - \frac{1}{2}\sigma^2\Delta t + o(\Delta t)}{-2\sigma\sqrt{\Delta t} + o(\Delta t)} \\ &= \frac{\sigma\sqrt{\Delta t} + \frac{1}{2}\sigma^2\Delta t + o(\Delta t)}{2\sigma\sqrt{\Delta t} + o(\Delta t)} \\ &= \frac{1}{2}(1 + \frac{1}{2}\sigma\sqrt{\Delta t}) + o(\sqrt{\Delta t}) \end{split}$$

Now let us find the \mathbb{Q}^S distribution of S_T as $N \to \infty$. We first define $X_T =$

 $log(\frac{S_T}{S_0})$, then we have:

$$X_T = \log(\frac{S_T}{S_0}) = \log(e^{\sum_{n=1}^N r\Delta t + \sigma\sqrt{\Delta t}\epsilon_n}) = \sum_{n=1}^N r\Delta t + \sigma\sqrt{\Delta t}\epsilon_n$$

Consider the m.g.f of X_T :

$$\mathbb{E}^{\mathbb{P}}[e^{uX_T}] = \mathbb{E}^{\mathbb{P}}[e^{\sum_{n=1}^{N}(ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n)}]$$

$$= \mathbb{E}^{\mathbb{P}}[\prod_{n=1}^{N}e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_n}]$$

$$= (\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}])^N \quad \text{as } \epsilon_n \text{ is i.i.d}$$

We then investigate the inner term $\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}]$:

$$\begin{split} \mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}] &= e^{ur\Delta t + u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 + \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &\quad + e^{ur\Delta t - u\sigma\sqrt{\Delta t}} \cdot \frac{1}{2}(1 - \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &= [1 + ur\Delta t + u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 + \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &\quad + [1 + ur\Delta t - u\sigma\sqrt{\Delta t} + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \cdot \frac{1}{2}(1 - \frac{1}{2}\sigma\sqrt{\Delta t} + o(\sqrt{\Delta t})) \\ &= \frac{1}{2}[1 + \frac{1}{2}\sigma\sqrt{\Delta t} + ur\Delta t + u\sigma\sqrt{\Delta t} + \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t \\ &\quad + 1 - \frac{1}{2}\sigma\sqrt{\Delta t} + ur\Delta t - u\sigma\sqrt{\Delta t} + \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t)] \\ &= 1 + ur\Delta t + \frac{1}{2}u\sigma^2\Delta t + \frac{1}{2}u^2\sigma^2\Delta t + o(\Delta t) \\ &= 1 + (u(r + \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t) \\ &= e^{(u(r + \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t)} \end{split}$$

Then we have the m.g.f of X_T to be:

$$\mathbb{E}^{\mathbb{P}}[e^{uX_T}] = (\mathbb{E}^{\mathbb{P}}[e^{ur\Delta t + u\sigma\sqrt{\Delta t}\epsilon_1}])^N$$

$$= (e^{(u(r + \frac{1}{2}\sigma^2) + \frac{1}{2}u^2\sigma^2)\Delta t + o(\Delta t)})^N$$

$$= e^{(u(r + \frac{1}{2}\sigma^2)T + \frac{1}{2}u^2\sigma^2T)} \quad \text{as } N \to \infty \text{ and } \Delta t = \frac{T}{N}$$

We notice that the m.g.f of the X_T is equal to the m.g.f of a random variable Y which follows the normal distribution with mean to be $(r + \frac{1}{2}\sigma^2)T$ and variance to be σ^2T . Thus,

$$S_T \overset{\mathbb{Q}^S}{\sim} lognormal((r + \frac{1}{2}\sigma^2)T + \log(S_0), \sigma^2 T)$$