

SFM 2017:

1.) ca) Let $(\Omega, \mathcal{F}, \mathbb{P})$ be probability space: $(\mathcal{F}_n)_{n \geq 0}$ be a filtration

If X_n is \mathcal{F}_n -measurable, $X_n \in L^1(\mathbb{P})$ and $(\forall n \geq 0)$;

$(n \geq 1)$ $\mathbb{E}[X_n | \mathcal{F}_{n-1}] = X_{n-1}$: $(X_n)_{n \geq 1}$ is a martingale.

cb) Integrable: $X_0 = \Delta_0 \in L^1(\mathbb{P})$

(Induct): If X_0, \dots, X_n integrable, $\mathbb{E}[|X_{n+1}|] \leq \mathbb{E}[|X_n|] + \sum_{k=0}^n \frac{1}{n+1} \mathbb{E}[|X_k|] \cdot \mathbb{E}[|\Delta_{n+1}|]$

Since X_n is a function of $\{\Delta_0, \dots, \Delta_n\}$, X_k, Δ_{n+1} independent for $k \leq n$

$$\therefore \mathbb{E}[|X_{n+1}|] \leq \mathbb{E}[|X_n|] + \frac{\mathbb{E}[|\Delta_{n+1}|]}{n+1} \sum_{k=0}^n \mathbb{E}[|X_k|] < \infty \quad (\text{induction hyp.})$$

$\therefore X_{n+1} \in L^1 \Rightarrow$ Induction complete.

X_n is clearly $\mathcal{G}(X_0, \dots, X_n)$ -measurable.

$$\mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n] = \sum_{k=0}^n \frac{1}{n+1} \mathbb{E}[\Delta_{n+1} \cdot X_k | \mathcal{F}_n]$$

$$= \mathbb{E}[\Delta_{n+1}] \cdot f(X_0, \dots, X_n) \quad (X_0, \dots, X_n \text{ is } \mathcal{F}_n\text{-measurable, } \Delta_{n+1} \text{ indep. of } \mathcal{F}_n)$$

$$= 0.$$

\therefore Valid martingale.

cc) If $T \leq N$ A.S.: $(X_n)_{n \geq 1}$ is a martingale.

Optional stopping: $\mathbb{E}[X_T] = \mathbb{E}[X_0]$

Let $Y_n = X_{\min\{T, n\}}$:

$$Y_n = \sum_{k=0}^{n-1} 1_{T \geq k} X_k + 1_{T \geq n} X_n \quad \text{is } \mathcal{G}(X_0, \dots, X_n) \text{ measurable}$$

$$\mathbb{E}[|Y_n|] \leq \sum_{k=0}^n \mathbb{E}[|X_k|] < \infty$$

$$\begin{aligned} \mathbb{E}[Y_{n+1} | \mathcal{F}_n] &= \sum_{k=0}^n 1_{T \geq k} X_k + \mathbb{E}[X_{n+1} 1_{T > n} | \mathcal{F}_n] \\ &= \sum_{k=0}^n 1_{T \geq k} X_k + 1_{T > n} \mathbb{E}[X_{n+1} | \mathcal{F}_n] \\ &= \sum_{k=0}^{n-1} 1_{T \geq k} X_k + (1_{T \geq n} X_n) = Y_n. \end{aligned}$$

\therefore Valid martingale.

Since $T \leq N$ A.S. $Y_N = Y_T$ (A.S.)

$$\begin{aligned} \therefore \mathbb{E}[Y_T] &= \mathbb{E}[Y_N] = \mathbb{E}[Y_0] \quad (\text{martingale}) \\ &= \mathbb{E}[X_0] \end{aligned}$$

2.) (a) Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space:

$(B_t)_{t \geq 0}$ is a Brownian Motion (BM) if:

$$\left. \begin{array}{l} B_0 = 0 \\ t \mapsto B_t \text{ is cont.} \end{array} \right\} \text{ A.S.}$$

$(B_t)_{t \geq 0}$ has independent increments, $\mathbb{E} \mathbb{P} (B_t - B_s) \sim N(0, t-s)$
for $t \geq s \geq 0$

(b) $\tilde{B}_0 = \frac{1}{c} B_0 = 0$ iff $B_0 = 0$

$t \mapsto \tilde{B}_t = \frac{1}{c} B_{c^2 t}$ is cont. iff $t \mapsto B_t$ is cont.

Since $(B_t)_{t \geq 0}$ has independent increments, $(B_{c^2 t})_{t \geq 0}$ increments must be independent $\Rightarrow \tilde{B}_t$ has independent increments.

$$(\tilde{B}_t - \tilde{B}_s) = \frac{1}{c} (B_{c^2 t} - B_{c^2 s}) \sim \frac{1}{c} \cdot N(0, c^2(t-s)) = N(0, t-s)$$

$$(c) \quad c \cdot Z = \sup_{t \geq 0} c \cdot B_t = \sup_{t \geq 0} c \cdot B_{t/c^2}$$

Since $(B_t)_{t \geq 0}$, $(c \cdot B_{t/c^2})_{t \geq 0}$ have the same distribution,
 $c \cdot Z \stackrel{(d)}{=} Z$

$$\therefore IP(Z \leq a) = IP(c \cdot Z \leq a) = IP(Z \leq a/c)$$

$\therefore IP(Z \leq a)$ is constant on $a \in (0, \infty)$.

Since $Z \in \{0, \infty\} \cup (0, \infty)$: $IP(Z \in (a, b]) = IP(Z \leq b) - IP(Z \leq a) = 0$
 for $a, b > 0$.

$$\therefore IP(Z \in (0, \infty)) = IP\left(\bigcup_{n \geq 1} \{Z \in (1/n, n]\}\right) \leq \sum_{n \geq 1} IP(Z \in (1/n, n]) = 0$$

$$= 0$$

$\therefore Z \in \{0, \infty\}$ A.S.

$$(d) \quad \text{If } Z = 0: \text{ Let } Z_1 = \sup_{t \geq 0} \{B_{t+1} - B_{t1} : t \geq 0\}$$

$$Z_1 \stackrel{(d)}{=} Z; \quad Z = 0 \Rightarrow B_{t1} \leq 0 \text{ and } B_{t+1} \leq B_{t1} \quad (t \geq 0)$$

$$\Rightarrow B_1 \leq 0 \text{ and } B_{Z_1} \leq B_1 < \infty$$

$$\Rightarrow B_1 \leq 0, \quad Z_1 \leq 0.$$

$$\therefore IP(Z = 0) \leq IP(B_1 \leq 0) \cdot IP\left(\sup_{t \geq 0} \{B_{t+1} - B_{t1} \geq 0\}\right)$$

($(B_{t+1} - B_{t1})_{t \geq 0}$ indep. of B_1)

$$= \frac{1}{2} \cdot IP(Z = 0)$$

$$\therefore IP(Z = 0) = 0 \Rightarrow Z = \infty \text{ A.S.}$$

3.) (a) Let S_t^k be price of asset k at time t , Fix $T > 0$.
Asset 0 be a riskless asset.

be

There is no arbitrage portfolios (*) iff \exists equivalent
martingale measure of $(S_t^k / S_t^0)_{0 \leq t \leq T}$

(*) $(\theta_k)_{k=0, \dots, T}$ is an arbitrage if it is a self-financing
P. pre-visible strategy and $\theta_0 \cdot S_0 \leq 0$, $\theta_T \cdot S_T \geq 0$ (A.S.)
 $IP(\theta_T \cdot S_T > 0) > 0$

(b) $(S_T - K)^+ - (K - S_T)^+ = S_T - K$

\therefore ~~market~~ $EC(S_0, k, T)$ be price of european call
with strike k , expiry T , current price S_0 .

$\therefore EC(S_t', k, T - t_0) - EP(S_t', k, T - t_0) = S_t' - K(1+r)^{-(T-t_0)}$

\therefore EC Pick call option iff $S_t' \geq K(1+r)^{-(T-t_0)}$

\therefore Payout = $(S_T' - K)^+ \mathbb{1}_{S_{t_0}' \geq K(1+r)^{t_0-T}} + (K - S_T')^+ \mathbb{1}_{S_{t_0}' \leq K(1+r)^{t_0-T}}$

(c) $C(k, t_0, T) = (S_T' - K)^+ = \mathbb{1}_{S_{t_0}' \leq K(1+r)^{t_0-T}} \{ (K - S_T')^+ - (S_T' - K)^+ \}$
 $= \mathbb{1}_{S_{t_0}' \leq K(1+r)^{t_0-T}} \{ K - S_T' \}$
 $= (K(1+r)^{t_0-T} - S_{t_0}')^+ + \mathbb{1}_{S_{t_0}' \leq K(1+r)^{t_0-T}} \{ S_{t_0}' - S_T' \}$
 $= (K(1+r)^{t_0-T} - S_{t_0}')^+ + \mathbb{1}_{S_{t_0}' \leq K(1+r)^{t_0-T}} \{ K - S_{t_0}' \}$

$$C(k, t_0, T) = (S_T' - k)^+ + \mathbb{1}_{S_{t_0}' \leq (1+r)^{t_0-T} \cdot k} \{ (k - S_T')^+ - (S_T' - k)^+ \}$$

$$= (S_T' - k)^+ + (1+r)^{t_0-T} (k - S_{t_0}')^+$$

$$+ \mathbb{1}_{S_{t_0}' \leq k \cdot (1+r)^{t_0-T}} \{ k - S_T + S_{t_0}' - (1+r)^{t_0-T} \cdot k \}$$

$$\begin{aligned} \mathbb{E}^G [\text{line 2} | F_{t_0}] &= \mathbb{1}_{S_{t_0}' \leq k \cdot (1+r)^{t_0-T}} \{ k - (1+r)^{t_0-T} k + S_{t_0}' - (1+r)^{T-t_0} S_{t_0}' \} \\ &= \mathbb{1}_{S_{t_0}' \leq k \cdot (1+r)^{t_0-T}} (1 - (1+r)^{t_0-T}) \{ k \cdot (1+r)^{t_0-T} - S_{t_0}' \} \end{aligned}$$

$$\begin{aligned} \text{Then } (1+r)^{-T} \cdot \mathbb{E}^G [\mathbb{1}_{S_{t_0}' \leq (1+r)^{t_0-T} \cdot k} \{ (k - S_T')^+ \} | F_{t_0}] &= (1+r)^{-T} \mathbb{1}_{S_{t_0}' \leq (1+r)^{t_0-T} \cdot k} \mathbb{E}^G [(k - S_T')^+ | S_{t_0}] \end{aligned}$$

definition

$$\begin{aligned} \text{Payoff}_A &= EC(S_t', k, T - t_0) + \text{Max}(k(1+r)^{-(T-t_0)} - S_t')^+ \\ &= EC(S_0, k, T - t_0) + EP(S_0, k(1+r)^{t_0-T}, t_0) \end{aligned}$$

$$\therefore \Pi = \Pi(EC(k, T)) + \Pi(EP(k(1+r)^{t_0-T}, t_0))$$

4) ca) $S_t^0 = (1+r)^t$, $S_t' = S_0' \cdot \prod_{k=1}^t (1+R_k)$ where $(R_k)_{k \geq 1}$ are iid,
 $IP(R_k = a) = p$, $IP(R_k = b) = q$, $-1 < a < b$. (CRR model)

Arbitrage free iff $a < r < b$.

$$G: ap^* + b(1-p^*) = r \Rightarrow (a-b)p^* = r-b \Rightarrow p^* = \frac{b-r}{b-a}$$

$G(R_k = a) = p^*$ is the martingale measure (unique)

(b) Assuming no arbitrage, market has unique martingale measure, \mathbb{Q} and all claims replicable

Let $(\Theta_n)_{n=0, \dots, T}$ be replicating strategy of C :

$$C = \Theta_{n,T} \cdot S_T \quad (\text{A.S.}) = (1+r)^T \Theta_T \cdot X_T$$

$$\therefore \mathbb{E}^{\mathbb{Q}}[C] = (1+r)^T \cdot \Theta_0 \cdot (S_0) = \Pi = (1+r)^{-T} \mathbb{E}^{\mathbb{Q}}[C]$$

is the fair price.

$$V(n; S_0, \dots, S_n) = \Theta_n(S_0, \dots, S_n) \cdot S_n =$$

$$(1+r)^{n-N} \mathbb{E}^{\mathbb{Q}}[C | \mathcal{F}_n](S_0, \dots, S_n)$$

$$\text{Hedge: } \Theta_{n+1} = \frac{V(n+1; S_0, \dots, S_n, S_n(1+b)) - V(n+1; S_0, \dots, S_n, S_n(1+a))}{(b-a) S_n}$$

$$(c) (1+b)^{T-t} \cdot \left(\frac{1+a}{1+b}\right)^n > K$$

$$\text{Let } N = \sup_{\inf} \left\{ n \geq n_0 : (1+b)^{T-t} \cdot \left(\frac{1+a}{1+b}\right)^n > K \right\}$$

\therefore We can take at most N down-steps after t

$$\mathbb{E}^{\mathbb{Q}}[C] = \sum_{k=0}^N \binom{T-t}{k} p^{+k} (1-p^{+})^{T-t-k} \left\{ \frac{(1+b)}{(1+a)} (1+b)^{T-t-k} (1+a)^k - K \right\}$$

$$\therefore \text{price} = (1+r)^{-T} \sum_{k=0}^N \binom{T-t}{k} p^{+k} (1-p^{+})^{T-t-k} \left\{ (1+b)^{T-t-k} (1+a)^k - K \right\}$$

$$CN = \left\lfloor \frac{\log(K (1+b)^{t-T})}{\log\left(\frac{1+a}{1+b}\right)} \right\rfloor$$

SFM 2018

1.) (a) Let $\forall n \geq 0$: M_n is F_n -measurable, $M_n \in L^1(\mathbb{P})$ and

$$\mathbb{E}[M_n | F_{n-1}] = M_{n-1}$$

(b)
$$\mathbb{E}[|M_n|] = \mathbb{E}\left[\prod_{k=1}^n |Y_k|\right] = \prod_{k=1}^n \mathbb{E}[|Y_k|] = 1 < \infty.$$

M_n is $\sigma(Y_1, \dots, Y_n)$ $\mathcal{G} = F_n$ -measurable.

$$\begin{aligned} \mathbb{E}[M_n | F_{n-1}] &= \prod_{k=1}^{n-1} Y_k \cdot \underbrace{\mathbb{E}[Y_n | F_{n-1}]}_{= \mathbb{E}[Y_n] \text{ as } Y_n \text{ indep. of } F_{n-1}} = M_{n-1} \end{aligned}$$

(c) Let $A_0 = 0$, $A_{n+1} - A_n = \mathbb{E}[X_{n+1} - X_n | F_n]$ (F_n measurable)

$$\therefore \mathbb{E}[A_{n+1}] = \sum_{k=1}^n \mathbb{E}[X_{k+1} - X_k] = \mathbb{E}[X_{n+1}] - X_0$$

$\therefore A_{n+1} - A_n$ is F_n -measurable.

(Induct): A_0 is F_0 -measurable constant, constant

~~$A_1 - A_0$~~ is $A_1 - A_0$ is F_0 measurable $\Rightarrow A_1$ is F_0 measurable

If A_n is F_{n-1} measurable: $A_{n+1} = A_{n+1} - A_n + A_n$ is F_n -measurable

Let $M_n = X_n - A_n$: $\mathbb{E}[|M_n|] \leq \mathbb{E}[|X_n|] + \mathbb{E}[|A_n|]$

$$\mathbb{E}[|X_n|] < \infty \quad (\text{by def.})$$

$$\begin{aligned} \mathbb{E}[|A_n|] &\leq \sum_{k=1}^n \mathbb{E}[\mathbb{E}[|X_k - X_{k-1}| | F_{k-1}]] \\ &= \sum_{k=1}^n \mathbb{E}[|X_k - X_{k-1}|] < \infty \end{aligned}$$

$$\therefore M_n \in L^1$$

M_n is clearly F_n -measurable

$$\mathbb{E}[M_{n+1} - M_n | F_n] = \mathbb{E}[X_{n+1} - X_n | F_n] - (A_{n+1} - A_n) = 0$$

$\therefore (M_n)_{n \geq 0}$ is a martingale.

(Unique): If $A_0 = 0$, $A_{n+1} - A_n = \mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n]$,
 $(A_n)_{n \geq 0}$ is uniquely defined.

$\therefore M_n = X_n - A_n$ is uniquely defined.

$$* \quad X_n = M_n + A_n = M_n' + A_n', \quad A_0 = A_0' = 0.$$

$$\mathbb{E}[M_n - M_n' | \mathcal{F}_{n-1}] = A_n' - A_n = Y_n, \quad (\text{This is a martingale.})$$

$$\mathbb{E}[Y_n | \mathcal{F}_{n-1}] = Y_{n-1} \quad (\text{martingale property of } M_n - M_n')$$

$$= Y_n \quad (A_n' - A_n \text{ is } \mathcal{F}_{n-1} \text{ measurable})$$

$$\therefore Y_n = Y_0 \text{ (A.S.)} = 0 \Rightarrow \text{Unique (A.S.)} \quad *$$

(Question badly phrased!!!)

(i) If $(X_n)_{n \geq 0}$ is a martingale:

$$\mathbb{E}[X_n] = \mathbb{E}[\mathbb{E}[X_n | \mathcal{F}_{n-1}]] \quad (\text{Tower})$$

$$= \mathbb{E}[X_{n-1}]$$

$$\therefore \mathbb{E}[X_0] = \mathbb{E}[X_1] = \dots = \mathbb{E}[X_n]$$

(ii) If $\mathbb{E}[X_n] = \mathbb{E}[X_0]$

$$\text{Let } X_n = M_n + A_n: \therefore \mathbb{E}[A_n] - \mathbb{E}[A_0] = 0.$$

$$\mathbb{E}[A_n] = \mathbb{E}[X_n] - \mathbb{E}[M_n] = \mathbb{E}[X_0] - \mathbb{E}[M_0] = 0.$$

But $(X_n)_{n \geq 0}$ super martingale $\Rightarrow \mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n] \leq 0$ A.S.

$$\therefore A_{n+1} - A_n \leq 0 \text{ A.S.}$$

$$\text{But } \mathbb{E}[A_{n+1} - A_n] = 0 \Rightarrow A_{n+1} = A_n \text{ (A.S.)}$$

$$\therefore A_0 = \dots = A_n \text{ A.S. } (=0) \Rightarrow X_n = M_n \Rightarrow \text{Martingale}$$

2.) (a) $S_t = S_0 e^{GB_t + Pt}$, $(B_t)_{t \geq 0}$ is a IP-brownian motion.

$\therefore X_t = S_0 e^{GB_t + (P-r)t}$ is discounted process.

(wrt G : $\frac{dG}{dIP} = e^{cB_T - c^2T/2}$) $(B_t)_{t \geq 0}$ is a c -drift B.M.
(Cameron Martin)

$\therefore X_t = S_0 e^{GW_t + Gct + (P-r)t}$

Pick c s.t. $(Gc + P - r) = -G^2/2 \Rightarrow c = (-G^2/2 + r - P)/G$

$\therefore X_t = S_0 e^{GW_t - G^2t/2}$; W_t is a G -B.M.

G is the equivalent martingale measure

(b) Price: $E^G [e^{-rT} f(S_T)]$ $\left\{ \begin{array}{l} S_T = S_0 e^{GW_T + (r-G^2/2)T} \\ W_T \sim N(0, T) \end{array} \right.$

$\therefore \pi = e^{-rT} \int_{-\infty}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} f(S_0 e^{G\sqrt{T}y + (r-G^2/2)T}) dy$

(c) $\pi_c = e^{-rT} \int_{-\infty}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} \cdot S_0^r e^{rG\sqrt{T}y} e^{r(r-G^2/2)T} dy$
 $= S_0^r e^{[r(r-G^2/2) + rG^2/2]T} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y - rG\sqrt{T})^2} e^{r^2G^2T} dy$
 $= S_0^r e^{-rT + r(r-G^2/2)T + r^2G^2T}$

(d) At time t : Replace $S_0 \rightarrow S_t$, $T \rightarrow T-t$

$\therefore S_t^r e^{-r(T-t) + r(r-G^2/2)(T-t) + r^2G^2(T-t)} = V(t)$

$G_t =$

Hedge: $\frac{\partial V}{\partial S_t} = r S_t^{r-1} e^{-r(T-t) + r(r-G^2/2)(T-t) + r^2G^2(T-t)}$ units of stock

replicates strategy.

" Hold $\sim \Theta_t$ at time t to hedge.

4.) ca)

$$\begin{aligned} \mathbb{E}[U'(X)(X-\mu)] &= \int_{\mathbb{R}} U'(x) \frac{e^{-\frac{1}{2\sigma^2}(x-\mu)^2}}{\sqrt{2\pi\sigma^2}} (x-\mu) dx \\ &= \underbrace{\left[-U'(x) \frac{e^{-\frac{1}{2\sigma^2}(x-\mu)^2}}{\sqrt{2\pi\sigma^2}} \cdot \sigma^2 \right]_{\mathbb{R}}}_{\rightarrow 0 \text{ as } U' \text{ has polynomial growth}} + \sigma^2 \int_{\mathbb{R}} U''(x) \frac{e^{-\frac{1}{2\sigma^2}(x-\mu)^2}}{\sqrt{2\pi\sigma^2}} dx \\ &= \mathbb{E}[U''(X)] \cdot \sigma^2 \end{aligned}$$

cb)

$$X = \mu + \sigma \cdot Z, \quad Z \sim N(0,1).$$

$$\frac{\partial f}{\partial \mu} = \mathbb{E} \left[\frac{\partial U(\mu + \sigma Z)}{\partial \mu} \right] = \mathbb{E}[U'(\mu + \sigma Z)] \geq 0.$$

Since $U' \neq 0$: Integral > 0 as well $\Rightarrow \frac{\partial f}{\partial \mu} > 0$.

$$\frac{\partial f}{\partial \sigma} = \mathbb{E}[U'(\mu + \sigma Z) \cdot Z] = \mathbb{E}[V''(Z)] = \sigma \mathbb{E}[U''(\mu + \sigma Z)]$$

$$\cancel{V'(Z) = U'(\mu + \sigma Z)} \\ V'(Z), \quad V(Z) = \frac{1}{\sigma} U(\mu + \sigma Z)$$

≤ 0 as U is concave.

IF μ ~~maximises~~ ^{expected} ~~mean~~ ^{of all} ~~U(X)~~ ^{must have the largest}

IF $X(\mu_0, \sigma_0)$ ~~maximises~~ $\mathbb{E}[U(X)]$: μ_0 For all

~~For fixed~~ we can
 f is fully determined by μ, σ^2 : For fixed μ_0 , maximise $\mathbb{E}[U(X)]$ by minimising σ^2 (mean-var efficient);

Also for fixed σ^2 , maximise $\mathbb{E}[U(X)]$ by maximising μ .

$$c1) \quad \frac{\partial^2 f}{\partial p^2} = E[U''(p + \theta Z)] \leq 0.$$

$$\frac{\partial^2 f}{\partial \theta^2} = \theta' E[U''(p + \theta Z) \cdot Z'] \leq 0.$$

$$\frac{\partial^2 f}{\partial \theta \partial p} = \cancel{\theta' E[U''(p + \theta Z)]} \theta' E[U''(p + \theta Z) \cdot Z]$$

$$\det(\nabla^2 f) = \theta' E[U''(p + \theta Z)] \cdot E[U''(p + \theta Z) \cdot Z'] - \theta' E[U''(p + \theta Z) \cdot Z]^2 \quad (1)$$

Cauchy let $d = E[-U''(p + \theta Z)]$.

$$\text{Nok: } -U''(p + \theta Z) \geq 0 \Rightarrow \frac{dG}{d1p} = -U''(p + \theta Z) \geq 0.$$

$$\therefore \text{det} \approx \theta' c1) / \theta' = (E^G[1] \cdot E^G[Z'] - E[Z']^2) \cdot d^2.$$

$$\therefore c1) / \theta' \cdot d^2 = E^G[Z'] - E[Z']^2 \geq 0 \quad (\text{Jensen's})$$

$$\therefore c1) \geq 0$$

semi

Hint: Negative ^{semi} definite 2nd derivative

4.) (a) $(\Theta_n)_{n=0, \dots, T}$ is an arbitrage if $\Theta_T \cdot S_T^X - \Theta_0 \cdot S_0 \geq 0$ (A.S.)
 $IP(\Theta_T \cdot S_T^X - \Theta_0 \cdot S_0 > 0) > 0$.

(b) FTAP I: There is no arbitrage opportunity iff discounted process has an equivalent martingale measure

(c) If \mathbb{Q} is a martingale measure:

$$V_n = \Theta_n \cdot S_n = \Theta_0 \cdot S_0 + \sum_{k=1}^n \Theta_k (S_k - S_{k-1}) \quad (\text{self financing})$$

$$\therefore \mathbb{E}^{\mathbb{Q}}[|V_n|] \leq |\Theta_0| \cdot |S_0| + \sum_{k=1}^n \mathbb{E}^{\mathbb{Q}}[|\Theta_k| (S_k - S_{k-1})] < \infty$$

Since Θ_n is \mathcal{F}_{n-1} measurable, $V_n = \Theta_n S_n$ is \mathcal{F}_n -measurable.

$$\mathbb{E}^{\mathbb{Q}}[V_n - V_{n-1} | \mathcal{F}_{n-1}] = \mathbb{E}^{\mathbb{Q}}[\Theta_n (S_n - S_{n-1}) | \mathcal{F}_{n-1}]$$

$$= \Theta_n \cdot \mathbb{E}^{\mathbb{Q}}[S_n - S_{n-1} | \mathcal{F}_{n-1}] = 0$$

$\therefore (V_n)_{n \geq 0}$ is a \mathbb{Q} -martingale

If $(V_n)_{n \geq 0}$ is a \mathbb{Q} -martingale: $\mathbb{E}[V_T] = \mathbb{E}[\mathbb{E}[V_T | \mathcal{F}_{T-1}]] = \mathbb{E}[V_{T-1}]$
 $\therefore \mathbb{E}[V_0] = \dots = \mathbb{E}[V_T]$

Claim: (iii) \Rightarrow (i)

WTP: $\mathbb{E}[X_n | \mathcal{F}_{n-1}] = X_{n-1}$

Since X_{n-1} is \mathcal{F}_{n-1} measurable, suffice to show:

$$\forall A \in \mathcal{F}_{n-1}: \mathbb{E}[X_n \cdot 1_A] = \mathbb{E}[X_{n-1} \cdot 1_A]$$

Consider the following strategy: (for \mathbb{S}_k asset k , $k \in \{1, \dots, d\}$)
 (k is fixed):

Stochastic Financial Models 2019

1.) (a) $(M_n, F_n)_{n \geq 0}$ is a martingale if:

$\forall n \geq 0: M_n \in L^1(\mathbb{P})$, M_n is F_n measurable

$$\mathbb{E}[M_{n+1} | F_n] = M_n$$

(b) $|X_n| \leq \max\{1, n \cdot |X_{n-1}|\}$

$$\therefore \mathbb{E}[|X_n|] \leq 1 + n \mathbb{E}[|X_{n-1}|] \quad (*)$$

Since $X_0 = 0 \Rightarrow X_0 \in L^1$: If $X_{k-1} \in L^1, \Rightarrow X_k \in L^1$ (from *)

\therefore (Induct): $X_n \in L^1 (\forall n \geq 0)$

X_n is a function of X_1 is $G(X_1, \dots, X_n)$ measurable.

$$\mathbb{E}[X_n | F_{n-1}] = \mathbb{E}[X_n | X_0, \dots, X_{n-1}] = \mathbb{E}[X_n | X_{n-1}]$$

$$\left. \begin{array}{l} \text{If } X_{n-1} = 0: \mathbb{E}[X_n | F_{n-1}] = \frac{1}{2^n} (1 + (-1)) = 0 = X_{n-1} \\ \text{If } X_{n-1} \neq 0: \mathbb{E}[X_n | F_{n-1}] = \frac{1}{n} (n \cdot X_{n-1}) + 0 = X_{n-1} \end{array} \right\} \mathbb{E}[X_n | F_{n-1}] = X_{n-1}$$

\therefore Valid martingale

(c) (i) \Rightarrow (ii)

$$M_n^T = \sum_{k=0}^{n-1} 1_{T=k} \cdot M_k + 1_{T \geq n} M_n \quad \left. \begin{array}{l} 1_{T=k} \text{ is } F_{n_k} \text{ measurable,} \\ 1_{T \geq n} \text{ is } F_{n-1} \text{ measurable.} \end{array} \right\}$$

\therefore If M_n^T is a sum/product of F_n -measurable R.V. $\Rightarrow M_n^T$ is F_n -measurable.

$$\mathbb{E}[|M_n^T|] \leq \sum_{k=0}^{n-1} \mathbb{E}[|M_k|] + \mathbb{E}[|M_n|] < \infty \quad (M_n \in L^1)$$

$$\mathbb{E}[M_n^T | F_{n-1}] = \underbrace{\sum_{k=0}^{n-1} 1_{T=k} M_k}_{F_{n-1} \text{-measurable}} + \underbrace{1_{T \geq n} \mathbb{E}[M_n | F_{n-1}]}_{\text{measurable}}$$

$$= \sum_{k=0}^{n-1} 1_{T \leq k} M_k + 1_{T > n-1} M_{n-1} = \sum_{k=0}^{n-2} 1_{T \leq k} M_k + 1_{T \geq n-1} M_{n-1} = M_{n-1}^T$$

(4/4) \therefore Martingale.

$$(ii) \Rightarrow (iii) : \mathbb{E}[M_n^2] = \mathbb{E}[M_n^2 | \mathcal{F}_{n-1}] \quad (\text{Tower}) = \mathbb{E}[M_{n-1}^2]$$

(2/2) \therefore Iterating: $\mathbb{E}[M_n^2] = \mathbb{E}[M_0^2] = \mathbb{E}[M_0]$

(iii) \Rightarrow (i) Fix $A \in \mathcal{F}_{n+1}$; $T = n 1_{\omega \in A} + (n+1) 1_{A^c}$

(Stopping time): If $k < n$, $\{T \leq k\} = \emptyset \in \mathcal{F}_k$

$$\{T \leq n\} = A \in \mathcal{F}_n$$

$$\{T \leq k\} = \Omega \in \mathcal{F}_k \quad \text{for } k \geq n+1$$

$$\therefore \mathbb{E}[M_{n+1}^2] = \mathbb{E}[M_n^2 1_A + M_{n+1}^2 1_{A^c}] \quad \left| \begin{array}{l} \mathbb{E}[(M_{n+1} - M_n)^2 1_{A^c}] = \\ \mathbb{E}[M_{n+1}^2 - M_n^2] = 0. \end{array} \right.$$

$$\mathbb{E}[M_{n+1}^2] = \mathbb{E}[M_n^2]$$

Since this is true $\forall A \in \mathcal{F}_n$: $\forall B \in \mathcal{F}_n$, $\mathbb{E}[M_{n+1} 1_B] = \mathbb{E}[M_n 1_B]$.

$\mathbb{E}[M_{n+1} | \mathcal{F}_n] = M_n$ is \mathcal{F}_n measurable.

\therefore (Uniqueness of conditional expectation): $\mathbb{E}[M_{n+1} | \mathcal{F}_n] = M_n$

Since $\mathbb{E}[M_n^2] = \sum_{k=0}^n \mathbb{E}[M_k^2] < \infty$, (M_n)

M_n integrable is given $\Rightarrow (M_n)_{n \geq 0}$ is a martingale wrt $(\mathcal{F}_n)_{n \geq 0}$

2.) (a) Let $(X_n)_{n \geq 0}$ be the discounted price process:

Q is an equivalent martingale measure iff: \mathbb{P}, Q have the same null sets and

(3/3) $(X_n)_{n \geq 0}$ is a Q -martingale.

(b) FTAP: \exists equivalent martingale measure iff there is no arbitrage opportunity.

(2/2)

(c) We want \tilde{G}, \tilde{m} s.t.:

$$\mathbb{E} \left[e^{\tilde{G}Z + \tilde{m}} \cdot e^{GZ + m} \right] = \mathbb{E} \left[e^{\tilde{G}Z + \tilde{m}} \right]$$

$$\text{LHS} = e^{(m + \tilde{m}) + \frac{1}{2}(G + \tilde{G})^2}, \quad \text{RHS} = e^{\tilde{m} + \frac{1}{2}\tilde{G}^2}$$

$$\text{Prek} = m + \frac{1}{2}(2G \cdot \tilde{G}) + \frac{1}{2}G^2 = 0$$

$$\therefore G \cdot \tilde{G} = -m - G^2/2 \Rightarrow \tilde{G} = -m/G - G/2$$

$$\text{Let } \tilde{m} = 0, \quad \tilde{G} = -m/G - G/2$$

$$dG/d\mathbb{P} = \frac{e^{\tilde{G}Z}}{\mathbb{E}[e^{\tilde{G}Z}]} \text{ is an equivalent measure}$$

$$\mathbb{E}^Q[S_1] = 1 = S_0 \Rightarrow \text{Equivalent martingale measure}$$

(9/9)

\therefore FTAP \Rightarrow No arbitrage.

(d) Let $\tilde{G}_k = -m_k/G_k - G_k/2$

$$dG/d\mathbb{P} = \prod_{k=1}^T \frac{e^{\tilde{G}_k Z_k}}{\mathbb{E}[e^{\tilde{G}_k Z_k}]} : Q, \mathbb{P} \text{ are equivalent}$$

say it is an measure as Z_k indep. ...

$$\mathbb{E}^G [S_{t+1}' | F_t] = S_t' \cdot \mathbb{E}^G [e^{G_{t+1} Z_{t+1} + M_{t+1}} | F_t] \quad (1)$$

~~Since changing measure preserves measurable sets, Z_{t+1} remains independent of F_t . (*) Z_{t+1} independent of F_t .~~

$$\therefore (1) = S_t' \cdot \mathbb{E}^G [e^{G_{t+1} Z_{t+1} + M_{t+1}}]$$

$$= S_t' \cdot \mathbb{E} \left[\prod_{k=1}^T e^{\tilde{G}_k Z_k} / \mathbb{E} [e^{\tilde{G}_k Z_k}] \cdot e^{\tilde{G}_{t+1} Z_{t+1} + G_{t+1} Z_{t+1} + M_{t+1}} / \mathbb{E} [e^{\tilde{G}_{t+1} Z_{t+1}}] \right]$$

$$= S_t' \quad (\text{independence of } Z_k)$$

(*) : Z_{t+1} & $(Z_1)^T_{n \geq 1}$ is G -independent.

$$\begin{aligned} \mathbb{P}^G(Z_1 \in A_1, \dots, Z_T \in A_T) &= \mathbb{E} \left[\prod_{k=1}^T \frac{1_{Z_k \in A_k} \cdot e^{\tilde{G}_k Z_k}}{\mathbb{E} [e^{\tilde{G}_k Z_k}]} \right] \\ &= \prod_{k=1}^T \left(\mathbb{E} \left[\frac{1_{Z_k \in A_k}}{\mathbb{E} [e^{\tilde{G}_k Z_k}]} \cdot e^{\tilde{G}_k Z_k} \right] \cdot \prod_{j \neq k} \mathbb{E} [e^{\tilde{G}_j Z_j}] \right) \\ &= \prod_{k=1}^T \mathbb{E}^G [1_{Z_k \in A_k}] = \prod_{k=1}^T \mathbb{P}^G(Z_k \in A_k) \end{aligned}$$

\therefore Independent

Why is S_t' integrable?

$\therefore Q$ is an equivalent martingale measure on (discounted) price process

FTAP \Rightarrow No arbitrage.

(516)

(192)

$$\begin{aligned}
 3.) \quad (a) \quad \pi(EC) &= e^{-rT} \int_{IR} e^{-y^2/2} \left[S_0 e^{(r-\sigma^2/2)T} (e^{6\sqrt{T}y} - K/S_0 e^{-(r-\sigma^2/2)T}) \right] dy \\
 &= S_0 e^{-\sigma^2 T/2} \int_{IR} \frac{1}{\sqrt{2\pi}} (e^{-y^2/2} e^{6\sqrt{T}y} - e^{-y^2/2} \frac{K e^{-(r-\sigma^2/2)T}}{S_0}) dy \\
 &= S_0 \cdot \left\{ \underbrace{\Phi(6\sqrt{T} - \frac{1}{6\sqrt{T}}(\log(K/S_0) - (r-\sigma^2/2)T))}_{d^+} - K e^{-rT} \underbrace{\Phi\left(\frac{(r-\sigma^2/2)T - \log(K/S_0)}{6\sqrt{T}}\right)}_{d^-} \right\}
 \end{aligned}$$

Payout: $(S_T - K)^+ - (K - S_T)^+ = S_T - K$

We need 1 stock, $-K e^{-rT}$ to replicate RHS

$$\therefore \pi(EC) - \pi(EP) = S_0 - K e^{-rT} \quad (\text{put-call parity})$$

$$\therefore \pi(EP) = K e^{-rT} \Phi(-d^-) - S_0 \Phi(-d^+)$$

(10/10)

$$\begin{aligned}
 (b) \quad \text{At } t=0: \quad \pi &= e^{-rT} \int_{IR} 1_{S_0 e^{6\sqrt{T}y + (r-\sigma^2/2)T} \geq K} \frac{e^{-y^2/2}}{\sqrt{2\pi}} dy \\
 &= e^{-rT} \int_{\frac{1}{6\sqrt{T}} [\log(K/S_0) - (r-\sigma^2/2)T]}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} dy \\
 &= e^{-rT} (\Phi(d^-))
 \end{aligned}$$

$$\text{At time } t: \quad \left. \begin{array}{l} S_0 \rightarrow S_t \\ T \rightarrow T-t \end{array} \right\} \pi = e^{-r(T-t)} \Phi\left(\frac{-\log(K/S_t) + (r-\sigma^2/2)(T-t)}{\sqrt{T-t} \cdot \sigma}\right)$$

$$\text{Hedge at } t: \quad \frac{\partial \pi}{\partial S_t} = e^{-r(T-t)} \Phi\left(\frac{-\log(K/S_t) + (r-\sigma^2/2)(T-t)}{\sigma \cdot \sqrt{T-t}}\right) \cdot \left(\frac{1}{S_t \cdot \sigma \sqrt{T-t}}\right)$$

replicate delta of option

\therefore Hold $-\partial \pi / \partial S_t$ units of stock to hedge.

(7/7)

(d) Payouts: $1_{S_T \geq K} + 1_{S_T < K} = 1$ ✓

1 can be replicated with e^{-rt}

$\therefore \Pi(C_{\text{dig Coll}}) + \Pi(C_{\text{dig Put}}) = e^{-rt}$ ✓ (put-call parity)

$\therefore \Pi(C_{\text{dig Put}}) = e^{-rt} \Phi(-d^-)$ ✓

(3/3)

(20/2)

4.) (a) Let $S_t^0 = (1+r)^t$, fix $-1 < a < b$:

$\{R_n\}_{n \geq 1}$ be iid; $IP(R_n = a) = p$, $IP(R_n = b) = 1-p = q$. ✓

$S_t^1 = S_0^1 \cdot \prod_{k=1}^t (1+R_k)$ (CRR model) ✓

(4/4)

Arbitrage free iff $a < r < b$:

$X_t^1 = S_0^1 \cdot \prod_{k=1}^t \left(\frac{1+R_k}{1+r} \right)$ ✓

$\therefore E^G[X_{t+1}^1 | \mathcal{F}_t^1] = X_t^1 \cdot E^G \left[\frac{1+R_{t+1}}{1+r} | \mathcal{F}_t^1 \right] = X_t^1$ ✓

iff: $IP^*(P^* = IP^G(G(R_{t+1} = a)), q^* = G(R_{t+1} = b | \mathcal{F}_t^1)$

$r = p^* a + (1-p^*) b$ ✓

iff $p^* = \frac{b-r}{b-a}$ (possible iff $b > r > a$) ✓

$\therefore \exists$ equivalent martingale measure iff $a < r < b$

$G\{a, R_1 = a_1, \dots, R_n = a_n\} = \left(\frac{b-r}{b-a} \right)^{\sum_{i=1}^n 1_{a_i=a}} \cdot \left(\frac{r-a}{b-a} \right)^{\sum_{i=1}^n 1_{a_i=b}}$ ✓

(3/3)

(b) Let $(\Theta_n)_{n=0, \dots, T}$ be replication strategy:

$$H = \Theta_T \cdot X_T \quad (\text{A.S.})$$

$$\text{At time } t: \Theta_t \cdot X_t = \mathbb{E}^G [\Theta_T \cdot X_T | \mathcal{F}_t] = \mathbb{E}^G [H | \mathcal{F}_t]$$

$$\therefore V_t(X_0, \dots, X_t) = \mathbb{E}^G [H | S_0 = x_0, \dots, S_t = x_t]$$

$$= \mathbb{E}^G \left[h(X_0, \dots, X_t; X_t \cdot \frac{(1+R_{t+1})}{(1+r)}, \dots, X_t \prod_{k=t+1}^T \frac{(1+R_k)}{(1+r)}) \right]$$

But $\left(\frac{(1+R_k)}{(1+r)} \right)_{k=t+1, \dots, T}$ have the same distribution as

$$\left(\frac{(1+R_k)}{(1+r)} \right)_{k=1, \dots, T-t}$$

$$\therefore V_t(X_0, \dots, X_t) = \mathbb{E}^G \left[h(X_0, \dots, X_t; X_t \left(\frac{(1+R_1)}{(1+r)} \right), \dots, X_t \prod_{k=1}^{T-t} \frac{(1+R_k)}{(1+r)}) \right]$$

$$= \mathbb{E}^G \left[h(X_0, \dots, X_t; X_t \cdot (S'_1/S_0), \dots, X_t (S'_{T-t}/S_0)) \right]$$

(5/5) $(S'_{T-k} = S'_0 \cdot \prod_{i=1}^k \frac{(1+R_i)}{(1+r)})$

(c) $V_t(X_0, \dots, X_t) = \mathbb{E}_G^G \left[h(X_t \cdot S'_{T-t}/S'_0) \right] \quad (\text{from (b)})$

(2/3)

$$\therefore V_t \text{ depend only on } X_t: V_t(X_t) = \mathbb{E}^G [h(X_t \cdot S'_{T-t}/S'_0)]$$

(d) $V_t(X_0, \dots, X_t) = \mathbb{E}^G \left[h(X_t (S'_{T-t}/S'_0), \max \{X_0, \dots, X_t, X_t (S'_1/S'_0), \dots, X_t (S'_{T-t}/S'_0)\}) \right]$

$$= \mathbb{E}^G \left[h(X_t (S'_{T-t}/S'_0), \max \{ \max \{X_0, \dots, X_t\}, \max \{X_t/S'_0 \cdot \max \{S'_1, \dots, S'_{T-t}\}\} \}) \right]$$

$$= \mathbb{E}^G \left[h(X_t (S'_{T-t}/S'_0), \max \{ \max \{X_0, \dots, X_t\}, X_t/S'_0 \cdot \max \{S'_0, \dots, S'_{T-t}\} \}) \right]$$

$$(X_t/S'_0 \cdot S'_0 = X_t)$$

$$X_T S_{T-t}^1$$

\therefore Function of $X = \{X_t\}$, $m = \max\{X_0, \dots, X_t\}$, S_0^1 , M_{T-t} .

$$g(x, m, S_0^1, S_{T-t}^1, M_{T-t}) = h\left(x \cdot S_{T-t}^1 / S_0^1, \max\{m, x / S_0^1 - M_{T-t}\}\right)$$

(5/5)

1/19 J

SFM 2020

1.) (a) $\bar{\theta} \cdot \bar{S}_1 = \bar{\theta} \cdot N(P, V) + \omega (1+r) (\omega_0 - \bar{\theta} \cdot S_0)$

$$= P(1+r) \omega_0 + \bar{\theta} (P - (1+r) S_0), \quad \text{where } \bar{\theta}^T V \bar{\theta}$$

$$\therefore \text{Min. } \frac{1}{2} \bar{\theta}^T V \bar{\theta} \quad \text{subject to} \quad (1+r) \omega_0 + \bar{\theta}^T d = \omega_1, \\ d = P - (1+r) S_0.$$

$$\text{Lagrangian: } \frac{1}{2} \bar{\theta}^T V \bar{\theta} - \lambda ((1+r) \omega_0 + \bar{\theta}^T d - \omega_1)$$

$$V \cdot \bar{\theta} = \lambda \cdot d \Rightarrow \bar{\theta} = \lambda \cdot (V^{-1} \cdot d)$$

$$\text{Constraint: } \omega_1 - (1+r) \omega_0 = \bar{\theta}^T d = \lambda (d^T V^{-1} d)$$

$$\therefore \lambda = \frac{\omega_1 - (1+r) \omega_0}{d^T V^{-1} d}$$

$$\therefore \text{Portfolio: } \bar{\theta} = \lambda \cdot V^{-1} (P - (1+r) S_0), \quad \lambda = \frac{\omega_1 - (1+r) \omega_0}{(P - (1+r) S_0)^T V^{-1} (P - (1+r) S_0)}.$$

(12/12) (b) $E[e^{S \cdot N(P, G)}] = e^{SP + S^T G / 2} \quad (\text{Gaussian MGF})$

$$\therefore E[u(\bar{\theta} \cdot \bar{S}_1)] = -E[e^{-r N((1+r) \omega_0 + \bar{\theta} \cdot d, \bar{\theta}^T V \bar{\theta})}] \\ = -e^{-r [(1+r) \omega_0 + \bar{\theta} \cdot d] + r/2 \bar{\theta}^T V \bar{\theta}}$$

$$\text{is maximised by maximising } ((1+r) \omega_0 + \bar{\theta} \cdot d - r/2 \bar{\theta}^T V \bar{\theta}) \quad (1)$$

$$(1) \text{ is a quadratic in } \bar{\theta} \Rightarrow \text{Unique global maximiser at} \\ \nabla (1) = 0.$$

$$\therefore \text{characteristic } d = r V \cdot \bar{\theta}$$

(8/8) (202) Portfolio = $\frac{1}{r} V^{-1} d$; Pick $r = \frac{d^T V^{-1} d}{\omega_1 - (1+r) \omega_0}$ for the same portfolio as in (a)

2.) ca) Let $X_n = S_n / S_n^0$:

$(\Theta_n)_{n=1, \dots, T}$ is an arbitrage if:

(9/2) $\Theta_T \cdot X_T - \Theta_0 X_0 \geq 0$ (A-S), $IP(\Theta_T \cdot X_T - \Theta_0 X_0 > 0) > 0$.

(1/2) cit) Market is complete if all claims can be replicated in the market *explain what this means?*

ciii) Let G be any equivalent measure:

~~Since~~ Since F is not changed, conditional expectation

$$\mathbb{E}^G \left[\frac{S_{n+1}}{S_{n+1}^0} \mid F_n \right] = \frac{S_n}{(Hr)^{n+1}} \cdot \mathbb{E}^G [e^{\log(Z_{n+1})} \mid F_n]$$

Pick: $\frac{dG}{dIP} = \prod_{k=1}^n \frac{e^{\tilde{G}_k \cdot N_k + \tilde{\mu}}}{\mathbb{E}[e^{\tilde{G}_k \cdot N_k}]}$, $\mathbb{E}(\log(Z_k)) = \mu + G N_k$

Claim: $(Z_n)_{n \geq 1}$ remains independent.

$$G(Z_1 \in A_1, \dots, Z_T \in A_T) = \mathbb{E} \left[\prod_{k=1}^T \frac{1_{Z_k \in A_k} \cdot e^{\tilde{G}_k N_k + \tilde{\mu}}}{\mathbb{E}[e^{\tilde{G}_k N_k}]} \right]$$

$$= \prod_{k=1}^T \mathbb{E} \left[1_{Z_k \in A_k} \cdot \frac{e^{\tilde{G}_k N_k + \tilde{\mu}}}{\mathbb{E}[e^{\tilde{G}_k N_k}]} \right] = \prod_{k=1}^T \left\{ \mathbb{E} \left[1_{Z_k \in A_k} \frac{e^{\tilde{G}_k N_k + \tilde{\mu}}}{\mathbb{E}[e^{\tilde{G}_k N_k}]} \right] \right\}$$

$$= \prod_{k=1}^T \mathbb{E} \left[1_{Z_k \in A_k} \frac{dG}{dIP} \right] = \prod_{k=1}^T \mathbb{E}^G [1_{Z_k \in A_k}]$$

$$\mathbb{E}^G \left[\frac{S_{n+1}}{S_n} \middle| \mathcal{F}_n \right] = \frac{S_n}{S_n} \cdot \frac{1}{\text{chr}} \cdot \mathbb{E}^G \left[e^{\frac{\log(Z_{n+1})}{S_{n+1}}} \right]$$

We want: $\mathbb{E}^G \left[e^{\frac{\log(Z_{n+1})}{S_{n+1}}} \right] = \text{chr}$

$$\begin{aligned} \text{LHS} &= \mathbb{E}^G \left[e^{\mu + G N_{n+1}} \cdot e^{\tilde{G}_{n+1} N_{n+1}} \right] \\ &= e^{\mu + \frac{1}{2}(G + \tilde{G}_{n+1})^2} \cdot \mathbb{E} \left[e^{\tilde{G}_{n+1} N_{n+1}} \right] \\ &= e^{\mu + \frac{1}{2} \tilde{G}_{n+1}^2} = \text{chr} \end{aligned}$$

$$\therefore \log(\text{chr}) = \mu + G' \frac{1}{2} + G \cdot \tilde{G}_{n+1}$$

\therefore Pick

Is $Z_k = e^{\tilde{G}_k N_k + \mu}$?

$$\text{Let } dG/dIP = \prod_{k=1}^T e^{\tilde{G}_k N_k + \mu} / \mathbb{E} \left[e^{\tilde{G}_k N_k + \mu} \right]$$

Since Z_1, \dots, Z_T is independent, we can write its measure

$$\text{as } \mu_1 \otimes \dots \otimes \mu_T = IP.$$

$$\therefore G = \left(\bigotimes_{k=1}^T \mu_k \left(e^{\tilde{G}_k N_k + \mu} / \mathbb{E} \left[e^{\tilde{G}_k N_k + \mu} \right] \right) \right) \Rightarrow (Z_1, \dots, Z_T)$$

remains independent

Martingale: $\mathbb{E}^G \left[e^{\mu + G N_{n+1}} \right] = \text{chr}$

$$\therefore \text{chr} \cdot \mathbb{E}^G \left[e^{\tilde{\mu} + \tilde{G} N_{n+1}} \right] = \mathbb{E} \left[e^{\mu + \tilde{\mu} + (G + \tilde{G}) N_{n+1}} \right]$$

$$\therefore \text{chr} \cdot e^{\tilde{\mu} + \tilde{G}^2/2} = e^{(\mu + \tilde{\mu}) + \frac{1}{2}(G + \tilde{G})^2}$$

$$\therefore \log(\text{chr}) = \tilde{\mu} + \mu + G \cdot \tilde{G} + G'^2/2$$

$$\therefore \tilde{G} = \frac{1}{G} \left[\log(\text{chr}) - \mu - G'^2/2 \right]$$

$$\therefore \mathbb{E}^G \left[\frac{S_{n+1}}{S_n} \mid \mathcal{F}_n \right] = \frac{S_n}{S_n^0}$$

$\frac{S_{n+1}}{S_n}$ is clearly \mathcal{F}_n measurable;

$$\mathbb{E}^G \left[\left| \frac{S_{n+1}}{S_n} \right| \right] = (1+r)^{-n} \mathbb{E}^G [S_n] = \text{const} < \infty.$$

(mgf of a gaussian $< \infty$)

(8/8) \therefore Martingale.

(iv) FTAP: No arbitrage iff \exists equivalent martingale measure.

(2/2) \therefore No arbitrage.

(v) Let $T=1$: If market is complete, all claims are replicable.

Suppose:

(θ_1^0, θ_1) be replication of S_1^2 .

$$\therefore \theta_1^0 + \theta_1 \cdot S_1 = S_1^2 \quad (\text{A.S.})$$

$$\theta_1^0 + S_0 (\theta_1 Z_1) = S_0^2 (\theta_1 Z_1)^2 \quad (\text{A.S.})$$

$$\begin{aligned} \therefore \theta_1 &= \frac{S_0 Z_1 - \theta_1^0 / S_0 Z_1}{S_0 Z_1 - \theta_1^0 / S_0 Z_1} \\ &= \frac{S_0 Z_1 e^{\mu + \sigma Z_1} - \theta_1^0 / S_0 Z_1 e^{-\mu - \sigma Z_1}}{S_0 Z_1 e^{\mu + \sigma Z_1} - \theta_1^0 / S_0 Z_1 e^{-\mu - \sigma Z_1}} \end{aligned}$$

Since $S_0 Z_1 \in \mathbb{C} \setminus \{0\}$: Let $d = S_0 Z_1$.

$$d^2 - \theta_1 d + \theta_1^0 = 0 \quad (\theta_1, \theta_1^0 \text{ are deterministic as } \mathcal{F}_0\text{-measurable})$$

$$\therefore d \in \frac{\theta_1 \pm \sqrt{\theta_1^2 - 4\theta_1^0}}{2} \quad \text{with } \mathbb{P} = 1$$

\therefore Contradiction as $\mathbb{P}(\log(N(\mu, \sigma^2)) \in A) = 0$ if $|A| > 0$.

\therefore Market not complete.

- 3.) (i) $B_0 = 0$ A.S. ✓
 $t \rightarrow B_t$ is contr. ✓ (A.S.)

(2/2) $(B_t)_{t \geq 0}$ has independent increments, $t \geq s \Rightarrow B_t - B_s \sim N(0, t-s)$ ✓

(ii) Let $W_t = \max_{0 \leq s \leq t} \{B_s\}$. ✓

(4/4) If $a, b \geq 0$: $IP(W_t \geq a, B_t \leq a-b) = IP(B_t \geq a+b)$
 (reflection principle)

better say $\tilde{B}_t = \begin{cases} B_t, & t \leq T_a \\ 2a - B_t, & t > T_a \end{cases}$ is BM.

(b) (i) $IP(M_t \leq a, B_t \leq b) = IP(B_t \leq b) - IP(M_t > a, B_t \leq b)$ ✓
 $= IP(B_t \leq b) - IP(M_t \geq a, B_t \leq a - (a-b))$
 $= IP(B_t \leq b) - IP(B_t \geq 2a-b)$ ✓

($0 \leq a, b \leq a$).

$\therefore f_{M, B}(a, b) = \frac{\partial^2 IP(M_t \leq a, B_t \leq b)}{\partial a \partial b} = -2 \int_{N(0, t)} (2a-b)$ ✓

$= (-2) \cdot \left(-\frac{1}{\sqrt{2\pi t}} \cdot 2(2a-b) \right) \cdot \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2t}(2a-b)^2}$

$= \frac{2(2a-b)}{t} \cdot \frac{1}{\sqrt{2\pi t}} e^{-\frac{1}{2t}(2a-b)^2}$ ✓

(ii) $IP(M_t \leq y, Z_t \leq z)$ $M_t, Z_t \in [0, \infty)$

$\begin{pmatrix} M_t \\ Z_t \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} M_t \\ B_t \end{pmatrix}$

$\therefore \int_{M_t, Z_t} (x, y-z) = \int_{M_t, B_t} (x, y) \cdot \begin{vmatrix} 1 & 0 \\ 1 & -1 \end{vmatrix} = 1$ ✓

$\therefore IP(M_t \in dy, Z_t \in dz) = \int_{M_t, B_t} (x, y, y-z) dy dz$ ✓

$= \frac{2}{\sqrt{2\pi t}} \cdot \frac{2y-(y-z)}{t} e^{-\frac{1}{2t}(2y-(y-z))^2} = \frac{2(y+z)}{\sqrt{2\pi t} \cdot t} e^{-\frac{1}{2t}(y+z)^2} dy dz$ ✓

(iii)

~~Claim $(M_1)_{t \geq 0}$~~

$$\begin{aligned}
 \mathbb{P}(Z_1 \leq x) &= \int_0^\infty \int_0^x \frac{2}{\sqrt{2\pi}} \frac{(y+z)}{1} e^{-(y+z)^2/2} dy dz \\
 &= \int_0^\infty \frac{2}{\sqrt{2\pi}} \left[e^{-(y+z)^2/2} \right]_0^x dz \\
 &= \int_0^\infty \frac{2}{\sqrt{2\pi}} \left[e^{-z^2/2} - e^{-(x+z)^2/2} \right] dz \\
 &= 1 - \int_x^\infty \frac{2}{\sqrt{2\pi}} e^{-(x+z)^2/2} dz
 \end{aligned}$$

$$\mathbb{P}(M_1 \geq 2, M_2 \leq 1) = 0 \quad \text{as } M_t \text{ is increasing.}$$

$$\text{But } \mathbb{P}(|B_1| \geq 2, |B_2| \leq 1) \geq \mathbb{P}(B_1 \in [2, 3], B_2 - B_1 \in [-3, -1]) > 0.$$

$$\mathbb{P}(M_1 - B_1 \geq 2, M_2 - B_2 \geq 1) \geq \mathbb{P}(M_1 \geq 2, B_1 \leq 0, B_2 \in [2, 3])$$

$$> 0.$$

\therefore ~~$(M_t)_{t \geq 0}$ cannot have~~ $(M_t)_{t \geq 0}$ cannot have the same distribution as $(B_t)_{t \geq 0}$, $(Z_t)_{t \geq 0}$.

(4/4)

$\therefore (B_t)_{t \geq 0}, (Z_t)_{t \geq 0}$ have the same dist.

(20/2)

* How to actually show this? *

4.) (a) Price $S_t = S_0 \cdot e^{G W_t + (r - G^2/2)t}$ is the price of a risky asset.

(1/1) risk free asset: $S_{rf} = e^{-rt}$ B_t is?

(ii) $\Pi(C) = e^{-rT} \cdot \mathbb{E}^G [C(S_0 e^{G W_t + (r - G^2/2)t} : 0 \leq t \leq T)]$,

where $(W_t)_{t \geq 0}$ is a G - brownian motion

(2/2)

* C is a function of $(S_t)_{0 \leq t \leq T}$.

(iii) Payout at time T: $(S_T - k)^+ \geq S_T - k$

To replicate $S_T - k$: We hold 1 unit of a risky asset, $-e^{-rT} \cdot k$ riskless asset at time 0.

$$\therefore \Pi(S_T - k \text{ payout}) = S_0 - e^{-rT} \cdot k$$

$$\text{Since } (S_T - k)^+ \geq S_T - k: \Pi((S_T - k)^+ \text{ payout}) \geq S_0 - e^{-rT} \cdot k$$

$$\therefore \Pi(C) \geq S_0 - e^{-rT} \cdot k$$

(2/2)

(iv) $e^{-rT} \mathbb{E}^G [(S_0 e^{G W_t + (r - G^2/2)t} - k)^+] =$

$$e^{-rT} \int_{\mathbb{R}} \frac{e^{-y^2/2}}{\sqrt{2\pi}} \cdot S_0 e^{G(r - G^2/2)t} \cdot (e^{G\sqrt{T} \cdot y} - \tilde{k}) dy, \quad \tilde{k} = k/S_0 e^{-(r - G^2/2)T}$$

$$= S_0 e^{-G^2 T/2} \int_{d^-}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} (e^{G\sqrt{T} \cdot y} - \tilde{k}) dy, \quad d^- = \log(\tilde{k})/G\sqrt{T}$$

$$= S_0 \cdot \int_{d^-}^{\infty} \frac{e^{-(y - G\sqrt{T})^2/2}}{\sqrt{2\pi}} dy - S_0 \tilde{k} e^{-G^2 T/2} \int_{d^-}^{\infty} \frac{e^{-y^2/2}}{\sqrt{2\pi}} dy \quad d^+ =$$

$$= S_0 \cdot \Phi(d^+) - e^{-rT} \cdot K \Phi(-d^-) \quad (\text{export } K)$$

$$= S_0 \cdot \Phi\left(\frac{(r + \sigma^2/2) \cdot T - \log(K/S_0)}{\sigma \sqrt{T}}\right) - K e^{-rT} \cdot \Phi\left(\frac{(r - \sigma^2/2) \cdot T - \log(K/S_0)}{\sigma \sqrt{T}}\right)$$

$\underbrace{\hspace{10em}}_{d^+} \qquad \qquad \qquad \underbrace{\hspace{10em}}_{d^-}$

(10/10)

relabel d^- here.vi) At time t :If $K > S_t$, we do not activate the contract.~~Value of activating: $(S_t - K)^+$~~ If $K \leq S_t$: If activate, payout = $S_t - K$

But value of keeping option to expiry

$$= EC(S_t, K, \sigma, r, T-t) \geq S_t - e^{-r(T-t)} \cdot K \quad (\text{iii})$$

$$\geq S_t - K \quad (r \geq 0)$$

$$= \text{Value of current activation}$$

 \therefore Do not activate. \therefore Do not activate at time t in all scenarios(5/5)
(20/20)

SFM 2021

- 1.) $\forall n \geq 0$: X_n is integrable, F_n -measurable, $\mathbb{E}[X_{n+1} | F_n] = X_n$
 $\Rightarrow (X_n)_{n \geq 0}$ is a martingale.

(2/2)

- cb) Distribution of ε_n : $\mathbb{E}[X_n - X_{n-1} | F_{n-1}] = 0 \Rightarrow \mathbb{E}[X_n - X_{n-1}] = 0$
 (Tower)

$$\therefore \mathbb{P}(\varepsilon_n = 1) - \mathbb{P}(\varepsilon_n = -1) = 0 \Rightarrow \frac{1}{2} = \mathbb{P}(\varepsilon_n = 1) = \mathbb{P}(\varepsilon_n = -1)$$

Claim: $(\varepsilon_n)_{n \geq 1}$ is independent.

~~$$\mathbb{P}(\varepsilon_1 = 1, \dots, \varepsilon_n = 1) = \mathbb{P}(\varepsilon_1)$$~~

Since $\varepsilon_n = X_n - X_{n-1}$: ε_n is F_n measurable.

$$\therefore \mathbb{E}[\varepsilon_n | F_{n-1}] = \mathbb{P}(\varepsilon_n = 1 | F_{n-1}) - \mathbb{P}(\varepsilon_n = -1 | F_{n-1}) = 0.$$

$$\therefore \frac{1}{2} = \mathbb{P}(\varepsilon_n = \pm 1 | F_{n-1})$$

$$\text{Since } \varepsilon_1, \dots, \varepsilon_{n-1} \text{ are } F_{n-1} \text{ measurable: } \mathbb{P}(\varepsilon_n = \pm 1 | \varepsilon_1, \dots, \varepsilon_{n-1}) = \mathbb{P}(\varepsilon_n = \pm 1)$$

$$\therefore \mathbb{P}(\varepsilon_n = a_n, \dots, \varepsilon_1 = a_1) = \mathbb{P}(\varepsilon_n) \cdot \mathbb{P}(\varepsilon_1, \dots, \varepsilon_{n-1})$$

$$\frac{1}{2} \cdot \mathbb{P}(\varepsilon_1 = a_1, \dots, \varepsilon_{n-1} = a_{n-1})$$

$$\text{Proceeding iteratively: } \mathbb{P}(\varepsilon_1 = a_1, \dots, \varepsilon_n = a_n) = \left(\frac{1}{2}\right)^n$$

\therefore True $\forall n \geq 1$ $(\varepsilon_n)_{n \geq 1}$ is independent

(4/4)

\therefore Random walk.

- cd) \hat{X}_n is a function of H_1, \dots, H_n ; X_0, \dots, X_n (all F_n -measurable)

$\therefore \hat{X}_n$ is F_n -measurable.

$$\mathbb{E}[|\hat{X}_n|] \leq \sum_{k=1}^n \|H_k\|_{\infty} \cdot \mathbb{E}[|X_k| + |X_{k-1}|]$$

$$< \infty \quad (M := \sup \{ |H_k| : 1 \leq k \leq n \})$$

$\therefore \hat{X}_n$ is integrable.

$$\mathbb{E}[\hat{X}_{n+1} - \hat{X}_n | \mathcal{F}_n] = \mathbb{E}[H_{n+1}(X_{n+1} - X_n) | \mathcal{F}_n] = H_{n+1} \cdot \mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n]$$

(H_{n+1} is \mathcal{F}_n -measurable)

$$= H_{n+1} \cdot 0 \quad (\text{martingale property of } (X_n)_{n \geq 0}) = 0$$

$\therefore (\hat{X}_n)_{n \geq 0}$ is a martingale.

(3/13)

(d) Claim: $(X_n)_{n \geq 0}$ is a martingale wrt $(\mathcal{F}_n)_{n \geq 0}$, $\mathcal{F}_n = \sigma(X_0, \dots, X_n)$

$$\mathbb{E}[|X_n|] \leq n < \infty, \quad X_n \text{ is } \mathcal{F}_n\text{-measurable}$$

$$\mathbb{E}[X_{n+1} - X_n | \mathcal{F}_n] = 0 \quad (\text{symmetric}) \quad \therefore \text{Done.}$$

$$X_n \text{ Define: } \left. \begin{array}{l} H_k = 1 \quad \text{if } k \leq T_a \\ H_k = -1 \quad \text{if } k > T_a \end{array} \right\} H_k = 1 - 2 \mathbb{1}_{k > T_a}$$

$$\text{Since } \{T_a \leq kn\} = \bigcap_{r=0}^{n-1} \{X_r \neq a\} \in \mathcal{F}_{n-1} : T_a \text{ is a stopping time,}$$

H_k is \mathcal{F}_{k-1} measurable.

$$\forall |H_k| \leq 1 \Rightarrow \text{Bounded.}$$

$\therefore \text{ c d c) } \Rightarrow (\hat{X}_n)_{n \geq 0} \quad (\hat{X}_0 = 0) \text{ is a martingale.}$

$$n \leq T_a: \quad \hat{X}_n = X_0 + \sum_{k=1}^n (X_k - X_{k-1}) = X_n$$

$$\begin{aligned} n > T_a: \quad \hat{X}_n &= \underbrace{\sum_{k=1}^{T_a} (X_k - X_{k-1})}_a + \underbrace{\sum_{r=T_a+1}^n (X_r - X_{r-1})}_{X_n - a} \\ &= 2a - X_n. \end{aligned}$$

$\therefore (\hat{X}_n)_{n \geq 0}$ is a martingale.

$$\forall n \geq 1: \begin{cases} X_n - X_{n-1} = 1 & (IP = 1/2) \\ X_n - X_{n-1} = -1 & (IP = 1/2) \end{cases} \quad \text{Symmetry of } (X_n)_{n \geq 0}$$

$\therefore (b) \Rightarrow$ Martingale Random Walk.

(9/4)

(c) $IP(M_n \geq a) = IP(\hat{M}_n = M_n \geq a)$ $(\hat{M}_n = M_n \{X_0, \dots, X_n\})$

$$IP(M_n \geq a, X_n \leq a-b) = IP(\hat{M}_n \geq a, X_n \leq a-b) \quad \text{since } \hat{M}_n \geq a \Rightarrow X_n \geq a$$

$$= IP(M_n \geq a, X_n \geq a+b) \quad (M_n \geq a \Leftrightarrow \hat{M}_n \geq a)$$

$$= IP(X_n \geq a+b) \quad (b \geq 0)$$

$$\therefore IP(M_n \geq a) = IP(X_n \geq a+0) + IP(\hat{M}_n \geq a, X_n > a)$$

$$= IP(X_n \geq a) + IP(X_n \geq a+1)$$

$$\therefore IP(M_n = a) = IP(X_n \geq a) - IP(X_n \geq a+1)$$

$$IP(M_n \geq a) - IP(M_n \geq a+1) = IP(X_n \geq a) - IP(X_n \geq a+2)$$

$$= IP(X_n \in [a, a+1])$$

$= ?$

(6/7)

(11/9, 2)

2.) (a) If $\phi \cdot (S_1 - S_0) \geq 0$ (A.S.), $IP(\phi \cdot (S_1 - S_0) \geq 0) > 0$:

(7/2)

ϕ is an arbitrage.

(b) If ϕ is maximising portfolio, $t=1$ payoff X_1 :

$$\phi \cdot (S_1 - S_0) \geq \phi \cdot (S_1 - S_0) \quad \text{maxi}$$

$$\phi \cdot (S_1 - S_0) \geq \phi \cdot (S_1 - S_0) + \phi \cdot X_0 \quad \text{maximised}$$

Let ϕ be an arbitrage portfolio:

$$\phi \cdot (X_2 + \dots) \geq 0$$

$$U(X_1) \leq U(X_1 + \phi(S_1 - (1+r)S_0)) \quad (A.S.)$$

$$U(X_1) < U(X_1 + \phi(S_1 - (1+r)S_0)) \quad \text{with } IP > 0$$

$$\therefore E[U(X_1)] < E[U(X_1 + \phi(S_1 - (1+r)S_0))]$$

\therefore Contradict optimal solution

(4/4)

\therefore No arbitrage exists

(c) Let X^* maximise $E[U(X^*)]$, $\Lambda = \{ \text{attainable } t=1 \text{ strategies} \}$

Λ is a vector space determined by $\underline{\theta}$ (# of risky shares held)

Pretend

$$\pi(Y) \text{ is the solution to: } E[U(X^*)] = \sup_{A \in \Lambda} \{ E[U(A + Y - \pi(Y))] \}$$

Fix $t \in (0, 1)$; Let Y_1, Y_2 be claims, A^* optimises

$$A \mapsto E[U(A + Y - \pi(Y))]$$

$$E[U(X^*)] = t E[U(A_1 + Y_1 - \pi(Y_1))] + (1-t) E[U(A_2 + Y_2 - \pi(Y_2))]$$

$$\leq E[U(tA_1 + (1-t)A_2 + tY_1 + (1-t)Y_2 - (t\pi(Y_1) + (1-t)\pi(Y_2)))]$$

$$\leq \sup_{A \in \Lambda} \{ E[U(A + tY_1 + (1-t)Y_2 - (t\pi(Y_1) + (1-t)\pi(Y_2)))] \}$$

But LHS: $\sup_{A \in \Lambda} \{ E[U(A + tY_1 + (1-t)Y_2 - (\pi(tY_1 + (1-t)Y_2)))] \}$

$$\therefore \pi(tY_1 + (1-t)Y_2) \geq t\pi(Y_1) + (1-t)\pi(Y_2)$$

(5/7)

\therefore Concave.

If $\pi(Y) > \sup \{p \in I\}$: Buying Y at $\pi(Y)$ is an (neg.) arbitrage. any?

$$\therefore \mathbb{E}[u(X^*)] \geq \mathbb{E}[u(X^* + Y - \pi(Y))] \quad (\text{A.S.})$$

$$u(X^*) \geq u(X^* + Y - \pi(Y)) \quad \text{with } IP > 0$$

$$\therefore \mathbb{E}[u(X^*)] \geq \mathbb{E}[u(X^* + Y - \pi(Y))]$$

Let A maximise $\mathbb{E}[u(A + Y - \pi(Y))]$

$$\therefore \mathbb{E}[u(A + Y - \pi(Y))] \leq \mathbb{E}[u(A)] \leq \mathbb{E}[u(X^*)]$$

$$\mathbb{E}[u(X^*)] \leq$$

\therefore Contradiction $\Rightarrow \pi(Y) \leq \sup \{p \in I\}$

(5/7)

18d

3) (a) Claim: $\mathbb{E}[M_{n+1}] = \mathbb{E}[M_{n+1} | \mathcal{F}_n] = \mathbb{E}[M_n]$

$$\therefore \mathbb{E}[M_0] = \mathbb{E}[M_1] = \dots = \mathbb{E}[M_n]$$

Since $\mathbb{E}[\hat{M}_{n+1} | \mathcal{F}_n] \leq \hat{M}_n$ (super martingale):

$$\mathbb{E}[\hat{M}_{n+1} - \hat{M}_n] = \mathbb{E}[\underbrace{\mathbb{E}[\hat{M}_{n+1} - \hat{M}_n | \mathcal{F}_n]}_{\leq 0 \text{ A.S.}}] \leq 0.$$

$\therefore (\mathbb{E}[\hat{M}_n])_{n \geq 0}$ is a decreasing sequence $\Rightarrow \mathbb{E}[\hat{M}_n] \leq \mathbb{E}[\hat{M}_0] = \mathbb{E}[M_0]$
 $= \mathbb{E}[M_n]$

~~$M_n = \sum_{k=0}^n 1_{T \geq k} \cdot M_k$~~ Let $T \leq N$ A.S.:

$$M_T = \sum_{k=0}^N 1_{T \geq k} \cdot M_k$$

$$\therefore \mathbb{E}[M_T - \hat{M}_T] = \sum_{k=0}^N \mathbb{E}[1_{T \geq k} \cdot \underbrace{(M_k - \hat{M}_k)}_{\pi_k}]$$

$$= \sum_{k=0}^N \mathbb{E}[1_{T \geq k} (M_k - M_{k-1})]$$

(3/7)

We have $M_0 - \hat{M}_0 = 0 : \Rightarrow E[M_0 - \hat{M}_0] = 0$

If $E[M_k - \hat{M}_k | F_k] \geq 0 :$

4.) (a) $(B_t)_{t \geq 0}$ is a brownian motion (BM) if:

$$B_0 = 0$$

$t \mapsto B_t$ is cont.

A.S.

$(B_t)_{t \geq 0}$ has stationary increments, $(s \leq t) \quad B_t - B_s \sim N(0, t-s)$

If $(W_t)_{t \geq 0}$ is gaussian: $W_0 = 0$, $t \mapsto W_t$ is cont. (A.S.)

$$(IE[W_0] = 0, IE[W_s] = 0 \Rightarrow W_0 = 0 \text{ A.S.})$$

Fix $0 = t_0 < t_1 \dots < t_n$:

~~$(W_{t_1}, \dots, W_{t_n})$~~ $(W_{t_1-t_0}, \dots, W_{t_n-t_{n-1}})$ is a Gaussian vector \Rightarrow determined by mean vector (0), Cov matrix ~~(t_i, t_j)~~

If $(B_t)_{t \geq 0}$ is BM: $B_{t_k} - B_{t_{k-1}} \sim N(0, t_k - t_{k-1})$

$$IE[B_s B_t] = s + IE[B_s (B_t - B_s)] = s \quad (s \leq t)$$

indepndent

$$\therefore \begin{pmatrix} B_{t_1} - B_{t_0} \\ \vdots \\ B_{t_n} - B_{t_{n-1}} \end{pmatrix} = \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix} \cdot \begin{pmatrix} B_{t_1} \\ \vdots \\ B_{t_n} \end{pmatrix} \quad (B_{t_1}, \dots, B_{t_n}) \sim N$$

$(B_{t_1}, \dots, B_{t_n})$ is normal, mean 0, $\text{Var} = (\text{Cov})_{ij} = \text{Min}\{t_i, t_j\}$

\therefore Since $\begin{pmatrix} B_{t_1} \\ \vdots \\ B_{t_n} \end{pmatrix}$, $\begin{pmatrix} W_{t_1} \\ \vdots \\ W_{t_n} \end{pmatrix}$ have the same mean, cov, are gaussian \Rightarrow Same distribution.

$\therefore (W_t)_{t \geq 0}$ has independent increments, $(s \leq t) \quad B_t - B_s \sim N(0, t-s)$

(6/b)

$\therefore (W_t)_{t \geq 0}$ is a BM

(b) $\hat{W}_0 = 0$, $IE[\hat{W}_0] = 0$, $IE[\hat{W}_t] = t \cdot 0 \quad (t \geq 0)$

$$IE[\hat{W}_s \hat{W}_t] = 0, \quad IE[\hat{W}_s \hat{W}_t] = st \quad IE[W_{1/s} W_{1/t}] = s \quad (s \leq t)$$

$(W_{t_1}, \dots, W_{t_n})$ gaussian $(\forall t_i \in [0, \infty)) \Rightarrow$

$(\frac{1}{\sqrt{t_1}} W_{1/t_1}, \dots, \frac{1}{\sqrt{t_n}} W_{1/t_n})$ gaussian ✓

(4/4)

$(\omega) = (W_t)_{t \geq 0}$ is gauss. BM ✓

cc1

Let $A_n \in \mathcal{B}(\mathbb{R})$:

$\mathbb{P}(2W_2 + c t_1 \in A_1, \dots, W_{t_n} + c t_n \in A_n)$

$$\text{LHS} = \int_{\mathbb{R} \dots} \int_{\mathbb{R}} P_{t_0=0, x_0, x_1}(t_1 - t_0) \dots P_{x_{n-1}, x_n}(t_n - t_{n-1}) \dots f(x_1, \dots, x_n) dx_n \dots dx_1$$

$$P_t(y) = \frac{1}{\sqrt{2\pi t}} e^{-y^2/2t}$$

$$= \int_{\mathbb{R} \dots} \int_{\mathbb{R}} f(x_1, \dots, x_n) P_{x_1 - c t_1}(t_1) \dots P_{x_{n-1}} \dots$$

$$P_{t_1}(y) = \frac{1}{\sqrt{2\pi t_1}} e^{-y^2/2t_1}$$

$$\text{LHS} = \int_{\mathbb{R} \dots} \int_{\mathbb{R}} f(x_1 + c t_1, \dots, x_n + c t_n) P_{t_1}(x_1) \dots P_{t_n - t_{n-1}}(x_n - x_{n-1}) dx_n \dots dx_1$$

$$= \int_{\mathbb{R} \dots} \int_{\mathbb{R}} f(x_1, \dots, x_n) P_{t_1}(x_1 - c t_1) \dots P_{t_n - t_{n-1}}(x_n - x_{n-1} - c(t_n - t_{n-1})) dx_n \dots dx_1$$

$$\stackrel{\text{why?}}{=} \int_{\mathbb{R} \dots} \int_{\mathbb{R}} f(x_1, \dots, x_n) P_{t_1}(x_1) \dots P_{t_n - t_{n-1}}(x_n - x_{n-1}) \cdot e^{+c x_1} \dots e^{+c(x_n - x_{n-1})} \cdot e^{-c^2/2 t_1} \dots e^{-c^2/2(t_n - t_{n-1})} dx_n \dots dx_1$$

$$= \mathbb{E} \left[f((W_t)_{0 \leq t \leq T}) \cdot e^{c W_{t_n} - c^2 t_n / 2} \right] = \text{RHS.} \quad \checkmark$$

(5/6)

$$\begin{aligned}
 (d) \quad f((W_t + x)_{t \geq T}) &= f\left\{t \cdot \left(\frac{1}{t} W_t + \frac{x}{t}\right)_{t \geq T}\right\} \\
 &= f\left\{\frac{1}{t} \cdot (\hat{W}_t + tx) : 0 \leq t \leq \frac{1}{T}\right\}
 \end{aligned}$$

∴ Cameron-Martin:

$$\begin{aligned}
 \text{LHS } \mathbb{E}[f] &= \mathbb{E}\left[e^{x \hat{W}_{1/T} - \frac{x^2}{2} (1/T)} \cdot f\left\{\frac{1}{t} \hat{W}_t : 0 \leq t \leq \frac{1}{T}\right\}\right] \\
 &= \mathbb{E}\left[e^{-\frac{x^2}{2T} + x \left(\frac{1}{T} W_T\right)} \cdot f\left\{\frac{1}{t} \cdot t W_{1/t} : 0 \leq t \leq \frac{1}{T}\right\}\right] \\
 &= \mathbb{E}\left[e^{-\frac{x^2}{2T} + \frac{x}{T} W_T} \cdot f\{W_t : t \geq T\}\right]
 \end{aligned}$$

(4/4)
(19.12)

3.) (a) Let $T \leq N$ A.S.

$$M_T = \sum_{k=0}^{N-1} 1_{T \leq k} X_k + 1_{T \leq N} X_N \quad (\text{similarly for } \hat{M}_T)$$

$$\mathbb{E}[M_T - \hat{M}_T] = \sum_{k=0}^N \mathbb{E}\left[1_{\substack{T \leq k \\ T \leq N}} (M_k - \hat{M}_k)\right]$$

$1_{T \leq k}$ is \mathcal{F}_k -measurable: (super) - Martingale property

$$\mathbb{E}[1_{T \leq k} M]$$

$$\text{Claim: } \mathbb{E}[\hat{M}_T] \leq \mathbb{E}[M_0]$$

$$\hat{M}_T = \hat{M}_0 + \sum_{k=1}^N 1_{T \geq k} (\hat{M}_k - \hat{M}_{k-1})$$

$$\therefore \mathbb{E}[\hat{M}_T - \hat{M}_0] = \sum_{k=1}^N \mathbb{E}[1_{T \geq k} \cdot (\hat{M}_k - \hat{M}_{k-1})]$$

$$1_{T \geq k} \text{ is } \mathcal{F}_{k-1} \text{ measurable} \Rightarrow \mathbb{E}[1_{T \geq k} (\hat{M}_k - \hat{M}_{k-1})] \leq 0.$$

$$\mathbb{E}[1_{T \geq k} (M_k - M_{k-1})] = 0.$$

(7/7) $\therefore \mathbb{E}[\hat{M}_T] \leq \mathbb{E}[\hat{M}_0] = \mathbb{E}[M_0] = \mathbb{E}[M_T]$ ✓

(3/3) (b) Hold θ_n stock: $X_{n+1} - \theta_n \cdot S_n$ cash left

$$\left. \begin{aligned} X_{n+1} - \theta_n S_{n+1} &\rightarrow (1+r) X_{n+1} - \theta_n S_n (1+r) \\ \theta_n S_{n+1} &\rightarrow \theta_n \otimes S_n \end{aligned} \right\} X_n = (1+r) X_{n+1} + \theta_n (S_n - (1+r) S_{n+1})$$

(c) Let θ_n be trading strategy:

$\forall n$ $V_n(X_n)$ is a function in $X_n \Rightarrow F_n$ -measurable.

$$\mathbb{E}[V_{n+1}(X_{n+1}) - V_n(X_n) | F_{n+1}] = V_{n+1}(X_{n+1}) - V_n(X_n)$$

$$\mathbb{E}[V_n((1+r)X_{n+1} + \theta_n(S_n - (1+r)S_{n+1})) | F_{n+1}]$$

Since ξ_n is independent of F_{n-1}

$$\text{But } \mathbb{E}[V_n((1+r)X_{n+1} + \theta_n S_{n+1}(\xi - (1+r))) | F_{n+1}] \leq \sup_{t \in \mathbb{R}} \left\{ \mathbb{E}[V_n((1+r)X_{n+1} + t(\xi - (1+r))) | F_{n+1}] \right\}$$

$$= V_n(X_{n+1})$$

(3/3) $\therefore \mathbb{E}[V_{n+1}(X_{n+1}) - V_n(X_n) | F_{n+1}] \leq 0 \Rightarrow \text{Super-martingale } (*)$ ✓

$$\mathbb{E}[U(X_N)] = \mathbb{E}[V_N(X_N)] \leq \mathbb{E}[V_0(X_0)] \quad (\text{super-martingale})$$

$$= V_0(X_0) \quad \text{deterministic} \quad - (1)$$

(True for all strategies).

If θ^* optimises gives martingale process: Equality in (1) is obtained in (1)

(7/7) \therefore Upper Bound is attained \Rightarrow Solution is optimal! ✓

(10/2)

(*) Note, did not mention L' properly. Can we ignore it?

seems you can assume it here ---

SFM 2022

1.) (a) $V_N(x) = x^2$

$$V_{n-1}(x) = \inf_u \left\{ u^2 + \mathbb{E}[V_n(X+u+\xi)] \right\}$$

(b) Try $V_n(x) = A_n + B_n \cdot x + C_n \cdot x^2$: $A_n = B_n = 0$, $C_n = 1$

$$u^2 + A_n + B_n(x+u) + C_n \left\{ \text{Var}(X+u+\xi) + \mathbb{E}[X+u+\xi]^2 \right\}$$

$$= u^2 + A_n + B_n(x+u) + C_n \left\{ \sigma^2 + (x+u)^2 \right\}$$

Diff. wrt u : $2u + B_n + 2C_n(x+u) = 0$

$$\therefore 2(1+C_n) \cdot u = -B_n - 2C_n x$$

$$\therefore u = -\frac{(B_n + 2C_n x)}{2(1+C_n)}$$

$$\therefore V_{n-1}(x) = (1+C_n)u^2 + (B_n + 2x C_n)u + (x B_n + \sigma^2 C_n + x^2 C_n)$$

$$= \frac{-(B_n + 2C_n x)^2}{4(1+C_n)} + (x B_n + \sigma^2 C_n + x^2 C_n) \quad \left. \begin{array}{l} B_n = 0 \Rightarrow B_{n-1} = 0 \\ B_n = 0 \Rightarrow B_{n-1} = 0 \end{array} \right\}$$

Since $B_n = 0 \Rightarrow B_{n-1} = 0$ and $B_n = 0$: $B_n = 0$ ($1 \leq n \leq N$)

$$\therefore V_{n-1}(x) = \frac{-C_n^2 x^2}{1+C_n} + C_n x^2 + \sigma^2 C_n$$

$$= x^2 \left\{ \frac{C_n + C_n^2 - C_n^2}{1+C_n} \right\} + \sigma^2 C_n = x^2 \left\{ \frac{C_n}{1+C_n} \right\} + \sigma^2 C_n$$

$$\therefore C_{n-1} = \frac{C_n}{1+C_n} :$$

$$C_N = 1, \quad C_{N-1} = \frac{1}{2}, \quad C_{N-2} = \frac{\frac{1}{2}}{1+\frac{1}{2}} = \frac{1}{3} = \frac{1}{3}$$

$$\text{If } C_n = \frac{1}{k}: \quad C_{n-1} = \frac{1}{k+1} \Rightarrow C_n = \frac{1}{1+(N-n)}$$

$$A_{n-1} = \sigma^2 C_n = \sigma^2 / (1+(N-n))$$

Optimal Control: $U_n^* = - \frac{C_n X}{1 + C_n} = - C_{n-1} \cdot X = - \frac{X_{n-1}}{2 + N - n}$

$\therefore U_1^* = - \frac{X_0}{N+1}$ (Base case)

(Induct): If $U_k^* = - \frac{X_0}{N+1} \dots - \frac{\varepsilon_{k-1}}{N-k+2}$

$$U_{k+1}^* = - \frac{1}{2+N-(k+1)} \left[\underbrace{X_{k+1}}_{X_{k+1}} + U_k^* + \varepsilon_{k+1} \right]$$

~~$X_k = X_{k+1}$~~ $X_{k+1} \left(1 - \frac{1}{2+N-k} \right) = X_{k+1} \left(\frac{1+N-k}{2+N-k} \right)$

$$\therefore U_{k+1}^* = - \frac{1}{2+N-k+1} \left[\frac{X_{k+1} (1+N-k)}{2+N-k} \right] - \frac{\varepsilon_{k+1}}{2+N-(k+1)}$$

$$= - \underbrace{\frac{X_{k+1}}{2+N-k}}_{U_k^*} - \frac{\varepsilon_{k+1}}{2+N-(k+1)}$$

$$= U_k^* - \frac{\varepsilon_{k+1}}{2+N-(k+1)}$$

\therefore Induction complete.

2) (a) Minimise : $\frac{1}{2} \theta^T V \theta$ subject to

$$\theta^T (p - c(r) \cdot S_0) + c(r) x = m$$

Lagrangian : $V \cdot \theta - \lambda (p - c(r) \cdot S_0) = 0$

$$\therefore \theta = \lambda V^{-1} d, \quad d = p - c(r) \cdot S_0$$

Constraint: $\lambda (d^T V^{-1} d) + c(r) x = m \Rightarrow \lambda = \frac{m - c(r) x}{d^T V^{-1} d}$

$$\therefore \theta = \underbrace{\left(\frac{m - c(r) x}{d^T V^{-1} d} \right)}_{\lambda} \cdot V^{-1} (p - c(r) S_0)$$

Minimise $\frac{1}{2} \theta^T V \theta$: subject to $E[X_1] \geq m$.

$$\begin{aligned} \text{If } E[X_1] = m: \quad \text{Min Var} &= \theta^T V \theta = \lambda^2 \left[d^T V^{-1} \cdot V \cdot V^{-1} d \right] \\ &= \frac{(m - c(r) x)^2}{d^T V^{-1} d} \end{aligned}$$

$$\begin{aligned} \therefore \text{Min} \{ \text{Var}(X_1) : E[X_1] \geq m \} &= \text{Min} \left\{ \frac{(k - c(r) x)^2}{d^T V^{-1} d} : k \geq m \right\} \\ &= 0 \quad \text{if} \quad m \leq c(r) \cdot x \\ &= \frac{(m - c(r) x)^2}{d^T V^{-1} d} \quad \text{if} \quad m > c(r) \cdot x. \end{aligned}$$

(b) Max. : $\frac{c(r) \cdot x + \theta^T (p - c(r) S_0)}{\sqrt{\theta^T V \theta}} = f(\theta)$

4 If $\lambda \neq 0$: $f(\lambda \theta) = f(\theta)$

So Max: $\theta^T \cdot (p - c(r) S_0)$, subj. $\sqrt{\theta^T V \theta} = 1$

$$L: (p - c) x_1 S_0 - 2\lambda (V \cdot \theta) = 0$$

($\lambda \geq 0$), else we can go to $-\infty$

$$\therefore \theta = V^{-1} (d/2\lambda)$$

$$\text{Constraint: } \frac{1}{4\lambda^2} \left[d^T V^{-1} \cdot V V^{-1} d \right] = d^T V^{-1} d / 4\lambda^2 = 1$$

$$\therefore \lambda = \frac{\sqrt{d^T V^{-1} d}}{2}$$

$$\therefore \theta = \cancel{d/2\lambda} \cdot c \cdot (V^{-1} \cdot d), \quad c > 0 \text{ is any constant}$$

$$E[X_1 | \theta] = c (d^T V^{-1} d) + c x_1 = M$$

$$\text{Var}(\theta) = c^2 (d^T V^{-1} d) = \frac{(M - c x_1)^2}{d^T V^{-1} d}$$

$$= \inf \{ \text{Var}(x_1) \mid E[x_1] = M \}$$

\therefore Mean - Var efficient. ($\forall c > 0$)

(c) V is symmetric $\Rightarrow \exists$ orthogonal P s.t. $V = P^T D P$,

$$D = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ & & \lambda_k & & 0 \end{pmatrix}, \quad \lambda_i \neq 0; \quad d = (p - c) x_1 S_0 = \begin{pmatrix} d_1 \\ \vdots \\ d_k \end{pmatrix}$$

d_i is 1st k coordinates

No solution is

$$P^T D P \cdot \theta = d \Rightarrow D \cdot (P \cdot \theta) = \underbrace{(P \cdot d)}_B \quad \text{has no solution}$$

$$\begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ & & \lambda_k & & 0 \end{pmatrix} \cdot \begin{pmatrix} P \cdot \theta \end{pmatrix} = \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_k \end{pmatrix} : \quad \beta_1 \text{ is 1st } k \text{ coordinates}$$

$$\text{If } \beta_k = 0, \quad \exists \text{ solution: } \quad \beta_1 \quad \left(\begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ & & \lambda_k & & 0 \end{pmatrix}^{-1} \beta_1 \right)$$

$$\therefore \beta_k \neq 0$$

$$\text{Pick } P \cdot \theta = \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_k \end{pmatrix}$$

Diagonal like V :

Let $V = P^T \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_k & & 0 \dots 0 \end{pmatrix} P$, P orthogonal.

$$\text{Min: } \theta^T V \theta \quad \text{subj. } \theta^T (E \cdot d + (1+r) X) = M$$

$$\equiv \text{Min: } \theta^T \cdot D \cdot \theta \quad \text{subj. } \theta^T \cdot (P d) = M - (1+r) X, \quad V = P^T D P$$

$$\therefore \text{No sh: to } V \cdot \theta = (M - (1+r) S) \quad \text{iff}$$

$$\text{No sh to: } \forall \theta \quad D \cdot \theta = P \cdot (M - (1+r) S)$$

$$\therefore \text{WLOG, } V \text{ diagonal} \Rightarrow V = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_k & & 0 \end{pmatrix}, \quad P \cdot (M - (1+r) S) = \begin{pmatrix} a_1 \\ \\ a_2 \end{pmatrix}$$

(a_1 is 1st k coordinates)

$$\text{If } a_2 = 0: \quad \begin{pmatrix} (\lambda_1 \dots \lambda_k)^+ \cdot a_1 \\ 0 \end{pmatrix} \text{ is a solution} \Rightarrow \text{reject.}$$

WLOG, $(a_2)_i \neq 0$:

$$\therefore \theta = \frac{m}{(a_2)_i} \delta_{i, (k+1)} \text{ is a mean } m \text{ stat.}$$

with 0 Var.

$$\therefore \text{Min } \{ \text{Var}(X_i) \mid \mathbb{E}[X_i] = m \} = 0$$

Pick $m > (1+r) \cdot S_0$: θ is a 0-var stat.

$$\therefore S_1 = m > (1+r) S_0 \quad (\text{A.S.})$$

\therefore Arbitrage.

3.) (a) $M_t \geq 0$:
$$\begin{aligned} \mathbb{E}[M_t | \mathcal{F}_s] &= M_s \cdot \mathbb{E}\left[e^{c(W_t - W_s) - \frac{c^2}{2}(t-s)} \middle| \mathcal{F}_s\right] \\ &= M_s \cdot \mathbb{E}\left[e^{c \cdot N(0, t-s) - \frac{c^2}{2}(t-s)}\right] = M_s \cdot e^{-\frac{c^2}{2}(t-s)} \cdot e^{0 + \frac{c^2}{2}(t-s)} \\ &= M_s \end{aligned}$$

$\therefore \mathbb{E}[M_t] = \mathbb{E}[M_t | \mathcal{F}_0] = M_0 = 1 \Rightarrow$ Integrable, measurable in \mathcal{F}_t .

\therefore Martingale

(b) Change measure : $\frac{dG}{dP} = e^{cW_T - \frac{c^2}{2}T}$

$(W_t)_{t \geq 0}$ is a c -drift Brownian motion (Cameron-Martin)

\therefore (Discounted price):
$$\begin{aligned} X_t &= S_0 e^{cW_t + (p-r)t} \\ &= S_0 e^{cB_t + (c^2 + p-r)t} \end{aligned}$$

Pick c st $c^2 + p - r = -\frac{\sigma^2}{2} \Rightarrow c = \frac{1}{\sigma} (r - p - \frac{\sigma^2}{2})$

X_t is a G -martingale (risk neutral measure G)

(c)
$$\begin{aligned} \pi &= e^{-rT} \cdot \mathbb{E}^G[S_T^P] = e^{-rT} \int_{\mathbb{R}} \frac{e^{-y^2/2}}{\sqrt{2\pi}} S_0^P e^{cP\sqrt{T}y + P(r-\frac{\sigma^2}{2})T} dy \\ &= e^{-rT} \cdot S_0^P \int_{\mathbb{R}} \frac{e^{-y^2/2}}{\sqrt{2\pi}} e^{cP\sqrt{T} \cdot y} e^{P(r-\frac{\sigma^2}{2}) \cdot T} dy \\ &= e^{-rT} S_0^P e^{P(r-\frac{\sigma^2}{2})T} \int_{\mathbb{R}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y - cP\sqrt{T})^2} e^{\frac{1}{2}c^2P^2T} dy \\ &= \cancel{e^{-rT}} \cdot S_0^P e^{-\cancel{(r-p)} \cdot rT} S_0^P e^{c^2T/2 [-p+p^2]} \end{aligned}$$

$$Y_1 = \max_{0 \leq t \leq T} \left\{ S_0 e^{G B_t + (r - G^2/2)t} \right\} = S_0 e^{G \max_{0 \leq t \leq T} \left\{ B_t + \frac{(r - G^2/2)}{G} t \right\}}$$

$$Y_2 = S_0^{2-p} S_0^p e^{(G B_T + (r - G^2/2)T)p} \quad \text{Min} \left\{ S_0 e^{G B_t + (r - G^2/2)t} \right\}$$

$$= S_0 e^{(G B_T + (r - G^2/2)T)p} e^{\max \left\{ -G B_t - (r - G^2/2)t \right\}}$$

$\therefore Y_2$ has the same distribution as:

$$S_0 e^{p(G B_T + (r - G^2/2)T) \cdot p} e^{G \cdot \max_{0 \leq t \leq T} \left\{ B_t - \frac{(r - G^2/2)}{G} t \right\}}$$

Let G_1 be measure: $\frac{dG_1}{dG} = e^{(r - G^2/2)d \cdot B_T - d^2 T/2}$, $d = \frac{(r - G^2/2)}{G}$

$$\therefore \mathbb{E}^{G_1} [Y_1] = \mathbb{E}^G \left[S_0 e^{G \max_{0 \leq t \leq T} \left\{ B_t + (rd)t - dt/2 \right\}} \right]$$

$$= \mathbb{E}^{G_1} \left[S_0 e^{G \max_{0 \leq t \leq T} \left\{ B_t - dt/2 \right\}} \right]$$

$$= \mathbb{E}^G \left[S_0 e^{d B_T - d^2 T/2} e^{G \max_{0 \leq t \leq T} \left\{ B_t - dt/2 \right\}} \right]$$

$$= \mathbb{E}^G \left[S_0 e^{p(G B_T + G d/2)} e^{G \max_{0 \leq t \leq T} \left\{ B_t - dt/2 \right\}} \right]$$

iff $GP = d$.

\therefore Pick $P = \frac{d}{G} = \frac{(r - G^2/2)}{G^2}$ = Claims have the same $t=0$ price.

4.) (a) Cost of buying Θ_n shares: $\Theta_n \cdot S_{n-1}$
 \therefore Cash held from $n-1 \rightarrow n$: $(X_{n-1} - \Theta_n S_{n-1}) \rightarrow (1+r) (X_{n-1} - \Theta_n S_{n-1})$
 Stock value: $\Theta_n S_{n-1} \rightarrow \Theta_n S_n$

$$\therefore X_n = (1+r) (X_{n-1} - \Theta_n S_{n-1}) + \Theta_n S_n$$

$$= \Theta_n (S_n - (1+r) S_{n-1}) + (1+r) X_{n-1}$$

(b)

~~$S_0, S_1, S_2, \dots, S_n, S_{n+1}$~~

Risk neutral: $E^Q [S_{n+1} / (1+r)^{n+1} | S_1, \dots, S_n] = S_n / (1+r)^n$

$$\therefore E^Q [\xi_n | S_1, \dots, S_n] = (1+r)$$

$$\therefore (1+r) = \mathbb{P}^Q(\xi_n = b | F_{n-1}) (1+b) + Q(\xi_n = a | F_{n-1}) (1+a)$$

$$\therefore Q(\xi_n = b | F_{n-1}) = \frac{r-a}{b-a} \quad \text{is independent of } n, F_{n-1}$$

Since Q is a unique measure, $(\xi_n)_{n \geq 1}$ remains independent.

$$Q(S_N = S_0 (1+b)^i (1+a)^{N-i}) = \left(\frac{r-a}{b-a}\right)^i \cdot \left(\frac{b-r}{b-a}\right)^{N-i} \cdot \binom{N}{i}$$

$$+ q = \frac{r-a}{b-a}$$

(c) Market is complete $\Rightarrow \exists$ replicating strategy $(\Theta_n)_{n=1, \dots, N}$.

$(\Theta_n \cdot X_n)_{n=1, \dots, N}$ is a martingale $(X_n = S_n / (1+r)^n)$, $\Theta_n = \Theta_N \cdot \frac{S_N}{S_n} = \frac{g(S_N)}{S_n}$ (A.S.)

$$V_n = (S_0, \dots, S_n) = E^Q[\Theta_n S_n | F_n] = (1+r)^{-n} E^Q[\Theta_n X_n | F_n]$$

$$= (1+r)^{-n} \cdot \frac{E^Q[\Theta_N \cdot X_N | F_n]}{g(S_N) \cdot (1+r)^{N-n}} = (1+r)^{-n} \cdot \frac{E^Q[\Theta_N \cdot X_N | F_n]}{g(S_N) \cdot (1+r)^{N-n}}$$

$$\begin{aligned} \therefore \Theta_n \cdot S_n (1+b) &= V_{n+1} (S_0, \dots, S_n; S_n (1+b)) \\ \Theta_n \cdot S_n (1+a) &= V_{n+1} (S_0, \dots, S_n; S_n (1+a)) \end{aligned} \quad \left. \begin{array}{l} \text{Either scenario, our} \\ \text{replicating strategy} \\ \text{must match valuation} \end{array} \right\}$$

$$\therefore \Theta_n = \frac{V_{n+1} (S_0, \dots, S_n; S_n (1+b)) - V_{n+1} (S_0, \dots, S_n; S_n (1+a))}{S_n (1+b - 1-a)}$$

cd) Payout = $(S_N - K)^+$

Pick $r = \inf \{ k \geq 0 : S_0 (1+b)^k (1-a)^{N-k} > K \}$

$$\mathbb{E}^G [(S_N - K)^+] = \sum_{n=r}^N \binom{N}{n} q^n (1-q)^{N-n} \left\{ S_0 (1+b)^r (1-a)^{N-n} - K \right\}$$

$$= \sum_{n=r}^N \binom{N}{n} (q \cdot (1+b))^n ((1-q)(1-a))^{N-n} - K \mathbb{Q}(S_N > K)$$

$$\therefore \Pi = Q(1+r)^{-N} \cdot \mathbb{E}^G [(S_N - K)^+]$$

$$= \sum_{n=r}^N \binom{N}{n} \underbrace{\left[\frac{q(1+b)}{(1+r)} \right]^n}_{q^+} \underbrace{\left[\frac{(1-q)(1-a)}{(1+r)} \right]^{N-n}}_{p^+} - K \mathbb{Q}(S_N > K)$$

$$q^+ + p^+ = (1+r)^{-1} \{ q(1+b) + (1-q)(1-a) \} = 1 \quad (\text{def of } G)$$

$$\begin{aligned} \hat{Q}(\xi_n = (1+b)) &= q^+ \\ \hat{Q}(\xi_n = (1-a)) &= (1-q^+) \end{aligned} \quad \left. \begin{array}{l} (\xi_n)_{n=1, \dots, N} \text{ iid wrt } \hat{Q} \end{array} \right\}$$

$$\therefore \Pi = \hat{Q}[S_N > K] - K(1+r)^{-N} \hat{Q}[S_N > K]$$

No.:

Date: