

## 12. Review problems

*Problem 12.1.* Consider the  $N$  period Binomial model with  $N = 5$ , and parameters  $0 < d < 1 + r < u$ . At maturity  $N = 5$ , a security pays \$1 if  $S_5 > (1 + r)S_4$ , and 0 otherwise. Find the arbitrage free price and trading strategy trading at time 0.

What not to do! LAST RESORT

Draws a tree of all coin tosses!

Better strategy: If payoff is a fn of Stock price then so is the AFP

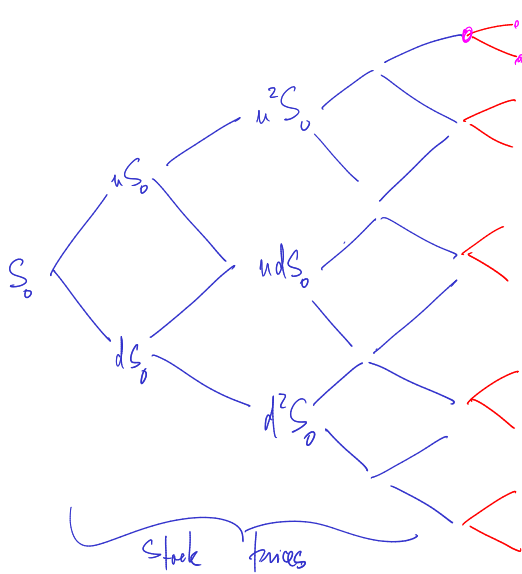
i.e. Drawing a tree of Stock prices may be enough  
(Much much less work!!)

Draw tree of Stock prices up to time 4  
& then coin tosses

$$\hat{p} = \frac{1+r - d}{u-d}$$

$$\hat{q} = \frac{u - (1+r)}{u-d}$$

$$V_4 = \frac{1}{1+r} E_4^2 V_5$$



$$\begin{matrix}
 V_5 = 1 \\
 0 \\
 | \\
 0 \\
 | \\
 0 \\
 | \\
 0 \\
 | \\
 0 \\
 | \\
 0
 \end{matrix}$$

$$\begin{aligned}
 V_4 &= \tilde{\phi} / (1+r) \\
 &= \tilde{\phi} / (1+r)
 \end{aligned}$$

$$\tilde{\phi} / (1+r)$$

.

.

.

.

.

time  $n=5$   
draws coin tosses

$$V_0 = \frac{1}{(1+r)^4} \stackrel{E}{=} V_4 = \frac{\tilde{\phi}/(1+r)}{(1+r)^4} = \frac{\tilde{\phi}}{(1+r)^5}$$

Trading strat  $\Delta_0 = \frac{V_1(1) - V_1(-1)}{S_0(u-d)} = 0$

Problem 12.2. Let  $f$  be a deterministic function, and define

$$X_t \stackrel{\text{def}}{=} \int_0^t \underline{f(s)} \underline{W_s} ds.$$

Find the distribution of  $X$ .

$$X \sim N$$

(

$$\mathbb{E} X_t = \mathbb{E} \int_0^t f(s) W_s ds = 0$$

think of this as

$$\lim_{\|P\| \rightarrow 0} \sum f(t_i) W_{t_i} (t_{i+1} - t_i)$$

h.c. of normals

→ Normal.

$$\textcircled{2} \quad \mathbb{E} X_t^2 = \mathbb{E} \left( \int_0^t f(s) W_s ds \right)^2$$

$$= \mathbb{E} \int_0^t f(s) W_s ds \int_0^t f(r) W_r dr$$

$$= \mathbb{E} \int_0^t \int_0^t f(s) f(r) W_s W_r ds dr$$

$$= \int_0^t \int_0^t f(s) f(r) (s \wedge r) ds dr$$

$$= \int_{r=0}^t \int_{s=0}^r f(s) f(r) s ds dr + \int_{r=0}^t \int_{s=r}^t f(s) f(r) r ds dr$$

& evaluate both!

*Problem 12.3.* Suppose  $\underline{\sigma}, \underline{\tau}, \underline{\rho}$  are three deterministic functions and  $M$  and  $N$  are two continuous martingales with respect to a common filtration  $\{\mathcal{F}_t\}$  such that  $M_0 = N_0 = 0$ , and

$$d[M, M]_t = \sigma_t dt, \quad d[N, N]_t = \tau_t dt, \quad \text{and} \quad d[M, N]_t = \rho_t dt.$$

- (a) Compute the joint moment generating function  $\mathbf{E} \exp(\lambda M(t) + \mu N(t))$ .
- (b) (*Lévy's criterion*) If  $\sigma = \tau = 1$  and  $\rho = 0$ , show that  $(M, N)$  is a two dimensional Brownian motion.

Let  $\underline{\varphi(t)} = \mathbf{E} e^{\lambda M_t + \mu N_t}$ .

Find  $\varphi(t)$ : Let  $X_t = e^{\lambda M_t + \mu N_t} = f(t, M_t, N_t)$



where  $f(t, x, y) = e^{\lambda x + \mu y}$

Compute ①  $\frac{\partial}{\partial t} f = 0$

②  $\frac{\partial}{\partial x} f = \lambda f$  ,  $\frac{\partial^2}{\partial x^2} f = \lambda^2 f$

③  $\frac{\partial}{\partial y} f = \mu f$  ,  $\frac{\partial^2}{\partial y^2} f = \mu^2 f$

④  $\frac{\partial^2}{\partial x \partial y} f = \lambda \mu f$ .

$$\Rightarrow dX_t = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dM + \frac{\partial f}{\partial y} dN +$$

$$+ \frac{1}{2} \left( \frac{\partial^2 f}{\partial x^2} d[M, M] + \frac{\partial^2 f}{\partial y^2} d[N, N] \right.$$

$$\left. + \frac{\partial^2 f}{\partial x \partial y} d[M, N] + \frac{\partial^2 f}{\partial y \partial x} d[N, M] \right)$$

$$= 0 + \lambda f dM + \mu f dN$$

$$+ \frac{1}{2} \left( \lambda^2 \sigma^2 + \mu^2 \tau^2 + 2\lambda\mu \rho \sigma\tau \right) dt$$

equal.

$$\Rightarrow e^{\lambda M_t - \mu N_t} - 1 = X_t - X_0$$

$$= \lambda \int_0^t \cancel{X} dM_s + \mu \int_0^t \cancel{X} dN_s + \frac{1}{2} \int_0^t \cancel{X} \left( \lambda^2 \sigma^2 + \mu^2 \tau + 2\lambda\mu\rho \right) ds$$

Not random.  
↓

$$\Rightarrow E(e^{\lambda M_t - \mu N_t} - 1) = E X_t - X_0 = \varphi(t) - 1$$

$$= 0 + 0 + \frac{1}{2} \int_0^t \varphi(s) \left( \lambda^2 \sigma^2 + \mu^2 \tau + 2\lambda\mu\rho \right) ds$$

$$\Rightarrow \partial_t \varphi = \frac{1}{2} \left( \lambda^2 \sigma(t) + \mu^2 \tau(t) + 2\lambda\mu \rho(t) \right) \varphi(t)$$

$$\text{Soln} \Rightarrow \varphi(t) = 1 \cdot \exp \left( \frac{1}{2} \lambda^2 \int_0^t \sigma ds + \mu^2 \int_0^t \tau + 2\lambda\mu \int_0^t \rho \right)$$

= MGF of normal with mean 0  
& cov matrix  $\begin{pmatrix} \int_0^t \sigma & \int_0^t \rho \\ \int_0^t \rho & \int_0^t \tau \end{pmatrix} \dots \textcircled{*}$

Note: If  $\rho = 1, \tau = 1, \rho = 0$

then  $d[M, M] = dt$   
 $d[W, N] = dt$   
 $d[M, N] = 0$  } ← Joint QV of a 2D B.M.

⊗ ⇒ MGF of  $(M, N) =$  MGF of Normal, mean 0  
& cov matrix  $\begin{pmatrix} t & 0 \\ 0 & t \end{pmatrix}$

Can check similarly

$$\begin{pmatrix} M_t - M_s \\ N_t - N_s \end{pmatrix} \sim \text{2D normal mean } 0 \text{ \& cov } \begin{pmatrix} t-s & 0 \\ 0 & t-s \end{pmatrix}$$

\& is ind of  $\mathcal{F}_s$

$\Rightarrow (M, N)$  is a 2D B.M. (Levy!)

*Problem 12.4.* Consider a financial market consisting of a risky asset and a money market account. Suppose the return rate on the money market account is  $r$ , and the price of the risky asset, denoted by  $S$ , is a geometric Brownian motion with mean return rate  $\alpha$  and volatility  $\sigma$ . Here  $r$ ,  $\alpha$  and  $\sigma$  are all deterministic constants. Compute the arbitrage free price of derivative security that pays

$$\rightarrow V_T = \frac{1}{T} \int_0^T S_t dt$$

at maturity  $T$ . Also compute the trading strategy in the replicating portfolio.

RNP formula: For any  $t \leq T$ ,

$$D_t = e^{-\int_0^t R_c ds} = e^{-rt}$$

AFP is  $V_t = \frac{1}{D_t} \mathbb{E}_t (D_T V_T)$

$$= e^{-r(T-t)} \mathbb{F}_t^r V_T$$

$$= e^{-r(T-t)} \left[ \mathbb{F}_t^r \frac{1}{T} \int_0^T S_s ds \right] \quad (v = T-t)$$

$$= \mathbb{F}_t^r \left( \int_0^t \mathbb{F}_t^r S_s ds + \int_t^T \mathbb{F}_t^r S_s ds \right)$$



$$= \frac{e^{-rT}}{T} \left( \int_0^t S_s ds + \int_t^T \underbrace{F_t^2}_{\text{green}} S_s ds \right)$$

Need to find this!

( $s > t$ )

Option 1:  $S_s = S_t \exp \left( \left( r - \frac{\sigma^2}{2} \right) (s-t) + \sigma (\tilde{W}_s - \tilde{W}_t) \right)$

$\left( \begin{array}{l} ds = \alpha S dt + \sigma S dW \\ \text{applying Ito to } \ln(s) \text{ \& solve} \end{array} \right)$

$$\hookrightarrow \Rightarrow \mathbb{F}_t^2 S_s = \mathbb{F}_t^2 \left( S_t \exp \left( \left( r - \frac{\sigma^2}{2} \right) (s-t) + \sigma \underbrace{\left( \tilde{W}_s - \tilde{W}_t \right)}_{N(0, s-t)} \right) \right)$$

(option 1)  
indefinite integral

$$\int_{y=-\infty}^{\infty} \left( S_t \exp \left( \left( r - \frac{\sigma^2}{2} \right) (s-t) + \sigma \sqrt{s-t} y \right) \frac{e^{-\frac{y^2}{2}}}{\sqrt{2\pi}} dy \right)$$

(option 2)

$$\equiv S_t e^{(r - \frac{\sigma^2}{2})(s-t)} \mathbb{F} \tilde{e}^{\sigma(\tilde{W}_s - \tilde{W}_t)}$$

$$= S_t e^{(r - \frac{\sigma^2}{2})(s-t)} e^{\frac{\sigma^2}{2}(s-t)}$$

(MGF of normal!)

( $s > t$ )

Option 2<sup>b</sup>: Find  $E_t[S_s]$  by discounting! (Much faster)

Know  $e^{-rs} S_s$  is a  $\tilde{P}$  mgf!

$$\begin{aligned}\Rightarrow \mathbb{E}_t^Q S_s &= e^{+rs} \mathbb{E}_t^Q (e^{-rs} S_s) \stackrel{\text{mg}}{=} e^{+rs} (e^{-rt} S_t) \\ &= e^{r(s-t)} S_t !\end{aligned}$$

---

Substitute back & get  $V_t$ !

Problem 12.5. Let  $X \sim N(0, 1)$ , and  $a, \alpha, \beta \in \mathbb{R}$ . Define a new measure  $\tilde{P}$  by

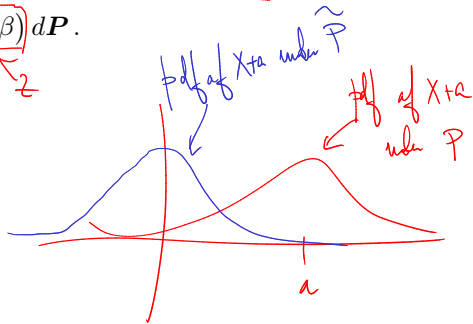
$$d\tilde{P} = \exp(\alpha X + \beta) dP.$$

Find  $\alpha, \beta$  such that  $X + a \sim N(0, 1)$  under  $\tilde{P}$ .

Compute MGF of  $X+a$  under  $\tilde{P}$

$$\text{Find } \tilde{E}(e^{\lambda(X+a)})$$

$$\text{Note } \tilde{E}(e^{\lambda(X+a)}) = E(e^{\lambda(X+a)} \cdot z)$$



$$= \mathbb{E} \left( e^{\lambda(X+a) + \alpha X + \beta} \right)$$

$$= \left( \mathbb{E} e^{(\lambda+\alpha)X} \right) \cdot e^{a\lambda + \beta}$$

$$= e^{(\lambda+\alpha)^2/2 + a\lambda + \beta}$$

Want

$$= e^{\lambda^2/2}$$

$$\text{Set } \frac{(\lambda + \alpha)^2}{2} + a\lambda + \beta = \frac{\lambda^2}{2} \quad (\text{for all } \lambda \in \mathbb{R})$$

$$\Leftrightarrow \frac{\lambda^2}{2} + \alpha\lambda + \frac{\alpha^2}{2} + a\lambda + \beta = \frac{\lambda^2}{2} \quad ( \quad )$$

$$\Leftrightarrow \underbrace{(\alpha + a)}_{\text{egrete to } 0} \lambda + \underbrace{\frac{\alpha^2}{2} + \beta}_{\text{egrete to } 0} = 0$$

$$\left. \begin{array}{l} \text{set } \alpha = -a \\ \beta = -\frac{a^2}{2} \end{array} \right\}$$

*Problem 12.6.* Let  $x_0, \mu, \theta, \sigma \in \mathbb{R}$ , and suppose  $X$  is an Itô process that satisfies

$$dX(t) = \theta(\mu - X_t) dt + \sigma dW_t,$$

with  $X_0 = x_0$ .

(a) Find functions  $f = f(t)$  and  $g = g(s, t)$  such that

$$X(t) = f(t) + \int_0^t g(s, t) dW_s.$$

The functions  $f, g$  may depend on the parameters  $x_0, \theta, \mu$  and  $\sigma$ , but should not depend on  $X$ .

(b) Compute  $\mathbf{E}X_t$  and  $\text{cov}(X_s, X_t)$  explicitly.



*Problem 12.7.* Let  $\theta \in \mathbb{R}$  and define

$$Z(t) = \exp\left(\theta W_t - \frac{\theta^2 t}{2}\right).$$

Given  $0 \leq s < t$ , and a function  $f$ , find a function such that

$$\mathbf{E}_s f(Z_t) = g(Z(s)).$$

Your formula for the function  $g$  can involve  $f$ ,  $s$ ,  $t$  and integrals, but not the process  $Z$  or expectations.

Problem 12.8. Let  $\underline{W}$  be a Brownian motion, and define

$$\underline{B}_t = \int_0^t \text{sign}(W_s) dW_s.$$

$$\text{sign}(x) = \begin{cases} 1 & x \geq 0 \\ -1 & x < 0 \end{cases}$$

- (a) Show that  $B$  is a Brownian motion.  
 (b) Is there an adapted process  $\sigma$  such that

$$W_t = \int_0^t \sigma_s dB_s?$$

If yes, find it. If no, explain why.

- (c) Compute the joint quadratic variation  $[B, W]$ .  
 (d) Are  $B$  and  $W$  uncorrelated? Are they independent? Justify.

(a)  $\rightarrow$  Levy: Noel

① Continuous

②  $B$  is a martingale (Ito int  $\checkmark$ )

③  $d[B, B] = dt$

( $\because d[B, B] = 1 dt \checkmark$ )

(b) Find  $\sigma$  so that  $W_t = \int_0^t \sigma_s dB_s$ .

Note  $dB = \overbrace{\text{sign}(W_s)} dW_s$

$\Rightarrow dW_s = \underbrace{\frac{1}{\text{sign}(W_s)}}_{\sigma_s} dB_s$

$\Rightarrow W_t = \int_0^t \text{sign}(W_s) dB_s$

$$\textcircled{c} d[B, W]_t = \text{sign}(W_t) \cdot 1 d[W, W]_t$$

$$= \text{sign}(W_t) dt$$

$$\Rightarrow [B, W]_t = \int_0^t \text{sign}(W_s) dt$$

$$\textcircled{d} \text{ find } E(B_t W_t) :$$

$$dB_t = \overbrace{\text{sign}(W_t)} dW_t$$

$$dW_t = \underbrace{1} dW_t$$

$$d(BW) = B dW + W dB + d[B, W]$$

$$\Rightarrow B_t W_t - 0 = \int_0^t B_s dW_s + \int_0^t W_s dB_s + \int_0^t \text{sign}(W_s) ds$$

$$\Rightarrow E(B_t W_t) = 0 + 0 + \int_0^t \underbrace{E \text{Sign}(W_s)}_0 ds$$

$\Rightarrow B$  &  $W$  are uncorrelated!

Note  $B_t$  &  $W_t$  are both normal, Not Jointly normal! need not  $\Rightarrow$  indep

If  $B$  &  $W$  were indep  $\Rightarrow [B, W] = 0$

$$\text{But } [B, W] = \int_0^t \text{Sign}(W_s) ds \neq 0$$

$\Rightarrow B$  &  $W$  can NOT be indep!

*Problem 12.9.* Let  $W$  be a Brownian motion. Does there exist an equivalent measure  $\tilde{P}$  under which the process  $tW_t$  is a Brownian motion? Prove it.

L

$$d \left( \int_0^t f(s) ds \right) = f(t) dt$$

$$\frac{d}{dt} \int_0^t f(s) ds = f(t)$$

$$M_t = \int_0^t W_s ds$$

---

$$\text{Find } E M_t^2$$



$$d(M^2) = \underbrace{2M dM}_{\text{nat ng!}} + \underbrace{d[M, M]}_0$$

$$M_t^2 = 2 \int_0^t \underbrace{M_s W_s}_{\downarrow} ds + 0$$

$$d(MW)_t = M_t dW_t + W_t^2 dt \Rightarrow E(M_t W_t) = 0 + \int_0^t s ds$$