

TUTORIAL 7 (SOLUTIONS)

1. $\{X_t\}$ is a Gaussian process: for any $0 \leq t_1 < t_2 < \dots < t_k$,

$$(X_{t_1}, \dots, X_{t_k}) = (t_1 W_{1/t_1}, \dots, t_k W_{1/t_k})$$

is a linear transform of a Gaussian random vector (of the W 's) and hence is Gaussian as well.

Therefore by Th^m on sl. (205), we just have to verify that $EX_t = 0$ (which is obvious) and $EX_s X_t = s \wedge t$.

As for $0 < s < t$

$$\begin{aligned} EX_s X_t &= E(s W_{1/s} \cdot t W_{1/t}) = st E \underbrace{W_{1/s} W_{1/t}}_{= \frac{1}{s} \wedge \frac{1}{t} = \frac{1}{t}} \\ &= st \times \frac{1}{t} = s = s \wedge t, \text{ done.} \end{aligned}$$

2. The RV $X = 2W_{t_1} - W_{t_2}$ is normal - as a linear combination of the components of a normal random vector. So we just have to compute

the mean & var of X :

$$EX = E(2W_{t_1} - W_{t_2}) = 2\underbrace{EW_{t_1}}_{=0} - \underbrace{EW_{t_2}}_{=0} = 0,$$

$$\begin{aligned} \text{var}(X) &= EX^2 = E(2W_{t_1} - W_{t_2})^2 = \\ &= 4\underbrace{EW_{t_1}^2}_{=t_1} - 4\underbrace{EW_{t_1}W_{t_2}}_{=t_1 \wedge t_2 = t_1} + \underbrace{EW_{t_2}^2}_{=t_2} = t_2, \end{aligned}$$

so that $X \sim N(0, t_2)$.

Alternatively, just use the independence of the increments of $\{W_t\}$:

$$X = 2W_{t_1} + (W_{t_1} + (W_{t_2} - W_{t_1})) = \underbrace{W_{t_1}}_{\sim N(0, t_1)} - \underbrace{(W_{t_2} - W_{t_1})}_{\substack{\text{indep't RV's} \\ \sim N(0, t_2 - t_1) \\ \text{by symmetry}}} \sim N(0, t_1 + (t_2 - t_1)) = N(0, t_2)$$

as the variance of the sum of two indep't RV's = sum of variances, and as the sum of indep't normal RV's is normal again (using ch.f.'s)

3. (a) Need $E(X_{t+h} | \mathcal{F}_t) = X_t$
for any $t \geq 0, h > 0$. Now

$$E(X_{t+h} | \mathcal{F}_t) = E\left(S_0 e^{(\mu-r)(t+h) + \sigma W_{t+h}} \mid \mathcal{F}_t\right)$$

CE2

$$\overbrace{S_0 e^{(\mu-r)t + \sigma W_t}} = X_t$$

as $e^{\sigma W_t}$ is \mathcal{F}_t -measurable.

$$E\left(e^{(\mu-r)h + \sigma(W_{t+h} - W_t)} \mid \mathcal{F}_t\right)$$

CE3 as $W_{t+h} - W_t$ is indepⁿ of \mathcal{F}_t

$$e^{(\mu-r)h} E e^{\sigma(W_{t+h} - W_t)}$$

" $e^{\sigma^2 h / 2}$ as $W_{t+h} - W_t \sim N(0, h)$ "

$$= X_t e^{(\mu-r)h + \frac{\sigma^2}{2} h}$$

must have $\Leftrightarrow (\mu - r + \frac{\sigma^2}{2})h = 0, \forall h > 0,$
so that $\mu = r - \frac{\sigma^2}{2}$.

(b) $E(S_T - K)^+ = E(S_T - K; S_T > K)$
 $= E(S_T; S_T > K) - E(K; S_T > K)$
 $= S_0 e^{(r - \frac{\sigma^2}{2})T} E(e^{\sigma W_T}; W_T > w) - KP(W_T > w)$

as $S_T > K \Leftrightarrow W_T > [\ln(K/S_0) - (r - \frac{\sigma^2}{2})T] / \sigma = w$

$$= S_0 e^{(r - \frac{\sigma^2}{2})T} E\left(e^{6\sqrt{T}Z}; Z > \frac{w}{\sqrt{T}}\right) - K D\left(Z > \frac{w}{\sqrt{T}}\right) \quad \text{HN-7}$$

as $w_T = \sqrt{T}Z$, $Z \sim \mathcal{N}(0, 1)$.

$$\begin{aligned} \text{Here } P\left(Z > \frac{w}{\sqrt{T}}\right) &= P\left(-Z > -\frac{w}{\sqrt{T}}\right) = P\left(Z < -\frac{w}{\sqrt{T}}\right) \\ &\stackrel{\text{as } z \stackrel{d}{=} -z \text{ (symmetric distr)}}{=} \mathcal{N}\left(-\frac{w}{\sqrt{T}}\right) = \mathcal{N}\left(\frac{-\ln(K/S_0) + (r - \frac{\sigma^2}{2})T}{6\sqrt{T}}\right) \\ &= \mathcal{N}\left(\frac{\ln(S_0/K) + (r + \frac{\sigma^2}{2})T}{6\sqrt{T}} - 6\sqrt{T}\right) \\ &= \mathcal{N}(h - 6\sqrt{T}), \quad \text{and} \end{aligned}$$

= h from slide 79

$$E\left(e^{6\sqrt{T}Z}; Z > \frac{w}{\sqrt{T}}\right) = \frac{1}{\sqrt{2\pi}} \int_{w/\sqrt{T}}^{\infty} e^{6\sqrt{T}x - x^2/2} dx$$

$$\begin{aligned} \text{as } -\frac{1}{2}(x^2 - 26\sqrt{T}x) &= -\frac{1}{2}(x - 6\sqrt{T})^2 + \frac{6^2 T}{2} \\ &= \frac{e^{6^2 T/2}}{\sqrt{2\pi}} \int_{w/\sqrt{T}}^{\infty} e^{-(x - 6\sqrt{T})^2/2} dx = \frac{e^{6^2 T/2}}{\sqrt{2\pi}} \int_{\frac{w}{\sqrt{T}} - 6\sqrt{T}}^{\infty} e^{-y^2/2} dy \\ &= e^{\frac{6^2 T}{2}} P(Z > -h) = e^{\frac{6^2 T}{2}} N(h) \quad (\text{as above}), \end{aligned}$$

$y = x - 6\sqrt{T}$ $\frac{w}{\sqrt{T}} - 6\sqrt{T} = -h$

so that

$$E(S_T - K)^+ = S_0 e^{rT} N(h) - K N(h - 6\sqrt{T}).$$

Cf. with the BS-formula for C on slide 76

A. Using the MG $X_t = W_t^2 - t$ & OST:

$$0 = EX_0 = EX_t = E(W_t^2 - t) = E(a + bt - t)$$

$$= a - (1-b)Et,$$

so that

$$Et = \frac{a}{1-b}.$$

$\overset{\sim}{=} (\sqrt{a+bt})^2$ when crossing u_t first;
 $= (-\sqrt{a+bt})^2 = (a+bt)$ when crossing v_t first.

Showing that $\{W_t^2 - t\}$ is an MG:

$$E(W_t^2 - t | \mathcal{F}_s) = E((W_s + (W_t - W_s))^2 - t | \mathcal{F}_s)$$

$$= \underbrace{E(W_s^2 | \mathcal{F}_s)}_{\overset{\sim}{=} W_s^2} + \underbrace{E((W_t - W_s)^2 | \mathcal{F}_s)}_{\overset{\sim}{=} E(W_t - W_s)^2} + 2 \underbrace{E(W_s(W_t - W_s) | \mathcal{F}_s)}_{\overset{\sim}{=} W_s E(W_t - W_s | \mathcal{F}_s)}$$

as W_s^2 is \mathcal{F}_s -meas. $\overset{\sim}{=} W_s^2$

as $W_t - W_s$ is indep⁺ of \mathcal{F}_s $\overset{\sim}{=} E(W_t - W_s)^2 = t - s$

as W_s is \mathcal{F}_s -m. $\overset{\sim}{=} W_s E(W_t - W_s | \mathcal{F}_s)$
 as $W_t - W_s$ is indep⁺ of \mathcal{F}_s $\overset{\sim}{=} W_s E(W_t - W_s) = 0$

$$- t =$$

$$= W_s^2 + (t - s) + 0 - t = W_s^2 - s = X_s, \text{ MG!}$$