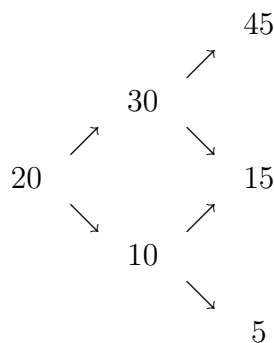


Please note that in the exam you are expected to (at least briefly) explain all steps in your solutions; in what follows, we often give only answers, without detailed solutions and/or explanations. The notation in the present paper is consistent with the one used in 2003 and may slightly differ from the one used in the 2002 exam paper.

1. (a) $C_2 = (S_2 - 20)^+ = \max\{S_2 - 20, 0\}$.
- (b) Use either (i) the valuation formula or (ii) the diagram method.
 - (i) $C_0 = \frac{1}{(1+r)^2} \mathbf{E}^* X$, where $X = C_2$ is the payoff.
Stock price diagram:



To compute \mathbf{E}^* , need to find $p^* = \frac{1+r-d}{u-d} = \frac{1+0.5}{1.5-0.5} = 0.5$, $1 - p^* = 0.5$. Now $C_0 = p^{*2}X(uu) + 2p^*(1 - p^*)X(ud) + (1 - p^*)^2X(dd) = 6.25$.

(ii) Or use the diagram method.

2. (a) Note that $S_0 = 10$. Then can take $f(x) = 10e^x$ and $X_t = -1.5t + 2W_t$ and use Itô's formula.
 - (b) As $X_t = 0.1e^{1.5t-2W_t}$, can proceed as in (a).
 - (c) Since $Y_t = Z_t X_t$ with $Z_t = K - T^{-1} \int_0^t S_u du$, can use the product rule of Itô's calculus (note that Z_t is a differentiable function!) to find dY_t : a routine computation.
3. (a) $m_X(t) = \frac{1}{\sigma\sqrt{2\pi}} \int e^{tx - (x-\mu)^2/2\sigma^2} dx$. When $\mu = 0$: complete the square in the exponential: $tx - x^2/2\sigma^2 = -(x - \sigma^2 t)^2/2\sigma^2 + \sigma^2 t^2/2$ to get $m_X(t) = e^{\sigma^2 t^2/2}$.
 - (b) As $X = X_0 + \mu$ (X_0 has zero mean and variance σ^2), $m_{X_0+\mu}(t) = e^{\mu t} m_{X_0}(t) = e^{\mu t + \sigma^2 t^2/2}$.
 - (c) $m_Z(t) = m_{2(X-Y)}(t) = m_{X-Y}(2t) = m_X(2t)m_{-Y}(2t) = m_X(2t)m_Y(-2t) = e^{(8\sigma^2)t^2/2}$, so $Z \sim N(0, 8\sigma^2)$.
 - (d) $\mathbf{E}(Z|Y) = 2(\mu - Y)$; $\mathbf{E}(e^{3Z}|Y) = e^{-6Y+6\mu+18\sigma^2}$.
 - (e) $\mathbf{E} e^{iuW} = \exp\{i\mu u - \sigma^2 u^2/2 + \lambda(e^{iu} - 1)\}$.
 - (f) Yes since $\mathbf{E} e^{iuW} = (\mathbf{E} e^{iuW_n})^n$, where W_n is a sum of two independent $N(\mu/n, \sigma^2/n)$ and $Po(\lambda/n)$ random variables.

- (g) $X_t = \mu t + \sigma W_t + N_t$, where $\{W_t\}$ is a std Brownian motion and $\{N_t\}$ a Poisson process with rate λ independent of $\{W_t\}$.
4. (a) See the lecture notes.
- (b) $\{W_t\}$ is adapted to \mathbf{F} by definition, $\mathbf{E}|W_t| < \infty$ as $W_t \sim N(0, t)$ (in fact, can easily compute this expectation!), and for $0 < s < t$,

$$\mathbf{E}(W_t | \mathcal{F}_s) = \mathbf{E}(W_s + (W_t - W_s) | \mathcal{F}_s) = \dots = W_s$$

(don't forget to explain all the steps!!).

- (c) $\{\tau \leq t\}$ means: the “future” values of S_s ($s \in (t, T]$) don't exceed the maximum on $[0, t]$ —and this requires the knowledge of the future, which is not contained in \mathcal{F}_t .
- (d) $\mathbf{E}W_\tau = \mathbf{E}W_0$ provided τ is a bounded stopping time for $\{W_t\}$. As

$$20e^{-\tau+5W_\tau} = S_\tau \geq S_0 = 20e^0 = 20,$$

must have $-\tau + 5W_\tau \geq 0$, i.e. $W_\tau \geq 0.2\tau$. So

$$0 = \mathbf{E}W_0 \neq \mathbf{E}W_\tau \geq 0.2\mathbf{E}\tau > 0$$

(as $\tau > 0$ with positive probability). Thus the assertion of the OST doesn't hold and hence τ can't be a stopping time (the only other condition of the theorem is met: $\tau \leq T$ is bounded).

5. (a) $W_t = \sigma^{-1}[\ln(\tilde{S}_t/S_0) - (\mu - r)t]$, i.e. each of \tilde{S}_t and W_t is a function of the other—hence the processes have common “histories”.
- (b) That the process is adapted and the expectation is finite is obvious (explain why). For $0 < s < t$,

$$\begin{aligned} \mathbf{E}(\tilde{S}_t | \mathcal{F}_s) &= S_0 e^{(\mu-r)t} \mathbf{E}(e^{\sigma W_t} | \mathcal{F}_s) = S_0 e^{(\mu-r)t} e^{\sigma W_s} \mathbf{E}(e^{\sigma(W_t - W_s)} | \mathcal{F}_s) \\ &= \dots = \tilde{S}_s e^{(\mu-r+\sigma^2/2)(t-s)}, \end{aligned}$$

hence the conclusion.

- (c) Payoff: $g(S_t) = 10^3 \mathbf{1}_{\{S_T > K\}}$.

Hence by the valuation formula, the time $t = 0$ price of the option is

$$\begin{aligned} e^{-rT} \mathbf{E}(10^3 \mathbf{1}_{\{S_T > K\}}) &= 10^3 e^{-rT} \mathbf{P}(S_T > K) = \dots \\ &= 10^3 e^{-rT} \mathbf{P}\left(W_T / \sqrt{T} > [\ln(K/S_0) - (r - \sigma^2/2)T] / \sigma \sqrt{T}\right) = \dots \end{aligned}$$

6. (a) For $u = u(t, y)$,

$$u'_t = -((4 - y)u)'_y + 2u''_{yy} = u + (y - 4)u'_y + 2u''_{yy}.$$

- (b) $0 = \pi(y) + (y - 4)\pi'(y) + 2\pi''(y)$.

(c) Substituting $\pi(y) = Ce^{-(y-\mu)^2/2\sigma^2}$ in the equation from (b) and calculating

$$\pi' = -\frac{y-\mu}{\sigma^2}\pi, \quad \pi'' = \frac{(y-\mu)^2}{\sigma^4}\pi - \frac{1}{\sigma^2}\pi,$$

we get

$$0 = \left[1 - \frac{(y-4)(y-\mu)}{\sigma^2} + \frac{2(y-\mu)^2}{\sigma^4} - \frac{2}{\sigma^2} \right] \pi(y).$$

The equation holds when $\mu = 4$, $\sigma^2 = 2$. Therefore the stationary distribution is normal indeed, with the above parameters values.

7. (a) Set $\psi(t, a) = 0$, $\psi(t, b) = 1$, $t \geq 0$; then

$$\begin{aligned} v(s, x) &= \mathbf{E}(\psi(\tau, X_\tau) | X_s = x) \\ &= \mathbf{P}(X_\tau = b | X_s = x) = \mathbf{P}(X_\tau = b | X_0 = x) = V(x) \end{aligned}$$

doesn't depend on s as the process is time-homogeneous. As it solves the BWKE

$$v'_s = -\mu v'_x - \frac{\sigma^2}{2} v''_{xx}$$

we use $v'_s = 0$, $v'_x = V'$, $v''_{xx} = V''$ and $\mu = -1$, $\sigma^2 = 1$ to get

$$0 = V' - \frac{1}{2}V'', \quad V(a) = 0, \quad V(b) = 1.$$

(b) Clearly,

$$X_\tau = \begin{cases} b & \text{with probability } V(x), \\ a & \text{with probability } 1 - V(x). \end{cases}$$

To find $V(x)$, solve the differential equation from part (a). Its characteristic equation is $\frac{1}{2}z^2 - z = 0$, roots: $z_1 = 0$, $z_2 = 2$. Hence the general solution $V(x) = C_1 e^{2x} + C_2$, and using the boundary conditions,

$$V(x) = \frac{e^{2x} - e^{2a}}{e^{2b} - e^{2a}}, \quad a \leq x \leq b.$$

[Now use this to write down the distribution of X_τ .]

(c) Clearly, $\mathbf{P}(\max_{t \geq 0} X_t > b) = \lim_{a \rightarrow -\infty} V(0)$. Letting $a \rightarrow -\infty$, we get $\lim_{a \rightarrow -\infty} V(x) = e^{-2(b-x)}$, $x < b$. Hence the answer.

8. (a) BWKE: for $v = v(s, x)$,

$$v'_s = -\mu v'_x - \frac{\sigma^2}{2} v''_{xx} = x(x-1) \left(v'_x + \frac{1}{2} v''_{xx} \right).$$

FWKE: for $u = u(t, y)$,

$$\begin{aligned} u'_t &= -(\mu u)'_y + \left(\frac{\sigma^2}{2} u \right)''_{yy} = -(y(1-y)u)'_y + \frac{1}{2} (y(1-y)u)''_{yy} \\ &= 2(y-1)u + (1-3y+y^2)u'_y + \frac{1}{2} y(1-y)u''_{yy}. \end{aligned}$$

(b) Similarly to 7(a), for $V(x) = \mathbf{P}(X_\tau = 1 | X_0 = x)$ with

$$\tau = \min\{t \geq 0 : X_t = 0 \text{ or } X_t = 1\},$$

we get from the BWKE

$$0 = -\mu V' - \frac{1}{2}V'' = -x(1-x)V' - \frac{1}{2}x(1-x)V''.$$

For $x \in (0, 1)$ this is equivalent to

$$0 = V' + \frac{1}{2}V'', \quad V(0) = 0, \quad V(1) = 1.$$

Characteristic equation: $\frac{1}{2}z^2 + z = 0$, roots: $z = 0$, $z = -2$, so the general solution: $V(x) = C_1 e^{-2x} + C_2$. From the boundary conditions:

$$V(x) = \frac{1 - e^{-2x}}{1 - e^{-2}}, \quad x \in (0, 1).$$