

Solutions to MTH 9831 Fall 2015 HW #7
Baruch College MFE Program

Please note that these are *possible* solutions, and that there are often many ways to approach and solve a problem or prove a theorem.

Problem 1 (PDE with terminal condition) $g(t, x)$ satisfies the PDE with terminal condition

$$g_t(t, x) + \frac{1}{2}\sigma^2 g_{xx}(t, x) = rg(t, x); \quad g(T, x) = x^4.$$

To apply discounted Feynman-Kac via the SDE

$$dX(t) = \beta(t, X(t))dt + \gamma(t, X(t))dW(t),$$

we have $\beta(t, x) = 0$, $\gamma(t, x) = \sigma$, $h(x) = x^4$ satisfying the PDE with terminal condition

$$g_t(t, x) + \beta(t, x)g_x(t, x) + \frac{1}{2}\gamma^2(t, x)g_{xx}(t, x) = rg(t, x); \quad g(T, x) = h(x).$$

By discounted Feynman-Kac, then,

$$g(t, x) = e^{-r(T-t)}E^{t,x}(h(X(T)))$$

satisfies the PDE and boundary condition above. We solve for $X(t)$ and evaluate the expected value:

$$\begin{aligned} dX(t) &= \sigma dW(t) \implies X(t) = \sigma W(t) \implies W(t) = \frac{X(t)}{\sigma} \\ \implies g(t, x) &= e^{-r(T-t)}E(h(X(T)) | X(t) = x) \\ &= e^{-r(T-t)}E(X^4(T) | \sigma W(t) = x) \\ &= e^{-r(T-t)}E(X^4(T) | W(t) = \frac{x}{\sigma}) \\ &= e^{-r(T-t)}\sigma^4 E(W^4(T) | W(t) = \frac{x}{\sigma}). \end{aligned}$$

Checking the terminal condition $g(T, x) = x^4$, we see that

$$g(T, x) = e^{-r(T-T)}\sigma^4 E\left(W^4(T) \middle| W(T) = \frac{x}{\sigma}\right) = x^4,$$

so our final answer is

$$g(t, x) = e^{-r(T-t)}\sigma^4 E\left(W^4(T) \middle| W(t) = \frac{x}{\sigma}\right),$$

which can be simplified tremendously: we decompose $W(T)$ into time- t -measurable and independent-of-time- t pieces. The odd powers of independent terms drop out (odd moments of Brownian motion are 0), kurtosis, etc. give our calculations: using $W(T) - W(t) \sim \sqrt{T-t}Z$, $Z \sim N(0, 1)$, and knowing the moments $E(Z) = E(Z^3) = 0$, $E(Z^2) = 1$, $E(Z^4) = 3$,

$$\begin{aligned} g(t, x) &= e^{-r(T-t)}\sigma^4 E(W^4(T) | W(t) = \frac{x}{\sigma}) \\ &= e^{-r(T-t)}\sigma^4 E([W(T) - W(t) + W(t)]^4 | W(t) = \frac{x}{\sigma}) \\ &= e^{-r(T-t)}\sigma^4 E([\sqrt{T-t}Z + x]^4) \\ &= e^{-r(T-t)}[\sigma^4(T-t)^2 E(Z^4) + 4(T-t)^{3/2}E(Z^3) + 6(T-t)E(Z^2) \\ &\quad + 4\sqrt{T-t}E(Z) + x^4] \\ &= e^{-r(T-t)}\left[\sigma^4\left(3(T-t)^2 + 6(T-t)\left(\frac{x}{\sigma}\right)^2 + \left(\frac{x}{\sigma}\right)^4\right)\right] \\ &= e^{-r(T-t)}[3\sigma^4(T-t)^2 + 6\sigma^2(T-t)x^2 + x^4]. \end{aligned}$$

To finish the check, take the partials and match the PDE:

$$\begin{aligned} g(t, x) &= e^{-r(T-t)} j(t, x) \\ g_t(t, x) + \frac{1}{2} \sigma^2 g_{xx}(t, x) &= r g(t, x) + e^{-r(T-t)} \left[j_t(t, x) + \frac{1}{2} j_{xx}(t, x) \right]; \\ j_t(t, x) + \frac{1}{2} \sigma^2 j_{xx}(t, x) &= -6\sigma^4(T-t)\sigma^2 x^2 - \frac{1}{2} \sigma^2 (12\sigma^2(T-t) + 12\sigma^2 x^2) = 0. \end{aligned}$$

Problem 2 (Shreve Vol. 2, #6.1, p. 282) We are given $X(t) = x$,

$$\begin{aligned} dX(u) &= (a(u) + b(u)X(u))du + (\gamma(u) + \sigma(u)X(u))dW(u) \\ a(u), b(u), \gamma(u), \sigma(u) &\in \mathcal{F}(u) \\ Z(u) &= \exp \left\{ \int_t^u \sigma(v) dW(v) + \int_t^u \left(b(v) - \frac{1}{2} \sigma^2(v) \right) dv \right\} \\ Y(u) &= x + \int_t^u \frac{a(v) - \sigma(v)\gamma(v)}{Z(v)} dv + \int_t^u \frac{\gamma(v)}{Z(v)} dW(v). \end{aligned}$$

(i) $Z(t) = 1$ is easy: $Z(t) = e^0 = 1$ since the integrals start at t .

$$\begin{aligned} dZ(u) &= Z(u) [\sigma(u)dW(u) + (b(u) - \frac{1}{2}\sigma^2(u))du + \frac{1}{2}\sigma^2(u)du] \\ &= Z(u) [\sigma(u)dW(u) + b(u)du]. \end{aligned}$$

- (a) isn't needed; it's already done, since we know what $Z(u)$ is.
(b) Let $Y(t) = x$; what we are given shows that

$$dY(u) = k(u)du + l(u)dW(u)$$

is solved by

$$k(u) = \frac{a(u) - \sigma(u)\gamma(u)}{Z(u)}, \quad l(u) = \frac{\gamma(u)}{Z(u)}.$$

(ii) $X(u) = Y(u)Z(u)$ satisfies $dX(u)$ written above: we're given

$$X(t) = Y(t)Z(t) = x,$$

$$dY(u) = \frac{a(u) - \sigma(u)\gamma(u)}{Z(u)} du + \frac{\gamma(u)}{Z(u)} dW(u), \quad u \geq t,$$

and by Itô's product rule, it's clear that

$$\begin{aligned} d(Y(u)Z(u)) &= Y(u)dZ(u) + Z(u)dY(u) + dY(u)dZ(u) \\ &= Y(u)Z(u) [\sigma(u)dW(u) + b(u)du] + (a(u) - \sigma(u)\gamma(u))du + \gamma(u)dW(u) \\ &\quad + \gamma(u)\sigma(u)du \\ &= (a(u) + b(u)X(u))du + (\gamma(u) + \sigma(u)X(u))dW(u) = dX(u). \end{aligned}$$

Therefore, the solution is

$$X(u) = Y(u)Z(u) = Z(u) \left[x + \int_t^u \frac{a(v) - \sigma(v)\gamma(v)}{Z(v)} dv + \int_t^u \frac{\gamma(v)}{Z(v)} dW(v) \right].$$

Problem 3 (which line is non-trivial?) Our original equation is

$$g(t, x) = E^{t,x}[h(X(T))] + \int_t^T E^{t,x}[f(X(u))]du$$

subject to the SDE

$$dX(u) = b(u, X(u))du + \sigma(u, X(u))dB(u), \quad X(t) = x.$$

The changes occurring from line to line:

- 1) to 2) moves “initial conditions” to “conditioning”, which is allowed by stochastic Fubini’s theorem in the integral term.
- 2) to 3) moves from fixed initial point $X(t) = x$ to a random one, changing the deterministic function $g(t, x)$ into the stochastic process $g(t, X(t))$.
- 3) to 4) moves from conditioning on only $X(t)$ to the entire history from times $[0, t]$ - this now takes into account the past, not only time t , but works because, as Corollary 6.3.2, p 267 states, solutions to SDEs are Markov processes. This means that the conditional expectation only looks at time t , not the entire past. (This is the line that we would most likely say is non-trivial and requires the most justification.) As I stated in the TA session, this is where we stop thinking about the Markov property of $g(t, X(t))$ and start thinking about what its associated martingale is. To do that, we need the entire history to measure on, not just the current time.
- 4) to 5) decomposes the integral $\int_0^T f(X(u))du = \int_0^t f(X(u))du + \int_t^T f(X(u))du$ into $\mathcal{F}(t)$ -measurable and independent terms.
- 5) to 6) substitutes the boundary condition $g(T, x) = h(x)$ at $X(T) = x$.

This last line shows us that the martingale in question is

$$G(t, X(t)) := g(t, X(t)) + \int_0^t f(X(u))du = E \left[g(T, X(T)) + \int_0^T f(X(u))du \middle| \mathcal{F}(t) \right].$$

Note that if $f(x) \equiv 0$, then we have our usual Feynman-Kac Theorem situation, and the PDE would be

$$g_t(t, x) + b(t, x)g_x(t, x) + \frac{1}{2}\sigma^2(t, x)g_{xx}(t, x) = 0; \quad g(T, x) = h(x).$$

We take the differential of our martingale $G(t, x)$; its dt term must be equal to 0. This dt term is hence the PDE satisfied by our original equation; it will be in terms of f , g , and h .

$$\begin{aligned} dG(t, X(t)) &= dg(t, X(t)) + f(X(t))dt \\ &= g_t(t, X(t))dt + g_x(t, X(t))dX(t) + \frac{1}{2}g_{xx}(t, X(t))dX(t)dX(t) + f(X(t))dt \\ &= g_t dt + b g_x dt + \sigma g_x dB(t) + \frac{1}{2}\sigma^2 g_{xx} dt + f dt \\ &= [g_t + b g_x + \frac{1}{2}\sigma^2 g_{xx} + f] dt + \sigma g_x dB(t). \end{aligned}$$

The PDE and boundary condition are, therefore,

$$g_t(t, x) + b(t, x)g_x(t, x) + \frac{1}{2}\sigma^2(t, x)g_{xx}(t, x) = -f(x); \quad g(T, x) = h(x).$$

Problem 4 (from PDE to SDE)

(a) $\mathcal{A} = (2 - x) \frac{\partial}{\partial x} + 2x^2 \frac{\partial^2}{\partial x^2}$ comes from the diffusion process

$$dX(t) = \beta(X(t))dt + \gamma(X(t))dW(t)$$

where $f(t, X(t))$ is a martingale, and so

$$\begin{aligned} df(t, X(t)) &= f_t dt + f_x dX(t) + \frac{1}{2} f_{xx} d[X, X](t) = \left(f_t + \beta f_x + \frac{1}{2} \gamma^2 f_{xx} \right) dt + \gamma f_x dW(t); \\ f_t + \beta f_x + \frac{1}{2} \gamma^2 f_{xx} &= 0. \end{aligned}$$

We know the generator fits the form

$$\mathcal{A} = \beta(x) \frac{\partial}{\partial x} + \frac{1}{2} \gamma^2(x) \frac{\partial^2}{\partial x^2}$$

which gives us $\beta(x) = 2 - x$ and $\gamma(x) = 2x$. Thus, the SDE is

$$dX(t) = (2 - X(t))dt + 2X(t)dW(t).$$

We can solve this by applying Problem 2 with $a(u) = 2$, $b(u) = -1$, $\gamma(u) = 0$, and $\sigma(u) = 2$.

(b) The two-dimensional SDE from generator

$$\mathcal{A} = \frac{1+x^2}{2} \frac{\partial^2}{\partial x^2} + x \frac{\partial^2}{\partial x \partial y} + \frac{1}{2} \frac{\partial^2}{\partial y^2} + 2x \frac{\partial}{\partial y} + 2y \frac{\partial}{\partial x}$$

is of the form

$$d \begin{pmatrix} X(t) \\ Y(t) \end{pmatrix} = \begin{pmatrix} \beta_1(X(t), Y(t)) \\ \beta_2(X(t), Y(t)) \end{pmatrix} dt + \begin{pmatrix} \gamma_{11}(X(t), Y(t)) & \gamma_{12}(X(t), Y(t)) \\ \gamma_{21}(X(t), Y(t)) & \gamma_{22}(X(t), Y(t)) \end{pmatrix} \begin{pmatrix} dW_1(t) \\ dW_2(t) \end{pmatrix},$$

where $W(t) = (W_1(t), W_2(t))$ is a standard 2-dimensional Brownian motion. The generator gives us the unique first derivative drift functions

$$\beta_1(x, y) = 2y, \beta_2(x, y) = 2x$$

but the second derivative diffusion functions are not unique; the generator gives

$$\frac{1}{2} \begin{pmatrix} \gamma_{11}(x, y) & \gamma_{12}(x, y) \\ \gamma_{21}(x, y) & \gamma_{22}(x, y) \end{pmatrix} \begin{pmatrix} \gamma_{11}(x, y) & \gamma_{12}(x, y) \\ \gamma_{21}(x, y) & \gamma_{22}(x, y) \end{pmatrix}^T = \frac{1}{2} \begin{pmatrix} 1+x^2 & x \\ x & 1 \end{pmatrix}$$

which yields infinitely many solution for $\gamma_{ij}(x, y)$. Two reasonably solvable solutions are

$$\begin{pmatrix} \gamma_{11}(x, y) & \gamma_{12}(x, y) \\ \gamma_{21}(x, y) & \gamma_{22}(x, y) \end{pmatrix} = \begin{pmatrix} \sqrt{1+x^2} & 0 \\ \frac{x}{\sqrt{1+x^2}} & \frac{1}{\sqrt{1+x^2}} \end{pmatrix}, \begin{pmatrix} 1 & x \\ 0 & 1 \end{pmatrix},$$

yielding the SDE systems

$$\begin{aligned} dX(t) &= 2Y(t)dt + \sqrt{1+X^2(t)}dW_1(t) \\ dY(t) &= 2X(t)dt + \frac{X(t)}{\sqrt{1+X^2(t)}}dW_1(t) + \frac{1}{\sqrt{1+X^2(t)}}dW_2(t) \end{aligned}$$

and

$$\begin{aligned} dX(t) &= 2Y(t)dt + dW_1(t) + X(t)dW_2(t) \\ dY(t) &= 2X(t)dt + dW_2(t). \end{aligned}$$

Problem 5 (F-K + Girsanov)

The Feynman-Kac theorem states that the PDE with terminal condition

$$g_t(u, x) + \beta(u, x)g_x(u, x) + \frac{1}{2}\gamma^2(u, x)g_{xx}(u, x) = 0, \quad g(T, x) = h(x), \quad t \leq u \leq T$$

is associated with the SDE

$$dX(u) = \beta(u, X(u))du + \gamma(u, X(u))dW(u), \quad X(t) = x,$$

where W is a standard Brownian motion on the measure P (using E for expected values). (This setup is just like in Problems 3 and 4, except here we're only ranging time over $u \in [t, T]$.) Here we are given $\beta(u, x) = \theta(x)$ and $\gamma(u, x) \equiv 1$, so the SDE reduces to

$$dX(u) = \theta(X(u))du + dW(u), \quad X(t) = x.$$

At this point, think about the structure of this SDE while considering Girsanov's theorem: the function $\theta(X(u))$ is an adapted function of u . Therefore, we can shift $X(u)$ in space and time in the following way, via Girsanov, and end up with a new Brownian motion on a new measure. Define

$$\tilde{B}(u) := X(u + t) - x, \quad 0 \leq u \leq T - t.$$

Then, integrating the $dX(u)$ SDE, we get the translation

$$\begin{aligned} dX(u) &= \theta(X(u))du + dW(u), \quad X(t) = x, \quad t \leq u \leq T \\ \implies X(u) - x &= \int_t^u \theta(X(v))dv + [W(u) - W(t)] \\ \implies \tilde{B}(u) &\stackrel{(d)}{=} \int_0^u \theta(x + \tilde{B}(v))dv + W(u), \quad 0 \leq u \leq T - t; \end{aligned}$$

where $\tilde{B}(u)$ is a standard Brownian motion on the measure \tilde{P} (\tilde{E}). (We say "equal in distribution" (d) in the last equality because I've swapped out $W(u + t)$ for an increment. Since we're only taking an expected value, it's ok to use this.) How do we shift between measures? We need the Radon-Nikodym derivative $\frac{d\tilde{P}}{dP} = Z(T - t)$ from the process

$$\begin{aligned} Z(u) &= \exp\left(-\int_0^u \theta(\tilde{B}(v) + x)dW(v) - \frac{1}{2}\int_0^u \theta^2(\tilde{B}(v) + x)dv\right), \quad 0 \leq u \leq T - t \\ &= \exp\left(-\int_0^u \theta(\tilde{B}(v) + x)d\tilde{B}(v) + \frac{1}{2}\int_0^u \theta^2(\tilde{B}(v) + x)dv\right) \end{aligned}$$

since $-dW(u) = -d\tilde{B}(v) + \theta(x + \tilde{B}(v))dv$. Finally, this yields the F-K solution for $g(t, x)$, with a change of measure from E with $W(t) = x$ to \tilde{E} with $\tilde{B}(0) = 0$, and at the end a change back to E since we're talking about a standard Brownian motion inside an expectation, so the notation no longer matters:

$$\begin{aligned} g(t, x) &= E^{t, x}(h(X(T))) = \tilde{E}^{0, 0}(Z^{-1}(T - t)h(x + \tilde{B}(T - t))) \\ &= \tilde{E}^{0, 0}\left(\exp\left(-\int_0^{T-t} \theta(\tilde{B}(v) + x)d\tilde{B}(v) + \frac{1}{2}\int_0^{T-t} \theta^2(\tilde{B}(v) + x)dv\right)h(x + \tilde{B}(T - t))\right) \\ &= E^{t, x}\left(\exp\left(-\int_t^T \theta(W(v))dW(v) + \frac{1}{2}\int_t^T \theta^2(W(v))dv\right)h(W(T))\right). \end{aligned}$$

Problem 6 (Exercise 6.10: implying the volatility surface)

This is merely walking through some regular calculus integration-by-parts calculations. No need to reproduce a solution here.