

Lecture #33, 34: The Characteristic Function for a Diffusion

Recall that the characteristic function of a random variable X is the function $\varphi_X : \mathbb{R} \rightarrow \mathbb{C}$ defined by $\varphi_X(\theta) = \mathbb{E}(e^{i\theta X})$. From Exercise 3.9, if $X \sim \mathcal{N}(\mu, \sigma^2)$, then the characteristic function of X is

$$\varphi_X(\theta) = \exp \left\{ i\mu\theta - \frac{\sigma^2\theta^2}{2} \right\}.$$

Suppose that $\{X_t, t \geq 0\}$ is a stochastic process. For each $T \geq 0$, we know that X_T is a random variable. Thus, we can consider $\varphi_{X_T}(\theta)$.

In the particular case that $\{X_t, t \geq 0\}$ is a diffusion defined by the stochastic differential equation

$$dX_t = \sigma(t, X_t) dB_t + \mu(t, X_t) dt \quad (21.1)$$

where $\{B_t, t \geq 0\}$ is a standard Brownian motion with $B_0 = 0$, if we can solve the SDE, then we can determine $\varphi_{X_T}(\theta)$ for any $T \geq 0$.

Example 21.1. Consider the case when both coefficients in (21.1) are constant so that

$$dX_t = \sigma dB_t + \mu dt$$

where $\{B_t, t \geq 0\}$ is a standard Brownian motion with $B_0 = 0$. In this case, the SDE is trivial to solve. If $X_0 = x$ is constant, then for any $T \geq 0$, we have

$$X_T = x + \sigma B_T + \mu T$$

which is simply arithmetic Brownian motion started at x . Therefore,

$$X_T \sim \mathcal{N}(x + \mu T, \sigma^2 T)$$

so that

$$\varphi_{X_T}(\theta) = \exp \left\{ i(x + \mu T)\theta - \frac{\sigma^2 T \theta^2}{2} \right\}.$$

Example 21.2. Consider the Ornstein-Uhlenbeck stochastic differential equation given by

$$dX_t = \sigma dB_t + aX_t dt$$

where σ and a are constants. As we saw in Lecture #31, if $X_0 = x$ is constant, then for any $T \geq 0$, we have

$$X_T = e^{aT} x + \sigma \int_0^T e^{a(T-s)} dB_s \sim \mathcal{N} \left(x e^{aT}, \frac{\sigma^2 (e^{2aT} - 1)}{2a} \right).$$

Therefore,

$$\varphi_{X_T}(\theta) = \exp \left\{ i x e^{aT} \theta - \frac{\sigma^2 (e^{2aT} - 1) \theta^2}{4a} \right\}.$$

Now it might seem like the only way to determine the characteristic function $\varphi_{X_T}(\theta)$ if $\{X_t, t \geq 0\}$ is a diffusion defined by (21.1) is to solve this SDE. Fortunately, this is not true. In many cases, the characteristic function for a diffusion defined by a SDE can be found using the *Feynman-Kac representation theorem* without actually solving the SDE.

Consider the diffusion

$$dX_t = \sigma(t, X_t) dB_t + \mu(t, X_t) dt. \quad (21.2)$$

We know from Version IV of Itô's formula (Theorem 14.12) that if $f \in C^1([0, \infty)) \times C^2(\mathbb{R})$, then

$$\begin{aligned} df(t, X_t) &= f'(t, X_t) dX_t + \frac{1}{2} f''(t, X_t) d\langle X \rangle_t + \dot{f}(t, x) dt \\ &= \sigma(t, X_t) f'(t, X_t) dB_t + \left[\mu(t, X_t) f'(t, X_t) + \frac{1}{2} \sigma^2(t, X_t) f''(t, X_t) + \dot{f}(t, x) \right] dt. \end{aligned}$$

We also know from Theorem 12.6 that any Itô integral is a martingale. Therefore, if we can find a particular function $f(t, x)$ such that the dt term is zero, then $f(t, X_t)$ will be a martingale. We define the *differential operator* (sometimes called the *generator* of the diffusion) to be the operator \mathcal{A} given by

$$(\mathcal{A}f)(t, x) = \mu(t, x) f'(t, x) + \frac{1}{2} \sigma^2(t, x) f''(t, x) + \dot{f}(t, x).$$

Note that Version IV of Itô's formula now takes the form

$$df(t, X_t) = \sigma(t, X_t) f'(t, X_t) dB_t + (\mathcal{A}f)(t, X_t) dt.$$

This shows us the first connection between stochastic calculus and differential equations, namely that if $\{X_t, t \geq 0\}$ is a diffusion defined by (21.2) and if $f \in C^1([0, \infty)) \times C^2(\mathbb{R})$, then $f(t, X_t)$ is a martingale if and only if f satisfies the partial differential equation

$$(\mathcal{A}f)(t, x) = 0.$$

The Feynman-Kac representation theorem extends this idea by providing an explicit formula for the solution of this partial differential equation subject to certain boundary conditions.

Theorem 21.3 (Feynman-Kac Representation Theorem). *Suppose that $u \in C^2(\mathbb{R})$, and let $\{X_t, t \geq 0\}$ be defined by the SDE*

$$dX_t = \sigma(t, X_t) dB_t + \mu(t, X_t) dt.$$

The unique bounded function $f : [0, \infty) \times \mathbb{R} \rightarrow \mathbb{R}$ satisfying the partial differential equation

$$(\mathcal{A}f)(t, x) = \mu(t, x) f'(t, x) + \frac{1}{2} \sigma^2(t, x) f''(t, x) + \dot{f}(t, x) = 0, \quad 0 \leq t \leq T, \quad x \in \mathbb{R},$$

subject to the terminal condition

$$f(T, x) = u(x), \quad x \in \mathbb{R},$$

is given by

$$f(t, x) = \mathbb{E}[u(X_T) | X_t = x].$$

Example 21.4. We will now use the Feynman-Kac representation theorem to derive the characteristic function for arithmetic Brownian motion satisfying the SDE

$$dX_t = \sigma dB_t + \mu dt$$

where σ , μ , and $X_0 = x$ are constants. Let $u(x) = e^{i\theta x}$ so that the Feynman-Kac representation theorem implies

$$f(t, x) = \mathbb{E}[u(X_T)|X_t = x] = \mathbb{E}[e^{i\theta X_T}|X_t = x]$$

is the unique bounded solution of the partial differential equation

$$\mu f'(t, x) + \frac{1}{2}\sigma^2 f''(t, x) + \dot{f}(t, x) = 0, \quad 0 \leq t \leq T, \quad x \in \mathbb{R}, \quad (21.3)$$

subject to the terminal condition

$$f(T, x) = e^{i\theta x}, \quad x \in \mathbb{R}.$$

Note that $f(0, x) = \mathbb{E}[e^{i\theta X_T}|X_0 = x] = \varphi_{X_T}(\theta)$ is the characteristic function of X_T .

In order to solve (21.3) we use separation of variables. That is, we guess that $f(t, x)$ can be written as a function of x only times a function of t only so that

$$f(t, x) = \chi(x)\tau(t), \quad 0 \leq t \leq T, \quad x \in \mathbb{R}. \quad (21.4)$$

Therefore, we find

$$f'(t, x) = \chi'(x)\tau(t), \quad f''(t, x) = \chi''(x)\tau(t), \quad \dot{f}(t, x) = \chi(x)\tau'(t)$$

so that (21.3) implies

$$\mu\chi'(x)\tau(t) + \frac{1}{2}\sigma^2\chi''(x)\tau(t) + \chi(x)\tau'(t) = 0, \quad 0 \leq t \leq T, \quad x \in \mathbb{R},$$

or equivalently,

$$\frac{\mu\chi'(x)}{\chi(x)} + \frac{\sigma^2\chi''(x)}{2\chi(x)} = -\frac{\tau'(t)}{\tau(t)}.$$

Since the left side of this equation which is a function of x only equals the right side which is a function of t only, we conclude that both sides must be constant. For ease, we will write the constant as $-\lambda^2$. Thus, we must solve the two ordinary differential equations

$$\frac{\mu\chi'(x)}{\chi(x)} + \frac{\sigma^2\chi''(x)}{2\chi(x)} = -\lambda^2 \quad \text{and} \quad -\frac{\tau'(t)}{\tau(t)} = -\lambda^2.$$

The ODE for τ is easy to solve; clearly $\tau'(t) = \lambda^2\tau(t)$ implies

$$\tau(t) = C \exp\{\lambda^2 t\}$$

where C is an arbitrary constant. The ODE for χ is

$$\mu\chi'(x) + \frac{\sigma^2\chi''(x)}{2} = -\lambda^2\chi(x),$$

or equivalently,

$$\sigma^2 \chi''(x) + 2\mu \chi'(x) + 2\lambda^2 \chi(x) = 0. \quad (21.5)$$

Although this ODE is reasonably straightforward to solve for χ , it turns out that we do not need to actually solve it. This is because of our terminal condition. We know that

$$f(T, x) = e^{i\theta x}$$

and we also have assumed that

$$f(t, x) = \chi(x)\tau(t).$$

This implies that

$$f(T, x) = \chi(x)\tau(T) = e^{i\theta x}$$

which means that

$$\tau(T) = 1 \quad \text{and} \quad \chi(x) = e^{i\theta x}.$$

We now realize that we can solve for the arbitrary constant C ; that is,

$$\tau(t) = C \exp\{\lambda^2 t\} \quad \text{and} \quad \tau(T) = 1$$

implies

$$C = \exp\{-\lambda^2 T\} \quad \text{so that} \quad \tau(t) = \exp\{-\lambda^2(T - t)\}.$$

We are also in a position to determine the value of λ^2 . That is, we know that $\chi(x) = e^{i\theta x}$ must be a solution to the ODE (21.5). Thus, we simply need to choose λ^2 so that this is true. Since

$$\chi'(x) = i\theta e^{i\theta x} \quad \text{and} \quad \chi''(x) = -\theta^2 e^{i\theta x},$$

we conclude that

$$-\sigma^2 \theta^2 e^{i\theta x} + 2i\mu \theta e^{i\theta x} + 2\lambda^2 e^{i\theta x} = 0,$$

and so factoring out $e^{i\theta x}$ gives

$$-\sigma^2 \theta^2 + 2i\mu \theta + 2\lambda^2 = 0.$$

Thus,

$$-\lambda^2 = i\mu \theta - \frac{\sigma^2 \theta^2}{2}$$

so that substituting in for $\tau(t)$ gives

$$\tau(t) = \exp\left\{i\mu(T - t)\theta - \frac{\sigma^2(T - t)\theta^2}{2}\right\}$$

and so from (21.4) we conclude

$$\begin{aligned} f(t, x) &= \chi(x)\tau(t) = e^{i\theta x} \exp\left\{i\mu(T - t)\theta - \frac{\sigma^2(T - t)\theta^2}{2}\right\} \\ &= \exp\left\{i(x + \mu(T - t))\theta - \frac{\sigma^2(T - t)\theta^2}{2}\right\}. \end{aligned}$$

Taking $t = 0$ gives

$$\varphi_{X_T}(\theta) = f(0, x) = \exp\left\{i(x + \mu T)\theta - \frac{\sigma^2 T \theta^2}{2}\right\}$$

in agreement with Example 21.1.

Example 21.5. We will now use the Feynman-Kac representation theorem to derive the characteristic function for a process satisfying the Ornstein-Uhlenbeck SDE

$$dX_t = \sigma dB_t + aX_t dt$$

where σ , a , and $X_0 = x$ are constants. Let $u(x) = e^{i\theta x}$ so that the Feynman-Kac representation theorem implies

$$f(t, x) = \mathbb{E}[u(X_T)|X_t = x] = \mathbb{E}[e^{i\theta X_T}|X_t = x]$$

is the unique bounded solution of the partial differential equation

$$axf'(t, x) + \frac{1}{2}\sigma^2 f''(t, x) + \dot{f}(t, x) = 0, \quad 0 \leq t \leq T, \quad x \in \mathbb{R}, \quad (21.6)$$

subject to the terminal condition

$$f(T, x) = e^{i\theta x}, \quad x \in \mathbb{R}.$$

Note that $f(0, x) = \mathbb{E}[e^{i\theta X_T}|X_0 = x] = \varphi_{X_T}(\theta)$ is the characteristic function of X_T .

If we try to use separation of variables to solve (21.6), then we soon discover that it does not produce a solution. Thus, we are forced to conclude that the solution $f(t, x)$ is not separable and is necessarily more complicated. Guided by the form of the terminal condition, we guess that $f(t, x)$ can be written as

$$f(t, x) = \exp\{i\theta\alpha(t)x + \beta(t)\}, \quad 0 \leq t \leq T, \quad x \in \mathbb{R}, \quad (21.7)$$

for some functions $\alpha(t)$ and $\beta(t)$ of t only satisfying $\alpha(T) = 1$ and $\beta(T) = 0$. Differentiating we find

$$\begin{aligned} f'(t, x) &= i\theta\alpha(t) \exp\{i\theta\alpha(t)x + \beta(t)\} = i\theta\alpha(t)f(t, x), \\ f''(t, x) &= -\theta^2\alpha^2(t) \exp\{i\theta\alpha(t)x + \beta(t)\} = -\theta^2\alpha^2(t)f(t, x), \quad \text{and} \\ \dot{f}(t, x) &= [i\theta\alpha'(t)x + \beta'(t)] \exp\{i\theta\alpha(t)x + \beta(t)\} = [i\theta\alpha'(t)x + \beta'(t)]f(t, x) \end{aligned}$$

so that (21.6) implies

$$i\theta a x \alpha(t) f(t, x) - \frac{\sigma^2 \theta^2}{2} \alpha^2(t) f(t, x) + [i\theta \alpha'(t)x + \beta'(t)] f(t, x) = 0, \quad 0 \leq t \leq T, \quad x \in \mathbb{R}.$$

Factoring out the common $f(t, x)$ reduces the equation to

$$i\theta a x \alpha(t) - \frac{\sigma^2 \theta^2}{2} \alpha^2(t) + i\theta \alpha'(t)x + \beta'(t) = 0,$$

or equivalently,

$$i\theta [a\alpha(t) + \alpha'(t)]x + \beta'(t) - \frac{\sigma^2 \theta^2}{2} \alpha^2(t) = 0.$$

Since this equation must be true for all $0 \leq t \leq T$ and $x \in \mathbb{R}$, the only way that is possible is if the coefficient of x is zero *and* the constant term is 0. Thus, we must have

$$a\alpha(t) + \alpha'(t) = 0 \quad \text{and} \quad \beta'(t) - \frac{\sigma^2 \theta^2}{2} \alpha^2(t) = 0. \quad (21.8)$$

This first equation in (21.8) involves only $\alpha(t)$ and is easily solved. That is, $\alpha'(t) = -a\alpha(t)$ implies $\alpha(t) = Ce^{-at}$ for some arbitrary constant C . The terminal condition $\alpha(T) = 1$ implies that $C = e^{aT}$ so that

$$\alpha(t) = e^{a(T-t)}.$$

Since we have solved for $\alpha(t)$, we can now solve the second equation in (21.8); that is,

$$\beta'(t) = \frac{\sigma^2\theta^2}{2}\alpha^2(t) = \frac{\sigma^2\theta^2}{2}e^{2a(T-t)}.$$

We simply integrate from 0 to t to find $\beta(t)$:

$$\beta(t) - \beta(0) = \frac{\sigma^2\theta^2}{2} \int_0^t e^{2a(T-s)} ds = \frac{\sigma^2\theta^2}{4a} (e^{2aT} - e^{2a(T-t)}).$$

The terminal condition $\beta(T) = 0$ implies that

$$\beta(0) = \frac{\sigma^2\theta^2}{4a} (1 - e^{2aT})$$

and so

$$\beta(t) = \frac{\sigma^2\theta^2}{4a} (1 - e^{2aT}) + \frac{\sigma^2\theta^2}{4a} (e^{2aT} - e^{2a(T-t)}) = \frac{\sigma^2(1 - e^{2a(T-t)})\theta^2}{4a} = -\frac{\sigma^2(e^{2a(T-t)} - 1)\theta^2}{4a}.$$

Thus, from (21.7) we are now able to conclude that

$$f(t, x) = \exp\{i\theta\alpha(t)x + \beta(t)\} = \exp\left\{i\theta e^{a(T-t)}x - \frac{\sigma^2(e^{2a(T-t)} - 1)\theta^2}{4a}\right\}$$

for $0 \leq t \leq T$ and $x \in \mathbb{R}$. Taking $t = 0$ gives

$$\varphi_{X_T}(\theta) = f(0, x) = \exp\left\{i\theta e^{aT}x - \frac{\sigma^2(e^{2aT} - 1)\theta^2}{4a}\right\}$$

in agreement with Example 21.2.