

Exercise 5.3

Remember that

$$\begin{aligned} c(0, x) &= \tilde{\mathbb{E}} \left[e^{-rT} (S(T) - K)^+ \right] \\ &= \tilde{\mathbb{E}} \left[e^{-rT} \left(x e^{\sigma \tilde{W}(T) + \left(r - \frac{\sigma^2}{2}\right)T} - K \right)^+ \right] \end{aligned}$$

- (1) Let $h(s) = (s - K)^+$. Use formula for $h'(s)$ to obtain a formula for $c_x(0, x)$.
- (2) Rewrite this formula in form of

$$c_x(0, x) = \hat{\mathbb{P}}(S(T) > K),$$

where $\hat{\mathbb{P}}$ is a probability measure equivalent to $\tilde{\mathbb{P}}$. Show that $\hat{W}(t) = \tilde{W}(t) - \sigma t$ is a Brownian motion under $\hat{\mathbb{P}}$.

- (3) Conclude

$$\hat{\mathbb{P}}(S(T) > K) = \hat{\mathbb{P}} \left(-\frac{\hat{W}(T)}{\sqrt{T}} < d_+(T, x) \right) = N(d_+(T, x)).$$

Proof

- (1) We have that

$$c(0, x) = \tilde{\mathbb{E}} \left[\underbrace{e^{-rT} h(S(T))}_{:=g(x)} \right]$$

Derivation w.r.t. x and noting that $g(x) \geq 0$, we obtain that

$$c(0, x) = \tilde{\mathbb{E}}[g(x)] \Rightarrow c_x(0, x) = \tilde{\mathbb{E}} \left[\frac{\partial g(x)}{\partial x} \right]$$

Also,

$$\frac{\partial g(x)}{\partial x} = e^{\sigma \tilde{W}(T) - \frac{\sigma^2 T}{2}} \cdot \mathbf{1}_{\{S(T) > K\}}$$

- (2) Define $Z_\sigma(\omega) = \exp \left(\sigma \tilde{W}(T) - \frac{\sigma^2 T}{2} \right)$. Consider the probability measure given by

$$d\hat{\mathbb{P}}(\omega) = Z_\sigma(\omega) d\tilde{\mathbb{P}}(\omega).$$

Note that $\hat{W}(t) = \tilde{W}(t) - \sigma t$ is a Brownian motion under $\hat{\mathbb{P}}$. We have that

$$\hat{\mathbb{P}}(S(T) > K) = \tilde{\mathbb{E}} \left[\mathbf{1}_{\{S(T) > K\}} \cdot Z_\sigma(\omega) \right] = \tilde{\mathbb{E}} \left[\frac{\partial g(x)}{\partial x} \right] = c_x(0, x).$$

- (3) Recall that

$$d_+(T, x) = \frac{1}{\sigma \sqrt{T}} \left[\log \frac{x}{K} + \left(r + \frac{1}{2} \sigma^2\right) T \right].$$

Therefore,

$$\begin{aligned} S(T) > K &\iff x e^{\sigma \tilde{W}(T) + \left(r - \frac{\sigma^2}{2}\right)T} > K \\ &\iff \log \frac{x}{K} + \sigma \tilde{W}(T) + \left(r - \frac{\sigma^2}{2}\right)T > 0 \\ &\iff \log \frac{x}{K} + \sigma \hat{W}(T) + \left(r + \frac{\sigma^2}{2}\right)T > 0 \\ &\iff \sigma \hat{W}(T) + \sigma \sqrt{T} d_+(T, x) > 0 \\ &\iff d_+(T, x) > -\frac{\hat{W}(T)}{\sqrt{T}} \end{aligned}$$

Since $\hat{W}(t)$ is a Brownian motion under $\hat{\mathbb{P}}$, $-\frac{\hat{W}(T)}{\sqrt{T}}$ is a standard normal distribution under $\hat{\mathbb{P}}$.