

STOCHASTIC PROCESSES AND DERIVATIVES

Sheet 2

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Exercise 1 (Stochastic volatility). Assume that the market offers interest rate 0. At time $T > 0$, consider a stock whose value is $S_T = S_0 \exp\left(-\frac{1}{2}\tilde{\sigma}^2 T + \tilde{\sigma} W_T\right)$, where $S_0 > 0$, $(W_t)_{t \geq 0}$ is a Brownian motion, and $\tilde{\sigma}$ is a r.v. independent of W_T .

1. Let $K > 0$ be a fixed strike. Determine explicitly the deterministic function $h : \mathbb{R} \rightarrow \mathbb{R}$ s.t. $C_0 := \mathbb{E}[(S_T - K)^+] = \mathbb{E}[h(\tilde{\sigma}^2)]$.
2. Discuss the accuracy of the approximation $C_0 \approx h(\mathbb{E}[\tilde{\sigma}^2]) + \frac{1}{2} \text{Var}[\tilde{\sigma}^2] \partial_x^2 h(x)|_{x=\mathbb{E}[\tilde{\sigma}^2]}$.
3. Show that $C_0 < C^{BS} := h(\mathbb{E}[\tilde{\sigma}^2])$ around the money, and $C_0 > C^{BS}$ in/out the money. Paying particular to $\sigma_I = \sqrt{\mathbb{E}[\tilde{\sigma}^2]}$.

Proof. 1. We have,

$$\begin{aligned} C_0 &\triangleq \mathbb{E} \left[\left(S_0 e^{-\frac{1}{2}\tilde{\sigma}^2 T + \tilde{\sigma} W_T} - K \right)^+ \right] \\ &= \mathbb{E} \left[\mathbb{E} \left[\left(S_0 e^{-\frac{1}{2}\tilde{\sigma}^2 T + \tilde{\sigma} W_T} - K \right)^+ \middle| \tilde{\sigma}^2 \right] \right] \\ &= \mathbb{E} [Call^{BS}(T, K, S_0, \tilde{\sigma}^2)]. \end{aligned}$$

So we identify h :

$$\begin{aligned} h(x) &= S_0 \Phi(d_+(x)) - K \Phi(d_-(x)), \quad \text{with} \\ d_{\pm}(x) &= \frac{\ln\left(\frac{S_0}{K}\right)}{\sqrt{T x}} \pm \frac{1}{2} \sqrt{T x}. \end{aligned}$$

2. Let us do a Taylor expansion of second order of $h(\tilde{\sigma}^2)$ around $\mathbb{E}[\tilde{\sigma}^2]$,

$$h(\tilde{\sigma}^2) = h(\mathbb{E}[\tilde{\sigma}^2]) + (\tilde{\sigma}^2 - \mathbb{E}[\tilde{\sigma}^2]) \partial_x h(x)|_{x=\mathbb{E}[\tilde{\sigma}^2]} + \frac{1}{2} (\tilde{\sigma}^2 - \mathbb{E}[\tilde{\sigma}^2])^2 \partial_x^2 h(x)|_{x=\mathbb{E}[\tilde{\sigma}^2]} + \varepsilon.$$

Now taking the expectation we have

$$\begin{aligned} C_0 &\triangleq \mathbb{E} [h(\tilde{\sigma}^2)] \\ &= h(\mathbb{E}[\tilde{\sigma}^2]) + \frac{1}{2} \text{Var}[\tilde{\sigma}^2] \partial_x^2 h(x)|_{x=\mathbb{E}[\tilde{\sigma}^2]} + \varepsilon, \end{aligned}$$

where ε is an error. This approximation works in the case were $\tilde{\sigma}^2$ is stochastic but “not so much”, and so $\tilde{\sigma}^2$ is near form its expected value.

3. If we consider that the previous approximation is an equality we have $C_0 = C^{BS} = \frac{1}{2} \text{Var}[\tilde{\sigma}^2] \partial_x^2 h(x)|_{x=\mathbb{E}[\tilde{\sigma}^2]}$. Let us check the sign of the second order derivative of h .

$$\begin{aligned} \partial_x h(x) &= S_0 \phi(d_+(x)) d'_+(x) - K \phi(d_-(x)) d'_-(x) \\ &= \frac{K}{\sqrt{2\pi}} e^{-\frac{d_{\pm}^2(x)}{2}} \frac{1}{2} \sqrt{\frac{T}{x}}, \end{aligned}$$

we have used the fact that $S_0\phi(d_+) = K\phi(d_-)$. And then after calculations

$$\partial_x^2 h(x) = \frac{K}{2\sqrt{2\pi}} e^{-\frac{d^2(x)}{2}} \sqrt{\frac{T}{x}} \left(\frac{1}{2x^2 T} \left(\ln \frac{S_0}{K} \right)^2 - \frac{T}{8} - \frac{1}{2x} \right).$$

So at the money and around, $h''(x) < 0$ and $C_0 < C^{BS}$. And if $K \ll S$ or $K \gg S$ we have $C_0 > C^{BS}$. This shows that we have a smile in the implied volatility. \square

Exercise 2 (Implicit volatility expansion). 1. Let $\lambda \in \mathbb{C}$. Let $h(x) = (e^k - e^x)^+$. Show that $\hat{h}(\lambda)$, the generalized Fourier transform of h at λ with $\text{Im}(\lambda) < -1$, i.e.

$$\hat{h}(\lambda) := \int_{\mathbb{R}} h(x) e^{i\lambda x} dx$$

is given by $\frac{e^{k-i\lambda k}}{i\lambda + \lambda^2}$.

2. Show that

$$\begin{aligned} u(t, x, k, \sigma_0) &:= \frac{1}{2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{t\phi_0(\lambda, \sigma_0)} dRe(\lambda) \\ &= e^x \Phi(d_+) - e^k \Phi(d_-), \end{aligned}$$

where $\phi_0(\lambda, \sigma_0) := -\frac{1}{2}\sigma_0^2(\lambda^2 + i\lambda)$, $d_{\pm} := \frac{x-k}{\sigma_0\sqrt{t}} \pm \frac{1}{2}\sigma_0\sqrt{t}$, and Φ is the distribution function of the standard Gaussian.

3. Deduce that $\sigma \mapsto u(t, x, k, \sigma)$ is analytic. Write down the Taylor expansion of $u(t, x, k, \sigma_0 + \delta)$ around σ_0 .

4. Let $\varepsilon > 0$ and define $\phi^\varepsilon(t, \lambda, \sigma_0) := (1 - \varepsilon)t\phi_0(\lambda, \sigma_0) + \varepsilon\phi(t, \lambda)$, where $i\lambda x + \phi(t, \lambda) := \ln(\mathbb{E}_x[e^{i\lambda X_t}])$ for some process X_t . Next define

$$u^\varepsilon(t, x, k, \sigma_0) := \frac{1}{2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{i\lambda x} e^{\phi^\varepsilon(t, \lambda, \sigma_0)} dRe(\lambda).$$

Thus, $u^\varepsilon(t, x, k, \sigma_0)|_{\varepsilon=1} = u^X$, where u^X is the price of payoff h under the process X_t and $u^\varepsilon(t, x, k, \sigma_0)|_{\varepsilon=1} = u^{BS}$, the price under Black-Scholes formula. Deduce that

$$u^\varepsilon(t, x, k, \sigma_0) = \sum_{k=0}^{\infty} \varepsilon^k u_n(t, x, k, \sigma_0),$$

where

$$u_n(t, x, k, \sigma_0) := \frac{1}{n!2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{i\lambda x} e^{t\phi_0(\lambda, \sigma_0)} (\phi(t, \lambda) - t\phi_0(\lambda, \sigma_0))^n dRe(\lambda).$$

Proof. 1. We have

$$\begin{aligned} \hat{h}(\lambda) &\triangleq \int_{\mathbb{R}} (e^x - e^k)^+ e^{-i\lambda x} dx \\ &= \int_k^{\infty} (e^x - e^k) e^{-i\lambda x} dx \\ &= \left[\frac{1}{1-i\lambda} e^{x(1-i\lambda)} + \frac{1}{i\lambda} e^{k-i\lambda x} \right]_k^{\infty} \\ &= \frac{-e^{k-i\lambda k}}{i\lambda + \lambda^2}. \end{aligned}$$

2. We set $X_t := x - \frac{1}{2}\sigma_0^2 t + \sigma_0 W_t$, where W is a Brownian motion. Hence

$$\begin{aligned} e^x \Phi(d_+) - e^k \Phi(d_-) &\triangleq \mathbb{E}_x[h(X_t)] \\ &= \mathbb{E}_x \left[\frac{1}{2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{i\lambda X_t} dRe(\lambda) \right] \\ &= \frac{1}{2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{i\lambda x} \mathbb{E}_x[e^{-\frac{1}{2}i\lambda\sigma_0^2 t + i\lambda\sigma_0 W_t}] dRe(\lambda). \end{aligned}$$

As $W_t \sim \mathcal{N}(0, t)$ we have $\mathbb{E}_x[e^{-\frac{1}{2}i\lambda\sigma_0^2 t + i\lambda\sigma_0 W_t}] = e^{-\frac{1}{2}i\lambda\sigma_0^2 t + \frac{1}{2}\sigma_0^2 \lambda^2 t} = e^{t\phi_0(\lambda, \sigma_0)}$.

3. \hat{h} is analytic and ϕ_0 as well, so by Fubini $\sigma \mapsto u(t, x, k, \sigma)$ too.

$$u(t, x, k, \sigma_0 + \delta) = \sum_{n \geq 1} \frac{\delta^n}{n!} \frac{\partial^n u}{\partial \sigma^n} \Big|_{\sigma = \sigma_0},$$

$$\text{and } \partial_\sigma^n u|_{\sigma = \sigma_0} = \frac{1}{2\pi} \int_{\mathbb{R}} \hat{h}(\lambda) e^{i\lambda x} \partial_\sigma^n e^{t\phi_0(\lambda, \sigma)} \Big|_{\sigma = \sigma_0} dRe(\lambda).$$

□

Exercise 3 (Bachelier model). Assume the market offers interest rate 0. The Bachelier model for the evolution of a stock price is given by

$$S_t^B = S_0 + \sigma^B W_t,$$

for $t \geq 0$, where $S_0 \geq 0$ is the spot price, $\sigma^B > 0$, and $(W_t)_{t \geq 0}$ a Brownian motion.

1. Prove that under the Bachelier model, the price of a call option on the stock with strike K at expiry time T is given by

$$C_0^B := (S_0 - K) \Phi\left(\frac{S_0 - K}{\sigma^B \sqrt{T}}\right) + \sigma^B \sqrt{T} \phi\left(\frac{S_0 - K}{\sigma^B \sqrt{T}}\right).$$

2. Let $\sigma^B = S_0 \sigma^{BS}$. using the relation $e^y \geq 1 + y$, prove that at the money option prices given by the Bachelier and Black-Scholes models satisfy

$$0 \leq C_0^B - C_0^{BS} \leq \frac{S_0}{24\sqrt{2\pi}} (\sigma^{BS})^3 T^{\frac{3}{2}}.$$

Comment on the accuracy of the Bachelier formula as an estimator of the Black-Scholes formula for the option price.

3. Suppose that one knows the option price is C_0 (model independent). Show that the at the money implied volatility σ_I^B and σ_I^{BS} yielded respectively by the Bachelier and Black-Scholes models satisfy

$$0 \leq \sigma_I^{BS} - \frac{\sigma_I^B}{S_0} \leq \frac{(\sigma_I^{BS})^3 T}{24}.$$

Comment on the accuracy of the Bachelier implied volatility as an estimator for the Black-Scholes implied volatility.

Proof. 1.

$$\begin{aligned} C_0^B &\triangleq \mathbb{E}[(S_0 + \sigma^B W_T - K)^+] \\ &= \int_{-d}^{\infty} (S_0 + \sigma^B \sqrt{T} z - K) \phi(z) dz, \quad \text{where } d := \frac{S_0 - K}{\sigma^B \sqrt{T}} \\ &= (S_0 - K) \int_{-d}^{\infty} \phi(z) dz + \sigma^B \sqrt{T} \int_{-d}^{\infty} z \phi(z) dz \\ &= (S_0 - K)(1 - \Phi(-d)) + \sigma^B \sqrt{T} [-\phi(z)]_{-d}^{\infty} \\ &= (S_0 - K) \Phi(d) + \sigma^B \sqrt{T} \phi(d). \end{aligned}$$

2. At the money we have $C_0^B = \frac{\sigma^B \sqrt{T}}{\sqrt{2\pi}}$. And

$$\begin{aligned} C_0^{BS} &= \frac{S_0}{\sqrt{2\pi}} \int_{-\frac{\sigma^{BS} \sqrt{T}}{2}}^{\frac{\sigma^{BS} \sqrt{T}}{2}} e^{-\frac{x^2}{2}} dx \\ &\leq \frac{S_0}{\sqrt{2\pi}} \sigma^{BS} \sqrt{T} = C_0^B. \end{aligned}$$

On the other hand we have

$$\begin{aligned}
 C_0^{BS} &\geq \frac{S_0}{\sqrt{2\pi}} \int_{-\frac{\sigma^{BS}\sqrt{T}}{2}}^{\frac{\sigma^{BS}\sqrt{T}}{2}} 1 + \frac{x^2}{2} dx \\
 &\geq C_0^B - \frac{S_0}{\sqrt{2\pi}} \left[\frac{x^3}{6} \right]_{-\frac{\sigma^{BS}\sqrt{T}}{2}}^{\frac{\sigma^{BS}\sqrt{T}}{2}} \\
 &\geq C_0^B - \frac{S_0}{24\sqrt{2\pi}} \left(\sigma^{BS}\sqrt{T} \right)^3.
 \end{aligned}$$

Which gives us the control on the difference between the pricing of the two models. When the volatility is low, Bachelier is a good approximation of BS.

3. we have the following relations

$$\begin{aligned}
 C_0 &= \frac{\sigma_I^B \sqrt{T}}{\sqrt{2\pi}} \\
 &= \frac{S_0}{\sqrt{2\pi}} \int_{-\frac{\sigma_I^{BS}\sqrt{T}}{2}}^{\frac{\sigma_I^{BS}\sqrt{T}}{2}} e^{-\frac{x^2}{2}} dx.
 \end{aligned}$$

Then we use the same inequalities as in the previous question to deduce the relation.

□