

MTH 9821 - L6

2017/10/12

Interview: ① Tell ~~you~~ sth. about yourself: sth. relevant to the job requirement
"... make me prepared for ...".

② Tell me what you're proud of: sth. that makes you really good.

For EY: Not much brain teasers.

Finite Difference Hedging and Valuation for European & American Options.

$V(S,t)$ satisfies the B-S PDE.

$$\frac{\partial V}{\partial t} + \frac{\sigma^2 S^2}{2} \frac{\partial^2 V}{\partial S^2} + (r-q) \cdot S \cdot \frac{\partial V}{\partial S} - rV = 0, \quad \forall S > 0, \quad \forall 0 < t < T.$$

Boundary condition: $V(S,T) = \text{option payoff}$: puts $\max(K-S, 0)$, call: $\max(S-K, 0)$

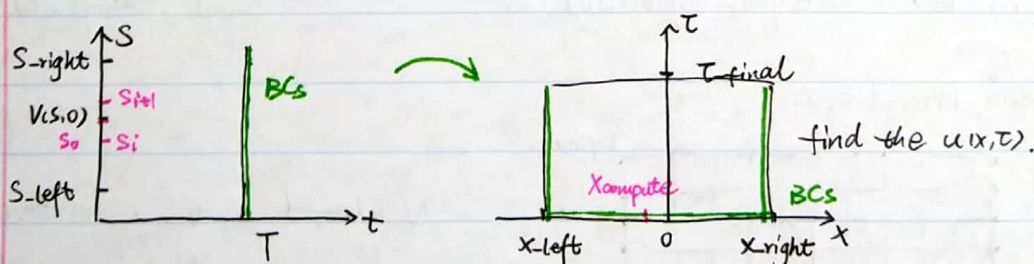
Change of variable:

$$x = \ln\left(\frac{S}{K}\right), \quad \tau = \frac{(T-t)\sigma^2}{2}, \quad V(S,t) = \exp(-ax-b\tau) u(x,\tau).$$

$$a = \frac{r-q}{\sigma^2} - \frac{1}{2}, \quad b = \left(\frac{r-q}{\sigma^2} + \frac{1}{2}\right)^2 + \frac{2q}{\sigma^2}.$$

$$\frac{\partial u}{\partial \tau} = \frac{\partial^2 u}{\partial x^2}, \quad \forall x \in \mathbb{R}, \quad \forall 0 < \tau < \tau_{\text{final}}, \quad \tau_{\text{final}} = \frac{\sigma^2 T}{2}$$

$$\text{Boundary conditions: } \begin{cases} Ke^{ax} \cdot \max(1-e^x, 0), \text{ puts} \\ Ke^{ax} \cdot \max(e^x-1, 0), \text{ calls} \end{cases} = u(x, 0).$$



$$S(t) = S(0) \cdot \exp\left[\left(r-q-\frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}Z\right].$$

$$S_{\text{left}} = S(0) \cdot \exp\left[\left(r-q-\frac{\sigma^2}{2}\right)T - 3\sigma\sqrt{T}\right] \Rightarrow x_{\text{left}} = \ln\left[\frac{S(0)}{K}\right] + \left(r-q-\frac{\sigma^2}{2}\right)T - 3\sigma\sqrt{T}$$

$$S_{\text{right}} = S(0) \cdot \exp\left[\left(r-q-\frac{\sigma^2}{2}\right)T + 3\sigma\sqrt{T}\right] \Rightarrow x_{\text{right}} = \ln\left[\frac{S(0)}{K}\right] + \left(r-q-\frac{\sigma^2}{2}\right)T + 3\sigma\sqrt{T}.$$

Put options Boundary Condition:

$$u(x, 0) = Ke^{ax} \max(1 - e^x, 0), \text{ for } x \in \mathbb{R}.$$

$$\text{As } s \rightarrow \infty: P(s, t) \rightarrow 0.$$

$$\text{As } s \rightarrow 0: P(s, t) \rightarrow ke^{-r(T-t)} Se^{-q(T-t)}.$$

$$[\text{Put-Call parity: } C(t) - P(t) = Se^{-q(T-t)} - ke^{-r(T-t)}]$$

$$u(x_{\text{left}}, \tau) = g_{\text{left}}(\tau) \quad u(x_{\text{right}}, \tau) = g_{\text{right}}(\tau) \quad [\text{Want}]$$

$$u(x_{\text{left}}, \tau) = [ke^{-r\frac{2\tau}{\sigma^2}} - Ke^{x_{\text{left}}} e^{-q\frac{2\tau}{\sigma^2}}] \exp(ax_{\text{left}} + b\tau).$$

$$(T-t = \frac{2\tau}{\sigma^2}, \quad S_{\text{left}} = Ke^{x_{\text{left}}}).$$

$$g_{\text{left}}(\tau) = K \cdot \exp(ax_{\text{left}} + b\tau) [e^{-\frac{2r\tau}{\sigma^2}} - e^{x_{\text{left}} - \frac{2q\tau}{\sigma^2}}]; \quad g_{\text{right}}(\tau) = 0.$$

Call options Boundary Condition:

$$u(x, 0) = Ke^{ax} \max(e^x - 1, 0), \text{ for } x \in \mathbb{R}.$$

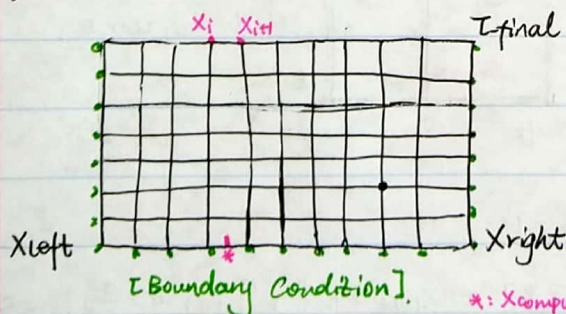
$$\text{As } s \rightarrow 0: C(s, t) \rightarrow 0 \Rightarrow C(s_{\text{left}}, t) = 0 \Rightarrow u(x_{\text{left}}, \tau) = 0 \Rightarrow g_{\text{left}}(\tau) = 0.$$

$$\text{As } s \rightarrow \infty: C(s, t) \simeq S_{\text{right}} e^{-q(T-t)} - ke^{-r(T-t)}$$

$$= \exp(-ax_{\text{right}} - b\tau) u(x_{\text{right}}, \tau).$$

$$g_{\text{right}}(\tau) = K \cdot \exp(ax_{\text{right}} + b\tau) [e^{x_{\text{right}} - \frac{2q\tau}{\sigma^2}} - e^{-\frac{2r\tau}{\sigma^2}}]; \quad g_{\text{left}}(\tau) = 0.$$

Domain Discretization.



N intervals in the x direction.

M intervals in the y direction.

$$\Delta x = \frac{x_{\text{right}} - x_{\text{left}}}{N}, \quad \Delta \tau = \frac{\tau_{\text{final}}}{M}.$$

$$\alpha = \Delta \tau / (\Delta x)^2 \quad \text{constant.}$$

$$\text{Start with } \alpha_{\text{temp}} \text{ and } M, \quad \Delta \tau = \frac{\tau_{\text{final}}}{M}.$$

$$\Delta x_{\text{temp}} = \sqrt{\frac{\Delta \tau}{\alpha_{\text{temp}}}}.$$

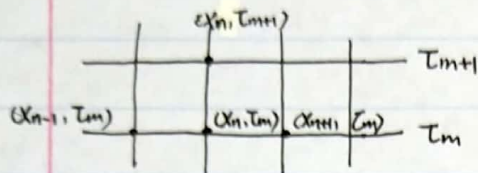
$$N = \text{floor} \left(\frac{x_{\text{right}} - x_{\text{left}}}{\Delta x_{\text{temp}}} \right), \quad \Delta x = \frac{x_{\text{right}} - x_{\text{left}}}{N}.$$

$$\alpha = \frac{\Delta \tau}{(\Delta x)^2}, \quad \text{Note } \alpha \leq \alpha_{\text{temp}}.$$

$$(x_n, \tau_m): \quad x_n = x_{\text{left}} + n \Delta x \quad (n = 0:N), \quad \tau_m = m \cdot \Delta \tau \quad (m \text{ in } 0:M).$$

Finite Difference approximation of the heat PDE.

$$\frac{\partial u}{\partial \tau}(x_i, \tau_m) = \frac{\partial^2 u}{\partial x^2}(x_i, \tau_m), \quad \forall x_{\text{left}} < x < x_{\text{right}}, \quad \forall 0 < \tau < \tau_{\text{final}}$$



[Forward Euler. explicit FD method].

$$\text{Forward FD approx: } \frac{\partial u}{\partial \tau}(x_n, \tau_m) = \frac{u(x_n, \tau_{m+1}) - u(x_n, \tau_m)}{\delta \tau} + O(\delta \tau).$$

$$\text{Central FD approx: } \frac{\partial^2 u}{\partial x^2}(x_n, \tau_m) = \frac{u(x_{n+1}, \tau_m) - 2u(x_n, \tau_m) + u(x_{n-1}, \tau_m))}{(\delta x)^2} + O((\delta x)^2).$$

$$\frac{u(x_n, \tau_{m+1}) - u(x_n, \tau_m)}{\delta \tau} + O(\delta \tau) = \frac{u(x_{n+1}, \tau_m) - 2u(x_n, \tau_m) + u(x_{n-1}, \tau_m))}{(\delta x)^2} + O((\delta x)^2).$$

$$\frac{u_n^{m+1} - u_n^m}{\delta \tau} = \frac{u_{n+1}^m - 2u_n^m + u_{n-1}^m}{(\delta x)^2}, \quad \text{keep } \alpha = \frac{\delta \tau}{(\delta x)^2} \text{ constant. Courant}$$

$$u_n^{m+1} - u_n^m = \alpha (u_{n+1}^m - 2u_n^m + u_{n-1}^m).$$

$$u_n^{m+1} = \alpha u_{n+1}^m + (1-2\alpha) u_n^m + \alpha u_{n-1}^m, \quad \forall n=1:(N-1).$$

For $m=0:(M-1)$, \downarrow , end.

For $m=0:(M-1)$:

$$U^{m+1} = A \cdot U^m + \begin{pmatrix} \alpha u_0^m \\ 0 \\ \vdots \\ 0 \\ \alpha u_N^m \end{pmatrix} \Bigg\}_{N \rightarrow} \quad A = \begin{bmatrix} 1-2\alpha & \alpha & & 0 \\ \alpha & \ddots & \ddots & \\ & \ddots & \ddots & \alpha \\ 0 & & \alpha & 1-2\alpha \end{bmatrix}$$

End.

$$U^m = \begin{pmatrix} u_1^m \\ \vdots \\ u_{N-1}^m \end{pmatrix} \quad \begin{matrix} u_0^m = g_{\text{left}}(m \delta \tau) \\ u_N^m = g_{\text{right}}(m \delta \tau) \end{matrix} \quad U^{m+1} = A \cdot U^m + \alpha \begin{pmatrix} g_{\text{left}}(m \delta \tau) \\ 0 \\ \vdots \\ 0 \\ g_{\text{right}}(m \delta \tau) \end{pmatrix}$$

Then, go back to the original domain. $x_{\text{compute}} = \ln(\frac{S_0}{K})$.

Let x_i and x_{i+1} such that $x_i \in x_{\text{compute}} < x_{i+1}$ [$S_i = K e^{x_i}$; $S_{i+1} = K e^{x_{i+1}}$].

Found u_i and u_{i+1} finite difference approximate solution for

$$u(x_i, \tau_{\text{final}}) \text{ and } u(x_{i+1}, \tau_{\text{final}}).$$

$$\text{approx of } V(S_i, 0) \leftarrow V_i = \exp(-\alpha x_i - b \tau_{\text{final}}) u_i^M.$$

$$\text{approx of } V(S_{i+1}, 0) \leftarrow V_{i+1} = \exp(-\alpha x_{i+1} - b \tau_{\text{final}}) u_{i+1}^M.$$

$$V_{\text{approx}}(S_0, 0) = \frac{V_i(S_{i+1} - S_0) + V_{i+1}(S_0 - S_i)}{S_{i+1} - S_i}$$

$$[\text{Linear Interpolation } f(x) = \frac{x-a}{b-a} f(b) + \frac{b-x}{b-a} f(a)]$$

For Greeks Computation:

$$\Delta_{\text{approx}}(S_0, 0) = \frac{V_{i+1} - V_i}{S_{i+1} - S_i} \quad (\Delta f_d)$$

p.s.: The domain discretization is uniform in x , but not in S .

$$P_{fd} = \frac{\frac{V_{i+2} - V_{i+1}}{S_{i+2} - S_{i+1}} - \frac{V_i - V_{i-1}}{S_i - S_{i-1}}}{\frac{S_{i+2} + S_{i+1}}{2} - \frac{S_i + S_{i-1}}{2}}$$

For American Put Option:

$$V(s, t) \geq K - S, \quad \forall S > 0, \quad \forall 0 < t < T. \quad [K - S = K(1 - e^x)]$$

$$V(s, t) = \exp(-ax - b\tau) u(x, \tau). \quad u(x, \tau) \geq K \exp(ax + b\tau) (1 - e^x).$$

for $m = 0 : M-1$;

$$U^{m+1} = \max \left(AU^m + \begin{pmatrix} \alpha U_0^m \\ 0 \\ \vdots \\ 0 \\ \alpha U_N^m \end{pmatrix}, (\text{early } \alpha\text{-premium } (x_n, \tau_{m+1}))_{m=0:N-1} \right)$$

End.

Implied Volatility:

Given V_{market} , find σ such that $V_{\text{numerical}}(\sigma) = V_{\text{market}}$.

$$f(\sigma) = V_{\text{numerical}}(\sigma) - V_{\text{market}}.$$

Start with $\sigma_0, 0$, until convergence

$$\sigma_{n+1} = \sigma_n - \frac{f(\sigma_n)(\sigma_n - \sigma_{n-1})}{f(\sigma_n) - f(\sigma_{n-1})}$$

End