

Numerical HW 4 Yun Sheng YS3171

(Q2.a) the value of option by PDE method, with θ set to 0, 0.5 and 1 respectively (Without k).

spot	120	spot	120	spot	120
USD zero rat	0.02	USD zero rat	0.02	USD zero rat	0.02
(domestic) JPY zero rate	0.005	(domestic) JPY zero rate	0.005	(domestic) JPY zero rate	0.005
strike	119	strike	119	strike	119
vol	0.1	vol	0.1	vol	0.1
time to expir	7	time to expir	7	time to expir	7
m	20	m	20	m	20
gamma max	24	gamma max	24	gamma max	24
method	Explicit	method	Implicit	method	Crack Nicolson
theta	0	theta	1	theta	0.5
x0	-0.05	x0	-0.05	x0	-0.05
x10	0.0083682	x10	0.0083682	x10	0.0083682
delta x	0.0058368	delta x	0.0058368	delta x	0.0058368
x20	0.0667365	x20	0.0667365	x20	0.0667365
tau max	9.589E-05	tau max	9.589E-05	tau max	9.589E-05
delta tau	3.995E-06	delta tau	3.995E-06	delta tau	3.995E-06
lambda	0.1172764	lambda	0.1172764	lambda	0.1172764
S min	113.1963	S min	113.1963	S min	113.1963
S max	127.21264	S max	127.21264	S max	127.21264
qdelta	-3	qdelta	-3	qdelta	-3
q	1	q	1	q	1
pde price	1.2585934	pde price	1.2695714	pde price	1.2629209
bs price	1.2520173	bs price	1.2520173	bs price	1.2520173
discrepancy	0.0065762	discrepancy	0.0175542	discrepancy	0.0109037

b)

theta=0	-0.05	-0.04	-0.03	-0.02	-0.01
x0	1.25859342	1.29659004	1.74613487	5.14399049	-1.954E+12
pde	0.11727639	0.17078243	0.27140632	0.49647668	1.18420909

We note when $x_0 = -0.01$, pde value is meaningless, and $\lambda = 1.18 > 0.5$ in this case.
When $\lambda \leq 0.5$, we see for other x_0 , it produce some meaningful value.
Therefore this coherent with explicit method is stable when $\lambda \leq 0.5$.

c)

theta=0.5	-0.05	-0.04	-0.03	-0.02	-0.01
x0	1.26292094	1.33384632	1.92110258	5.69517187	21.9115686
pde	0.11727639	0.17078243	0.27140632	0.49647668	1.18420909

When $\theta = 0.5$, it employs crack nicolson method which is unconditionally stable.
Therefore all x_0 and λ produce meaningful stable results.

$$(Q3a) \quad G\vec{v}^{(k)} = \begin{pmatrix} \alpha & \beta & & \\ \gamma & \alpha & \ddots & \\ & \ddots & \ddots & \beta \\ & & \gamma & \alpha \end{pmatrix} \times \begin{pmatrix} \sqrt{\frac{x}{\beta}} \sin \frac{kx}{N+1} \\ \vdots \\ \sqrt{\frac{x}{\beta}} \sin \frac{kx}{N+1} \end{pmatrix} = \begin{pmatrix} \alpha \sqrt{\frac{x}{\beta}} \sin \frac{kx}{N+1} + \beta \sqrt{\frac{x}{\beta}} \sin \frac{(k+1)x}{N+1} \\ \vdots \\ \gamma \times \sqrt{\frac{x}{\beta}} \sin \frac{(N-1)x}{N+1} + \beta \sqrt{\frac{x}{\beta}} \sin \frac{Nx}{N+1} \end{pmatrix} = \begin{pmatrix} \gamma \sqrt{\frac{x}{\beta}} \sin \frac{kx}{N+1} + \alpha \sqrt{\frac{x}{\beta}} \sin \frac{(k+1)x}{N+1} + \beta \sqrt{\frac{x}{\beta}} \sin \frac{2(k+1)x}{N+1} \\ \vdots \\ \gamma \sqrt{\frac{x}{\beta}} \sin \frac{(N-1)x}{N+1} + \alpha \sqrt{\frac{x}{\beta}} \sin \frac{Nx}{N+1} + \beta \sqrt{\frac{x}{\beta}} \sin \frac{(N+1)x}{N+1} \end{pmatrix}$$

$$\text{Then } G\vec{v}^{(k)}_{ci} = \gamma \times \sqrt{\frac{x}{\beta}} \sin \frac{(i-1)x}{N+1} + \alpha \sqrt{\frac{x}{\beta}} \sin \frac{ix}{N+1} + \beta \sqrt{\frac{x}{\beta}} \sin \frac{(i+1)x}{N+1} \text{ for } i=1 \dots N.$$

$$\begin{aligned} M_k \vec{v}^{(k)\top}_{ci} &= (\alpha + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1}) \left(\sqrt{\frac{x}{\beta}} \sin \frac{ix}{N+1} \right) \\ &= \alpha \sqrt{\frac{x}{\beta}} \sin \frac{ikx}{N+1} + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1} \cdot \sin \frac{ikx}{N+1} \\ &= \alpha \sqrt{\frac{x}{\beta}} \sin \frac{ikx}{N+1} + \sqrt{\frac{x}{\beta}}^{i+1} \cdot \beta \cdot \left(\sin \frac{(i-1)x}{N+1} + \sin \frac{(i+1)x}{N+1} \right) \\ &= \gamma \sqrt{\frac{x}{\beta}}^{i-1} \sin \frac{(i-1)x}{N+1} + \alpha \sqrt{\frac{x}{\beta}}^i \sin \frac{ix}{N+1} + \beta \sqrt{\frac{x}{\beta}}^{i+1} \sin \frac{(i+1)x}{N+1} \end{aligned}$$

Therefore we have proved $M_k \vec{v}^{(k)\top}_{ci} = (G\vec{v}^{(k)}_{ci})^\top \Rightarrow M_k \vec{v}^{(k)\top} = G\vec{v}^{(k)\top}$
 Thus $\vec{v}^{(k)\top}$ is an eigenvector of G for the eigenvalue M_k .

$$b) \quad (G - \lambda I) = 0 \text{ when } \beta = 0. \text{ then } |G - \lambda I| = (\alpha - \lambda)^N = 0$$

$$\text{Thus } \lambda = \alpha.$$

If $\beta = 0$ then any non zero vector $X = \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix}$ satisfy $GX = \lambda X$.

If $\beta \neq 0$, then $(G - \alpha I) \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} = 0$. thus $\begin{cases} x_1 = 0 \\ x_2 = 0 \\ \vdots \\ x_N = 0 \end{cases}$

then only vector $X = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ x_N \end{pmatrix}$ satisfy $GX = \lambda X$ for any $x_N \in \mathbb{R}$

Thus eigenvector will be $\begin{cases} \begin{pmatrix} x_1 \\ \vdots \\ x_N \end{pmatrix} \text{ if } \beta = 0 \\ \begin{pmatrix} 0 \\ \vdots \\ 0 \\ x_N \end{pmatrix} \text{ if } \beta \neq 0. \end{cases}$

c) To make G singular we want $|G| = 0$. that is G has an eigenvalue 0.

① If $\beta \neq 0$ and $x_i = 0$ then $|G| = \alpha^N = 0$ then $\alpha = 0$

② If $\beta \neq 0$, then $M_k = \alpha + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1} = 0$ for some $k \in \{1, \dots, N\}$.

③ If $\beta \neq 0$, $M_k = \alpha + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1}$ then $\alpha = 0$ and $2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1} = 0$. in order to have $\cos \frac{kx}{N+1} = 0$. N is odd.

thus for N is even, set is $\{ \beta \neq 0, \alpha = 0 \} \cup \{ \beta \neq 0, \exists k \in \{1, \dots, N\} \text{ s.t. } M_k = \alpha + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1} = 0 \}$

for N is odd, set is $\{ \beta \neq 0, \alpha = 0 \} \cup \{ \beta \neq 0, \exists k \in \{1, \dots, N\} \text{ s.t. } M_k = \alpha + 2\beta \sqrt{\frac{x}{\beta}} \cos \frac{kx}{N+1} = 0 \} \cup \{ \beta \neq 0, \alpha = 0 \}$

$$Q4(a) \quad \frac{\partial V}{\partial X} = \frac{\partial V}{\partial S} \cdot \frac{\partial S}{\partial X} = \frac{\partial V}{\partial S} \cdot k \ell^X = \frac{\partial V}{\partial S} \cdot S$$

$$\frac{\partial^2 V}{\partial X^2} = \left(\frac{\partial V}{\partial S} \cdot S \right) \frac{\partial}{\partial X} = \left(\frac{\partial V}{\partial S} \right) \frac{\partial V}{\partial S} S + \frac{\partial V}{\partial S} \cdot S = \frac{\partial^2 V}{\partial S^2} S^2 + \frac{\partial V}{\partial S} \cdot S$$

$$(b) \quad \frac{\partial V(X, T)}{\partial T} = \beta k e^{\alpha X + \beta T} y(X, T) + k e^{\alpha X + \beta T} \frac{\partial y(X, T)}{\partial T}$$

$$\frac{\partial V(X, T)}{\partial X} = \alpha k e^{\alpha X + \beta T} y(X, T) + k e^{\alpha X + \beta T} \frac{\partial y(X, T)}{\partial X}$$

$$\frac{\partial^2 V(X, T)}{\partial X^2} = \alpha^2 k e^{\alpha X + \beta T} y(X, T) + 2\alpha k e^{\alpha X + \beta T} \frac{\partial y(X, T)}{\partial X} + k e^{\alpha X + \beta T} \frac{\partial^2 y(X, T)}{\partial X^2}$$

Substitute into equation and divided by $k e^{\alpha X + \beta T}$ on both side

$$\beta y + \frac{\partial y}{\partial T} = \alpha^2 y + 2\alpha \frac{\partial y}{\partial X} + \frac{\partial^2 y}{\partial X^2} + \alpha b y + b \frac{\partial y}{\partial X} + c y.$$

$$\text{Thus } \frac{\partial y}{\partial T} = (\alpha^2 + ab + (-\beta))y + (2\alpha + b) \frac{\partial y}{\partial X} + \frac{\partial^2 y}{\partial X^2}$$

$$Q4(C) \quad C(S, T) = \max\{S - K, 0\} = \max\{Ke^{-x} - K, 0\}$$

Also, $C(S, T) = V(X, 0) = Ke^{-\frac{x}{2}(q^s-1)} y(X, 0) \Leftrightarrow \alpha = -\frac{q^s-1}{2}$

$$y(X, 0) = \frac{\max\{Ke^{-x} - K, 0\}}{Ke^{-\frac{x}{2}(q^s-1)}} = \max\left\{e^{\frac{x}{2}(q^s+1)} - e^{\frac{x}{2}(q^s-1)}, 0\right\}$$

$$(d) \quad P(S, T) = \max\{S - K, 0\} = \max\{K - Ke^{-x}, 0\}$$

$$P(S, T) = V(X, 0) = Ke^{-\frac{x}{2}(q^s-1)} y(X, 0)$$

$$y(X, 0) = \frac{\max\{K - Ke^{-x}, 0\}}{Ke^{-\frac{x}{2}(q^s-1)}} = \max\left\{e^{\frac{x}{2}(q^s-1)} - e^{\frac{x}{2}(q^s+1)}, 0\right\}$$

forward = Call - Put = $\max\left\{e^{\frac{x}{2}(q^s+1)} - e^{\frac{x}{2}(q^s-1)}\right\} - \max\left\{e^{\frac{x}{2}(q^s-1)} - e^{\frac{x}{2}(q^s+1)}\right\}$

$$= e^{\frac{x}{2}(q^s+1)} - e^{\frac{x}{2}(q^s-1)}$$

$$Q5. \quad -\lambda \theta w_{i-1, r+1} + (1+2\lambda\theta)w_{i, r+1} - \lambda \theta w_{i+1, r+1} = \lambda(1-\theta)w_{i-1, r} + (-2\lambda(1-\theta))w_{i, r} + \lambda(1-\theta)w_{i+1, r}$$

when $i = 1$

$$(1+2\lambda\theta)w_{1, r+1} - \lambda \theta w_{2, r+1} = \lambda(-2\lambda(1-\theta))w_{1, r} + \lambda(1-\theta)w_{2, r} + \lambda(1-\theta)w_{0, r+1}$$

By $A \stackrel{(r+1)}{\rightarrow} B \stackrel{(r)}{\rightarrow} d \stackrel{(r)}{\rightarrow}$, thus $r_i(a, T_r) = \lambda(1-\theta)w_{0, r} + \lambda\theta w_{1, r+1}$

when $i = N-1$.

$$-\lambda \theta w_{N-1, r+1} + (1+2\lambda\theta)w_{N, r+1} = \lambda(1-\theta)w_{N-2, r} + (-2\lambda(1-\theta))w_{N-1, r} + \lambda(1-\theta)w_{N, r} + \lambda\theta w_{N, r+1}$$

$$\text{Thus } r_2(b, T_r) = \lambda(1-\theta)w_{N, r} + \lambda\theta w_{N, r+1}$$

for $i = 2 \text{ to } N-2$. $d_i, r = 0$.

$$d^{(r)} = \begin{pmatrix} r_1(a, T_r) \\ 0 \\ \vdots \\ r_2(b, T_r) \end{pmatrix} \quad \text{and} \quad r_i(a, T_r) = \begin{cases} 0 & \text{Call} \\ e^{\frac{x}{2}(q^s-1) + \frac{\tau}{q}(q^s-1)^2} & \text{Put} \\ e^{\frac{x}{2}(q^s+1) + \frac{\tau}{q}(q^s+1)^2} & \end{cases}$$

$$\vec{A}^{(r)} = \begin{cases} \begin{pmatrix} 0 & \dots & 0 & \lambda\theta w_{N, r_{\max}} + \lambda(1-\theta)w_{N, r_{\max}-1} \\ 0 & \dots & 0 & \lambda\theta w_{N, r_{\max}-1} + \lambda(1-\theta)w_{N, r_{\max}-2} \\ \vdots & \vdots & \vdots & \\ 0 & \dots & 0 & \lambda\theta w_{N, 1} + \lambda(1-\theta)w_{N, 0} \end{pmatrix} & \text{for European call} \\ \begin{pmatrix} \lambda\theta w_{0, r_{\max}} + \lambda(1-\theta)w_{0, r_{\max}-1} & 0 & \dots & 0 \\ \lambda\theta w_{0, r_{\max}-1} + \lambda(1-\theta)w_{0, r_{\max}-2} & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \\ \lambda\theta w_{0, 1} + \lambda(1-\theta)w_{0, 0} & 0 & \dots & 0 \end{pmatrix} & \text{for European put} \end{cases}$$