

數值偏微分法

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授課大綱

- 回顧
 - 微分方程和商品評價的關係
 - Black Scholes PDE
- 偏微分方程求解
 - Explicit method:和樹狀結構相似
 - Stable conditions
 - 變數變換
 - Implicit method
 - Crank-Nicolson Method:前兩者的結合

使用Ito's formula

推導衍生性商品價格P的隨機過程

- By Black-Scholes assumption $dS = (\mu - q)Sdt + \sigma Sdz$ where dz is a standard Brownian motion.
- By Ito lemma

$$dP = \frac{\partial P}{\partial t} dt + \frac{\partial P}{\partial S} dS + \frac{1}{2} \frac{\partial^2 P}{\partial S^2} (dS)^2$$

$$= [(\mu - q)S \frac{\partial P}{\partial S} + \frac{\partial P}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 P}{\partial S^2}] dt + \sigma S \frac{\partial P}{\partial S} dz$$

Black-Scholes Differential Equation

$$\begin{cases} dS = (\mu - q)Sdt + \sigma Sdz \\ dP = \left((\mu - q)S \frac{\partial P}{\partial S} + \frac{\partial P}{\partial t} + \frac{\sigma^2 S^2}{2} \frac{\partial^2 P}{\partial S^2} \right) dt + \sigma S \frac{\partial P}{\partial S} dz \end{cases}$$

$$\rightarrow dP - \frac{\partial P}{\partial S} dS = \left(\frac{\partial P}{\partial t} + \frac{\sigma^2 S^2}{2} \frac{\partial^2 P}{\partial S^2} \right) dt$$

short: 1 derivative security

long: $\frac{\partial P}{\partial S}$ share underlying asset

the value of this portfolio $\Pi = -P + S \frac{\partial P}{\partial S}$

$$d\Pi = -dP + \frac{\partial P}{\partial S} dS = \left(-\frac{\partial P}{\partial t} - \frac{\sigma^2 S^2}{2} \frac{\partial^2 P}{\partial S^2} \right) dt$$

$$d\Pi + qS \frac{\partial P}{\partial S} dt \rightarrow \left(-\frac{\partial P}{\partial t} - \frac{\sigma^2 S^2}{2} \frac{\partial^2 P}{\partial S^2} \right) dt + qS \frac{\partial P}{\partial S} dt = r\Pi dt$$

$$\frac{1}{2} \sigma^2 S^2 \frac{\partial^2 P}{\partial S^2} + (r - q)S \frac{\partial P}{\partial S} - rP + \frac{\partial P}{\partial t} = 0$$

選擇權價格滿足此偏微分方程⁴

美式賣權的PDE及邊界條件

The Black-Scholes differential equation for American puts is

$$\frac{1}{2}\sigma^2 S^2 \frac{\partial^2 P}{\partial S^2} + (r - q)S \frac{\partial P}{\partial S} - rP + \frac{\partial P}{\partial t} = 0$$

where $P(S, T) = \max(X - S, 0)$ and

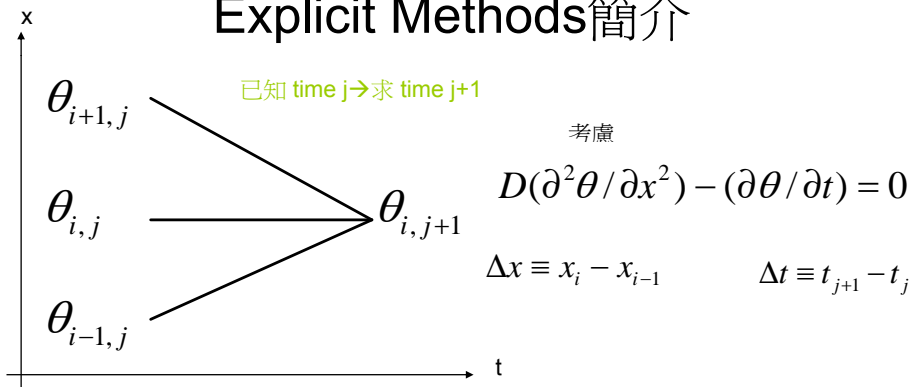
$P(S, t) = \max(\bar{P}(S, t), X - S)$ for $t < T$.

\bar{P} denotes the option value at time t if it is not exercised for the next instant of time.

使用數值方法解偏微分方程

- 並非所有的偏微分方程都有公式解
- 對於無法求公式解的方程,可用數值方法逼近
 - Explicit method: 和樹狀結構評價法接近
 - Implicit method: 解方程式
 - Crank-Nicolson Method: 上述兩種方法的混合

Explicit Methods簡介



$$\frac{\partial \theta(x, t)}{\partial t} \Big|_{t=t_j} = \frac{\theta(x_i, t_{j+1}) - \theta(x_i, t_j)}{\Delta t} + O(\Delta t)$$

$$\frac{\partial^2 \theta(x, t)}{\partial x^2} \Big|_{x=x_i} = \frac{\theta(x_{i+1}, t_j) - 2\theta(x_i, t_j) + \theta(x_{i-1}, t_j))}{(\Delta x)^2} + O[(\Delta x)^2]$$

代入 $D(\partial^2 \theta / \partial x^2) - (\partial \theta / \partial t) = 0$

$$\frac{\theta_{i,j+1} - \theta_{i,j}}{\Delta t} = D \frac{\theta_{i+1,j} - 2\theta_{i,j} + \theta_{i-1,j}}{(\Delta x)^2}$$

Rearrange as

$$\theta_{i,j+1} = \frac{D\Delta t}{(\Delta x)^2} \theta_{i+1,j} + [1 - \frac{2D\Delta t}{(\Delta x)^2}] \theta_{i,j} + \frac{D\Delta t}{(\Delta x)^2} \theta_{i-1,j}$$

The explicit method is numerically unstable unless $\Delta t \leq (\Delta x)^2 / (2D)$

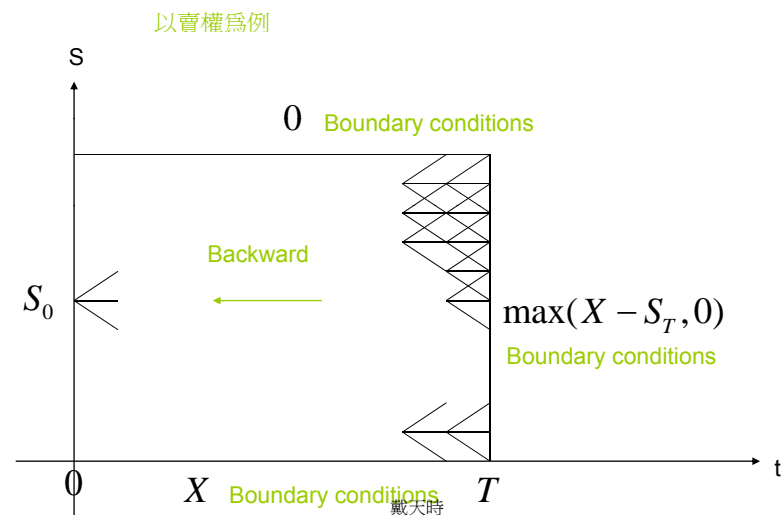
$$\begin{array}{l} \theta_{i+1,j} \frac{D\Delta t}{(\Delta x)^2} \\ \theta_{i,j} \frac{1 - \frac{2D\Delta t}{(\Delta x)^2}}{(\Delta x)^2} \\ \theta_{i-1,j} \frac{D\Delta t}{(\Delta x)^2} \end{array} \rightarrow \theta_{i,j+1}$$

When the stability condition is satisfied, the coefficients for $\theta_{i+1,j}$, $\theta_{i,j}$, and $\theta_{i-1,j}$ all lie between zero and one and sum to one. They can therefore be interpreted as probabilities.

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9

使用Explicit Method 評價選擇權



10

改寫Black Scholes PDE

The Black-Scholes differential equation is

$$\frac{1}{2} \sigma^2 S^2 \frac{\partial^2 P}{\partial S^2} + (r - q)S \frac{\partial P}{\partial S} - rP + \frac{\partial P}{\partial t} = 0$$

$$\frac{1}{2} \sigma^2 S^2 \frac{P_{i+1,j} - 2P_{i,j} + P_{i-1,j}}{(\Delta S)^2} + (r - q)S \frac{P_{i+1,j} - P_{i-1,j}}{2\Delta S} - rP_{i,j} + \frac{P_{i,j} - P_{i,j-1}}{\Delta t} = 0$$

for $1 \leq i \leq N - 1$ $P_{i,j-1} = \max(aP_{i-1,j} + bP_{i,j} + cP_{i+1,j}, X - S_{i,j-1})$

where

已知 time j → 求 time j-1

$$a \equiv \left[\left(\frac{\sigma S}{\Delta S} \right)^2 - \frac{(r - q)S}{\Delta S} \right] \frac{\Delta t}{2} \quad b \equiv 1 - r\Delta t - \left(\frac{\sigma S}{\Delta S} \right)^2 \Delta t \quad c \equiv \left[\left(\frac{\sigma S}{\Delta S} \right)^2 + \frac{(r - q)S}{\Delta S} \right] \frac{\Delta t}{2}$$

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11

Stable Conditions

Stable → $a > 0$ $b > 0$ $c > 0$

$$\frac{S}{\Delta S} > \frac{r - q}{\sigma^2} \quad \frac{1}{r + \left(\frac{\sigma S}{\Delta S} \right)^2} > \Delta t$$

Since $a > 0$

Since $b > 0$

如果boundary condition 沒取好 → S 的值太大或太小導致 a 或 b 小於零 → 出現不合理的結果

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12

	A	B	C	D	E	F	G	H	I	J	K	L
1	S	X	r	vol	T	delta S	delta t	q				
2	50	50	0.1	0.4	0.41667	5	0.04167	0				
3	Stock price				Time to maturity (months)							
4	(dollars)	5	4.5	4	3.5	3	2.5	2	1.5	1	0.5	0
5	100	0	0	0	0	0	0	0	0	0	0	0
6	95	0.06	0	0	0	0	0	0	0	0	0	0
7	90	-0.11	0.05	0	0	0	0	0	0	0	0	0
8	85	0.28	-0.046542	0.05	0	0	0	0	0	0	0	0
9	80	-0.13	0.2	0	0.05	0	0	0	0	0	0	0
10	75	0.46	0.06	0.2	0.041625	0.06	0	0	0	0	0	0
11	70	0.32	0.46	0.23	0.25	0.1	0.09	0	0	0	0	0
12	65	0.91	0.68	0.63	0.44	0.37	0.21	0.14	0	0	0	0
13	60	1.48	1.37	1.17	1.021225	0.81	0.65	0.42	0.27	0	0	0
14	55	2.59	2.39	2.21	1.99	1.77	1.5	1.24	0.9	0.59	0	0
15	50	4.26	4.08	3.89	3.68	3.44	3.18	2.87	2.53	2.07	1.5625	0
16	45	6.76	6.61	6.47	6.31	6.15	5.96	5.75	5.5	5.24	5	5
17	40	10.28	10.2	10.13	10.06	10.01	10	10	10	10	10	10
18	35	15	15	15	15	15	15	15	15	15	15	15
19	30	20	20	20	20	20	20	20	20	20	20	20
20	25	25	25	25	25	25	25	25	25	25	25	25
21	20	30	30	30	30	30	30	30	30	30	30	30
22	15	35	35	35	35	35	35	35	35	35	35	35
23	10	40	40	40	40	40	40	40	40	40	40	40
24	5	45	45	45	45	45	45	45	45	45	45	45
25	0	50	50	50	50	50	50	50	50	50	50	50

S太大導致b值為負

	a	b	c
C8	0.927017	-0.930833	0.99875
K15	0.455	0.035833	0.505
E10	0.71875	-0.504167	0.78125

13

如何修改?

$$\frac{1}{r + \left(\frac{\sigma S}{\Delta S}\right)^2} > \Delta t = 0.04167 \quad \text{當 } S \text{ 太大, 左式不能滿足}$$

- 將切的期數增加($\Delta t \downarrow$)
- 參見 hull18.2-2.xls

$$\frac{1}{r + \left(\frac{\sigma S}{\Delta S}\right)^2} \xrightarrow{\text{最大值 } (S=95)} \frac{1}{0.1 + \left(\frac{0.4 \times 95}{5}\right)^2} = 0.017283 > \Delta t = 0.0104$$

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14

PDE的變數變換

Change of variable $V \equiv \ln S$ $U(V, t) \equiv P(e^V, t)$

$$\frac{\partial P}{\partial t} = \frac{\partial U}{\partial t} \quad \frac{\partial P}{\partial S} = \frac{1}{S} \frac{\partial U}{\partial V} \quad \frac{\partial^2 P}{\partial S^2} = \frac{1}{S^2} \frac{\partial^2 U}{\partial V^2} - \frac{1}{S^2} \frac{\partial U}{\partial V}$$

$$\frac{1}{2} \sigma^2 \frac{\partial^2 U}{\partial V^2} + (r - q - \frac{\sigma^2}{2}) \frac{\partial U}{\partial V} - rU + \frac{\partial U}{\partial t} = 0$$

subject to $U(V, T) = \max(X - e^V, 0)$ and

$$U(V, t) = \max(\bar{U}(V, t), X - e^V) \quad t < T$$

繼續持有的價值

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提前履約的價值

15

程式撰寫

Along the V axis, the grid will span from V_{\min} to $V_{\min} + N \times \Delta V$ at ΔV apart for some suitably small V_{\min} ; hence boundary conditions at the lower ($V = V_{\min}$) and upper ($V = V_{\min} + N \times \Delta V$) boundaries will have to be specified.

請參見 hull18.3.xls

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16

The Explicit Scheme

The Black-Scholes differential equation is

$$\frac{1}{2}\sigma^2 \frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{(\Delta V)^2} + (r - q - \frac{\sigma^2}{2}) \frac{U_{i+1,j} - U_{i-1,j}}{2\Delta V} - rU_{i,j} + \frac{U_{i,j} - U_{i,j-1}}{\Delta t} = 0$$

for $1 \leq i \leq N - 1$

$$U_{i,j-1} = aU_{i-1,j} + bU_{i,j} + cU_{i+1,j}$$

where

$$a \equiv [(\frac{\sigma}{\Delta V})^2 - \frac{r - q - \sigma^2/2}{\Delta V}] \frac{\Delta t}{2} \quad b \equiv 1 - r\Delta t - (\frac{\sigma}{\Delta V})^2 \Delta t \quad c \equiv [(\frac{\sigma}{\Delta V})^2 + \frac{r - q - \sigma^2/2}{\Delta V}] \frac{\Delta t}{2}$$

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17

Stable $\rightarrow a > 0 \quad b > 0 \quad c > 0$

$$\frac{\sigma^2}{r - q - \frac{\sigma^2}{2}} > \Delta V \quad \frac{1}{r + (\frac{\sigma}{\Delta V})^2} > \Delta t$$

Since $a > 0$

Since $b > 0$

優點: a,b,c>0的條件和S (or V) 無關, 取boundary condition不用考慮a or b < 0 的問題
只要 ΔV 和 Δt 夠小即可

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18

Deep in the Money 邊界條件

由於取指數原因, S 的lower bound > 0

$U_{0,j}$ 不能直接設成 X

Deep in the money

Put : $\frac{\partial P}{\partial S}$ 趨近於 -1 $\frac{U_{1,j} - U_{0,j}}{\Delta S} = -1$

for j : $\Delta S = U_{0,j} - U_{1,j} = e^{V_{\min} + \Delta V} - e^{V_{\min}}$

$$U_{0,j} = U_{1,j} + (e^{V_{\min} + \Delta V} - e^{V_{\min}})$$

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19

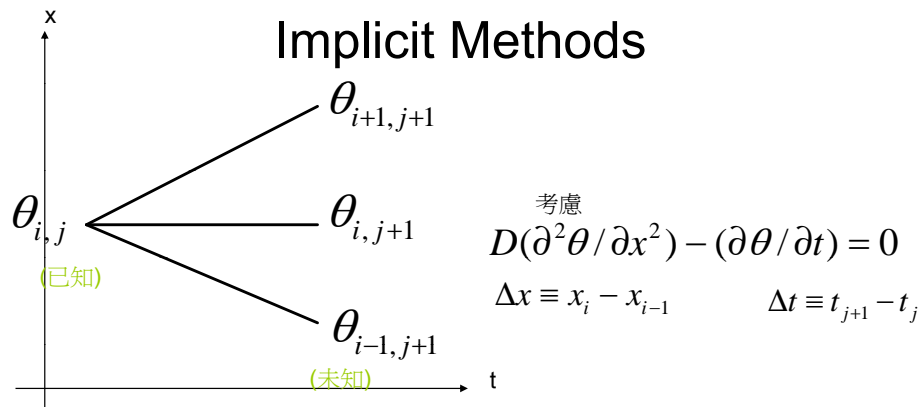
Deep OTM 邊界條件

For the upper boundary, we set $U_{N,j-1} = 0$.

$U_{i,j}$ is set to the greater of the value derived above and $X - e^{V_{\min} + i \times \Delta V}$ for early-exercise considerations.

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20



考慮

$$D(\partial^2 \theta / \partial x^2) - (\partial \theta / \partial t) = 0$$

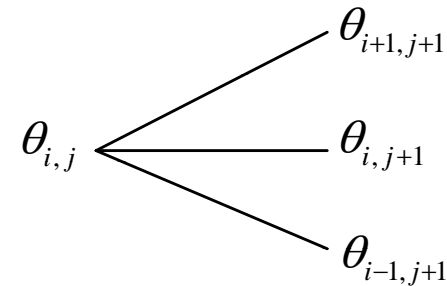
$$\Delta x \equiv x_i - x_{i-1} \quad \Delta t \equiv t_{j+1} - t_j$$

$$\frac{\theta_{i,j+1} - \theta_{i,j}}{\Delta t} = D \frac{\theta_{i+1,j+1} - 2\theta_{i,j+1} + \theta_{i-1,j+1}}{(\Delta x)^2}$$

Rearrange as

$$\theta_{i-1,j+1} - (2 + \gamma)\theta_{i,j+1} + \theta_{i+1,j+1} = -\gamma\theta_{i,j}$$

where $\gamma \equiv (\Delta x)^2 / (D\Delta t)$ 戴天時



The value of any one of the three quantities at t_{j+1} cannot be calculated unless the other two are known.

使用解方程式的方法求值

Suppose the boundary conditions are given at $x = x_0$ and $x = x_{N+1}$.

$$\begin{bmatrix} a & 1 & 0 & \cdots & \cdots & \cdots & 0 \\ 1 & a & 1 & 0 & \cdots & \cdots & 0 \\ 0 & 1 & a & 1 & 0 & \cdots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & \cdots & 0 & 1 & a & 1 \\ 0 & \cdots & \cdots & \cdots & 0 & 1 & a \end{bmatrix} \begin{bmatrix} \theta_{1,j+1} \\ \theta_{2,j+1} \\ \theta_{3,j+1} \\ \vdots \\ \vdots \\ \theta_{N,j+1} \end{bmatrix} = \begin{bmatrix} -\gamma\theta_{1,j} - \theta_{0,j+1} \\ -\gamma\theta_{2,j} \\ -\gamma\theta_{3,j} \\ \vdots \\ \vdots \\ -\gamma\theta_{N-1,j} \\ -\gamma\theta_{N,j} - \theta_{N+1,j+1} \end{bmatrix}$$

where $a \equiv -2 - \gamma$

Crank-Nicolson Method

Taking the average of explicit method and implicit method results in

$$\frac{\theta_{i,j+1} - \theta_{i,j}}{\Delta t} = \frac{1}{2} \left[D \frac{\theta_{i+1,j+1} - 2\theta_{i,j+1} + \theta_{i-1,j+1}}{(\Delta x)^2} + D \frac{\theta_{i+1,j} - 2\theta_{i,j} + \theta_{i-1,j}}{(\Delta x)^2} \right]$$

Rearrange as

$$\gamma\theta_{i,j+1} - \frac{\theta_{i+1,j+1} - 2\theta_{i,j+1} + \theta_{i-1,j+1}}{2} = \gamma\theta_{i,j} + \frac{\theta_{i+1,j} - 2\theta_{i,j} + \theta_{i-1,j}}{2}$$

使用解方程式的方法求值

Suppose the boundary conditions are given

at $x = x_0$ and $x = x_{N+1}$.

$$\begin{bmatrix} a & -0.5 & 0 & \dots & \dots & \dots & 0 \\ -0.5 & a & -0.5 & 0 & \dots & \dots & 0 \\ 0 & -0.5 & a & -0.5 & 0 & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & \dots & 0 & -0.5 & a & -0.5 \\ 0 & \dots & \dots & \dots & 0 & 0.5 & a \end{bmatrix} \begin{bmatrix} \theta_{1,j+1} \\ \theta_{2,j+1} \\ \theta_{3,j+1} \\ \vdots \\ \vdots \\ \vdots \\ \theta_{N,j+1} \end{bmatrix} = \begin{bmatrix} \frac{\theta_{0,j} + (2\gamma - 2)\theta_{1,j} + \theta_{2,j} + \theta_{0,j+1}}{2} \\ \frac{\theta_{1,j} + (2\gamma - 2)\theta_{2,j} + \theta_{3,j}}{2} \\ \frac{\theta_{2,j} + (2\gamma - 2)\theta_{3,j} + \theta_{4,j}}{2} \\ \vdots \\ \vdots \\ \frac{\theta_{N-2,j} + (2\gamma - 2)\theta_{N-1,j} + \theta_{N,j}}{2} \\ \frac{\theta_{N-1,j} + (2\gamma - 2)\theta_{N,j} + \theta_{N+1,j} + \theta_{N+1,j+1}}{2} \end{bmatrix}$$

where $a \equiv 1 + \gamma$

使用Implicit Method解變數變換後的 Black Scholes PDE

$$\frac{1}{2}\sigma^2 \frac{\partial^2 U}{\partial V^2} + (r - q - \frac{\sigma^2}{2}) \frac{\partial U}{\partial V} - rU + \frac{\partial U}{\partial t} = 0$$

改寫

$$\frac{1}{2}\sigma^2 \frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{(\Delta V)^2} + (r - q - \frac{\sigma^2}{2}) \frac{U_{i+1,j} - U_{i-1,j}}{2\Delta V} - rU_{i,j} + \frac{U_{i,j+1} - U_{i,j}}{\Delta t} = 0$$

for $1 \leq i \leq N - 1$

$$aU_{i-1,j} + bU_{i,j} + cU_{i+1,j} = U_{i,j+1}$$

where

$$a \equiv -\left[\left(\frac{\sigma}{\Delta V}\right)^2 + \frac{r - q - \sigma^2/2}{\Delta V}\right] \frac{\Delta t}{2} \quad b \equiv 1 + r\Delta t + \left(\frac{\sigma}{\Delta V}\right)^2 \Delta t$$

$$c \equiv -\left[\left(\frac{\sigma}{\Delta V}\right)^2 + \frac{r - q - \sigma^2/2}{\Delta V}\right] \frac{\Delta t}{2}$$

The system of equations can be written in matrix form:

$$\begin{bmatrix} b^* & c & 0 & \dots & \dots & \dots & 0 \\ a & b & c & 0 & \dots & \dots & 0 \\ 0 & a & b & c & 0 & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & \dots & 0 & a & b & c \\ 0 & \dots & \dots & \dots & 0 & a & b \end{bmatrix} \begin{bmatrix} U_{1,j} \\ U_{2,j} \\ U_{3,j} \\ \vdots \\ \vdots \\ \vdots \\ U_{N-1,j} \end{bmatrix} = \begin{bmatrix} U_{1,j+1} - K \\ U_{2,j+1} \\ U_{3,j+1} \\ \vdots \\ \vdots \\ U_{N-2,j+1} \\ U_{N-1,j+1} \end{bmatrix}$$

where $b^* \equiv a + b$ and $K \equiv a(e^{V_{\min} + \Delta V} - e^{V_{\min}})$.

(See the proof in the next slide)

$$aU_{0,j} + bU_{1,j} + cU_{2,j} = U_{1,j+1}$$

$$(a + b)U_{1,j} + cU_{2,j} = U_{1,j+1} + a(U_{1,j} - U_{0,j})$$

Deep in the money $\frac{\partial P}{\partial S} \rightarrow -1$ $\frac{U_{1,j} - U_{0,j}}{\Delta S} = -1$

$$(a + b)U_{1,j} + cU_{2,j} = U_{1,j+1} - a(e^{V_{\min} + \Delta V} - e^{V_{\min}})$$

Define $b^* \equiv a + b$ $K \equiv a(e^{V_{\min} + \Delta V} - e^{V_{\min}})$

$$b^*U_{1,j} + cU_{2,j} = U_{1,j+1} - K$$