

Explicit method for Black-Scholes equation

The application of finite difference methods developed in the previous sections for the Black-Scholes equation is straightforward. For instance, we consider the application of the explicit method but the implicit methods would be similar. We consider the truncated region \mathcal{R}_V^T defined in (4), for a sufficiently large value of S^* and in order to replace the terminal value problem associated with Black-Scholes equation by an initial value problem we perform a change of variables $U(S, t) := V(S, T - t)$ and consider the problem

$$\left\{ \begin{array}{ll} \frac{\partial U}{\partial t} = \frac{\sigma^2}{2} S^2 \frac{\partial^2 U}{\partial S^2} + rS \frac{\partial U}{\partial S} - rU & \text{in } \mathcal{R}_V^T \\ U(S, 0) = V(S, T) & S \in [0, S^*] \\ U(0, t) = V(0, T - t) & t \in [0, T] \\ U(S^*, t) = V(S^*, T - t) & t \in [0, T]. \end{array} \right. \quad (42)$$

Then, we define the grid of points of the form (S_i, t_j) , where

$$S_i = h_S i, \quad i = 0, 1, \dots, N_S, \quad \text{where} \quad h_S = \frac{S^*}{N_S},$$

$$t_j = h_t j, \quad j = 0, 1, \dots, N_t, \quad \text{where} \quad h_t = \frac{T}{N_t}$$

and will use the notation $U_{i,j} = U(S_i, t_j)$.

We consider the finite difference approximations

$$\frac{\partial^2 U}{\partial S^2}(S_i, t_j) \approx \frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{h_S^2},$$

$$\frac{\partial U}{\partial S}(S_i, t_j) \approx \frac{U_{i+1,j} - U_{i-1,j}}{2h_S},$$

and

$$\frac{\partial U}{\partial t}(S_i, t_j) \approx \frac{U_{i,j+1} - U_{i,j}}{h_t}.$$

Therefore, at a general point (S_i, t_j) , we impose

$$\frac{U_{i,j+1} - U_{i,j}}{h_t} = \frac{\sigma^2}{2} S_i^2 \left(\frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{h_S^2} \right) + rS_i \left(\frac{U_{i+1,j} - U_{i-1,j}}{2h_S} \right) - rU_{i,j}$$

that can be written as

$$\begin{aligned} U_{i,j+1} &= U_{i,j} + \frac{\sigma^2}{2} S_i^2 h_t \left(\frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{h_S^2} \right) + rS_i h_t \left(\frac{U_{i+1,j} - U_{i-1,j}}{2h_S} \right) - rh_t U_{i,j} \\ &= U_{i-1,j} \left(\frac{\sigma^2}{2} h_t \frac{S_i^2}{h_S^2} - r \frac{h_t S_i}{2h_S} \right) + U_{i,j} \left(1 - \frac{\sigma^2}{2} h_t \frac{S_i^2}{h_S^2} - rh_t \right) + U_{i+1,j} \left(\frac{\sigma^2}{2} h_t \frac{S_i^2}{h_S^2} + r \frac{h_t S_i}{2h_S} \right) \\ &= U_{i-1,j} \frac{h_t}{2} (\sigma^2 i^2 - ri) + U_{i,j} (1 - \sigma^2 i^2 h_t - rh_t) + U_{i+1,j} \frac{h_t}{2} (\sigma^2 i^2 + ri) \end{aligned}$$

because $\frac{S_i}{h_S} = i$.

Figure 16 shows the plot of the solution in the case of an European call option, obtained with the parameters $r = 0.06$, $\sigma = 0.3$, $T = 1$ and $K = 10$. We took $S^* = 15$, $N_S = 50$ and $N_t = 10000$.

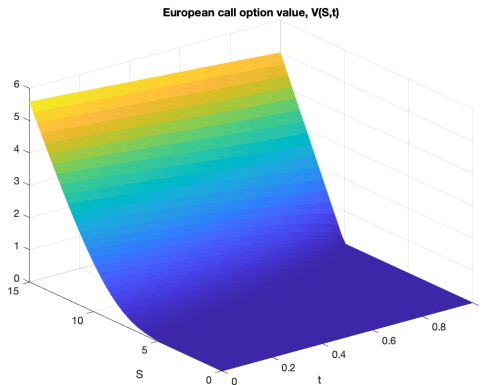


Figure: Plot of $V(S, t)$ in the case of an European call option, obtained with the parameters $r = 0.06$, $\sigma = 0.3$, $T = 1$ and $K = 10$. We took $S^* = 15$, $N_S = 50$ and $N_t = 10000$.

Figure 17 shows the plot of the solution for $t = 0$, $t = 0.5$ and $t = 1$, for which we have the payoff function.

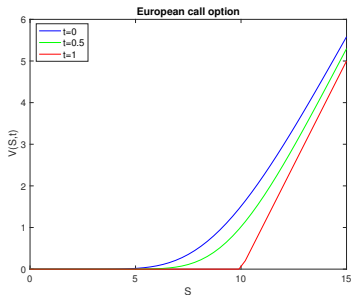


Figure: Plot of $V(S, t)$ for $t = 0$, $t = 0.5$ and $t = 1$, for which we have the payoff function.

The stability analysis would be similar to the analysis performed for the heat equation, but the calculations involved in the case of Black-Scholes equation are much more difficult.

Crank-Nicolson method for Black-Scholes equation

In the Crank-Nicolson method we define the scheme

$$\begin{aligned} \frac{U_{i,j+1} - U_{i,j}}{h_t} = & \frac{1}{2} \left[\frac{\sigma^2}{2} S_i^2 \left(\frac{U_{i+1,j} - 2U_{i,j} + U_{i-1,j}}{h_S^2} \right) + rS_i \left(\frac{U_{i+1,j} - U_{i-1,j}}{2h_S} \right) - rU_{i,j} \right] \\ & + \frac{1}{2} \left[\frac{\sigma^2}{2} S_i^2 \left(\frac{U_{i+1,j+1} - 2U_{i,j+1} + U_{i-1,j+1}}{h_S^2} \right) + rS_i \left(\frac{U_{i+1,j+1} - U_{i-1,j+1}}{2h_S} \right) - rU_{i,j+1} \right] \end{aligned}$$

that can be written as

$$\begin{aligned} & U_{i-1,j+1} \left(-\frac{\sigma^2}{4} h_t \frac{S_i^2}{h_S^2} + \frac{rh_t}{4} \frac{S_i}{h_S} \right) + U_{i,j+1} \left(1 + \frac{\sigma^2}{2} h_t \frac{S_i^2}{h_S^2} + \frac{rh_t}{2} \right) + U_{i+1,j+1} \left(-\frac{\sigma^2}{4} h_t \frac{S_i^2}{h_S^2} - \frac{rh_t}{4} \frac{S_i}{h_S} \right) \\ & = U_{i-1,j} \left(\frac{\sigma^2}{4} h_t \frac{S_i^2}{h_S^2} - \frac{rh_t}{4} \frac{S_i}{h_S} \right) + U_{i,j} \left(1 - \frac{\sigma^2}{2} h_t \frac{S_i^2}{h_S^2} - \frac{rh_t}{2} \right) + U_{i+1,j} \left(\frac{\sigma^2}{4} h_t \frac{S_i^2}{h_S^2} + \frac{rh_t}{4} \frac{S_i}{h_S} \right) \end{aligned}$$

that is

$$\begin{aligned} U_{i-1,j+1} \left(-\frac{\sigma^2}{4} h_t i^2 + \frac{r h_t}{4} i \right) + U_{i,j+1} \left(1 + \frac{\sigma^2}{2} h_t i^2 + \frac{r h_t}{2} \right) + U_{i-1,j+1} \left(-\frac{\sigma^2}{4} h_t i^2 - \frac{r h_t}{4} i \right) \\ = U_{i-1,j} \left(\frac{\sigma^2}{4} h_t i^2 - \frac{r h_t}{4} i \right) + U_{i,j} \left(1 - \frac{\sigma^2}{2} h_t i^2 - \frac{r h_t}{2} \right) + U_{i-1,j} \left(\frac{\sigma^2}{4} h_t i^2 + \frac{r h_t}{4} i \right) \end{aligned}$$

and defining $a_i = -\frac{\sigma^2}{4} h_t i^2 + \frac{r h_t}{4} i$, $b_i = 1 + \frac{\sigma^2}{2} h_t i^2 + \frac{r h_t}{2}$,
 $c_i = -\frac{\sigma^2}{4} h_t i^2 - \frac{r h_t}{4} i$ and $d_i = 1 - \frac{\sigma^2}{2} h_t i^2 - \frac{r h_t}{2}$ we get

$$U_{i-1,j+1} a_i + U_{i,j+1} b_i + U_{i-1,j+1} c_i = -U_{i-1,j} a_i + U_{i,j} d_i - U_{i-1,j} c_i$$

and the values $U_{i+1,j+1}$, $U_{i,j+1}$ and $U_{i-1,j+1}$ are calculated by solving a linear system involving the values $U_{i+1,j}$, $U_{i,j}$ and $U_{i-1,j}$, which are assumed to be already calculated,

$$\begin{bmatrix} b_1 & c_1 & 0 & \dots & \dots & \dots & 0 \\ a_2 & b_2 & c_2 & 0 & \dots & \dots & 0 \\ 0 & a_3 & b_3 & c_3 & 0 & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & a_{N_S-3} & b_{N_S-3} & c_{N_S-3} & 0 \\ 0 & \dots & \dots & 0 & a_{N_S-2} & b_{N_S-2} & c_{N_S-2} \\ 0 & \dots & \dots & \dots & 0 & a_{N_S-1} & b_{N_S-1} \end{bmatrix} \begin{bmatrix} U_{1,j+1} \\ U_{2,j+1} \\ U_{3,j+1} \\ \vdots \\ U_{N_S-3,j+1} \\ U_{N_S-2,j+1} \\ U_{N_S-1,j+1} \end{bmatrix} =$$

$$\begin{bmatrix} -a_1 & d_1 & -c_1 & 0 & \dots & \dots & \dots & \dots & 0 \\ 0 & -a_2 & d_2 & -c_2 & 0 & \dots & \dots & \dots & 0 \\ 0 & 0 & -a_3 & d_3 & -c_3 & \dots & \dots & \dots & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \ddots & \vdots \\ 0 & \dots & 0 & 0 & -a_{N_S-3} & d_{N_S-3} & -c_{N_S-3} & 0 & 0 \\ 0 & \dots & \dots & \dots & 0 & -a_{N_S-2} & d_{N_S-2} & -c_{N_S-2} & 0 \\ 0 & \dots & \dots & \dots & \dots & 0 & -a_{N_S-1} & d_{N_S-1} & -c_{N_S-1} \end{bmatrix} \begin{bmatrix} U_{0,j} \\ U_{1,j} \\ U_{2,j} \\ \vdots \\ U_{N_S-2,j} \\ U_{N_S-1,j} \\ U_{N_S,j} \end{bmatrix} + \begin{bmatrix} -a_1 U_{0,j+1} \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ -c_{N_S-1} U_{N_S,j+1} \end{bmatrix}.$$

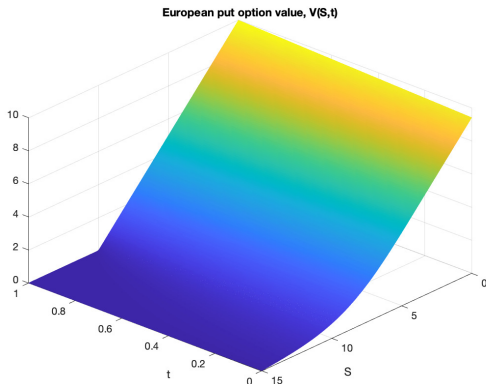


Figure: Plot of $V(S, t)$ in the case of an European put option, calculated by Crank-Nicolson method, obtained with the parameters $r = 0.06$, $\sigma = 0.3$, $T = 1$ and $K = 10$. We took $S^* = 15$, $N_S = 200$ and $N_t = 1000$.