

4. Black-Scholes Equation and Finite Difference Schemes

Session 22
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4.1 Derivation of Black-Scholes Equation

Let the stock price process be geom. BM: $dS = \mu S dt + \sigma S dW$

Solution: $S(t) = S_0 e^{(\mu - \frac{\sigma^2}{2})t + \sigma W(t)}$

Let $C(S, t)$ be the price of an option

Ito's Lemma gives $dC = \left(\frac{\partial C}{\partial t} + \frac{\partial C}{\partial S} \mu S + \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 \right) dt + \frac{\partial C}{\partial S} \sigma S dW$

(Recall Ito: $dX = f dt + g dW$ then $dF(x, t) = \left(\frac{\partial F}{\partial t} + f \frac{\partial F}{\partial x} + \frac{1}{2} g^2 \frac{\partial^2 F}{\partial x^2} \right) dt + \frac{\partial F}{\partial x} g dW$)

Merton's trick: consider the right portfolio to eliminate risk

value of portfolio $\Pi = -C + \frac{\partial C}{\partial S} S$ $= r(-C + \frac{\partial C}{\partial S} S) dt$

$d\Pi = -dC + \frac{\partial C}{\partial S} dS = r\Pi dt$ (such that $\Pi = \Pi_0 e^{rt}$)

such a portfolio grows with riskless rate r (deterministically)

$= - \left(\frac{\partial C}{\partial t} + \frac{\partial C}{\partial S} \mu S + \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 \right) dt - \frac{\partial C}{\partial S} \sigma S dW$
 $+ \frac{\partial C}{\partial S} \mu S dt + \frac{\partial C}{\partial S} \sigma S dW$

$= \left(-\frac{\partial C}{\partial t} - \frac{1}{2} \frac{\partial^2 C}{\partial S^2} \sigma^2 S^2 \right) dt$

$$\Rightarrow \frac{\partial C}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 C}{\partial S^2} + rS \frac{\partial C}{\partial S} = rC \quad \text{Black-Scholes Equation}$$

Notes:

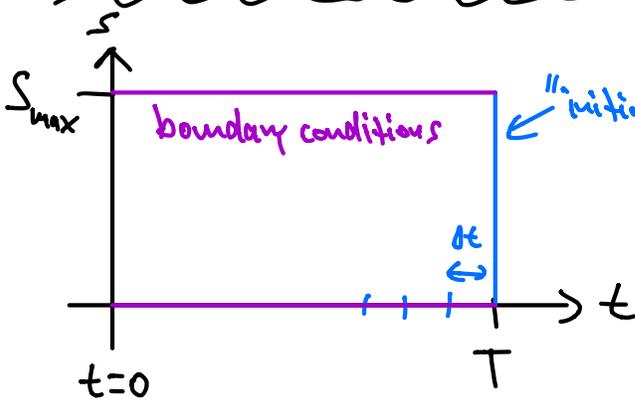
- independent of μ (option price only depends on volatility!)
- second order PDE
- by a change of variables this eq. can be transformed into a heat equation

$$\frac{\partial u}{\partial t} = \kappa \frac{\partial^2 u}{\partial x^2} \quad (\text{or } \kappa \Delta u \text{ for higher dim.})$$

- backward drift-diffusion equation (specify $C(S, T)$, solve for $C(S, 0)$)

We specify $C(S, T) = \text{payoff} \stackrel{\text{European calls}}{=} \max(0, S - K)$ ↙ strike price

4.2 Discrete Finite Differences



boundary conditions:

$$C(t, 0) = 0$$

$$C(t, S_{\max}) = S_{\max} - Ke^{-r(T-t)}$$

(actually a sol. to BS eq.)

or $C(t, S_{\max}) = S_{\max}$ (since S_{\max} should be chosen much bigger than K)

partition $[0, T]$ into M steps of size $\Delta t = \frac{T}{M}$, $t_j = j\Delta t$

partition $[0, S_{\max}]$ into N steps of size $\Delta s = \frac{S_{\max}}{N}$, $s_i = i\Delta s$

call $C(t_j, s_i) = C_i^j$

$$\frac{\partial C}{\partial t} \approx \frac{C_i^{j+1} - C_i^j}{\Delta t} + \mathcal{O}(\Delta t)$$

one could choose $\frac{\partial C}{\partial s} \approx \frac{C_{i+1}^j - C_i^j}{\Delta s} + \mathcal{O}(\Delta s)$

alternatively (fix j):

$$(1) C(s_i + \Delta s) = C(s_i) + \frac{\partial C}{\partial s}(s_i) \Delta s + \frac{1}{2} \frac{\partial^2 C}{\partial s^2}(s_i) \Delta s^2 + \frac{1}{3!} \frac{\partial^3 C}{\partial s^3} \Delta s^3 + \mathcal{O}(\Delta s^4)$$

$$(2) C(s_i - \Delta s) = C(s_i) - \frac{\partial C}{\partial s}(s_i) \Delta s + \frac{1}{2} \frac{\partial^2 C}{\partial s^2}(s_i) \Delta s^2 - \frac{1}{3!} \frac{\partial^3 C}{\partial s^3} \Delta s^3 + \mathcal{O}(\Delta s^4)$$

$$(1) - (2) \Rightarrow C(s_i + \Delta s) - C(s_i - \Delta s) = 2 \frac{\partial C}{\partial s}(s_i) \Delta s + \mathcal{O}(\Delta s^3)$$

"centralized" derivative $\frac{\partial C}{\partial s} \approx \frac{C_{i+1}^j - C_{i-1}^j}{2\Delta s} + \mathcal{O}(\Delta s^2)$ (better error)

second derivative: (1) + (2)

$$\Rightarrow C(s_i + \Delta s) + C(s_i - \Delta s) = 2C(s_i) + \frac{\partial^2 C}{\partial s^2}(s_i) \Delta s^2 + \mathcal{O}(\Delta s^4)$$

$$\Rightarrow \frac{\partial^2 C}{\partial s^2} \approx \frac{C_{i+1}^j - 2C_i^j + C_{i-1}^j}{\Delta s^2} + \mathcal{O}(\Delta s^2) \quad (\text{also } \Delta s^2 \text{ error})$$

4.3 Stability of Time-stepping Methods

Ex.: $\frac{dy}{dt} = \lambda y$, solution: $y(t) = y_0 e^{\lambda t}$

consider $\lambda < 0$, say also $|\lambda| \gg 1$.

Explicit Euler method: $\frac{y^{j+1} - y^j}{\Delta t} = \lambda y^j$ (r.h.s. evaluated at j)

$$\Rightarrow y^{j+1} = \lambda y^j \Delta t + y^j = (1 + \lambda \Delta t) y^j$$

$$\Rightarrow y^j = (1 + \lambda \Delta t)^j y_0$$

here: if $y^j \rightarrow \lim_{t \rightarrow \infty} y_0 e^{\lambda t} = 0$ ($\lambda < 0$), then scheme is stable

$$\Rightarrow \text{we need } |1 + \lambda \Delta t| < 1$$

$$\Rightarrow 1 + \lambda \Delta t < 1 \quad \text{and} \quad -1 - \lambda \Delta t < 1$$

$$\Rightarrow \lambda < 0$$

$$\Rightarrow \Delta t < \frac{2}{-\lambda} \quad \text{Condition on smallness of } \Delta t \text{ for scheme to be stable}$$

Implicit Euler method: $\frac{y^{j+1} - y^j}{\Delta t} = \lambda y^{j+1}$ (r.h.s. evaluated at $j+1$)

$$\Rightarrow (1 - \lambda \Delta t) y^{j+1} = y^j \Rightarrow y^{j+1} = \left(\frac{1}{1 - \lambda \Delta t} \right) y^j$$

$$\Rightarrow y^j = \left(\frac{1}{1 - \lambda \Delta t} \right)^j y^0$$

now stability condition is $\left| \frac{1}{1-\lambda\Delta t} \right| < 1$; this always holds here (since $\lambda < 0$)

↳ unconditionally stable

4.4 Application to Heat Equation

consider $\frac{\partial V}{\partial t} = \frac{\partial^2 V}{\partial x^2}$, initial value $V(x, 0)$ (want to know $V(x, T)$)
(for BS: backwards)

$$\frac{\partial V}{\partial t}(x_i, t_j) = \frac{V(x_i, t_{j+1}) - V(x_i, t_j)}{\Delta t} + O(\Delta t)$$

$$\frac{\partial^2 V}{\partial x^2}(x_i, t) = \frac{V(x_{i+1}, t) - 2V(x_i, t) + V(x_{i-1}, t))}{\Delta x^2} + O(\Delta x^2)$$

$t = t_j$: explicit scheme

$t = t_{j+1}$: implicit scheme

denote $V(x_i, t_j) = V_i^j$

explicit:
$$\frac{V_i^{j+1} - V_i^j}{\Delta t} = \frac{V_{i+1}^j - 2V_i^j + V_{i-1}^j}{\Delta x^2}$$

$$\Rightarrow V_i^{j+1} = \frac{\Delta t}{\Delta x^2} V_{i+1}^j + \left(1 - \frac{2\Delta t}{\Delta x^2}\right) V_i^j + \frac{\Delta t}{\Delta x^2} V_{i-1}^j$$

note: need $\frac{\Delta t}{\Delta x^2} < 1$ for stability

implicit:
$$\frac{V_i^{j+1} - V_i^j}{\Delta t} = \frac{V_{i+1}^{j+1} - 2V_i^{j+1} + V_{i-1}^{j+1}}{\Delta x^2}$$

$$\Rightarrow V_i^j = -\frac{\Delta t}{\Delta x^2} V_{i+1}^{j+1} + \left(1 + \frac{2\Delta t}{\Delta x^2}\right) V_i^{j+1} - \frac{\Delta t}{\Delta x^2} V_{i-1}^{j+1}$$

$$\begin{pmatrix} (1+2a) & -a & & & 0 \\ -a & (1+2a) & -a & & \\ & -a & \ddots & \ddots & \\ 0 & & \ddots & \ddots & -a \\ & & & -a & (1+2a) \end{pmatrix} \begin{pmatrix} V_1^{j+1} \\ V_2^{j+1} \\ \vdots \\ V_n^{j+1} \end{pmatrix} = \begin{pmatrix} V_1^j \\ V_2^j \\ \vdots \\ V_n^j \end{pmatrix}$$

$n \times n$ matrix A vector \vec{V}^{j+1} vector \vec{V}^j

\Rightarrow need to solve tridiagonal system of equations to get \vec{V}^{j+1} from \vec{V}^j

what happens at the boundary?

V_0^{j+1} and V_{n+1}^{j+1} are given by fixed boundary conditions

we have $V_1^j = -a V_2^{j+1} + (1+2a) V_1^{j+1} - a V_0^{j+1}$

$V_n^j = -a V_{n+1}^{j+1} + (1+2a) V_n^{j+1} - a V_{n-1}^{j+1}$

so with boundary conditions the tridiagonal system is

$$\begin{pmatrix} \ddots & \ddots & \ddots & & \\ & \ddots & (1+2a) & -a & \\ & -a & \ddots & \ddots & \\ & & & -a & \ddots \\ & & & & & \ddots \end{pmatrix} \begin{pmatrix} V_1^{j+1} \\ \vdots \\ V_n^{j+1} \end{pmatrix} = \begin{pmatrix} V_1^j + a V_0^{j+1} \\ V_2^j \\ \vdots \\ V_{n-1}^j \\ V_n + a V_{n+1}^{j+1} \end{pmatrix}$$

← fixed
↖ fixed