



Weierstrass Institute for  
Applied Analysis and Stochastics



# Computational finance – Lecture 11

Christian Bayer

### 1 Pricing partial differential equations

### 2 The finite difference method

### 3 Finite element method

$$dS_t = rS_t dt + \sigma S_t dW_t$$

### European option pricing PDE

$$u(t, x) := E \left[ e^{-r(T-t)} f(S_T) \mid S_t = x \right],$$

$$\frac{\partial}{\partial t} u(t, x) + \frac{1}{2} \sigma^2 x^2 \frac{\partial^2}{\partial x^2} u(t, x) + rx \frac{\partial}{\partial x} u(t, x) - ru(t, x) = 0, \quad u(T, x) = f(x).$$

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### American option pricing PDE (HJB equation in form of variational inequality)

$$v(t, x) := \sup_{\tau \in \mathcal{T}_{t,T}} E \left[ e^{-r(\tau-t)} f(S_\tau) \mid S_t = x \right], \quad \mathcal{T}_{t,T} := \{ \text{Stopping times } t \leq \tau \leq T \},$$

$$\frac{\partial}{\partial t} v(t, x) + \frac{1}{2} \sigma^2 x^2 \frac{\partial^2}{\partial x^2} v(t, x) + rx \frac{\partial}{\partial x} v(t, x) - rv(t, x) \leq 0, \quad v(t, x) \geq f(x), \quad v(T, x) = f(x),$$

with equality in the PDE on  $\{ x \mid v(t, x) > f(x) \}$ .

$$\frac{\partial}{\partial t}u(t, x) + \frac{1}{2}\sigma^2 x^2 \frac{\partial^2}{\partial x^2}u(t, x) + rx \frac{\partial}{\partial x}u(t, x) - ru(t, x) = 0, \quad u(T, x) = (K - x)^+$$

- Change of variables:  $y := \log(x/K)$ ,  $\tau := \frac{1}{2}\sigma^2(T - t)$ ,  $q := 2r/\sigma^2$  and

$\tilde{u}(\tau, y) := \frac{1}{K} \exp\left(\frac{1}{2}(q - 1)y + \left(\frac{1}{4}(q - 1)^2 + q\right)\tau\right) u(t, x)$ , satisfying the **heat equation**

$$\frac{\partial}{\partial \tau}\tilde{u}(\tau, y) = \frac{\partial^2}{\partial y^2}\tilde{u}(\tau, y), \quad \tilde{u}(0, y) = \left(e^{\frac{1}{2}(q-1)y} - e^{\frac{1}{2}(q+1)y}\right)_+.$$

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- Boundary conditions: Natural boundary condition for put option at  $y \rightarrow \infty$  and from put-call-parity:

$$\tilde{u}(\tau, y) = \exp\left(\frac{1}{2}(q - 1)y + \frac{1}{4}(q - 1)^2\tau\right) \text{ for } y \rightarrow -\infty, \quad \tilde{u}(\tau, y) = 0 \text{ for } y \rightarrow \infty$$

$$\frac{\partial}{\partial t}v(t, x) + \frac{1}{2}\sigma^2 x^2 \frac{\partial^2}{\partial x^2}v(t, x) + rx \frac{\partial}{\partial x}v(t, x) - rv(t, x) \leq 0, \quad v(t, x) \geq (K-x)^+, \quad v(T, x) = (K-x)^+$$

The same transformation as above gives:

$$\left( \frac{\partial}{\partial \tau} \tilde{v}(\tau, y) - \frac{\partial^2}{\partial y^2} \tilde{v}(\tau, y) \right) (\tilde{v}(\tau, y) - g(\tau, y)) = 0, \quad \frac{\partial}{\partial \tau} \tilde{v}(\tau, y) - \frac{\partial^2}{\partial y^2} \tilde{v}(\tau, y) \geq 0,$$

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Replace derivatives in time and space by finite difference quotients based on grids.

- ▶ Notation:  $t_i := i\Delta t$ ,  $i = 0, \dots, N$ ,  $\Delta t := T/N$ .  $x_j := a + j\Delta x$ ,  $j = 0, \dots, M$ ,  $\Delta x := (b-a)/M$ .
- ▶ Solving heat equation with appropriate boundary conditions on  $[a, b]$ , setting  $u_{i,j} := u(t_i, x_j)$  – and similarly its FD approximation  $\bar{u}_{i,j}$ .

Based on the approximations:

$$\frac{\partial}{\partial t} u(t_i, x_j) = \frac{u_{i+1,j} - u_{i,j}}{\Delta t} + O(\Delta t), \quad \frac{\partial^2}{\partial x^2} u(t_i, x_j) = \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{\Delta x^2} + O(\Delta x^2).$$



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### Explicit FD scheme for the heat equation

With  $\lambda := \frac{\Delta t}{\Delta x^2}$  set  $\bar{u}_{i+1,j} := \bar{u}_{i,j} + \lambda(\bar{u}_{i,j+1} - 2\bar{u}_{i,j} + \bar{u}_{i,j-1})$ ,  $i = 0, \dots, N-1$ ,  $j = 1, \dots, M-1$ .

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► Boundary and initial conditions for the European put option:

$$\bar{u}_{0,j} = \left( e^{\frac{1}{2}(q-1)x_j} - e^{\frac{1}{2}(q+1)x_j} \right)^+, \quad \bar{u}_{i+1,0} = \exp\left( \frac{1}{2}(q-1)a + \frac{1}{4}(q-1)^2 t_{i+1} \right), \quad \bar{u}_{i+1,M} = 0$$

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- ▶ Up to the boundary conditions,

$$\bar{u}_{i+1,:} = A(\lambda)\bar{u}_{i,:}, \quad A(\lambda) := \begin{pmatrix} 1-2\lambda & \lambda & 0 & \dots & 0 \\ \lambda & 1-2\lambda & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \lambda \\ 0 & \dots & 0 & \lambda & 1-2\lambda \end{pmatrix}.$$

### Example (Instability of the explicit FD scheme)

Consider the heat equation with  $u(0, x) = \sin(\pi x)$ ,  $x \in [0, 1]$ ,  $u(t, 0) \equiv u(t, 1) \equiv 1$ . Then  $u(t, x) = \sin(\pi x)e^{-\pi^2 t}$ . Compute  $u(0.5, 0.2)$ :

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- ▶ FD with  $\Delta x = 0.1$ ,  $\Delta t = 0.0005$ :  $u(0.5, 0.2) \approx \bar{u}_{1000,2} = 0.00435$ .
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- ▶ The explicit FD scheme is prone to **instability**, i.e., explosive error propagation.
- ▶  $x \mapsto Ax$  is **stable** iff the **spectral radius is bounded by 1**. For  $A(\lambda)$  this can be proved to be the case when  $\lambda \leq 1/2$ .

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### Theorem

*The explicit FD scheme converges is stable and converges when  $\Delta t \leq \frac{1}{2}\Delta x^2$  (plus technical conditions). In this case, the error behaves like  $O(\Delta t) + O(\Delta x^2)$ .*

Based on the approximations:

$$\frac{\partial}{\partial t} u(t_i, x_j) = \frac{u_{i,j} - u_{i-1,j}}{\Delta t} + O(\Delta t), \quad \frac{\partial^2}{\partial x^2} u(t_i, x_j) = \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{\Delta x^2}$$



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### Implicit FD scheme for the heat equation

Define  $\bar{u}_{i,:}$  as solution of the system  $\bar{u}_{i-1,j} = \bar{u}_{i,j} + \frac{\Delta t}{\Delta x^2} (-\bar{u}_{i,j+1} + 2\bar{u}_{i,j} - \bar{u}_{i,j-1})$ ,  $i = 1, \dots, N$ ,  $j = 1, \dots, M - 1$ .

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► Up to boundary conditions:

$$A \bar{u}_{i,:} = \bar{u}_{i-1,:}, \quad A := \begin{pmatrix} 1 + 2\lambda & -\lambda & 0 & \cdots & 0 \\ -\lambda & 1 + 2\lambda & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & -\lambda \\ 0 & \cdots & 0 & -\lambda & 1 + 2\lambda \end{pmatrix}.$$

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### Theorem

The implicit FD scheme is **unconditionally stable** and converges with  $O(\Delta t) + O(\Delta x^2)$ .

- ▶ Assume that the Cauchy problem is **well-posed**.
- ▶ Notation:  $u$  ... solution of the PDE,  $u^i$  ...  $u$  at time  $t_i$  discretized on the  $x$ -grid,  $\bar{u}^i$  ... FD approximation, given by  $B_1 \bar{u}^{i+1} = B_0 \bar{u}^i + f^i$ .

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- ▶ Consistency:  $\|B_1 u^{i+1} - (B_0 u^i + f^i)\| \rightarrow 0$  as  $\Delta t = T/N, \Delta x = (b - a)/M \rightarrow 0, i = 0, \dots, N$ .

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- ▶ **Consistency**:  $\|B_1 u^{i+1} - (B_0 u^i + f^i)\| \rightarrow 0$  as  $\Delta t = T/N, \Delta x = (b-a)/M \rightarrow 0, i = 0, \dots, N$ .
- ▶ **Stability**: there is a constant  $C$  s.t.  $\|(B_1^{-1} B_0)^N\| \leq C$  uniformly in  $N$ . (Note:  $B_1, B_0$  depend on  $N$  via  $\Delta t, \Delta x$ .)

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- ▶ **Convergence**: Consider  $i(N)$  s.t.  $t_i \rightarrow t$  as  $N \rightarrow \infty$ . Then  $\|\bar{u}^{i(N)} - u^{i(N)}\| \rightarrow 0$  as  $N \rightarrow \infty$ .

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### Theorem (Lax–Richtmyer; Lax equivalence principle; Fundamental theorem of numerical analysis)

For a **consistent** scheme, **stability** is equivalent to **convergence**, provided the problem is linear and well-posed.



$$\frac{\partial}{\partial t} u(t_{i+1}, x_j) = \frac{u_{i+1,j} - u_{i,j}}{\Delta t} + O(\Delta t)$$

can be seen as **forward** or **backward** difference quotient, leading to

$$\frac{\bar{u}_{i+1,j} - \bar{u}_{i,j}}{\Delta t} = \frac{\bar{u}_{i,j+1} - 2\bar{u}_{i,j} + \bar{u}_{i,j-1}}{\Delta x^2} \text{ or } \frac{\bar{u}_{i+1,j} - \bar{u}_{i,j}}{\Delta t} = \frac{\bar{u}_{i+1,j+1} - 2\bar{u}_{i+1,j} + \bar{u}_{i+1,j-1}}{\Delta x^2}$$

$$\frac{\partial}{\partial t} u(t_{i+1}, x_j) = \frac{u_{i+1,j} - u_{i,j}}{\Delta t} + O(\Delta t)$$

can be seen as forward or backward difference quotient, leading to

$$\frac{\bar{u}_{i+1,j} - \bar{u}_{i,j}}{\Delta t} = \frac{\bar{u}_{i,j+1} - 2\bar{u}_{i,j} + \bar{u}_{i,j-1}}{\Delta x^2} \quad \text{or} \quad \frac{\bar{u}_{i+1,j} - \bar{u}_{i,j}}{\Delta t} = \frac{\bar{u}_{i+1,j+1} - 2\bar{u}_{i+1,j} + \bar{u}_{i+1,j-1}}{\Delta x^2}$$

Instead, take the **mean** of the right hand sides:

### Crank–Nicolson scheme

Define  $\bar{u}_{i+1,:}$  as solution of the system

$$\bar{u}_{i+1,j} - \bar{u}_{i,j} = \frac{\Delta t}{2\Delta x^2} (\bar{u}_{i,j+1} - 2\bar{u}_{i,j} + \bar{u}_{i,j-1} + \bar{u}_{i+1,j+1} - 2\bar{u}_{i+1,j} + \bar{u}_{i+1,j-1}).$$

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- ▶ System of equations of the form  $A\bar{u}_{i+1,:} = B\bar{u}_{i,:}$ .
- ▶ Unconditionally stable and converges with error  $O(\Delta t^2) + O(\Delta x^2)$ .

$$\left( \frac{\partial}{\partial \tau} \tilde{v}(\tau, y) - \frac{\partial^2}{\partial y^2} \tilde{v}(\tau, y) \right) (\tilde{v}(\tau, y) - g(\tau, y)) = 0, \quad \frac{\partial}{\partial \tau} \tilde{v}(\tau, y) - \frac{\partial^2}{\partial y^2} \tilde{v}(\tau, y) \geq 0,$$

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- ▶ Have to solve linear inequality systems of the form  $Aw - b \geq 0, w \geq g$ ,  $(aw - b)^T (w - g) = 0$  for the approximate solution  $w$ . **Projection SOR** (Successive over-relaxation algorithm).

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- ▶ Poor man's algorithm: Use standard FD iterations, but take **maximum with payoff** function at each iteration step.

**1** Pricing partial differential equations

**2** The finite difference method

**3** Finite element method

Consider, for simplicity, the **Poisson equation**  $\Delta u = f$  on  $[0, 1]$  with  $u(0) = u(1) = 0$ .



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1. **Variational (weak) formulation:**  $u$  is the only element of  $V := H_0^1$  such that for every **test function**  $v \in V$ :

$$A(u, v) = L(v), \quad A(u, v) := - \int_0^1 u'(x)v'(x) dx, \quad L(v) := \int_0^1 f(x)v(x) dx$$

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2. **Projection onto finite dimensional space:** Choose  $V_h \subset V$ ,  $\dim V_h < \infty$ ,  $h > 0$ , and consider the projected problem  $\forall v \in V_h : A(u_h, v) = L(v)$ , with solution  $u_h \in V_h$ . E.g.,

$$V_h := \left\{ v \in C([0, 1]) \mid v|_{[x_i, x_{i+1}]} \text{ affine, } i = 0, \dots, N, v(0) = v(1) = 0 \right\}, \quad x_i := ih, \quad h := \frac{1}{N+1}.$$

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- 4. Solve the system:** For  $u_h = \sum_{i=1}^N \xi_i \phi_i$ ,  $\bar{A}\xi = \bar{L}$ , where  $\bar{A}_{i,j} := A(\phi_i, \phi_j)$ ,  $\bar{L}_i := L(\phi_i)$ ,  $i, j = 1, \dots, N$ .