

Applications Computational Finance, II

Goals

- | Overview of methods to price options
- | Directions on how to proceed (resources, research, programming)
- | Using C++ code and libraries to 'bootstrap' your efforts
- | Learning to map algorithms to code

Main Methods

- | PDE methods (FDM, FEM, meshless, spectral)
- | Integral methods (Conv, Cos, Fourier)
- | Simulation methods (Monte Carlo, Lattice (binomial, trinomial))
- | Exact/closed solutions

3

Some Use Cases

- | Price options (equity, fixed income, interest rate, hybrid)
- | Hedging and option sensitivities
- | Calibration and parameter estimation

4

PDE Methods (FDM) (1/2)

- | Option model described as a partial differential equation (PDE) defined in some region of (S,t) space
- | Can be one-factor or multi-factor equations
- | Describe any *contingent claim* that 'depends' on some *underlying variable*
- | Most PDEs are of *convection-diffusion* type

5

PDE Methods (FDM) (2/2)

- | Approximate PDE by finite difference methods (FDM)
- | Solve discrete assembled equations by matrix solvers
- | We discretise in space and time (different methods)
- | Stability and accuracy of finite difference schemes

6

Main FDM Categories

- | Alternating Direction Implicit (ADI)
- | Method of Fractional Steps ("Soviet Splitting")
- | Alternating Direction Explicit (ADE)
- | Others, sub-themes and variations

7

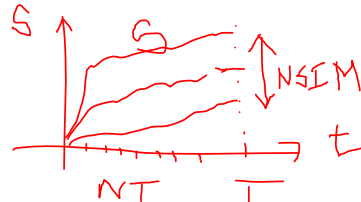
Stochastic Differential Equations (SDE)

- | Describe random behaviour of underlying variables (one-factor, n-factor)
- | No closed solution in general
- | Use FDM to approximate the SDE
- | Use SDE as basis for Monte Carlo and lattice methods

8

Monte Carlo

- | A simulation method; based in multiple times doing a 'SDE-FDM' trajectory
- | Geometric Brownian Motion (GBM); random walk on $[0, T]$
- | Take multiple samples
- | Discount summed values from $t = T$ to $t = 0$



9

Lattice Methods

- | Similar to Monte Carlo in spirit and structure
- | We trace evolution of the underlying in time (it can go up or down) up to $t = T$ (*forward induction*)
- | We then carry out *backward induction* from $t = T$ to $t = 0$
- | The price at $t = 0$ is the option price

10

How do we evaluate a Method?

- | Applicability to a wide range of derivatives types
- | Robustness (works for all parameters), accuracy and efficiency
- | Ease of use/implementation
- | Maintainability (becomes important when software system starts to stabilise)

11

FDM

- | Accuracy and efficiency are advantages
- | Difficulty in setting up the schemes and fine-tuning
- | FDM for n-factor problems difficult for $n = 4, 5, \dots$
- | Ideally, applied/numerical mathematics (or similar) background needed

12

Monte Carlo

- | Applicable to a wide range of option types
- | Easy to apply (a *blunt instrument*)
- | Difficulty with computation of option sensitivities and early exercise
- | Slow convergence, not very accurate method
- | Mathematical foundations not as strong as with PDE methods

13

Exact Solutions

- | No numerical approximations needed (in principle)
- | In a form that is easily computed, in principle
- | Only applicable to a given range of the parameter set
- | It might forever to come up with a solution

14

FDM Building Blocks

- | Creating meshes in S and t
- | Vectors and matrices
- | Data structures for PDE and FDM entities

15

MC Building Blocks

- | Random number generators
- | Data structures for SDE and FDM
- | Vectors and matrices
- | Payoff functions

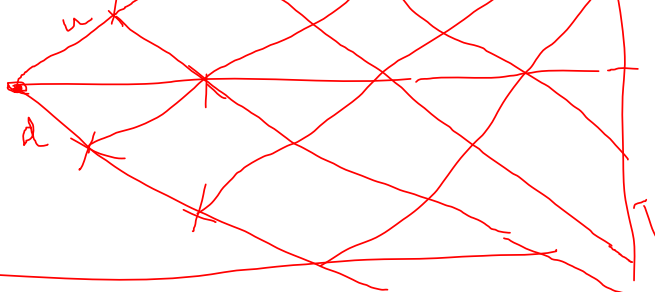
16

Lattice Building Blocks

- | Lattice data structures
- | Algorithms to compute jumps
- | Payoff entities

Bwd

Fwd



17