# Scientific Computing Understanding and solving stochastic PDE

USHAKOVA Oxana

August 5, 2016

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#### **Notations**

- *S* Stock price, also called Spot price (or any underlying asset)
- V(S,t) value of an option, depending on time and spot price
- K Strike price
- r risk-free rate
- d dividend yield
- ullet  $\mu$  drift rate of S the rate at which the average of S changes
- $\bullet$   $\sigma$  volatility of the stock, standard deviation of log(S) return on stock
- T<sub>0</sub>, T initial and final time
- $\theta$  long variance : as t tends to infinity, the expected value of  $\nu$  tends to  $\theta$
- ullet  $\kappa$  the rate at which u reverts to heta
- $\xi$  the volatility of volatility



## Financial mathematics

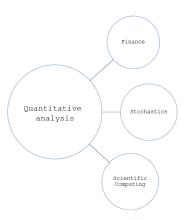


Figure: Financial mathematics



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# What is an option?

An option - a contract that gives the buyer the right, **but not the obligation**, to buy or sell an underlying asset (a stock, a bond, gold, other option) at a specific price, called Strike price, on a certain date, called maturity.

What right is proposed?

- The right to buy Call option
- The right to sell Put option



# What are option parameters?

#### Parameters to fix at $t_0$ :

- Who buys (long), who sells (short)?
- What is the underlying asset ?
- What is the maturity T of the contract ?
- Does the contract give the right to buy (call option) or to sell(put option)?
- What is the Strike price K?
- What is the price of the option itself, i.e. premium?



# Long Call Payoff

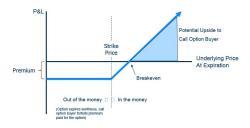


Figure: Options Payoffs

Buying Call Option brings profit (in the money), if S at maturity T is higher than the Breakeven point. So the payoff of this option is: (premium ignored):

$$Payoff = \begin{cases} S - K & \text{if } S(T) > K \\ 0 & \text{if not} \end{cases}$$

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# Other Payoffs

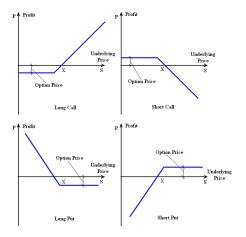


Figure: Other Payoffs



## Vanilla VS Exotics

## Vanilla Option

European option

#### **Exotic Option**

- American option (Bermudian)
- Barrier option (Paris)
- Asian option
- Lookback option (Russian)
- Binary option
- Cliquer option
- etc, etc..



## Greeks

**Delta** - the rate of change of the option price w.r.t. the price of the underlying asset:  $\delta = \frac{\partial C}{\partial S}$ 

**Gamma** - the rate of change of the  $\delta$  w.r.t. the price of the underlying asset:  $\Gamma = \frac{\partial^2 \Pi}{\partial C^2}$ 

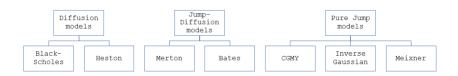
**Theta** - the rate of change of the portfolio price w.r.t. the time:  $\theta = \frac{\partial \Pi}{\partial t}$ 

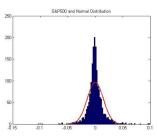
**Vega** - the rate of change of the portfolio price w.r.t. the volatility of the underlying asset:  $v = \frac{\partial \Pi}{\partial \sigma}$ 

**Rho** - the rate of change of the portfolio price w.r.t. the interest rate: $v=\frac{\partial\Pi}{\partial r}$ 



# Pricing models

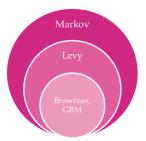




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# What is a Wiener process

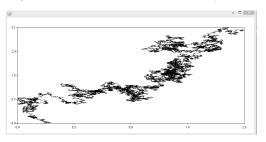
- Lévy processes: independent, stationary increments
- Markov processes: "memoryless"



**BUT**: no jumps ?  $\rightarrow$  add Compound Poisson process!

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Figure: Wiener and Wiener-Poisson process



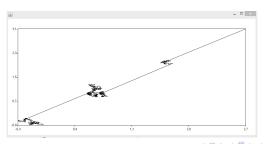
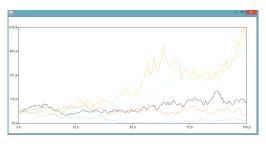
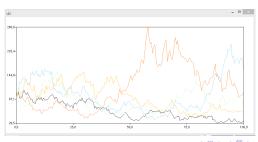


Figure: Black Scholes and Merton





## Ito's lemma and Black-Scholes

#### Ito's lemma:

 $X_t$  given by:  $dX_t = udt + vdB_t$ ,  $f(t,x) \in C^2$ ,  $Y_t = f(t,X_t)$ .

$$dY_t = \frac{\partial f}{\partial t}(t, X_t)dt + \frac{\partial f}{\partial x}(t, X_t)dB_t + \frac{1}{2}\frac{\partial^2 f}{\partial x^2}(t, X_t)(dB_t)^2$$
 (1)

#### Applying to Black-Scholes:

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$$dS_t = \mu S_t dt + \sigma S_t dB_t, X_0 > 0$$
 (2)

with  $f(t,x) = Inx, f \in C^2$  and  $Y_t = InS_t$ , gives:

$$\int_0^T dY_t = \left(\mu - \frac{1}{2}\sigma^2\right)T + \sigma B_t \tag{3}$$

Or

$$S_T = e^{Y_T} = S_0 e^{(\mu - \frac{1}{2}\sigma^2)T + \sigma dB_t}$$
 (4)

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## Technical aspects

- Derivation
  - Economic approach
  - From Heat PDE
- Numerical methods:
  - Monte-Carlo (Box-Muller Algorithm)
  - Tree methods (Cox-Ross-Rubenstein model)
  - Solving PDE
- $S_t$  follows Geometric Brownian motion:  $dS_t = S_t \mu dt + S_t \sigma dB_t$
- Put-Call Parity:

$$C + Ke^{-rT} = P + S_0 (5)$$

# Why use the Geometric Brownian motion?

- The expected returns of GBM are independent of the value of the process (stock price), which agrees with what we would expect in reality.
- A GBM process only assumes positive values, just like real stock prices.
- A GBM process shows the same kind of 'roughness' in its paths as we see in real stock prices.
- Calculations with GBM processes are relatively easy.



# GBM path in C++ with gnuplot

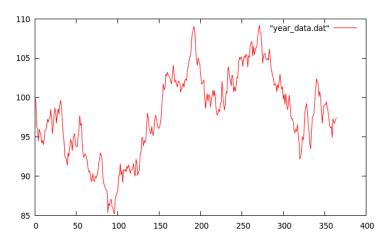


Figure : GBM path



# Black-Scholes analytical solution

With Ito's lemma we find:

$$S_T = e^{Y_T} = e^{Y_0 + (\mu - \frac{1}{2}\sigma^2)T + \sigma dB_t} = S_0 e^{(\mu - \frac{1}{2}\sigma^2)T + \sigma dB_t}$$
 (6)

and

$$C = e^{-dT} S_0 \mathcal{N}(d_1) - e^{-rT} \mathcal{K} \mathcal{N}(d_2)$$
 (7)

$$P = e^{-rT} K \mathcal{N}(-d_2)) - e^{-dT} S_0 \mathcal{N}(-d_1)$$
 (8)

with

$$d_1 = \frac{\ln(\frac{S_0}{K}) + (r - d + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$$

$$d_2 = \frac{\ln(\frac{S_0}{K}) + (r - d - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}} = d_1 - \sigma\sqrt{T}$$

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## Weak Formulation

Linear parabolic PDE with non-constant coefficients and non-homogenous boundary conditions and, possibly, non-differentiable or discontinuous final conditions:

- Choosing space : Weighted Sobolev  $\forall u \in V, V = \{v \in L^2(\mathbb{R}_+) : S \frac{\partial v}{\partial S} \in L^2(\mathbb{R}_+)\}$
- Bilinear form a:

$$a(u,w) = \int_0^\infty \frac{\partial u}{\partial S} \frac{\sigma^2}{2} \frac{S^2 w}{\partial S} - rSw \frac{\partial u}{\partial S} + ruw$$
 (9)

Weak formulation (for put):

$$(\frac{\partial u}{\partial t}, w) + a(u, w) = 0, \forall w \in V$$
 (10)

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## Black-Scholes PDE - Call and Put

$$\frac{\partial C}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 C}{\partial S(t)^2} + rS(t) \frac{\partial C}{\partial S(t)} - rC = 0$$
 (11)

$$C(0,t) = 0$$
  
 $C(S,t) = e^{-dT}S_0 - e^{-rT}K$  when  $S \longrightarrow \infty$   
 $C(S,T) = max(S - K,0)$ 

$$\frac{\partial P}{\partial t} - \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 P}{\partial S(t)^2} - rS(t) \frac{\partial P}{\partial S(t)} + rP = 0$$
 (12)

$$P(0,t) = e^{-rT}K$$
  
 $P(S,t) = 0$  when  $S \longrightarrow \infty$   
 $P(S,T) = max(K - S, 0)$ 

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## Black-Scholes PDE terms

$$\frac{\partial C}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 C}{\partial S(t)^2} + rS(t) \frac{\partial C}{\partial S(t)} - rC = 0$$
 (13)

- Diffusion term:  $\frac{1}{2}\sigma^2 S^2 \frac{\partial^2 C}{\partial S(t)^2}$
- Convection term:  $rS(t) \frac{\partial C}{\partial S(t)}$

Peclet number :  $\frac{diff}{conv} = \frac{rS}{\frac{1}{2}\sigma^2S^2} = \frac{2rS}{\sigma^2S^2} = \frac{2r}{\sigma^2S}$ 

**Remarque** :  $Pe < \frac{r}{\sigma^2}$ , if not, then small  $\sigma$  is not compensated by a small r.

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## Existence and Uniqueness

#### Martingales + Filtration + Ito Calculus

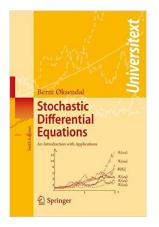


Figure: Oksendal - SDE (6ed)

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# Mesh adaptation and Delaunay Triangulation

Delanay algorithm: interpolation error is bounded by:

$$||u - u_h|| < C||\nabla(\nabla u)h^2$$
 (14)

- no obtuse triangles
- neighbour triangles with the same size

For each edge the circle circumscribing one triangle does not contain the fourth vertex.



# Vanilla put

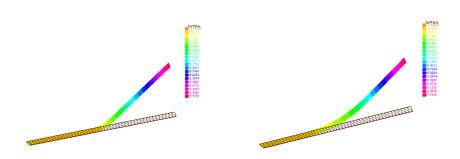
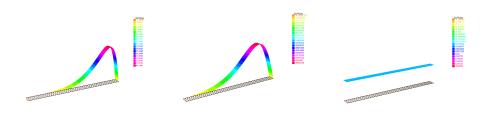


Figure :  $\sigma = 0.1$  and  $\sigma = 0.3$ 

# Barrier put : 30, 90, 100 with K=100



 $\textbf{Figure}: \ \mathsf{barriers} = 30, 90, 100$ 

# Asian put

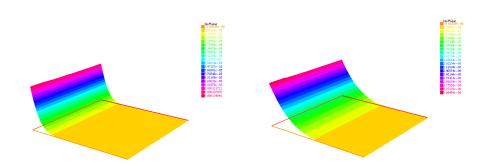
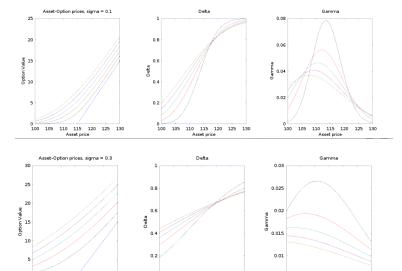


Figure :  $\mu = 0.1$  and  $\mu = 0.3$ 

## Delta and Gamma for $\sigma = 0.1, 0.3$





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110 115 120 125 130

Asset price

110 115 120 125 130

105

110 115 120 125 130

Asset price

0.005

## Delta for $\sigma = 0.1$

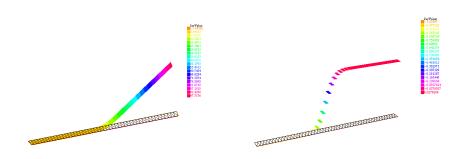


Figure :  $\sigma = 0.1$  and  $\sigma = 0.3$ 

## Delta for $\sigma = 0.3$

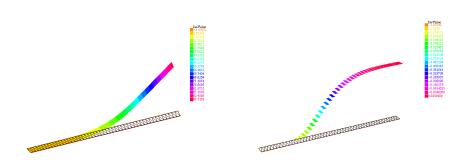
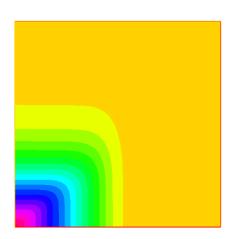
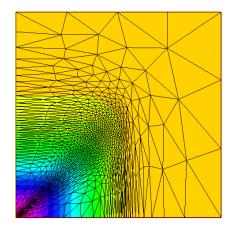


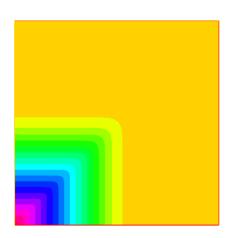
Figure : Delta for  $\sigma=0.1$  and  $\sigma=0.3$ 

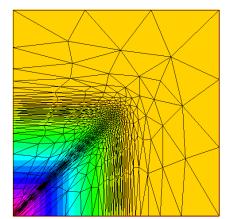
## Classic asymmetric data





## Low volatility with high correlation

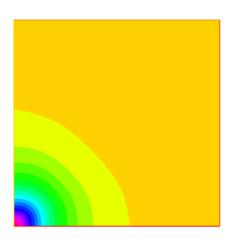


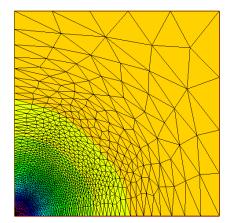


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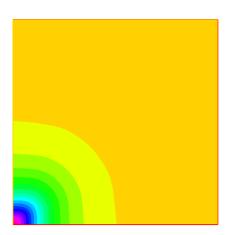
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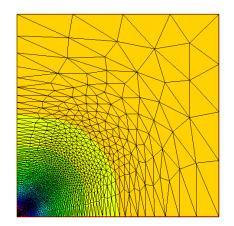
## High volatility but low correlation





## High volatility with high correlation





#### **Basics**

- Forward-Time Central-Space (FTCS) or Explicit
- Backward-Time Central-Space (BTCS)or Implicit
- Central-Time Central-Space (CTCS) or Crank-Nicolson

$$V_j^i = \frac{1}{1 + r\Delta t} (AV_{j+1}^{i+1} + BV_j^{i+1} + CV_{j-1}^{i+1})$$
 (15)

with

$$A = (\frac{1}{2}\sigma^2 j^2 + \frac{1}{2}rj)\Delta t$$

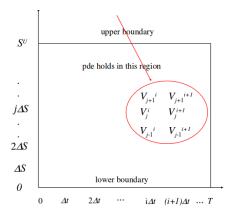
$$\bullet \ B = 1 - \sigma^2 j^2 \Delta t$$

$$C = (\frac{1}{2}\sigma^2 j^2 - \frac{1}{2}rj)\Delta t$$

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# "Explicit" grid

- ullet Space interval  $[0,S^U]$  in jmax intervals of length  $\Delta S=S^U/jmax$
- ullet Time interval [0,T] in imax intervals of length  $\Delta t = T/imax$
- ullet The value at each node  $V(j\Delta S,i\Delta t)\longrightarrow V^i_j$



# Stability

Explicit method  $\Rightarrow$  unstable!

Conditions of stability derived by using probabilistic approach.

If A, B, C as probabilities  $\Rightarrow$  positive.

- A and  $C o j > |rac{r}{\sigma^2}|$
- $B o \Delta t < \frac{1}{\sigma^2 j^2}$

Increase jmax by  $10 \rightarrow$  Increase imax by 100.



# FDM and Analytical solutions in C++

```
🙆 🖨 📵 ushakova@ushakova-SATELLITE-L70-B: ~/Desktop/Options/src
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/src$ q++ main.cpp
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/srcS ./a.out
S0=100 K=115 Imax=100 Jmax=10
PDE solution = 8.50637
Analytical solution = 8.33779
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/src$ q++ main.cpp
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/src$ ./a.out
S0=100 K=115 Imax=500 Jmax=50
PDE solution = 8.33924
Analytical solution = 8.33779
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/src$ q++ main.cpp
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/src$ ./a.out
S0=100 K=115 Imax=1000 Jmax=100
PDE solution = 8.34046
Analytical solution = 8.33779
ushakova@ushakova-SATELLITE-L70-B:~/Desktop/Options/srcS
```

Figure: FDM and Analytical solutions

# Comparing FDM and FEM for SDE

- Why use FEM?
  - Much more accurate
  - Lots of free libraries
- Why use FDM?
  - No usual boundary problems
  - Easier to implement