

Scientific Computing

Understanding and solving stochastic PDE

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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

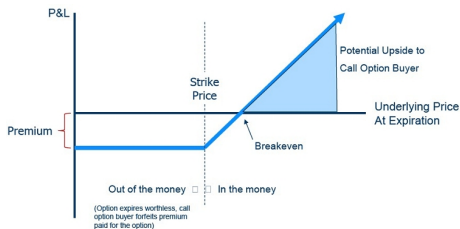


Figure : Options Payoffs

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):

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

Other Payoffs

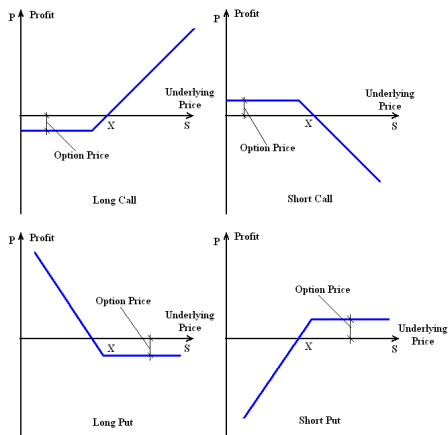


Figure : Other Payoffs

Vanilla VS Exotics

Vanilla Option

- European option

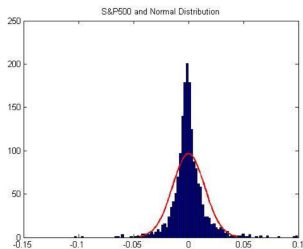
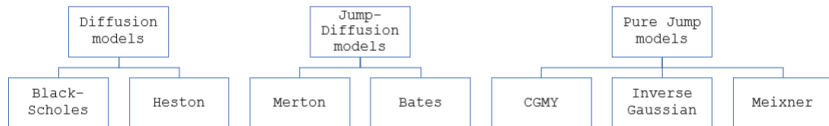
Exotic Option

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

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
- μ - drift rate of S - the rate at which the average of S changes
- σ - volatility of the stock, standard deviation of $\log(S)$ - return on stock
- T_0, T - initial and final time
- θ - long variance : as t tends to infinity, the expected value of ν tends to θ
- κ - the rate at which ν reverts to θ
- ξ - the volatility of volatility

Pricing models



What is a Wiener process

Wiener Process (Brownian Motion) is a stochastic process that lives in family of:

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

BUT: no jumps \rightarrow add Compound Poisson process

Figure : Wiener and Poisson process

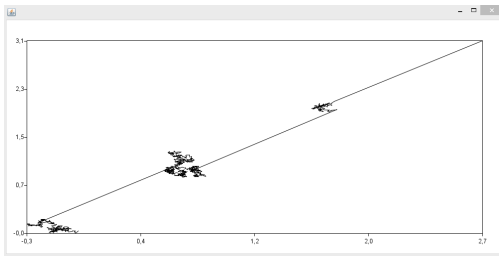
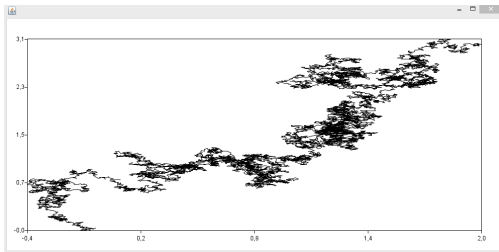
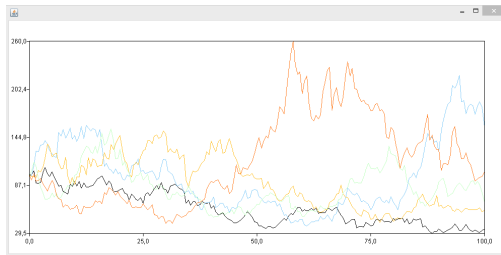
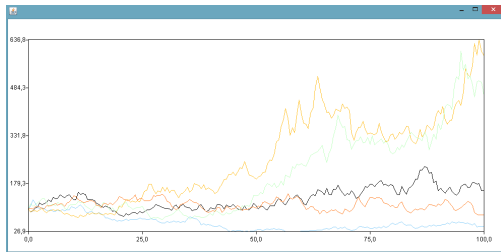


Figure : Black Scholes and Merton



Ito's lemma - derivation

- 1 X_t - stochastic process, that satisfies $dX_t = \mu dt + \sigma dB_t$.
- 2 Taylor series expansion for $f(t, x) \in C^2$:

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dx + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} dx^2 \dots \quad (1)$$

- 3 Change $x \longrightarrow X_t$, $dx \longrightarrow \mu dt + \sigma dB_t$:

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} (\mu dt + \sigma dB_t) + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} (\mu^2 dt^2 + 2\mu\sigma dt dB_t + \sigma^2 dB_t^2) + \dots \quad (2)$$

- 4 Using $dt * dt = dt * dB_t = dB_t * dt = 0$, $dB_t * dB_t = dt$, we get:

$$df = \left(\frac{\partial f}{\partial t} + \mu \frac{\partial f}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial x^2} \right) dt + \sigma \frac{\partial f}{\partial x} dB_t \quad (3)$$

Technical aspects

- 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:

$$C + Ke^{-rT} = P + S_0 \quad (4)$$

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.

Analytic Solution

With Ito's lemma we find :

$$S_T = e^{Y_T} = e^{Y_0 + (\mu - \frac{1}{2}\sigma^2)T + \sigma dB_t} = S_0 e^{(\mu - \frac{1}{2}\sigma^2)T + \sigma dB_t} \quad (5)$$

and

- $C = e^{-rT} \mathbb{E}[\max(S_0 e^{(r-d-\frac{\sigma^2}{2})T + \sigma B_T} - K, 0)]$

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

- $P = e^{-rT} \mathbb{E}[K - \max(S_0 e^{(r-d-\frac{\sigma^2}{2})T + \sigma B_T}, 0)]$

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

with

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

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

Black-Scholes PDE - Call and Put

$$\frac{\partial C}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 C}{\partial S(t)^2} + rS(t) \frac{\partial C}{\partial S(t)} - rC = 0 \quad (6)$$

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

$$C(S, t) = e^{-dT} S_0 - e^{-rT} K \text{ when } S \rightarrow \infty$$

$$C(S, T) = \max(S - E, 0)$$

$$\frac{\partial P}{\partial t} - \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 P}{\partial S(t)^2} - rS(t) \frac{\partial P}{\partial S(t)} + rP = 0 \quad (7)$$

$$P(0, t) = e^{-rT} K$$

$$P(S, t) = 0 \text{ when } S \rightarrow \infty$$

$$P(S, T) = \max(E - S, 0)$$

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

$$\forall u \in V, V = \{v \in L^2(\mathbb{R}_+) : S \frac{\partial v}{\partial S} \in L^2(\mathbb{R}_+)\}$$

- Coercive form a :

$$a(u, w) = \int_0^\infty \frac{\partial u}{\partial S} \frac{\sigma^2}{2} \frac{S^2 w}{\partial S} - rSw \frac{\partial u}{\partial S} + ruw \quad (8)$$

- Weak formulation (for put):

$$\left(\frac{\partial u}{\partial t}, w\right) + a(u, w) = 0, \forall w \in V \quad (9)$$

Existence and Uniqueness

Martingales + Filtration + Ito Calculus

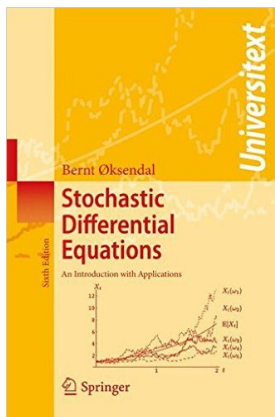


Figure : Øksendal - SDE (6ed)

Discretization

ADD!!!!!!!!!!!!!!

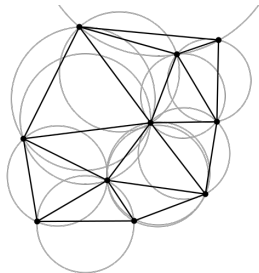
Mesh adaptation and Delaunay Triangulation

Delaunay algorithm keeps the error of interpolation bounded by:

$$\|u - u_h\| < C \|\nabla(\nabla u) h^2 \quad (10)$$

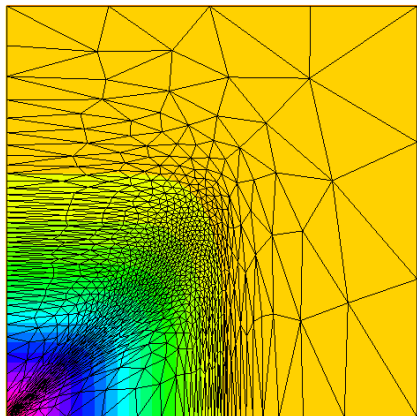
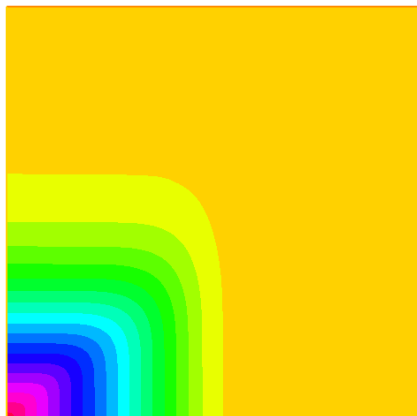
Delaunay triangulation helps to create a "good" mesh : no obtuse triangles, neighbor triangles have more or less the same size.

In other words, the Delaunay triangulation create a mesh where for each edge the circle circumscribing one triangle does not contain the fourth vertex.



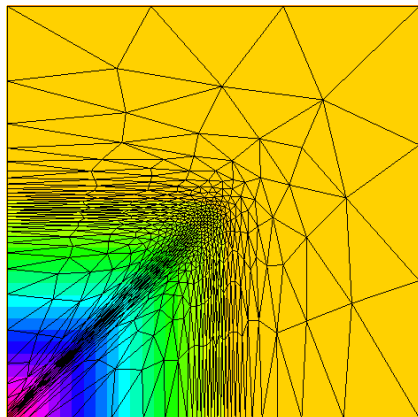
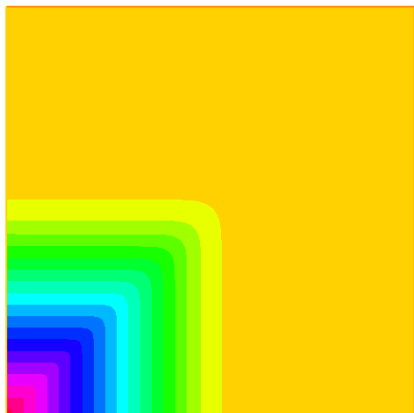
Freefem++ output

Classic asymmetric data



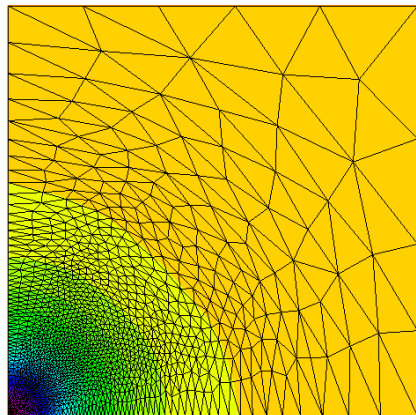
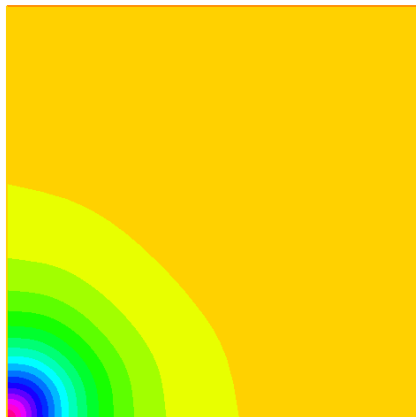
Freefem++ output

Low volatility with high correlation



Freefem++ output

High volatility but low correlation



Freefem++ output

High volatility with high correlation

