

## Heston Model

$S_t = \text{price}$

$$dS_t = \mu S_t dt + \sqrt{V_t} S_t dW_t$$

$$dV_t = \kappa(\theta - V_t) dt + \sigma \sqrt{V_t} dz_t$$

$$E(dW_t dz_t) = \rho dt$$

$W_t, z_t \sim \text{BM}^1$   
correlated  
w/  $\rho$

$\kappa > 0$  speed of mean reversion  
 $\theta$  mean reversion parameter  
 $\sigma$  volatility of variance

$$\left( \begin{array}{l} \frac{dS_t}{S_t} = - \\ X_t = \log S_t \\ \text{Ito adj for } X_t \end{array} \right.$$

CIR model (Cox-Ingersoll-Ross)

$$dV_t = k(\theta - V_t)dt + \sigma\sqrt{V_t}dz_t$$

$$\Delta t \rightarrow V_{t+\Delta t} - V_t = \underbrace{k(\theta - V_t)\Delta t} + \underbrace{\sigma\sqrt{V_t}\Delta z_t}_{\text{noise}}$$

$$\theta \frac{V_t \cdot V_{t+\Delta t}}{V_t}$$

$$dr_t = k(\theta - r_t)dt + \sigma\sqrt{r_t}dz_t$$

Vasicek model



$$dr_t = k(\theta - r_t)dt + \sigma r_t dz_t$$

For CIR model please read lecture notes.

$$dV_t = \kappa(\theta - V_t) dt + \sigma \sqrt{V_t} dz_t$$

calculate  $E(V_t)$   
 $E(V_t^2)$

$$V_t - V_0 = \int_0^t \kappa(\theta - V_0) ds + \int_0^t \sigma \sqrt{V_0} dz_s$$

$$E(V_t) - E(V_0) = E\left[\int_0^t \kappa(\theta - V_0) ds\right] + 0$$

$$E(V_t) - E(V_0) = \int_0^t E(\kappa(\theta - V_0)) ds$$

$$E[V_t] = \mu_t$$

$$\mu_t - \mu_0 = \int_0^t \kappa(\theta - \mu_0) ds$$

$$\frac{d\mu_t}{dt} = \kappa(\theta - \mu_t)$$

$$d\mu_t + \kappa \mu_t dt = \kappa \theta dt \quad | \quad e^{\kappa t}$$

... etc

Hint for  $V_t^2$   
 use Ito's lemma  
 first to get  $V_t^2$   
 dynamics

$$\mathbb{E}(V_t | V_0) = \theta + (V_0 - \theta)e^{-\kappa(t-\Delta)}$$

Heston PDE

$C(S, r, t)$  = price of an option at time  $t$   
 Stock value =  $S$   
 Variance value of  $V_t = v$

$$x = \ln S$$

$$u(x, r, t) = C(e^x, r, t)$$

$$\begin{aligned} \frac{\partial u}{\partial t} + \left(r - \frac{1}{2}v\right) \frac{\partial u}{\partial x} + \frac{1}{2}\sigma^2 \frac{\partial^2 u}{\partial x^2} + \rho\sigma v \frac{\partial^2 u}{\partial x \partial v} + \frac{1}{2}\sigma^2 v \frac{\partial^2 u}{\partial v^2} \\ - ru + (\kappa(\theta - r)) \frac{\partial u}{\partial v} = 0 \end{aligned}$$

SABR

$$dS_t = \alpha_t S_t^\beta dW_t$$

$$d\alpha_t = \nu \alpha_t dZ_t$$

$\rho$  corr ~~coef~~ between  $W$  &  $Z$  -

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Option price is given by plugging into the B-S formula a volatility value  $\sigma_{SABR}$ .

## Option Sensitivities

$$C(S, t, K, r, \sigma)$$

$$\delta = \frac{\partial C}{\partial S} =$$

$$\delta = \frac{\partial C}{\partial S}(S_0) = \frac{C(S_0 + \Delta S) - C(S_0)}{\Delta S}$$

## Quadrature methods

Methodology to approximate a definite integral

$$dx_t = \mu(t, x_t) dt + \sigma(t, x_t) dw_t$$

price of an European type derivative is:

$$F(t, x_t) = E \left[ e^{-\int_t^T r(s) ds} F(T, x_T) \mid \mathcal{F}_t \right]$$

$X_t$   $p(\Delta, x, t, y) =$  density of r.v.  $X_T \mid X_x = x$

$$F(t, x_t) = \int_{-\infty}^{\infty} e^{-\int_t^T r(s) ds} F(T, y) p(\Delta, x, t, y) dy$$

Problem: Consider  $I(f) = \int_A f(x) dx$

Def: A quadrature rule of order  $n$  is an expression:

$$I_n(f) = \sum_{i=1}^n w_i^n f(x_i^n)$$

$w_i^n = \text{weight}$

$x_i^n = \text{quadrature nodes}$

$$I_n(f) \longrightarrow I(f) \text{ as } n \rightarrow \infty$$

Basic rule of constructing a quadrature rule is

- approximate  $f$  using some interpolating function  $p_n$  (polynomials) s.t.

$$p_n(x_i^n) = f(x_i^n) \text{ for all } x_i^n$$

- Integrate  $p_n$  and return  $I(p_n)$  as  $I_n$  approx



## Rectangle rule (order 0)

Idea: approximate  $f$  with piecewise constant fct.

$$\int_a^b f(x) dx$$

$$x_0 = a$$

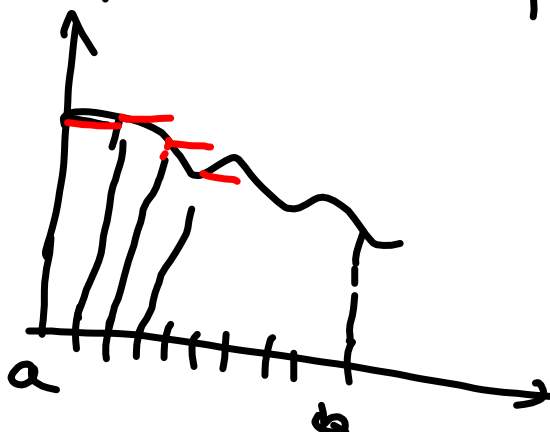
$$x_1 = a + \frac{b-a}{n}$$

$$x_2 = a + 2 \frac{b-a}{n}$$

$$x_n = a + n \frac{b-a}{n} = b$$

$$h = \frac{b-a}{n}$$

$$I_n(f) = \sum_{i=1}^n h f(x_{i-1}^n)$$



① If  $a = \infty$  or  $b = \infty$  we need a modification



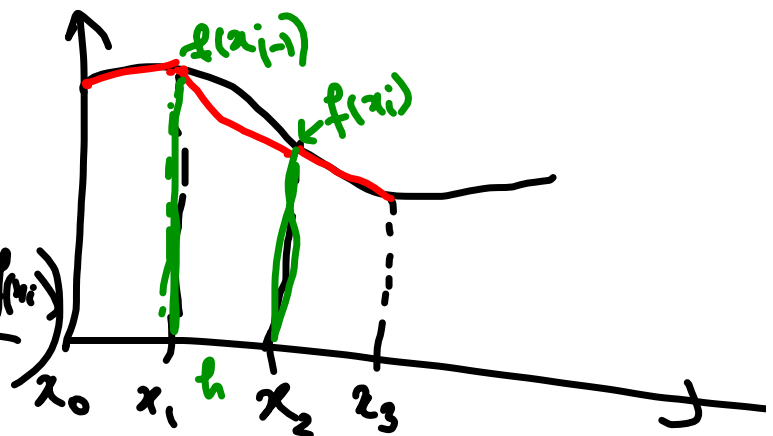
② there is nothing special about left hand point

Midpoint rule

$$I_n(f) = \sum_{i=1}^n h f\left(\frac{x_i + x_{i-1}}{2}\right)$$

## Trapezoid rule

$$I_n(f) = \sum_{i=1}^n h \cdot \left( \frac{f(x_{i-1}) + f(x_i)}{2} \right)$$



$$= h \left( \frac{f(x_0)}{2} + f(x_1) + f(x_2) + \dots + f(x_{n-1}) + \frac{1}{2} f(x_n) \right)$$

## Simpson's Rule

Use a quadratic polynomial

Construct a 2<sup>nd</sup> order polynomial so that

$$f(x_{i-1}) = P(x_{i-1})$$

$$f(x_i) = P(x_i)$$

$$f(x_{i+1}) = P(x_{i+1})$$

## Lagrange interpolating polynomials.

given  $(x_1, y_1) (x_2, y_2) \dots (x_k, y_k)$   
 $x_1 \neq x_2 \dots \neq x_k$

$$l_j(x) = \prod_{\substack{i=1 \\ i \neq j}}^k \frac{x - x_i}{x_j - x_i} = \frac{x - x_1}{x_j - x_1} \dots \frac{x - x_{j-1}}{x_j - x_{j-1}} \cdot \frac{x - x_{j+1}}{x_j - x_{j+1}} \dots \frac{x - x_k}{x_j - x_k}$$

$$l_j(x_i) = 0 \text{ for all } i \neq j$$

$$l_j(x_j) = 1$$

$$L(x) = \sum_{j=1}^k y_j \cdot l_j(x) = y_1 l_1(x) + \dots + y_k l_k(x)$$

Our problem

$$P(x) = f(x_{i-1}) \frac{x-x_i}{x_{i-1}-x_i} \frac{x-x_{i+1}}{x_{i-1}-x_{i+1}} + \dots$$

Next integrate  $\int_{x_{i-1}}^{x_{i+1}} P(x) dx$

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx \approx \frac{x_{i+1}-x_{i-1}}{6} (f(x_{i-1}) + 4f(x_i) + f(x_{i+1}))$$

Divide interval in  $x_0, x_1, \dots, x_n$

use  $x_{i-1}, \frac{x_{i-1}+x_i}{2}, x_i$

$$I(f) = \frac{h}{6} \sum_{i=1}^n \left( f(x_{i-1}) + 4f\left(\frac{x_{i-1}+x_i}{2}\right) + f(x_i) \right)$$