

Black-Scholes Equations

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Itô Process: $dx_t = a dt + \sigma dB_t$

Special Case: $a(t, x) = a(t, x_t)$

$\sigma(t, x) = \sigma(t, x_t)$

ii) \tilde{A} $dx_t = a(t, x_t)dt + \sigma(t, x_t)dB_t$ SDE Stochastic Diff Eq

B-S: $a(t, x) = rx$ $r \in \mathbb{R}$
 $\sigma(t, x) = \sigma_0 x$ $\sigma_0 \in \mathbb{R}$

Using SDE \tilde{A} $dx_t = rx_t dt + \sigma_0 x_t dB_t$

Wf: $\frac{dx_t}{x_t} = r dt + \sigma_0 dB_t$

$\int_0^t \int_0^s (dx_t)^2 = \sigma_0^2 x_t^2 dt$

Then $d \log x_t = \frac{dx_t}{x_t} - \frac{1}{2} \sigma_0^2 dt$

Then $d \log x_t + \frac{1}{2} \sigma_0^2 dt = \frac{dx_t}{x_t} = r dt + \sigma_0 dB_t$

Integrate $\log \frac{x_T}{x_0} = (r - \frac{\sigma_0^2}{2})T + \sigma_0 B_T$

Wf: $x_T = x_0 e^{(r - \frac{\sigma_0^2}{2})T + \sigma_0 B_T}$

log normal Dist

Call Option: $(x_T - K)_+$

Price = $e^{-rT} \mathbb{E}(x_T - K)_+$

= $x_0 N(d_1) - Ke^{-rT} N(d_2)$

$d_1 = \frac{\log(x_0/K) + \sigma^2 T/2}{\sigma \sqrt{T}}$

$d_2 = d_1 - \sigma \sqrt{T}$

N : cumulative normal function

$\frac{dx_t}{x_t} \neq d \log x_t$!

$f(t, x) = \log x$

$\Rightarrow f_t = 0$

$f_x = \frac{1}{x}$

$f_{xx} = -\frac{1}{x^2}$

$\Rightarrow d \log x_t = \frac{dx_t}{x_t} - \frac{1}{2x_t^2} (dx_t)^2$

Itô \rightarrow

$dx_t = rx_t dt$

$\Rightarrow x_T = x_0 e^{rT}$