# QF620 - Stochastic Modelling in Finance

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#### **Abstract**

This report encapsulates the fundamentals of Stochastic Modelling in a condensed report by applying theories including but not limited to, Black Scholes model, Bachelier model, Black76 model, Displaced-diffusion model. Through the application of computational and programming methodologies, the team managed to construct the various models and calibrate them to match the option prices. Using the output generated from the various models, the team analysed the results and elaborated on its underlying significance.

### 1. Analytical Option Formulae

### 1.1 Black-Scholes Vanilla Call Option

The Black-Scholes model for the stock price process is defined as

$$dS_t = rS_t dt + \sigma S_t dW_t$$

By solving the stochastic differential equation using Itô's formula, we can get

$$S_t = S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma W_t}$$

Vanilla European call option price is derived as below.

$$V_o^c = e^{-rT} E[(S_t - K)^+]$$

$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} - K)^+ e^{-\frac{x^2}{2}} dx$$

Here,

$$S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} > K$$

$$x > \frac{\log\left(\frac{K}{S_0}\right) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} = x^*$$

Then,

$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \left[ S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T} \int_{x^*}^{\infty} e^{\sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx - K \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \right]$$

$$= e^{-rT} \left[ S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T} \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2 - 2\sigma\sqrt{T}x + \sigma^2T - \sigma^2T}{2}} dx - K\Phi(-x^*) \right]$$

$$= S_0 \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{(x - \sigma\sqrt{T})^2}{2}} dx - Ke^{-rt}\Phi(-x^*)$$

Let

$$y = x - \sigma\sqrt{T} \Rightarrow dy = dx$$
$$x = x^*, y = x^* - \sigma\sqrt{T}$$

Then,

$$\begin{split} V_o^c &= S_0 \frac{1}{\sqrt{2\pi}} \int_{x^* - \sigma\sqrt{T}}^{\infty} e^{-\frac{y^2}{2}} dy - Ke^{-rT} \Phi(-x^*) \\ &= S_0 \Phi(-x^* + \sigma\sqrt{T}) - Ke^{-rT} \Phi(-x^*) \\ V_o^c &= S_0 \Phi\left(\frac{\log(\frac{S_0}{K}) + (r + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}\right) - \\ Ke^{-rT} \Phi\left(\frac{\log(\frac{S_0}{K}) + (r - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}\right) \end{split}$$

#### 1.2 Black-Scholes Vanilla Put Option

$$V_o^p = e^{-rT} E[(K - S_t)^+]$$

$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (K - S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x})^+ e^{-\frac{x^2}{2}} dx$$

Here,

$$S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} < K$$

$$x < \frac{\log\left(\frac{K}{S_0}\right) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} = x^*$$

Then,

$$V_o^p = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} \left( K - S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} \right) e^{-\frac{x^2}{2}} dx$$
$$= e^{-rT} K \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} e^{-\frac{x^2}{2}} dx - \frac{1}{2\pi} \int_{-\infty}^{x^*} e^{-\frac{x^2}{$$

$$e^{-rT}S_0 \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx$$

$$= e^{-rT}K\Phi(x^*) - S_0 e^{-\frac{1}{2}\sigma^2T} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} e^{\sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx$$

$$= e^{-rT}K\Phi(x^*) - S_0 \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} e^{-\frac{(x - \sigma\sqrt{T})^2}{2}} dx$$

Let

$$y = x - \sigma\sqrt{T} \Rightarrow dy = dx$$
$$x = x^*, y = x^* - \sigma\sqrt{T}$$

Then,

$$\begin{split} V_o^p &= e^{-rT} K \Phi(x^*) - S_0 \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^* - \sigma\sqrt{T}} e^{-\frac{y^2}{2}} dy \\ &= e^{-rT} K \Phi(x^*) - S_0 [\Phi(x^* - \sigma\sqrt{T})] \\ V_o^p &= e^{-rT} K \Phi\left(\frac{\log(\frac{K}{S_0}) - (r - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}\right) - \\ S_0 \Phi\left(\frac{\log(\frac{K}{S_0}) - (r + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}\right) \end{split}$$

### 1.3 Black-Scholes Digital Cash-or-Nothing Call

$$V_{CashDigital}^{c}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \mathbf{1}_{S_{T} > K} e^{-\frac{x^{2}}{2}} dx$$
$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{x^{*}}^{\infty} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}$$

Therefore,

$$V^{c}_{CashDigital}(0) = e^{-rT} \Phi \left( \frac{\log(\frac{S_0}{K}) + \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} \right)$$

### 1.4 Black-Scholes Digital Cash-or-Nothing Put

$$V_{CashDigital}^{p}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \mathbf{1}_{K > S_{T}} e^{-\frac{x^{2}}{2}} dx$$
$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^{*}} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}$$

Therefore,

$$V_{CashDigital}^{p}(0) = e^{-rT} \Phi\left(\frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}\right)$$

### 1.5 Black-Scholes Digital Asset-or-Nothing Call

$$V_{AssetDigital}^{c}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} S_{T} \mathbf{1}_{S_{T} > K} e^{-\frac{x^{2}}{2}} dx$$
$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{x^{*}}^{\infty} S_{T} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}$$

Therefore,

$$V^{c}_{AssetDigital}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx$$

From the same process as Black-Scholes vanilla call option,

$$V^{c}_{AssetDigital}(0) = S_0 \Phi \left( \frac{\log(\frac{S_0}{K}) + (r + \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right)$$

#### 1.6 Black-Scholes Digital Asset-or-Nothing Put

$$V_{AssetDigital}^{p}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} S_{T} \mathbf{1}_{K > S_{T}} e^{-\frac{x^{2}}{2}} dx$$
$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^{*}} S_{T} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}$$

Therefore,

$$V_{AssetDigital}^{p}(0) = e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx$$

From the same process as Black-Scholes vanilla put option,

$$V_{AssetDigital}^{p}(0) = S_0 \Phi \left( \frac{\log(\frac{K}{S_0}) - (r + \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right)$$

#### 1.7 Bachelier Vanilla Call Option

The Bachelier model for the stock price process is defined as

$$dS_t = \sigma S_0 dW_t$$
  $S_T = S_0 (1 + \sigma W_T), \qquad W_T \sim N(0, T)$ 

Vanilla European call option price is derived as below.

$$V_o^c = E[(S_t - K)^+]$$

$$= E[(S_0(1 + \sigma W_T) - K)^+]$$

$$= E[(S_0(1 + \sigma \sqrt{T}x) - K)^+], \qquad X \sim N(0, 1)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (S_0 + \sigma S_0 \sqrt{T} x - K)^+ e^{-\frac{x^2}{2}} dx$$

Here,

$$S_0 + \sigma S_0 \sqrt{T}x - K > 0$$
$$x > \frac{K - S_0}{\sigma S_0 \sqrt{T}} = x^*$$

Then,

$$V_o^c = \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} (S_0 + \sigma S_0 \sqrt{T} x - K) e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} \left( \int_{x^*}^{\infty} (S_0 - K) e^{-\frac{x^2}{2}} dx + \int_{x^*}^{\infty} \sigma S_0 \sqrt{T} x e^{-\frac{x^2}{2}} dx \right)$$

$$= (S_0 - K) \left[ \Phi(\infty) - \Phi(x^*) \right] + \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} \sigma S_0 \sqrt{T} x e^{-\frac{x^2}{2}} dx$$

$$\text{Let } u = -\frac{x^2}{2}, du = -x dx$$

$$V_o^c = (S_0 - K) \Phi(-x^*) - \sigma S_0 \sqrt{T} \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^u du$$

$$= (S_0 - K) \Phi(-x^*) - \sigma S_0 \sqrt{T} \frac{1}{\sqrt{2\pi}} \left[ e^{-\frac{x^2}{2}} \right]_{x^*}^{\infty}$$

$$= (S_0 - K) \Phi(-x^*) + \sigma S_0 \sqrt{T} \phi(-x^*)$$

### 1.8 Bachelier Vanilla Put Option

$$V_o^p = E[(K - S_t)^+]$$

$$= E[(K - S_0(1 + \sigma W_T))^+]$$

$$= E[(K - S_0(1 + \sigma \sqrt{T}x))^+], \qquad X \sim N(0, 1)$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (K - S_0 - \sigma S_0 \sqrt{T}x)^+ e^{-\frac{x^2}{2}} dx$$

 $=(S_0-K)\Phi\left(\frac{S_0-K}{\sigma S_0\sqrt{T}}\right)+\sigma S_0\sqrt{T}\phi\left(\frac{S_0-K}{\sigma S_0\sqrt{T}}\right)$ 

Here,

$$K - S_0 - \sigma S_0 \sqrt{T}x > 0$$
$$x < \frac{K - S_0}{\sigma S_0 \sqrt{T}} = x^*$$

Then,

$$V_o^p = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} (K - S_0 - \sigma S_0 \sqrt{T}x) e^{-\frac{x^2}{2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} \left( \int_{-\infty}^{x^*} (K - S_0) e^{-\frac{x^2}{2}} dx - \int_{-\infty}^{x^*} \sigma S_0 \sqrt{T}x e^{-\frac{x^2}{2}} dx \right)$$

$$= (K - S_0) \Phi(x^*) - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} \sigma S_0 \sqrt{T}x e^{-\frac{x^2}{2}} dx$$

Let 
$$u = -\frac{x^2}{2}$$
,  $du = -xdx$   

$$V_o^p = (K - S_0)\Phi(x^*) + \sigma S_0 \sqrt{T} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{x^*} e^u du$$

$$= (K - S_0)\Phi(x^*) + \sigma S_0 \sqrt{T} \frac{1}{\sqrt{2\pi}} \left[ e^{-\frac{x^2}{2}} \right]_{-\infty}^{x^*}$$

$$= (K - S_0)\Phi(x^*) + \sigma S_0 \sqrt{T} \phi(x^*)$$

$$= (K - S_0)\Phi\left(\frac{K - S_0}{\sigma S_0 \sqrt{T}}\right) + \sigma S_0 \sqrt{T} \phi\left(\frac{K - S_0}{\sigma S_0 \sqrt{T}}\right)$$

# 1.9 Bachelier Digital Cash-or-Nothing Call Option

$$V_{CashDigital}^{c}(0) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \mathbf{1}_{S_T > K} e^{-\frac{x^2}{2}} dx$$
$$= \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx$$

where,

$$x^* = \frac{K - S_0}{\sigma S_0 \sqrt{T}}$$

Therefore,

$$V_{CashDigital}^{c}(0) = \Phi\left(\frac{S_0 - K}{\sigma S_0 \sqrt{T}}\right)$$

### 1.10 Bachelier Digital Cash-or-Nothing Put Option

$$V_{CashDigital}^{p}(0) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \mathbf{1}_{K>S_T} e^{-\frac{x^2}{2}} dx$$
$$= \frac{1}{\sqrt{2\pi}} \int_{\infty}^{x^*} e^{-\frac{x^2}{2}} dx$$

where,

$$x^* = \frac{K - S_0}{\sigma S_0 \sqrt{T}}$$

Therefore,

$$V_{CashDigital}^{p}(0) = \Phi\left(\frac{K - S_0}{\sigma S_0 \sqrt{T}}\right)$$

### 1.11 Bachelier Digital Asset-or-Nothing Call Option

$$V_{AssetDigital}^{c}(0) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} S_{T} \mathbf{1}_{S_{T} > K} e^{-\frac{x^{2}}{2}} dx$$
$$= \frac{1}{\sqrt{2\pi}} \int_{x^{*}}^{\infty} S_{T} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{K - S_0}{\sigma S_0 \sqrt{T}}$$

Therefore,

$$V^{c}_{Asset Digital}(0) = rac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} S_0(1+\sigma\sqrt{T}x) e^{-rac{x^2}{2}} dx$$

From the same process as Bachelier vanilla call option,

$$V^{c}_{AssetDigital}(0) = S_0 \Phi \left( \frac{S_0 - K}{\sigma S_0 \sqrt{T}} \right) + \sigma S_0 \sqrt{T} \phi \left( \frac{S_0 - K}{\sigma S_0 \sqrt{T}} \right)$$

#### 1.12 Bachelier Digital Asset-or-Nothing Put Option

$$V_{AssetDigital}^{p}(0) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} S_{T} \mathbf{1}_{K>S_{T}} e^{-\frac{x^{2}}{2}} dx$$
$$= \frac{1}{\sqrt{2\pi}} \int_{\infty}^{x^{*}} S_{T} e^{-\frac{x^{2}}{2}} dx$$

where,

$$x^* = \frac{K - S_0}{\sigma S_0 \sqrt{T}}$$

Therefore,

$$V_{Asset Digital}^{p}(0) = \frac{1}{\sqrt{2\pi}} \int_{\infty}^{x^*} S_0(1 + \sigma\sqrt{T}x)e^{-\frac{x^2}{2}} dx$$

From the same process as Bachelier vanilla put option,

$$V_{AssetDigital}^{p}(0) = S_0 \Phi\left(\frac{K - S_0}{\sigma S_0 \sqrt{T}}\right) + \sigma S_0 \sqrt{T} \phi\left(\frac{K - S_0}{\sigma S_0 \sqrt{T}}\right)$$

#### 1.13 Black76 Vanilla Call Option

$$dF_t = \sigma F_t dW_t$$
$$F_t = F_0 e^{-\frac{\sigma^2 T}{2} + \sigma W_t}$$

$$F_0 = S_0 e^{rT}$$

Vanilla European call option price is derived as below.

$$V_o^c = e^{-rT} E[(F_t - K)^+]$$

$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (F_0 e^{-\frac{\sigma^2 T}{2} + \sigma\sqrt{T}x} - K)^+ e^{-\frac{x^2}{2}} dx$$

$$= e^{-rT} \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} - K)^+ e^{-\frac{x^2}{2}} dx$$

The result above is exactly same as Black-Scholes model. Therefore, when we rearrange Black-Scholes model for forward price, we can get Black76 model.

Black-Scholes Model:

$$V_o^c = S_0 \Phi \left( \frac{\log(\frac{S_0}{K}) + (r + \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right) -$$

$$Ke^{-rT}\Phi\left(\frac{\log(\frac{S_0}{K})+\left(r-\frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}}\right)$$

Black76 Model:

$$V_o^c = e^{-rT} \left[ F_0 \Phi \left( \frac{\log(\frac{F_0}{K}) + \frac{1}{2}\sigma^2 T}{\sigma\sqrt{T}} \right) - K \Phi \left( \frac{\log(\frac{F_0}{K}) - \frac{1}{2}\sigma^2 T}{\sigma\sqrt{T}} \right) \right]$$

#### 1.14 Black76 Other options

As shown in Black76 Vanilla call option, when we rearrange Black-Scholes models for forward price, we can get Black76 model.

Black76 Vanilla Put Option:

$$\begin{split} V_o^p &= \\ e^{-rT} K \Phi \left( \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right) T}{\sigma \sqrt{T}} \right) - \\ S_0 \Phi \left( \frac{\log(\frac{K}{S_0}) - \left(r + \frac{\sigma^2}{2}\right) T}{\sigma \sqrt{T}} \right) \\ &= e^{-rT} \left[ K \Phi \left( \frac{\log(\frac{K}{F_0}) + \frac{1}{2}\sigma^2 T}{\sigma \sqrt{T}} \right) - F_0 \Phi \left( \frac{\log(\frac{K}{F_0}) - \frac{1}{2}\sigma^2 T}{\sigma \sqrt{T}} \right) \right] \end{split}$$

Black76 Digital Cash-or-Nothing Call Option:

$$\begin{split} V^{c}_{CashDigital}(0) &= e^{-rT} \Phi \left( \frac{\log(\frac{S_{0}}{K}) + \left(r - \frac{\sigma^{2}}{2}\right)T}{\sigma \sqrt{T}} \right) \\ &= e^{-rT} \Phi \left( \frac{\log(\frac{F_{0}}{K}) - \frac{1}{2}\sigma^{2}T}{\sigma \sqrt{T}} \right) \end{split}$$

Black76 Digital Cash-or-Nothing Put Option:

$$\begin{split} V_{CashDigital}^{p}(0) &= e^{-rT} \Phi \left( \frac{\log(\frac{K}{S_0}) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma \sqrt{T}} \right) \\ &= e^{-rT} \Phi \left( \frac{\log(\frac{K}{F_0}) + \frac{1}{2}\sigma^2 T}{\sigma \sqrt{T}} \right) \end{split}$$

Black76 Digital Asset-or-Nothing Call Option:

$$V^{c}_{AssetDigital}(0) = S_0 \Phi \left( \frac{\log(\frac{S_0}{K}) + (r + \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right)$$

$$= e^{-rT} F_0 \Phi \left( \frac{\log(\frac{F_0}{K}) + \frac{1}{2} \sigma^2 T}{\sigma \sqrt{T}} \right)$$

Black76 Digital Asset-or-Nothing Put Option:

$$V_{AssetDigital}^{p}(0) = S_0 \Phi \left( \frac{\log(\frac{K}{S_0}) - (r + \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right)$$
$$= e^{-rT} F_0 \Phi \left( \frac{\log(\frac{K}{F_0}) - \frac{1}{2}\sigma^2 T}{\sigma \sqrt{T}} \right)$$

### 1.15 Displaced-diffusion Model

The Displaced-diffusion model for the forward price process is defined as

$$dF_t = \sigma[\beta F_t + (1 - \beta)F_0]dW_t$$

Let

$$X_{t} = \log[\beta F_{t} + (1 - \beta)F_{0}] = f(F_{t})$$

$$f'(F_{t}) = \frac{\beta}{\beta F_{t} + (1 - \beta)F_{0}}, \quad f''(F_{t}) = -\frac{\beta^{2}}{(\beta F_{t} + (1 - \beta)F_{0})^{2}}$$

Applying Itô's formula,

$$dX_t = f'(F_t)dF_t + \frac{1}{2}f''(F_t)(dF_t)^2$$

$$= \beta \sigma W_t - \frac{\beta^2 \sigma^2}{2}dt$$

$$\int_0^T dX_t = \int_0^T \beta \sigma dW_t - \int_0^T \frac{\beta^2 \sigma^2}{2}dt$$

$$X_T - X_0 = \beta \sigma W_T - \frac{\beta^2 \sigma^2}{2}T$$

$$\log[\beta F_{t} + (1-\beta)F_{0}] - \log[\beta F_{0} + (1-\beta)F_{0}] = \beta \sigma W_{T} - \frac{\beta^{2}\sigma^{2}}{2}$$

$$\log\left[\frac{\beta F_{T} + (1-\beta)F_{0}}{F_{0}}\right] = \beta \sigma W_{T} - \frac{\beta^{2}\sigma^{2}}{2}$$

$$\frac{\beta F_{T} + (1-\beta)F_{0}}{F_{0}} = e^{-\frac{\beta^{2}\sigma^{2}T}{2} + \beta \sigma W_{T}}$$

$$\beta F_{T} + (1-\beta)F_{0} = F_{0}e^{-\frac{\beta^{2}\sigma^{2}T}{2} + \beta \sigma W_{T}}$$

$$F_{T} = \frac{F_{0}}{\beta}e^{-\frac{\beta^{2}\sigma^{2}T}{2} + \beta \sigma W_{T}} - \frac{1-\beta}{\beta}F_{0}$$

Forward price of Black76 :  $F_t = F_0 e^{-\frac{\sigma^2 T}{2} + \sigma W_t}$ 

Comparing this equation with Black76's  $F_t$ , we know that Displaced Diffusion( $F_0$ , k,  $\sigma$ ,  $\beta$ , T) is equal to Black76( $\frac{F_0}{B}$ ,

 $K + \frac{1-\beta}{\beta}F_0$ ,  $\sigma\beta$ , T). When we substitute Black76's parameters with those of Displaced-diffusion model, we derive Displaced-diffusion model.

Vanilla Call Option:

$$\begin{split} V_0^c &= e^{-rT} \left[ \frac{F_0}{\beta} \Phi \left( \frac{\log \frac{F_0}{\beta K + (1-\beta)F_0} + \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) - \\ \left( K + \frac{1-\beta}{\beta} F_0 \right) \Phi \left( \frac{\log \frac{F_0}{\beta K + (1-\beta)F_0} - \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) \right] \end{split}$$

Vanilla Put Option:

$$= e^{-rT} \left[ \left( K + \frac{1-\beta}{\beta} F_0 \right) \Phi \left( \frac{\log \frac{\beta K + (1-\beta)F_0}{F_0} + \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) - \frac{F_0}{\beta} \Phi \left( \frac{\log \frac{\beta K + (1-\beta)F_0}{F_0} - \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) \right]$$

Digital Cash-or-Nothing Call Option:

$$= e^{-rT} \Phi \left( \frac{\log \frac{F_0}{\beta K + (1-\beta)F_0} - \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right)$$

Digital Cash-or-Nothing Put Option :

$$e^{-rT}\Phi\left(rac{\lograc{eta K+(1-eta)F_0}{F_0}+rac{1}{2}\sigma^2eta^2T}{\sigmaeta\sqrt{T}}
ight)$$

Digital Asset-or-Nothing Call Option:

$$= e^{-rT} \left[ \frac{F_0}{\beta} \Phi \left( \frac{\log \frac{F_0}{\beta K + (1-\beta)F_0} + \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) - \frac{1-\beta}{\beta} F_0 \Phi \left( \frac{\log \frac{F_0}{\beta K + (1-\beta)F_0} - \frac{1}{2}\sigma^2 \beta^2 T}{\sigma \beta \sqrt{T}} \right) \right]$$

Digital Asset-or-Nothing Put Option:

$$=e^{-rT}\left[\frac{F_0}{\beta}\Phi\left(\frac{\log\frac{\beta K+(1-\beta)F_0}{F_0}-\frac{1}{2}\sigma^2\beta^2T}{\sigma\beta\sqrt{T}}\right)-\frac{1-\beta}{\beta}F_0\Phi\left(\frac{\log\frac{\beta K+(1-\beta)F_0}{F_0}+\frac{1}{2}\sigma^2\beta^2T}{\sigma\beta\sqrt{T}}\right)\right]$$

#### 2. Model Calibration

### 2.1 Displaced-Diffusion Model

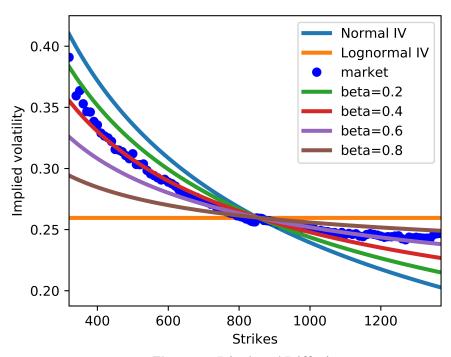


Figure 1. Displaced Diffusion

In the chart with strike price as the x axis and implied volatility as the y axis, when  $\beta$  increases from 0.2 to 0.8,  $\sigma$  decreases from about 0.42 to about 0.28. The displaced-diffusion model can be regarded as the weighted average of a lognormal and a normal model.  $\beta$  can be regarded as the weights of the lognormal and normal model in displaced-diffusion model. When  $\beta$  is equal to zero, the whole displaced-diffusion model becomes a normal model. When  $\beta$  is equal to 1, the whole displaced-diffusion model becomes a lognormal model. When we change  $\beta$  between 0 and 1, we are making "tradeoff" between a lognormal and a normal model. When the value of  $\beta$  is close to 1, the line drawn in the chart is close to the lognormal model line, which is a horizontal line. When the value of  $\beta$  is close to 0, the line drawn in the chart is close to normal model.

#### 2.2 SABR Model

 $\alpha$  is 0.98,  $\rho$  is -0.23 and v is 0.37. In SABR Model, the correlation parameter  $\rho$  is proportional to the skewness of stock returns. Positive correlation between stock and volatility will give us the return distribution with positive skewness. On the contrary, negative correlation between stock and volatility will give us the return distribution with negative skewness. In this case, when  $\rho$  is negative, the curve is downward sloping and steep for the reason that negative correlation increases the price of out-of-money put options and at the same time, decreases the price of out-of-money call options. When  $\rho$  is positive, the curve is more flattened.

v is called volatility of volatility in SABR Model. It is related to the kurtosis of stock return. When v increases, the kurtosis of the stock return also increases. When v is 0, the volatility is deterministic. When it comes to the curve, the increasing v will move the right tail and left tail of the curve upward for the reason that larger volatility of volatility increases the price of out-of-money call options as well as the out-of-money put options.

# 3. Static Replication

Payoff function:  $h(S_T) = S_T^3 + 2.5 \log S_T + 10.0$ 

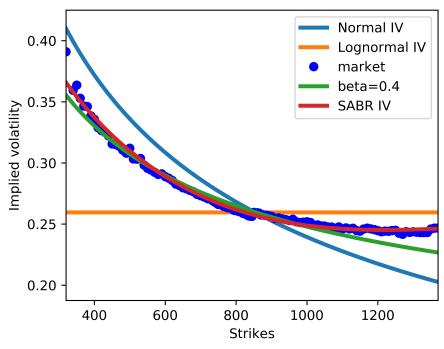


Figure 2. SABR

#### 3.1 Black Scholes Model - Payoff Function

Under Black Scholes stock price process,

$$S_T = S_0 e^{(r - \frac{\sigma^2}{2})T + \sigma W_T}$$
  $S_T^3 = S_0^3 e^{3(r - \frac{\sigma^2}{2})T}$   $\log S_T = \log S_0 + (r - \frac{\sigma^2}{2})T + \sigma W_T$ 

Using risk free rate as the numeraire and given  $B_0 = 0$   $B_T = e^{-rT}$ 

$$\frac{V_0}{B_0} = E^{Q^*} \left[ \frac{V_T}{B_T} \right]$$

$$V_0 = e^{-rT} E[V_T] = e^{-rT} \left[ S_0^3 e^{3T(r - \frac{\sigma^2}{2})} E\left[ e^{3\sigma W_T} \right] + 2.5 \left[ (r - \frac{\sigma^2}{2})T + \log S_0 \right] + 10 \right]$$

$$= e^{-rT} \left[ S_0^3 e^{3T(r + \sigma^2)} + 2.5 \left[ (r - \frac{\sigma^2}{2})T + \log S_0 \right] + 10 \right]$$

Inserting  $\sigma_{ATM} = 0.26$  and  $S_0, r, T$ , we are able to calculate  $V_0 = 813409453.499285$ 

#### 3.2 Bachelier Model - Payoff Function

Under Bachelier stock price process, after some derivaton, we are able to get the value of payoff as:

$$V_0 = S_0^3 + 3\sigma^2 S_0^3 T + 2.5 \left[ \log S_0 + \frac{1}{\sqrt{2\pi}} \int_{x^*}^{\infty} \log(1 + \sigma\sqrt{T}x) e^{-\frac{x^2}{2}} dx \right] + 10$$

where  $x^* = \frac{-1}{\sigma\sqrt{T}}$  and python calculates  $\mathbf{V_0} = 773441382.5428648$  by using  $\sigma_{ATM}$  as the input

#### 3.3 Static Replication - Payoff Function

We used single value of  $\sigma_{ATM}$  as the input for derivative valuation for the above two scenarios. We can also value the derivative by static-replicating option available in the market.

$$V_0 = e^{rT} E[h(S_T)]$$
  
=  $e^{-rT} h(F) + \int_0^F h''(K) P(K) dK + \int_F^\infty h''(K) C(K) dK$ 

where  $h'(S_T) = 3S_T^2 + \frac{2.5}{S_T}$ ,  $h''(S_T) = 6S_T - \frac{2.5}{S_T^2}$ ,  $F = S_0 e^{rT}$  By setting strike price K as a random variable whose value is used as input for implied vol calculated by calibrated SABR, we are able to value the derivative by integrating the static-replication formula above. The value we obtained is **805098821.1389331**.

"Model-free" integrated variance:

$$\sigma_{MF}^2 T = E \left[ \int_0^T \sigma_t^2 dt \right]$$

#### 3.4 Black Scholes Model - "Model-Free" Integrated Variance

Under Black Scholes, since it is model-free, we can just use  $\sigma_{ATM}$  as the input on the left hand side of the equation to calculate integrated variance. The value calculated is **0.093528767**. Alternatively, we can do static-replication using right hand side to verify our result.

$$\left[\int_0^T \sigma_t^2 dt\right] = 2e^{rT} \int_0^F \frac{P(K)}{K^2} dK + 2 * e^{rT} \int_F^\infty \frac{C(K)}{K^2} dK$$

The result of static replication is **0.093215189**, quite close to what we get by just using  $\sigma_{ATM}$ .

#### 3.5 Bachelier Model - "Model-Free" Integrated Variance

Under Bachelier, we use the same  $\sigma_{ATM}$  and arrive at the same integrated variance of **0.093528767**.

#### 3.6 Static Replication - "Model-Free" Integrated Variance

Similar to what have been done in 3.3, we can let implied vol computed from SABR be dependent of strike price and integrate the formula below:

$$\left[\int_0^T \sigma_t^2 dt\right] = 2e^{rT} \int_0^F \frac{P(K)}{K^2} dK + 2e^{rT} \int_F^\infty \frac{C(K)}{K^2} dK$$

The value of integrated variance is 0.103380242 by applying static replication.

## 4. Dynamics Hedging

Using the Black-Scholes formula, we can calculate the "fair" value of the option price. When the hedger uses dynamic hedging to rehedge continuously, the difference between the option value from Black-Scholes formula and the actual option value, which is the difference between the stock price and strike price at time t, should be zero. However, if the replication strategy deviates from the Black-Scholes method, the difference between the stock price and the strike price at time t may deviate from zero. The accumulated difference between the "fair value" from the Black-Sholes formula and the actual option value is the final profit and loss. According to the charts, if we rehedge a large number of times, the mean of the final profit and loss is 0. Moreover, the distribution of the final profit and loss is a normal distribution. Furthermore, when we hedge more frequently, the standard deviation decreases.

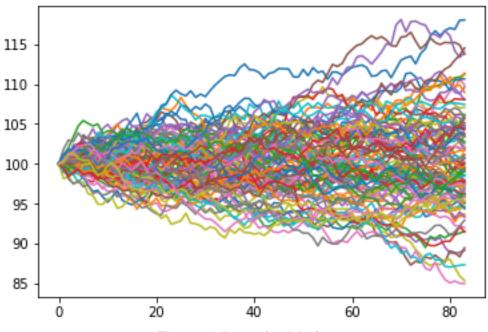


Figure 3. Brownian Motion

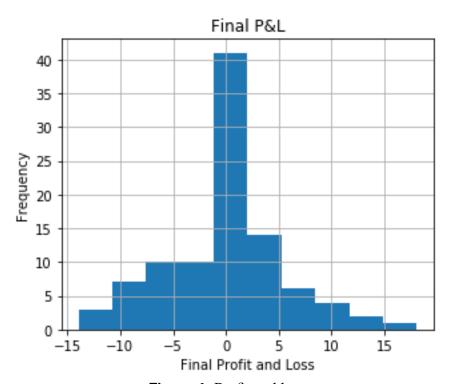


Figure 4. Profit and loss