

Bachelier Model (1900)

Louis Bachelier was the first mathematician to use random walk to analyse stock prices in 1900.

In Bachelier model, the stock price process is a **symmetrical random walk**, correspond to a market under **equilibrium**. This follows a **normal distribution**:

$$S_T = S_0 + \sigma W_T, \quad W_T \sim N(0, T).$$

Given this definition, we can then proceed to derive valuation formulas for vanilla European options.

However, a shortcoming of this model is that the lack of a lower bound at 0.

In other words, while this is a reasonable model for interest rates, it leads to non-zero probability for negative stock prices.

Bachelier Model – Arithmetic Brownian Process

The Bachelier model for the stock price process is defined as

$$dS_t = \sigma dW_t. \Rightarrow \int_0^T dS_t = \sigma \int_0^T dW_t$$

Integrating this stochastic equation, we can show that the terminal stock price is normally distributed as

$$S_T - S_0 = \sigma(W_T - W_0)$$

$$S_T \sim N(S_0, \sigma^2 T).$$

$$S_T = S_0 + \sigma W_T = S_0 + \sigma \sqrt{T} X,$$

Let V_c denote the price of a European call option, we have:

$$X \sim N(0, 1)$$

$$\begin{aligned} V_c &= e^{-rT} \mathbb{E}[(S_T - K)^+] = e^{-rT} \mathbb{E}[(S_0 + \sigma W_T - K)^+] \\ &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} (S_0 + \sigma \sqrt{T}x - K)^+ e^{-\frac{x^2}{2}} dx \end{aligned}$$

Note that $(S_0 + \sigma \sqrt{T}x - K)^+ = 0$ whenever $S_0 + \sigma \sqrt{T}x - K < 0$, and will only take on non-zero values when

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

$$\bar{\Phi}(\infty) - \bar{\Phi}(x^*) = 1 - \bar{\Phi}(x^*) = \bar{\Phi}(-x^*)$$

Bachelier Model – Arithmetic Brownian Process

Hence, we now write

$$\begin{aligned} V_c &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} \left(S_0 + \sigma\sqrt{T}x - K \right) e^{-\frac{x^2}{2}} dx \\ &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} (S_0 - K) e^{-\frac{x^2}{2}} dx + \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} \sigma\sqrt{T}x e^{-\frac{x^2}{2}} dx \\ &= e^{-rT} (S_0 - K) \left[\Phi(\infty) - \Phi(x^*) \right] - \frac{e^{-rT} \sigma\sqrt{T}}{\sqrt{2\pi}} \left[e^{-\frac{x^2}{2}} \right]_{x^*}^{\infty} \\ &= e^{-rT} \left[(S_0 - K) \Phi(-x^*) + \frac{\sigma\sqrt{T}}{\sqrt{2\pi}} e^{-\frac{(x^*)^2}{2}} \right] \\ &= e^{-rT} \left[(S_0 - K) \Phi(-x^*) + \frac{\sigma\sqrt{T}}{\sqrt{2\pi}} e^{-\frac{(x^*)^2}{2}} \right] \\ &= e^{-rT} \left[(S_0 - K) \Phi \left(\frac{S_0 - K}{\sigma\sqrt{T}} \right) + \sigma\sqrt{T} \phi \left(\frac{S_0 - K}{\sigma\sqrt{T}} \right) \right]. \end{aligned}$$

For **at-the-money (ATM)** options, we have $K = S_0$, and this formula reduces

$$\text{to } V_c = e^{-rT} \sigma \sqrt{\frac{T}{2\pi}}.$$



$$\int_{x^*}^{\infty} x e^{-\frac{x^2}{2}} dx$$

$$\text{let } u = \frac{x^2}{2}$$

$$\frac{du}{dx} = \frac{2x}{2} = x \quad \Rightarrow \quad du = x dx$$

$$\begin{aligned} \int_{x^*}^{\infty} x e^{-\frac{x^2}{2}} dx &= \int e^{-u} du = -[e^{-u}] = -\left[e^{-\frac{x^2}{2}}\right]_{x^*}^{\infty} \\ &= -\left[0 - e^{-\frac{x^{*2}}{2}}\right] \\ &= e^{-\frac{x^{*2}}{2}} \end{aligned}$$

$$C_{all} = e^{-rT} \left[(S_0 - K) \Phi\left(\frac{S_0 - K}{\sigma\sqrt{T}}\right) + \sigma\sqrt{T} \phi\left(\frac{S_0 - K}{\sigma\sqrt{T}}\right) \right]$$

$$ATM : S_0 = K$$

$$ATM \text{ Call} = e^{-rT} \sigma\sqrt{T} \phi(0) = e^{-rT} \sigma\sqrt{T} \cdot \frac{1}{\sqrt{2\pi}}$$

$$\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

$$\approx \sigma\sqrt{T} \cdot 0.4$$

Black-Scholes Model (1973)

In a landmark 1973 paper, Fischer Black and Myron Scholes introduced the **Black-Scholes model**, which models the stock price as

$$S_T = S_0 \exp \left[\left(r - \frac{\sigma^2}{2} \right) T + \sigma W_T \right], \quad W_T \sim N(0, T).$$

Given this definition, we can readily verify that $\mathbb{E}[S_T] = S_0 e^{rT}$. Rearranging, we can write it as

$$\frac{S_T}{S_0} = \exp \left[\left(r - \frac{\sigma^2}{2} \right) T + \sigma N(0, T) \right].$$

Consequently,

$$\log \frac{S_T}{S_0} = \left(r - \frac{\sigma^2}{2} \right) T + \sigma N(0, T) \sim N \left(\left(r - \frac{\sigma^2}{2} \right) T, \sigma^2 T \right).$$

Alternatively, we can also write S_T as

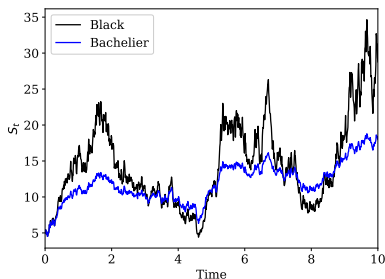
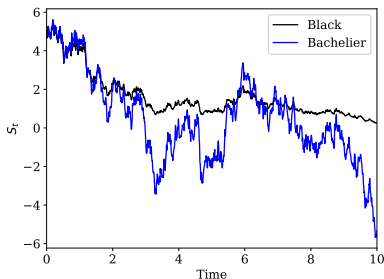
$$\log S_T = \log S_0 + \left(r - \frac{\sigma^2}{2} \right) T + \sigma \sqrt{T} N(0, 1).$$

Black-Scholes vs. Bachelier

Below are 2 sample paths (same Brownian motion) from the 2 models:

$$\text{Black-Scholes: } S_{t+\Delta t} = S_t \exp \left[\left(r - \frac{\sigma^2}{2} \right) \Delta t + \sigma \cdot (W_{t+\Delta t} - W_t) \right]$$

$$\text{Bachelier: } S_{t+\Delta t} = S_t + \sigma \cdot (W_{t+\Delta t} - W_t)$$



Question How do the two models compare?

Black-Scholes Model – Geometric Brownian Process

Under the Black-Scholes model, the stock price process follows the stochastic differential equation:

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

Note that a direct integration does not allow us to solve the SDE:

$$S_T = S_0 + \int_0^T rS_t dt + \int_0^T \sigma S_t dW_t.$$

However, we can solve the SDE by first applying **Itô's formula** to the function $X_t = f(S_t) = \log(S_t)$:

$$dX_t = \left(r - \frac{\sigma^2}{2}\right) dt + \sigma dW_t$$

Integrating both sides and substituting for X_t , we arrive at

$$S_T = S_0 \exp \left[\left(r - \frac{\sigma^2}{2}\right) T + \sigma W_T \right].$$

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

$$X_t = f(S_t) = \log(S_t)$$

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

$$f'(x) = \frac{1}{x}, \quad f''(x) = -\frac{1}{x^2}$$

Ito's formula:

$$dX_t = f'(S_t) dS_t + \frac{1}{2} f''(S_t) (dS_t)^2$$

$$= \frac{1}{\cancel{S_t}} \left(r \cancel{S_t} dt + \sigma \cancel{S_t} dW_t \right) - \frac{1}{2} \cdot \frac{1}{\cancel{S_t}^2} \cdot \sigma^2 \cancel{S_t}^2 dt$$

$$= r dt + \sigma dW_t - \frac{\sigma^2}{2} dt$$

$$\int_0^T dX_t = \int_0^T \left(r - \frac{\sigma^2}{2} \right) dt + \int_0^T \sigma dW_t$$

$$X_T - X_0 = (r - \frac{\sigma^2}{2})T + \sigma W_T$$

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

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

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

Black-Scholes Model – Option Pricing

Now let us derive the option pricing formula for a European call option under Black-Scholes model.

$$\begin{aligned} V_c &= e^{-rT} \mathbb{E}[(S_T - K)^+] \\ &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \left(S_0 e^{\left(r - \frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} - K \right)^+ e^{-\frac{x^2}{2}} dx \end{aligned}$$

Again, the terms in the $(\cdot)^+$ operator will need to be positive for it take non-zero values

$$S_T - K > 0 \quad \Rightarrow \quad x > \frac{\log\left(\frac{K}{S_0}\right) - \left(r - \frac{\sigma^2}{2}\right)T}{\sigma\sqrt{T}} = x^*.$$

Now we can proceed to evaluate the integral

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

Black-Scholes Model – Option Pricing

Next, we have:

$$\begin{aligned}
 V_c &= \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} S_0 e^{\left(-\frac{\sigma^2}{2}\right)T + \sigma\sqrt{T}x} e^{-\frac{x^2}{2}} dx - \frac{e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} K e^{-\frac{x^2}{2}} dx \\
 &= \frac{S_0 e^{-\frac{\sigma^2 T}{2}}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2 - 2\sigma\sqrt{T}x}{2}} dx - \frac{K e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\
 &= \frac{S_0 e^{-\frac{\sigma^2 T}{2}}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2 - 2\sigma\sqrt{T}x + \sigma^2 T - \sigma^2 T}{2}} dx - \frac{K e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\
 &= \frac{S_0 e^{-\frac{\sigma^2 T}{2}}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{(x - \sigma\sqrt{T})^2}{2}} e^{\frac{\sigma^2 T}{2}} dx - \frac{K e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx \\
 &= \frac{S_0}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{(x - \sigma\sqrt{T})^2}{2}} dx - \frac{K e^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx
 \end{aligned}$$

where we have used the **completing the square** trick.

$$\int_{x^*}^{\infty} e^{-\frac{(x - \sigma\sqrt{T})^2}{2}} dx = \int_{x^* - \sigma\sqrt{T}}^{\infty} e^{-\frac{y^2}{2}} dy$$

$$\begin{array}{l|l} \text{let } y = x - \sigma\sqrt{T} & x = x^* \Rightarrow y = x^* - \sigma\sqrt{T} \\ dy = dx & x \rightarrow \infty \Rightarrow y \rightarrow \infty \end{array}$$

Black-Scholes Model – Option Pricing

Finally, we obtain

$$\begin{aligned}V_c &= S_0 \left[\Phi(\infty) - \Phi(x^* - \sigma\sqrt{T}) \right] - Ke^{-rT} \Phi(-x^*) \\&= S_0 \Phi(-x^* + \sigma\sqrt{T}) - Ke^{-rT} \Phi(-x^*) \\&= S_0 \Phi \left(\frac{\log \frac{S_0}{K} + \left(r + \frac{\sigma^2}{2}\right) T}{\sigma\sqrt{T}} \right) - Ke^{-rT} \Phi \left(\frac{\log \frac{S_0}{K} + \left(r - \frac{\sigma^2}{2}\right) T}{\sigma\sqrt{T}} \right)\end{aligned}$$

In many references, it is common to let

$$d_1 = \frac{\log \frac{S_0}{K} + \left(r + \frac{\sigma^2}{2}\right) T}{\sigma\sqrt{T}}, \quad d_2 = d_1 - \sigma\sqrt{T},$$

leading to

$$V_c = S_0 \Phi(d_1) - Ke^{-rT} \Phi(d_2).$$

Black Model (1976) – Forward Price Process

So far we have regarded the underlying as the stochastic variable, and derived a stochastic differential equation to describe its price dynamic.

In 1976, Fischer Black proposed **modeling the forward price** instead of the underlying price. We have the definition of the forward price

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

$$F_t = e^{r(T-t)} S_t = f(t, S_t)$$

and the underlying price process of

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

Applying Itô's formula to the function $f(t, x) = e^{r(T-t)} x$ allows us to write down the stochastic differential equation for the forward price

$$dF_t = \sigma F_t dW_t,$$

which is a more compact equation—it is driftless and is therefore a **martingale**.

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

$$F_t = e^{r(T-t)} \cdot S_t = f(t, S_t)$$

$$f(t, x) = e^{r(T-t)} \cdot x$$

$$f_t = -r e^{r(T-t)} \cdot x$$

$$f_{xx} = e^{r(T-t)}, \quad f_{xx} = 0$$

Ito's formula:

$$dF_t = f_t(t, S_t) dt + f_x(t, S_t) dS_t + \frac{1}{2} f_{xx}(t, S_t) (dS_t)^2$$

$$= -r e^{r(T-t)} S_t dt + e^{r(T-t)} \cdot (r S_t dt + \sigma S_t dW_t) + 0$$

$$= \sigma S_t e^{r(T-t)} dW_t$$

$$= \sigma F_t dW_t$$

Black Model (1976) – Forward Price Process

The Black model is defined on the forward price and is given by

$$dS_t = rS_t dt + \sigma S_t dW_t, \quad dF_t = \sigma F_t dW_t.$$

As this is also a geometric process, we can solve this stochastic differential equation by applying Itô's formula to $X_t = f(F_t)$ where $f(x) = \log(x)$.

The solution is given by:

$$S_T = S_0 e^{(r - \frac{\sigma^2}{2})T + \sigma W_T} \quad F_T = F_0 e^{-\frac{\sigma^2 T}{2} + \sigma W_T}$$

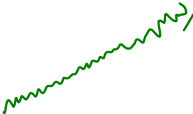
Let $D(0, T) = e^{-rT}$ denote the **discount factor**, under this model the price of a European call option is given by


$$V_c = D(0, T) \left[F_0 \Phi \left(\frac{\log \frac{F_0}{K} + \frac{1}{2} \sigma^2 T}{\sigma \sqrt{T}} \right) - K \Phi \left(\frac{\log \frac{F_0}{K} - \frac{1}{2} \sigma^2 T}{\sigma \sqrt{T}} \right) \right].$$

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

\uparrow mean-reversion speed \nwarrow long-run average

$$\theta = r_t \quad : \quad dr_t = k(0) dt + \sigma dW_t$$


$$\theta > r_t \quad : \quad dr_t = k(+ve) dt + \sigma dW_t$$


$$\theta < r_t \quad : \quad dr_t = k(-ve) dt + \sigma dW_t$$


Mean-reverting Process – Vasicek Model

The **Ornstein-Uhlenbeck process** is used in solid-state physics to model gas molecules under the influence of pressure and temperature.

Oldrich Vasicek adapted this model in 1977 to model interest rate as a **mean reverting stochastic process**, given by

$$dr_t = \kappa(\theta - r_t)dt + \sigma dW_t.$$

Applying Itô formula to $X_t = e^{\kappa t} r_t = f(t, r_t)$, we obtain

$$\begin{aligned} d(e^{\kappa t} r_t) &= \kappa e^{\kappa t} r_t dt + e^{\kappa t} dr_t \\ &= \kappa \theta e^{\kappa t} dt + \sigma e^{\kappa t} dW_t. \end{aligned}$$

Integrating both sides from 0 to t , we can obtain a solution to the stochastic differential equation

$$\begin{aligned} \int_0^t d(e^{\kappa u} r_u) &= \int_0^t \kappa \theta e^{\kappa u} du + \int_0^t \sigma e^{\kappa u} dW_u \\ r_t &= r_0 e^{-\kappa t} + \theta(1 - e^{-\kappa t}) + \sigma \int_0^t e^{\kappa(u-t)} dW_u. \end{aligned}$$

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

$$X_t = e^{kt} \cdot r_t = f(t, r_t)$$

$$f(t, x) = e^{kt} \cdot x$$

$$f_t = ke^{kt} \cdot x$$

$$f_{xx} = e^{kt}, \quad f_{rx} = 0$$

Ito's formula:

$$\begin{aligned} dX_t &= f_t(t, r_t) dt + f_x(t, r_t) dr_t + \frac{1}{2} f_{xx}(t, r_t) (dr_t)^2 \\ &= ke^{kt} \cdot r_t dt + e^{kt} [k(\theta - r_t) dt + \sigma dw_t] + 0 \end{aligned}$$

$$dX_t = \theta ke^{kt} dt + \sigma e^{kt} dw_t$$

$$\int_0^t dX_u = \theta k \int_0^t e^{ku} du + \sigma \int_0^t e^{ku} dW_u$$

$$X_t - X_0 = \theta [e^{kt}]_0^t + \sigma \int_0^t e^{ku} dW_u$$

$$e^{kt} r_t - e^{k \cdot 0} \cdot r_0 = 0(e^{kt} - 1) + \sigma \int_0^t e^{ku} dW_u$$

$$r_t = r_0 e^{-kt} + 0(1 - e^{-kt}) + \sigma \int_0^t e^{k(u-t)} dW_u$$

Mean-reverting Process – Vasicek Model

Taking expectation on both sides gives us the mean

$$\mathbb{E}[r_t] = r_0 e^{-\kappa t} + \theta(1 - e^{-\kappa t}).$$

Recall **Itô's Isometry theorem** states that

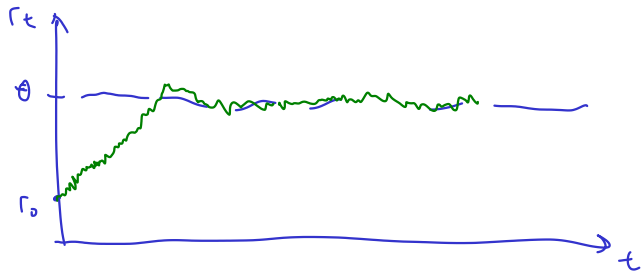
$$\mathbb{E} \left[\left(\int_0^T X_t dW_t \right)^2 \right] = \mathbb{E} \left[\int_0^T X_t^2 dt \right].$$

Applying it to our case,

$$\begin{aligned} V[r_t] &= \mathbb{E} \left[\left(\sigma \int_0^t e^{\kappa(u-t)} dW_u \right)^2 \right] \\ &= \mathbb{E} \left[\sigma^2 \int_0^t e^{2\kappa(u-t)} du \right] = \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa t}). \end{aligned}$$

The distribution of r_t is therefore given by

$$r_t \sim N \left(r_0 e^{-\kappa t} + \theta(1 - e^{-\kappa t}), \frac{\sigma^2}{2\kappa} (1 - e^{-2\kappa t}) \right).$$



$$\int_0^T dW_t = \lim_{n \rightarrow \infty} \sum_{i=1}^n (W_{t_i} - W_{t_{i-1}})$$

$$= \lim_{n \rightarrow \infty} \left[(\cancel{W_{t_1}} - W_{t_0}) + (\cancel{W_{t_2}} - \cancel{W_{t_1}}) \right. \\ \left. + \dots + (W_{t_n} - \cancel{W_{t_{n-1}}}) \right]$$

$$= \lim_{n \rightarrow \infty} [W_{t_n} - W_{t_0}]$$

$$= W_T - W_0$$



Session 6: Valuation Framework and Stochastic Volatility Models

Tee Chyng Wen

QF620 Stochastic Modelling in Finance

Pricing Models vs Reporting Models

So far we have been formulating our models as **pricing models**.

- ⇒ As the name suggests, pricing models are used to price and risk-manage derivatives.
- ⇒ The **dynamic** of the pricing models ought to conform to the modeler's intuition of the underlying asset's **evolution over time**.
- ⇒ The **Greeks** of the pricing models should accurately capture the **sensitivities** of the derivatives.

All financial institutions with a trading desk tend to have their own choices of pricing models, with the model parameters (e.g. σ , β , etc.) calibrated to the liquid option markets.

Pricing Models vs Reporting Models

Apart from pricing models, many (option) exchanges have also adopted the notion of **reporting models**.

- ⇒ A reporting model is used merely to report market option prices—these prices are driven by **supply and demand**.
- ⇒ Since it is often more elegant to report implied volatilities instead of prices (why?), a reporting model is required to perform this **conversion from price to volatility**.
- ⇒ Reporting models tend to make simplifying assumption about the asset dynamics, given that the primary objective is to arrive at an analytical tractable pricing formula for **price-volatility conversion**.
- ⇒ This reporting model's parameters (e.g. implied volatilities) can then be displayed on brokers' screens to communicate live option prices.

Example SPX index option chain, expiration on 15-Oct-2021.

Implied Volatility

Based on the observed option prices traded in the market, we can calculate the **implied volatilities**:

⇒ they are defined as the volatility parameter (σ) that we need to substitute into the Black-Scholes formula to match the option prices we observe.

In general, for each strike K , we will need to have an implied volatility parameter σ :

Strikes	Prices	Implied Volatilities
K_1	$C(K_1), P(K_1)$	$\text{BlackScholes}(S, K_1, r, \sigma_{K_1}, T)$
K_2	$C(K_2), P(K_2)$	$\text{BlackScholes}(S, K_2, r, \sigma_{K_2}, T)$
K_3	$C(K_3), P(K_3)$	$\text{BlackScholes}(S, K_3, r, \sigma_{K_3}, T)$
K_4	$C(K_4), P(K_4)$	$\text{BlackScholes}(S, K_4, r, \sigma_{K_4}, T)$
\vdots	\vdots	\vdots