

Black-Scholes (1973)

$$S_T = S_0 + \sigma W_T$$

$$\mathbb{E}[S_T] = S_0$$

equilibrium

supply = demand

Black-Scholes (1973)

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

$$\mathbb{E}[S_T] = S_0 e^{(r - \frac{\sigma^2}{2})T} \mathbb{E}[e^{\sigma W_T}]$$

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

$$= S_0 e^{rT}$$

no-arbitrage

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

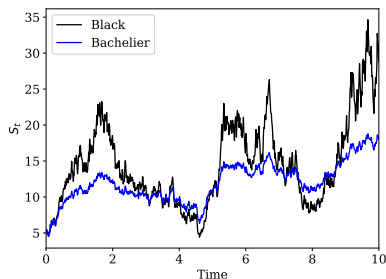
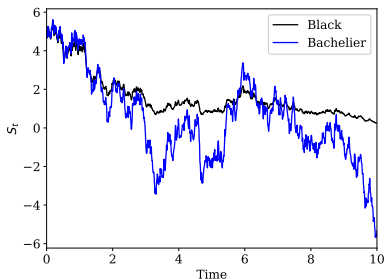
$$\log S_T = \log S_0 + \left(r - \frac{\sigma^2}{2} \right) T + \sigma \sqrt{T} N(0, 1).$$

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

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

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

$$\text{let } X_t = \log S_t = f(S_t)$$

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

$$f'(x) = \frac{1}{x}$$

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

Ito's formula:

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

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

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

$$X_T - X_0 = \left(r - \frac{\sigma^2}{2} \right) (T - \cancel{0}) + \sigma \cdot (W_T - \cancel{W_0})$$

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.

$$\text{let } \begin{array}{l} y = x - \sigma\sqrt{T} \\ dy = dx \end{array} \quad \left| \quad \begin{array}{l} \text{if } x = x^* , \quad y = x^* - \sigma\sqrt{T} \\ x \rightarrow \infty , \quad y \rightarrow \infty \end{array} \right.$$

$$= \frac{S_0}{\sqrt{2\pi}} \int_{x^* - \sigma\sqrt{T}}^{\infty} e^{-\frac{y^2}{2}} dy - \frac{Ke^{-rT}}{\sqrt{2\pi}} \int_{x^*}^{\infty} e^{-\frac{x^2}{2}} dx$$

$$= S_0 \left[\cancel{\Phi}^1(\infty) - \Phi(x^* - \sigma\sqrt{T}) \right] - Ke^{-rT} \left[\cancel{\Phi}^1(\infty) - \Phi(x^*) \right]$$

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

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

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

$$\mathbb{P}[\mathbb{1}_{S_T > K}] = \mathbb{P}(S_T > K)$$

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

$$\rightarrow F_t = e^{r(T-t)} S_t$$

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

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**.

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

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

$I_{t_0}^1$, formula:

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

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

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

$$= \sigma F_t dW_t$$

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

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

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

Black-Scholes

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

↓

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

Black

$$dF_t = \sigma F_t dW_t$$

↓

$$F_T = F_0 e^{-\frac{\sigma^2 T}{2} + \sigma W_T}$$

Black Model (1976) – Forward Price Process

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

$$dF_t = \sigma F_t dW_t.$$

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:

$$F_T = F_0 e^{-\frac{\sigma^2 T}{2} + \sigma W_T}$$

$$V_c = e^{-rT} \mathbb{E}[(F_T - K)^+]$$

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

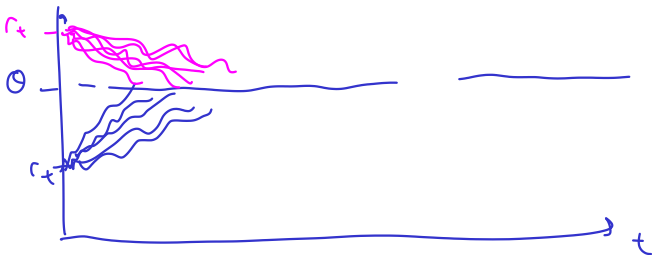
$$V_c = D(0, T) \left[F_0 \Phi \left(\frac{\log \frac{F_0}{K} + \frac{1}{2} \sigma^2 T}{\sigma \sqrt{T}} \right) - K \Phi \left(\frac{\log \frac{F_0}{K} - \frac{1}{2} \sigma^2 T}{\sigma \sqrt{T}} \right) \right].$$

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

↑
mean-reversion
speed

long-run
average of
interest rate

if $r_t < \theta$: $dr_t = \underbrace{+ve}_{\text{positive}} dt + \sigma dW_t$
 $r_t > \theta$: $dr_t = \underbrace{-ve}_{\text{negative}} dt + \sigma dW_t$



KMV model

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

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

$$dr_t = \kappa(0 - r_t) dt + \sigma dW_t$$

$$f(t, r_t) = e^{\kappa t} x$$

$$\text{let } X_t = e^{\kappa t} r_t = f(t, r_t)$$

$$f_t = \kappa e^{\kappa t} x$$

Ito's formula:

$$f_x = e^{\kappa t}, \quad f_{xx} = 0$$

$$dX_t = f_t(t, r_t) dt + f_x(t, r_t) dr_t + \frac{1}{2} f_{xx}(t, r_t) (dr_t)^2$$

$$= \kappa e^{\kappa t} r_t dt + e^{\kappa t} [\kappa(0 - r_t) dt + \sigma dW_t] + 0$$

$$= \cancel{\kappa e^{\kappa t} r_t dt} + \left[\cancel{\kappa e^{\kappa t} 0 dt} - \cancel{\kappa e^{\kappa t} r_t dt} + \sigma e^{\kappa t} dW_t \right]$$

$$dX_t = \kappa 0 e^{\kappa t} dt + \sigma e^{\kappa t} dW_t$$

$$\int_0^t dX_u = \int_0^t \kappa \theta e^{\kappa u} du + \int_0^t \sigma e^{\kappa u} dW_u$$

$$X_t - X_0 = \left[\theta e^{\kappa u} \right]_0^t + \sigma \int_0^t e^{\kappa u} dW_u$$

$$e^{\kappa t} \cdot r_t - e^0 \cdot r_0 = \mathcal{O}\left(e^{\kappa t} - e^0\right) + \sigma \int_0^t e^{\kappa u} dW_u$$

$$r_t = r_0 e^{-\kappa t} + \mathcal{O}\left(1 - e^{-\kappa t}\right) + \sigma \int_0^t e^{\kappa(u-t)} dW_u$$

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

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



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**.

$$C = S \mathbb{I}(d_1) - Ke^{-rT} \mathbb{I}(d_2)$$

- ⇒ 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 :

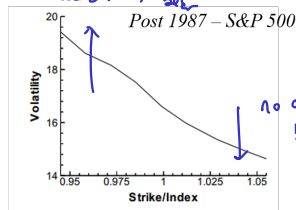
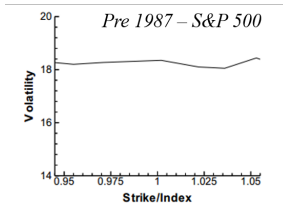
| Strikes | Prices | Implied Volatilities |
|----------|------------------|---|
| K_1 | $C(K_1), P(K_1)$ | $\text{BlackScholes}(S, K_1, r, \sigma_{K_1}, T)$ |
| K_2 | $C(K_2), P(K_2)$ | $\text{BlackScholes}(S, K_2, r, \sigma_{K_2}, T)$ |
| K_3 | $C(K_3), P(K_3)$ | $\text{BlackScholes}(S, K_3, r, \sigma_{K_3},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

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

Volatility Smile

Black-Scholes model assumes that the volatility of stock returns is **constant through time and strikes**. Is this true?

If the Black-Scholes assumptions are correct, then the implied volatilities of options should fall on a horizontal line when plotted against strikes.



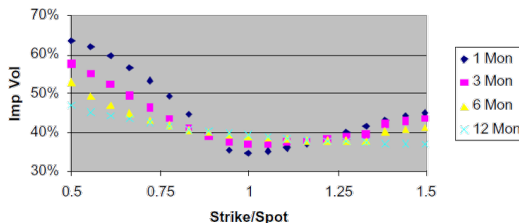
Prior to the 1987 Black Monday crash, this was roughly valid empirically. However, a distinct **volatility smile** manifested after the 1987 crash across a wide range of market—in anticipation of extreme market moves.

Volatility Smile

Background

- According to classical formulation, the Black-Scholes implied volatility of an option should be independent of its strike and expiration.
- Prior to the stock market crash of October 1987, the volatility smile of equity index options was indeed approximately flat.
- The Black-Scholes model assumes that a stock's return volatility is a constant, independent of strike and time to expiration.
- The volatility smile's appearance after the 1987 crash was due to the market's shock of discovering, for the first time since 1929, that a huge market could drop by 20% or more in a short period of time.
- In a *liquid* option market, option prices are determined by supply and demand, not by a valuation formula.

Volatility Smile



Volatility smile is generally steepest for short expiries, and is flatter for longer expiries.

Higher implied volatilities translate to higher option prices. The figure above shows that lower strike options are more in demand.

Market generally trades **out-of-the-money (OTM)** and **at-the-money (ATM) options**. **In-the-money (ITM)** options are relatively less liquid. This translates to more demand in the market for equity index put options.

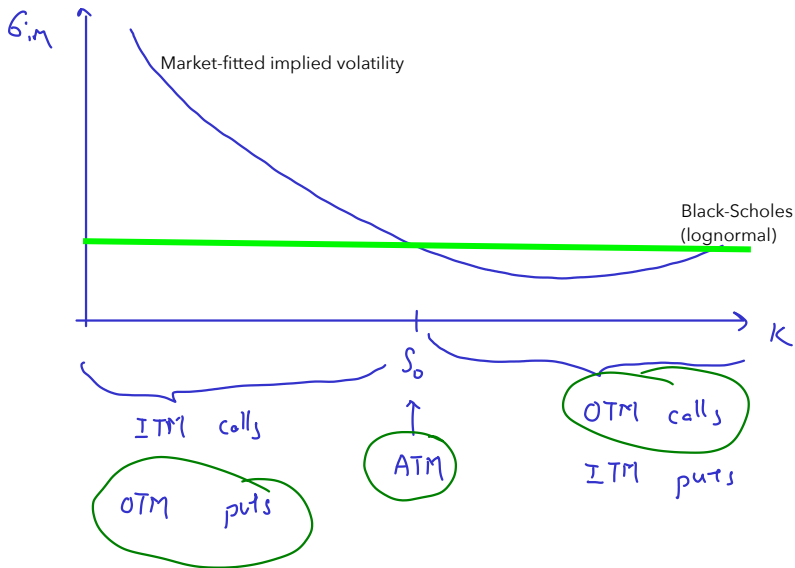
Fitting Market Prices

Suppose we are using the Black-Scholes or Bachelier model. The only model parameter we can vary is the volatility parameters (σ_{LN} or σ_N).

Since the at-the-money option (ATM) is the most important, we should choose the volatility parameter to fit the ATM option.

BS as rpt model

| Strikes | Imp-Vol | Black-Scholes | Bachelier |
|----------|----------------|---|---|
| K_1 | σ_{K_1} | $\text{BlackScholes}(S, K_1, \sigma_{LN}, T)$ | $\text{Bachelier}(S, K_1, \sigma_N, T)$ |
| K_2 | σ_{K_2} | $\text{BlackScholes}(S, K_2, \sigma_{LN}, T)$ | $\text{Bachelier}(S, K_2, \sigma_N, T)$ |
| K_3 | σ_{K_3} | $\text{BlackScholes}(S, K_3, \sigma_{LN}, T)$ | $\text{Bachelier}(S, K_3, \sigma_N, T)$ |
| K_4 | σ_{K_4} | $\text{BlackScholes}(S, K_4, \sigma_{LN}, T)$ | $\text{Bachelier}(S, K_4, \sigma_N, T)$ |
| \vdots | \vdots | \vdots | |



Fitting Market Implied Volatilities

Example Consider Google's call and put options on 2013-08-30. We look at options expiring on 2015-01-17, the spot stock price is 846.9, and the at-the-money volatility is ≈ 0.26 .

