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Block 1: Model Specification and Pricing PDE

We already know:

For a given market, described by the equations:

$$\begin{cases} dM(t) &= rM(t)dt \\ dS(t) &= \mu S(t)dt + \sigma S(t)dW^{\mathbb{P}}(t), \end{cases}$$

and a contingent claim of the form

$$\chi = V(T, S(T)),$$

the arbitrage free price is given, via Itô's Lemma, by $V(t, S)$, where function $V(t, S)$ satisfies the Black-Scholes equation:

$$\begin{cases} \frac{\partial V}{\partial t} + rS \frac{\partial V}{\partial S} + \frac{1}{2} S^2 \sigma^2 \frac{\partial^2 V}{\partial S^2} - rV &= 0 \\ V(T, S) &= \chi. \end{cases}$$

Block 2: Relation to Monte Carlo via Feynman-Kac Theorem

$V(t, S)$ is the unique solution of the final condition problem

$$\begin{cases} \frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0, \\ V(T, S) = \text{given} \end{cases}$$

This solution can also be obtained as

$$V(t, S) = e^{-r(T-t)} \mathbb{E}^{\mathbb{Q}}\{V(T, S(T)) | S(t)\}$$

with the sum of the first derivatives of the option square integrable.
Given the system of stochastic differential equations:

$$dS(t) = rS(t)dt + \sigma S(t)dW^{\mathbb{Q}}(t).$$

Block 3: Pricing via Integrals

- ▶ Pricing under risk-neutral measure:

$$V(t, S) = e^{-r(T-t)} \mathbb{E}^{\mathbb{Q}} \{ V(T, S(T)) | S(t) \}$$

- ▶ Quadrature:

$$V(t, S) = e^{-r(T-t)} \int_{\mathbb{R}} V(T, S(T)) f(S(T) | S(t)) dS(t)$$

- ▶ Transitional PDF, $f(S(T) | S(t))$, typically not available, but the characteristic function often is.

Analytic Solution of BS prices

Theorem (Black-Scholes formula)

The price of a European call option with strike price K and maturity T is given by the formula:

$$V_c(t, S(t)) = S(t) \cdot \mathcal{N}(d_1(t, S)) - e^{-r(T-t)} K \cdot \mathcal{N}(d_2(t, S(t))), \text{ with}$$

$$d_1(t, S(t)) = \frac{1}{\sigma\sqrt{T-t}} \left(\log \frac{S(t)}{K} + \left(r + \frac{1}{2}\sigma^2 \right) (T-t) \right),$$

$$d_2(t, S(t)) = d_1(t, S(t)) - \sigma\sqrt{T-t},$$

where \mathcal{N} is the cumulative distribution function for standard normal distribution i.e., $\mathcal{N}(0, 1)$.

Summary of Black-Scholes model

In the Black-Scholes formula we have:

- ▶ time to maturity: T (known)
- ▶ strike : K (known)
- ▶ risk free rate: r (known)
- ▶ current underlying price: S_0 (known)

What about σ ? Risk is a driving factor for options so under normal circumstances the option's theoretical value is a monotonically increasing function of the volatility. This means there is a one-to-one relationship between **the option price** and **the volatility**.

How we can test whether this holds in the reality?

Option Quotations in the Market

- Let us refresh how options are quoted.

| Calls | | | | | | | | | Puts | | | | | | | | | |
|---------------|-------|-------|--------|--------|----------|----------|------------|-----------|--------|-------|-------|--------|------|--------|----------|----------|------------|-----------|
| | Bid | Ask | Last | Volume | Open Int | Open Nch | ImpVol Mid | Delta Mid | Strike | | Bid | Ask | Last | Volume | Open Int | Open Nch | ImpVol Mid | Delta Mid |
| + 31-Mar-2021 | | | | | | | | | | | | | | | | | | |
| ☞ | 143.4 | 144 | 111.05 | | 41 | 0 | 20.4368 | 0.5176 | 3750 ☞ | 143.1 | 143.7 | 225.67 | | 43 | 0 | 20.4118 | -0.4795 | |
| ☞ | 140.2 | 141 | 115.5 | | 70 | 0 | 20.3359 | 0.5119 | 3755 ☞ | 144.9 | 145.6 | 213.73 | | 1 | 0 | 20.3041 | -0.4852 | |
| ☞ | 137.3 | 137.5 | 144.81 | 9 | 6.951 | -159 | 20.2454 | 0.5063 | 3800 ☞ | 146.9 | 147.6 | 142.6 | | 876 | 400 | 20.2282 | -0.4908 | |
| ☞ | 134 | 134.6 | 142.34 | | 17 | 1 | 20.1083 | 0.5005 | 3805 ☞ | 148.9 | 149.6 | 149.24 | 1 | 6 | 0 | 20.1199 | -0.4966 | |
| ☞ | 131.1 | 131.8 | 134.9 | 1 | 444 | 0 | 20.0313 | 0.4947 | 3810 ☞ | 150.9 | 151.6 | 143.46 | 4 | 22 | 1 | 20.0219 | -0.5023 | |
| + 30-Jun-2021 | | | | | | | | | | | | | | | | | | |
| ☞ | 210.5 | 211.6 | | | 0 | 0 | 21.0256 | 0.5176 | 3750 ☞ | 220.1 | 221.6 | 218.81 | 3 | 0 | 0 | 20.9539 | -0.4763 | |
| ☞ | 207.3 | 208.5 | | | 0 | 0 | 20.9436 | 0.5139 | 3755 ☞ | 222 | 223.6 | 236.44 | | 58 | 0 | 20.8820 | -0.4801 | |
| ☞ | 204.2 | 205.4 | 197.62 | 10 | 3.920 | 0 | 20.8647 | 0.5100 | 3800 ☞ | 223.8 | 225.4 | 215.1 | 2 | 1,573 | 350 | 20.8085 | -0.4839 | |
| ☞ | 201.2 | 202.3 | | | 0 | 0 | 20.7889 | 0.5062 | 3805 ☞ | 225.8 | 227.4 | 235.54 | | 58 | 0 | 20.7234 | -0.4878 | |
| ☞ | 198.1 | 199.3 | | | 0 | 0 | 20.7114 | 0.5023 | 3810 ☞ | 227.7 | 229.3 | 227.75 | | 10 | 0 | 20.6415 | -0.4916 | |
| + 30-Sep-2021 | | | | | | | | | | | | | | | | | | |
| ☞ | 281.8 | 286.1 | 278.29 | | 905 | 2 | 21.6095 | 0.5430 | 3750 ☞ | 262.3 | 266.2 | 261.41 | 1 | 299 | 1 | 21.6301 | -0.4478 | |
| ☞ | 266 | 270.1 | 262.92 | | 231 | 0 | 21.2738 | 0.5281 | 3775 ☞ | 271.2 | 275.3 | 252.62 | | 98 | 0 | 21.2921 | -0.4627 | |
| ☞ | 250.5 | 254.5 | 194.3 | | 951 | 0 | 20.9421 | 0.5129 | 3800 ☞ | 280.6 | 284.8 | 308.49 | | 27 | 0 | 20.9630 | -0.4780 | |
| ☞ | 235.4 | 239.3 | 203.73 | | 252 | 0 | 20.6141 | 0.4972 | 3825 ☞ | 290.3 | 294.6 | 308.64 | | 2 | 0 | 20.6299 | -0.4936 | |
| ☞ | 220.7 | 224.5 | 221.8 | | 599 | 0 | 20.2888 | 0.4811 | 3850 ☞ | 300.4 | 304.9 | 326.72 | | 3 | 0 | 20.3034 | -0.5097 | |
| + 31-Dec-2021 | | | | | | | | | | | | | | | | | | |
| ☞ | 319.1 | 326.9 | 326.21 | 1 | 448 | 0 | 21.6803 | 0.5400 | 3750 ☞ | 309.2 | 316.2 | 303.34 | | 16 | 0 | 21.5757 | -0.4476 | |
| ☞ | 303.5 | 311.1 | 308.11 | | 70 | 25 | 21.3895 | 0.5271 | 3775 ☞ | 318.3 | 325.5 | 315.1 | | 46 | 0 | 21.2776 | -0.4606 | |
| ☞ | 288.2 | 295.7 | 284.8 | | 114 | 0 | 21.0950 | 0.5139 | 3800 ☞ | 327.7 | 335.1 | 322 | | 8 | 6 | 20.9803 | -0.4738 | |
| ☞ | 273.2 | 280.5 | 276.8 | 100 | 104 | 0 | 20.8013 | 0.5005 | 3825 ☞ | 337.4 | 345 | 327.7 | | 11 | 10 | 20.6831 | -0.4874 | |
| ☞ | 258.6 | 265.8 | 268.45 | 3 | 41 | 0 | 20.5173 | 0.4868 | 3850 ☞ | 347.4 | 355.2 | 361.52 | | 2 | 0 | 20.3853 | -0.5012 | |

Figure: Call and Put options for S&P index, spot is about 3800.

Market Implied volatility

Example:

Suppose we have given a standard call option V_C on 100 shares of company Z. The strike is \$75 and it expires in 55 days. The risk free rate is 5%. The current stock price is \$85, and from historical data we have obtained $\sigma_{hist} = 0.25$. So, the call price is given by BS model is:

$$BS(\sigma_{hist}, r, T, K, S_0) = BS(25\%, 5\%, \frac{55}{365}, 75, 85) = 10.8667$$

But in the market the price of such a call option is \$12.25.

What does it mean? An arbitrage?

Market Implied volatility

Example Continuation

Based on the standard BS pricing model, we find the volatility **implied** by the market price V_C to be **43.89%** i.e.,

$$\sigma_{market} = g(V_C^{mkt}) = 43.89\%.$$

In order to check the calculation we substitute the σ_{market} to the pricing model, i.e.,

$$\begin{aligned} V_C(t, S) &= BS(\sigma_{market}, r, T, K, S_0) \\ &= BS(43.89\%, 5\%, \frac{55}{365}, \$75, \$85) = \$12.25 \end{aligned}$$

How to find implied volatility, $\sigma_{impl} = \sigma_{market}$?

Market Implied volatility

Implied Volatility: "the wrong number in the wrong formula to get the right price". [Rebonato 1999]

Mathematically we have:

$$V_c(t, S) = BS(\sigma, r, T, K, S_0)$$

where BS is monotonically increasing in σ (higher volatility corresponds to higher prices). Now, assume the existence of some inverse function

$$g_\sigma(\cdot) = BS^{-1}(\cdot)$$

so that

$$\sigma_{impl} = g_\sigma(V_c^{mkt}, r, T, K, S_0)$$

By computing the implied volatility for traded options with different strikes and maturities, we can test the Black-Scholes model.

Solving the inverse pricing model function

How to find implied volatility?

The BS pricing function BS does not have a closed-form solution for its inverse $g_{\sigma}(\cdot)$. Instead, a root finding technique is used to solve the equation:

$$BS(\sigma_{impl}, r, T, K, S_0) - V_C^{mkt} = 0.$$

There are many ways to solve this equation, one of the most popular method are methods of "Newton" and "Brent"¹. Since the options prices can move very quickly, it is often important to use the most efficient method when calculating implied volatilities.

¹http://en.wikipedia.org/wiki/Brent's_method

Method of Newton

- Suppose an approximation x_n . Assume that g is differentiable, and write $\epsilon_n = x_{\text{ex}} - x_n$, a Taylor series expansion gives:

$$0 = g(x_{\text{ex}}) = g(x_n) + \epsilon_n g'(x_n) + \frac{\epsilon_n^2}{2} g''(x_n) + \dots$$

- Ignoring the terms of second order and higher,

$$x_{\text{ex}} \approx x_n - \frac{g(x_n)}{g'(x_n)}.$$

- Use this expression as our new approximation:

$$x_{n+1} = x_n - \frac{g(x_n)}{g'(x_n)}$$

- How fast does the error reduce with this approximation? With one additional term in the Taylor series expansion, we get:

$$\epsilon_{n+1} = x_{\text{ex}} - x_{n+1} = \epsilon_n + \frac{g(x_n) - g(x_{\text{ex}})}{g'(x_n)} \approx -\epsilon_n^2 \frac{g''(x_n)}{2g'(x_n)}$$

Method of Newton

One starts with an initial guess which is reasonably close to the true root. The function is then approximated by its tangent line, and one computes the x-intercept of this tangent line.

Suppose $g : [a, b] \rightarrow \mathbb{R}$ is a differentiable function. From basic calculus we have:

$$g'(x_n) = \frac{g(x_n) - 0}{x_n - x_{n+1}} = \frac{0 - g(x_n)}{x_{n+1} - x_n}, \quad n = 0, 1, \dots \quad (1)$$

which gives us the iteration:

$$x_{n+1} = x_n - \frac{g(x_n)}{g'(x_n)}, \quad n = 0, 1, \dots \quad (2)$$

with some arbitrary initial value x_0 . In the case of BS we have:

$$\sigma_{n+1} = \sigma_n - \frac{BS(\sigma_n, \cdot) - V_C^{mkt}}{\frac{\partial BS(\sigma_n, \cdot)}{\partial \sigma_n}}, \quad n = 0, 1, \dots$$

Method of Newton

In Equation (1) we ignore the $\mathcal{O}(h^2)$ term, so we expect the error $x_n - x_{\text{ex}}$ squares as n increases to $n + 1$; that is: if

$$x_n - x_{\text{ex}} = \mathcal{O}(h)$$

then

$$x_{n+1} - x_{\text{ex}} = \mathcal{O}(h^2)$$

The analysis can be formalized to give the following result:

Theorem

Suppose g has a continuous second derivative, and suppose $x_{\text{ex}} \in \mathbb{R}$ satisfies $g(x_{\text{ex}}) = 0$ and $g'(x_{\text{ex}}) \neq 0$. Then there exists a $\delta > 0$ such that for $|x_0 - x_{\text{ex}}| < \delta$ the sequence given in (2) is defined for all $n > 0$, $\lim_{n \rightarrow +\infty} |x_n - x_{\text{ex}}| = 0$, and there exists a constant C such that:

$$|x_{n+1} - x_{\text{ex}}| \leq C|x_n - x_{\text{ex}}|^2.$$

Method of Newton

- ▶ Implementation in Python.

Implied volatility and Black-Scholes

Why is the implied volatility important?

Implied volatility: Model

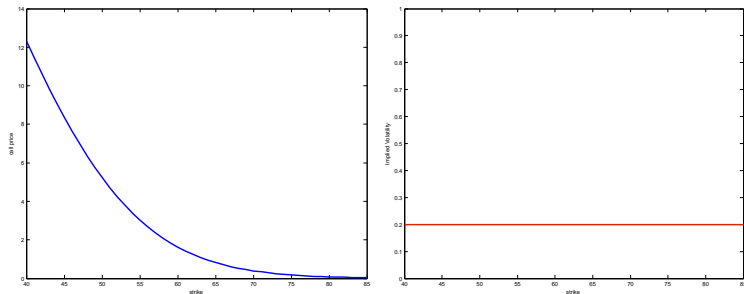


Figure: MODEL–LEFT: BS Call Prices, RIGHT: Implied Volatilities.

Implied volatility and Black-Scholes

Why is the implied volatility important?

Implied volatility: Market

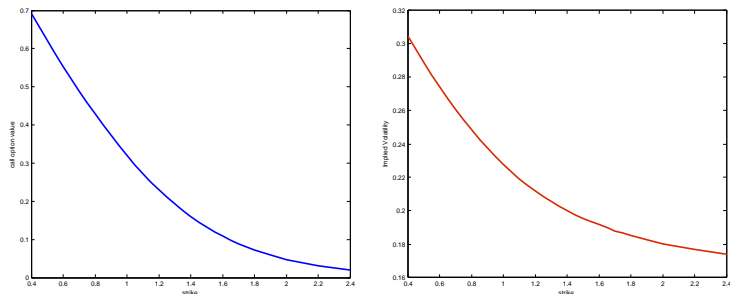


Figure: MARKET DATA– LEFT: Market Call Prices, RIGHT: Implied Volatilities.

Implied Volatility Shapes

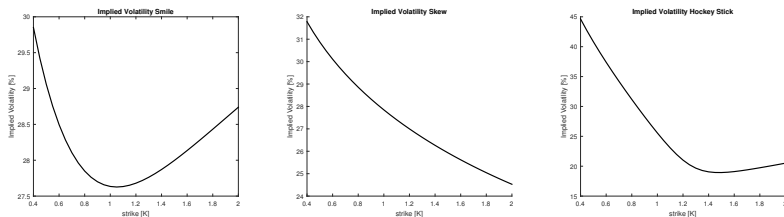


Figure: Typical implied volatility shapes: a smile, a skew and the so-called hockey stick. The hockey stick can be seen as a combination of the implied volatility smile and the skew.

Implied Volatility Term Structure

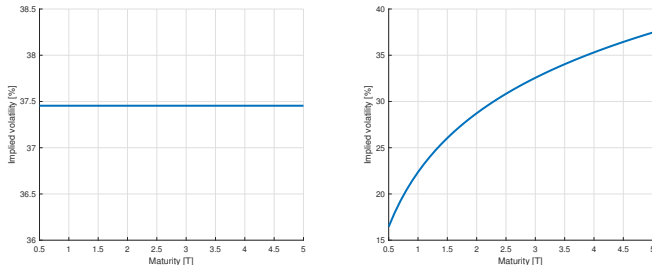


Figure: Comparison of the volatility term structure (for ATM volatilities) for Black-Scholes model with constant volatility σ_* (left-side picture) versus a model with time-dependent volatility $\sigma(t)$ (right-side picture).

Implied Volatility Surface

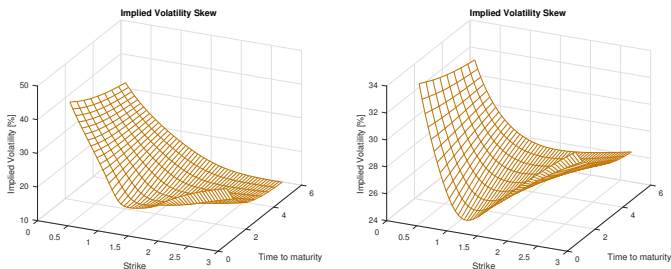


Figure: Implied volatility surfaces. A pronounced smile for the short maturity and a pronounced skew for longer maturities T (left side figure), and a pronounced smile over all maturities (right side).

Deficiencies of the Black-Scholes Model

- ▶ The Black-Scholes model, and its notion of hedging option contracts by stocks and money, forms the foundation of modern finance.

However:

- ▶ Delta hedging is supposed to be a continuous process, but in practice it is a discrete process (a hedged portfolio is typically updated once a week or so).
- ▶ Empirical studies of financial time series have revealed that the normality assumption for asset prices cannot capture *heavy tails* and *asymmetries*, present in log-asset returns in practice.
- ▶ The volatility is supposed to be a known deterministic function of time, which is inconsistent since numerical inversion of the Black-Scholes equation based on market prices from different strikes and fixed maturity, produces a so-called *volatility skew or smile*.

Deficiencies of the Black-Scholes Model

- ⇒ The idea of implied volatility does not fit to the Black-Scholes model
 - ▶ Look for market consistent asset price models.
- ⇒ Use a **local volatility**, a **model with jumps**, or **stochastic volatility** to better fit market data, and incorporate smile effects