

## CQF Lecture 3.3 Understanding Volatility

### Solutions

1. Explain what actual and implied volatilities are, and what is their relationship? Name three assumptions made in estimation of actual volatility from the market option prices.

**Solution:** *Actual volatility* is the measure of the amount of randomness in *asset returns* at any particular time. *Implied volatility* is the key input into the Black-Scholes option pricing formulae that gives the market price of an option.

Implied volatility is often described as the market's view of the future actual volatility over the lifetime of an option. However, one must be aware that it can be influenced by other effects, such as expectation of a crash, supply and demand.

Each implied volatility number has timescale associated with it. Actual volatility exists in a very instant, it is an instantaneous process. **Actual, not implied, volatility is supposed to be an input to all option pricing formulae.**

If the actual volatility were known it would straightforward to figure out the implied, but the inverse does not hold. Bootstrapping the actual volatility from the market option prices (expressed in the Black-Scholes implied volatility) requires making big assumptions:

- (a) Option prices now have full information about future volatility.
- (b) This inverse problem is uniquely solvable (in fact, multiple solutions are possible).
- (c) Actual volatility (a continuous and instantaneous process) is fitted with piecewise constant values.

2. The market price for a European put with strike  $E = 100$ , expiration one year, interest rate is 5% p.a. is quoted at \$5.57 for stock value at \$100. How do you find its implied volatility?

**Solution:** The Black-Scholes formula for the European put option on a non-dividend paying stock with  $t = 0$  is

$$P = K e^{-rT} N(-d_2) - S_0 N(-d_1)$$

$$d_1 = \frac{\ln(\frac{S_0}{E}) + (r + \frac{1}{2}\sigma_{BS}^2)T}{\sigma_{BS}\sqrt{T}}$$

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

**It is not possible to invert option price formulae in order to express the implied volatility  $\sigma_{BS}$  as a function of option price  $P$ .** Think about 'extracting'  $\sigma_{BS}$  from within *CDF* of the Normal distribution, which is an integral.

Therefore, the problem has to be solved numerically using **a root-finding method**. The methods range from a 'simple' bisection to Newton-Raphson method with variations that are specific to Black-Scholes formulae (calculation speed is improved by the choice of initial value for  $\sigma_{BS}$ ).

Root-finding methods would implement an iterative search for such value of  $\sigma$  that gives  $f(\sigma) = P_{BS}(\sigma) - P_{Market} = 0$

$$f(\sigma_{i+1}) = f(\sigma_i) - f'(\sigma_i)(\sigma_{i+1} - \sigma_i)$$

One can start with  $\sigma_0$  and produce successively better estimates  $\sigma_1, \sigma_2, \dots$  by moving 'up' or 'down' the function. The use of root-finding methods relies on option price monotonically increasing *wrt* volatility parameter. This guarantees a unique solution.

$$\sigma_{i+1} = \sigma_i - \frac{f(\sigma_i)}{f'(\sigma_i)} \quad \Rightarrow \quad \sigma_{i+1} = \sigma_i - \frac{V(\sigma_i)}{Vega(\sigma_i)}$$

The iteration stops as soon as  $f(x_n)$  reaches pre-determined tolerance level. This numerical method of finding implied volatility can be easily implemented in VBA. Further explanation of root-finding methods is available in CQF Extra on Further Mathematical Methods.

**The numerical answer for the European put is  $\sigma \approx 20\%$ .**

3. Assume a **time-dependent** volatility function  $\sigma(t)$ . Consistent with Black-Scholes framework, the implied volatility  $\sigma_i(t, T)$  measured at time  $t$  of an European option expiring at time  $T$  must satisfy

$$\sigma_i(t, T) = \sqrt{\frac{1}{T-t} \int_t^T \sigma^2(s) ds}$$

Solve the inverse problem (an integral equation) to show that, at calibration time  $t^*$ , the volatility function  $\sigma(t)$  must be consistent with implied volatility  $\sigma_i$

$$\sigma^2(t) = 2(t - t^*) \sigma_i(t^*, t) \frac{\partial \sigma_i(t^*, t)}{\partial t} + \sigma_i^2(t^*, t)$$

**Solution:** We know that the implied volatility is the square root of variance, so

$$\sigma_i^2(t, T) = \frac{1}{T-t} \int_t^T \sigma^2(s) ds \quad (1)$$

We solve by differentiating both sides *wrt*  $T$  (use partial derivatives and chain rule). To make solution easier we leave the integral on the *rhs*:

$$\frac{d}{dT} ((T-t) \sigma_i^2(t, T)) = \frac{d}{dT} \int_t^T \sigma^2(s) ds \quad (2)$$

$$\sigma_i^2(t, T) + (T-t) 2\sigma_i(t, T) \frac{\partial \sigma_i(t, T)}{\partial T} = \sigma^2(T) \quad (3)$$

$$\underline{\sigma^2(T)} = \sigma_i^2(t, T) + 2(T-t) \sigma_i(t, T) \frac{\partial \sigma_i(t, T)}{\partial T}$$

But  $\sigma^2$  is ‘not allowed’ to be a function of expiration  $T$ . At this stage, **variables became parameters**. Let  $t^*$  be the calibration time, then  $t \rightarrow t^*$  and  $T \rightarrow t$

$$\sigma^2(t) = \sigma_i^2(t^*, t) + 2(t - t^*) \sigma_i(t^*, t) \frac{\partial \sigma_i(t^*, t)}{\partial t}$$

In the inverse notation  $\sigma_i(T; t^*)$ , to mean the implied volatility measured at time  $t^*$  of a European option expiring at time  $T \rightarrow t$ , the solution is expressed as

$$\sigma(t) = \sqrt{\sigma_i^2(t; t^*) + 2(t - t^*) \sigma_i(t; t^*) \frac{\partial \sigma_i(t; t^*)}{\partial t}}$$

To formulate the inverse-time calibration solution, we had to change notation twice!

4. Suppose implied volatilities are observable at  $T_i, i = 0, 1, 2, \dots, n$ , with  $T_0 = t^*$  is the date of calibration (fitting). Assuming that the actual volatility function is **piecewise constant**, show that for  $T_{i-1} < t < T_i$  the total variance is

$$\sigma^2(t) = \frac{(T_i - t^*) \sigma_i^2(t^*, T_i) - (T_{i-1} - t^*) \sigma_i^2(t^*, T_{i-1})}{T_i - T_{i-1}}$$

**Solution:** The task is to solve a discretised integral equation.

$$\sigma_i^2(t^*, T_i) = \frac{1}{T_i - t^*} \int_{T_0=t^*}^{T_i} \sigma^2(s) ds \quad (\text{splitting the integral}) \quad (4)$$

$$(T_i - t^*) \sigma_i^2(t^*, T_i) = \int_{T_0=t^*}^{T_{i-1}} \sigma^2(s) ds + \int_{T_{i-1}}^{T_i} \sigma^2(s) ds \quad (5)$$

The assumption of piecewise constant actual volatility  $\forall t \in (T_{i-1}, T_i)$  means that the second integral on the *rhs* is over a constant

$$\int_{T_{i-1}}^{T_i} \sigma^2(s) ds = \sigma^2(t) \int_{T_{i-1}}^{T_i} ds = \underline{\sigma^2(t)(T_i - T_{i-1})} \quad (6)$$

We saw the first integral on *rhs* as implied volatility. Plugging in result (6) gives

$$(T_i - t^*) \sigma_i^2(t^*, T_i) = (T_{i-1} - t^*) \sigma_i^2(t^*, T_{i-1}) + \underline{\sigma^2(t)(T_i - T_{i-1})} \quad (7)$$

rearrange to obtain the expression for  $\sigma^2(t)$

$$\sigma^2(t) = \frac{(T_i - t^*) \sigma_i^2(t^*, T_i) - (T_{i-1} - t^*) \sigma_i^2(t^*, T_{i-1})}{T_i - T_{i-1}} \quad (8)$$

for  $T_{i-1} < t < T_i$ , where  $t^* = T_0$  is earlier time of calibration  $t^* < T_{i-1} < T_i$ .

The implication from the interim result (7) is that we can decompose the implied volatility into a time-weighted average of actual (local) volatilities:

$$\begin{aligned} (T_i - t^*) \sigma_i^2(t^*, T_i) &= (T_{i-2} - t^*) \sigma_i^2(t^*, T_{i-2}) + \dots + \sigma^2(t)(T_{i-1} - T_{i-2}) + \sigma^2(t)(T_i - T_{i-1}) \\ &\approx (T_{i-2} - T_{i-3}) \sigma_{1M}^2(t) + (T_{i-1} - T_{i-2}) \sigma_{2M}^2(t) + (T_i - T_{i-1}) \sigma_{3M}^2(t) \\ &\approx \quad (\text{as seen in the Lecture example with three-month term structure}). \end{aligned}$$

5. Denote the actual volatility by  $\sigma_a$  and implied volatility by  $\sigma_i$ , where subscript ‘a’ means actual and ‘i’ means implied. Similarly,  $\Delta_a$  means  $\Delta$  is calculated using actual volatility, and  $\Delta_i$  means  $\Delta$  is calculated by using implied volatility. The Black-Scholes Delta is calculated as  $\Delta = e^{-DT} N(d_1)$  where  $t = 0$  and volatility parameter is used to calculate  $d_1$ .

Assume that an asset follows the GBM with continuous dividend rate  $D$ , and an option written on this asset is denoted by  $V(S, t; \sigma)$ .

- Within the Black-Scholes framework, what is Mark-to-Market profit if one hedges the option by using actual volatility to calculate Delta. How is this profit going to be realised (guaranteed or not)?
- What about the Mark-to-Market profit if hedging with the implied volatility?

**Solution:** At time  $t$ , we construct the following hedging portfolio

- (a) Buy one unit of option:  $V_i$
- (b) Short  $\Delta_a$  unit of stock:  $-\Delta_a S$
- (c) Leftover cash borrowed/received:  $-(V_i - \Delta_a S)$

At time  $t + dt$ , the portfolio value becomes

- (a) Option:  $V_i + dV_i$
- (b) Stock:  $-\Delta_a S - \Delta_a dS$
- (c) Cash:  $-(V_i - \Delta_a S)(1 + rdt) - \Delta_a DSdt$

The Mark-to-Market profit over time  $dt$  will be the difference of the portfolios

$$dV_i - \Delta_a dS - (V_i - \Delta_a S)rdt - \Delta_a DSdt \quad \dagger$$

Because the option would be correctly valued at  $V_a$  its MtM profit will be offset by delta hedging exactly, so we have the equality

$$dV_a - \Delta_a dS - (V_a - \Delta_a S)rdt - \Delta_a DSdt = 0$$

Subtracting term equal to zero from  $\dagger$  gets cancellation of all terms multiplied by  $\Delta_a$ . The **P&L over timestep**  $dt$  is

$$dV_i - dV_a - r(V_i - V_a)dt \quad \Rightarrow \quad \boxed{e^{rt} d(e^{-rt}(V_i - V_a))}$$

$$\begin{aligned} d(e^{-rt}V) &= e^{-rt}dV - re^{-rt}Vdt \\ &= e^{-rt}(dV - rVdt) \end{aligned}$$

We have used a Factor  $e^{-rt}$ .

To calculate the **total profit**, sum up all the small-step profits and apply a discount factor  $e^{-r(t-t_0)}$ .

$$\begin{aligned}
& \int_{t_0}^T e^{-r(t-t_0)} e^{rt} d(e^{-rt}(V_i - V_a)) \\
&= e^{rt_0} e^{-rT} (V_i(T) - V_a(T)) - e^{rt_0} e^{-rt_0} (V_i(0) - V_a(0)) \\
&= V_a(0) - V_i(0) \quad \text{because} \quad V_i(T) - V_a(T) \equiv 0
\end{aligned}$$

**Conclusion:** if we hedge with actual volatility, then the total profit is a guaranteed amount, equal to the difference between Black Scholes option values  $V_a - V_i$ .

To see how this guaranteed profit is realised on the MtM basis over  $dt$ , we now **invoke Itô lemma directly** to expand  $dV_i$ , the first term of  $\dagger$  and write using familiar Greeks from  $dV = (\frac{\partial V}{\partial t} + \frac{1}{2}\sigma_a^2 S^2 \frac{\partial^2 V}{\partial S^2})dt + \frac{\partial V}{\partial S}dS$ , where  $\frac{1}{2}\frac{\partial^2 V}{\partial S^2}dS^2 \rightarrow \frac{1}{2}\frac{\partial^2 V}{\partial S^2}\sigma_a^2 S^2 dt$ .

$$\begin{aligned}
& \Theta_i dt + \frac{1}{2}\sigma_a^2 S^2 \Gamma_i dt + \Delta_i dS - \Delta_a dS - (V_i - \Delta_a S)r dt - \Delta_a D S dt \\
& \text{insert our model } dS = \mu S dt + \sigma_a S dX \\
& \underbrace{(\Theta_i - r V_i)} dt + \frac{1}{2}\sigma_a^2 S^2 \Gamma_i dt + (\Delta_i - \Delta_a)S(\mu dt + \sigma_a dX) + (r - D)\Delta_a S dt
\end{aligned}$$

$$V_i \text{ satisfies BSE with } \sigma = \sigma_i \text{ for } \underbrace{\Theta_i - r V_i} = - \left( \Delta_i(r - D)S + \frac{1}{2}\sigma_i^2 S^2 \Gamma_i \right)$$

$$\begin{aligned}
&= -\Delta_i(r - D)S dt - \frac{1}{2}\sigma_i^2 S^2 \Gamma_i dt + \frac{1}{2}\sigma_a^2 S^2 \Gamma_i dt + (\Delta_i - \Delta_a)S(\mu dt + \sigma_a dX) \\
&\quad + (r - D)\Delta_a S dt \\
&= \frac{1}{2}(\sigma_a^2 - \sigma_i^2)S^2 \Gamma_i dt + (\Delta_i - \Delta_a)S(\mu dt + \sigma_a dX) + (\Delta_a - \Delta_i)(r - D)S dt \\
&= \frac{1}{2}(\sigma_a^2 - \sigma_i^2)S^2 \Gamma_i dt + (\Delta_i - \Delta_a)[(\mu - r + D)S dt + \sigma_a S dX]
\end{aligned}$$

With the asset following a risk-neutral drift  $\mu = r$  and no dividends  $D = 0$ ,

$$\frac{1}{2}(\sigma_a^2 - \sigma_i^2)S^2 \Gamma_i dt + \underline{(\Delta_i - \Delta_a)\sigma_a S dX}$$

how the profit is achieved is random due to the diffusion term. See Figure 1 below.

If hedging with the implied volatility, the Mark-to-Market profit over time  $dt$  will look similar to † with the obvious change  $\Delta_a \rightarrow \Delta_i$

$$\begin{aligned}
& \frac{dV_i - \Delta_i dS - r(V_i - \Delta_i S) dt}{\text{using BSE first time with } \sigma = \sigma_a} \\
&= \Theta_i dt + \frac{1}{2} \sigma_a^2 S^2 \Gamma_i dt - rV_i dt + r\Delta_i S dt \\
& \quad \text{using BSE second time with } \sigma = \sigma_i \text{ for } \Theta_i dt = rV_i dt - r\Delta_i S dt - \frac{1}{2} \sigma_i^2 S^2 \Gamma_i dt \\
&= \frac{1}{2} (\sigma_a^2 - \sigma_i^2) S^2 \Gamma_i dt
\end{aligned}$$

We understand this result as the gain from the curvature  $\frac{1}{2} \sigma_a^2 S^2 \Gamma_i dt$  being cancelled by the loss from time decay  $\frac{1}{2} \sigma_i^2 S^2 \Gamma_i dt$  because the simplified Black-Scholes is  $\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} = 0 \Rightarrow \Theta_i = -\frac{1}{2} \sigma_i^2 S^2 \Gamma_i$  (remember the heat equation).

Add up the present value of all of the small-step profits to get the total profit of

$$\frac{1}{2} (\sigma_a^2 - \sigma_i^2) \int_0^T e^{-rt} S^2 \Gamma_i dt$$

This is always positive but path-dependent. Please see Figure 2.

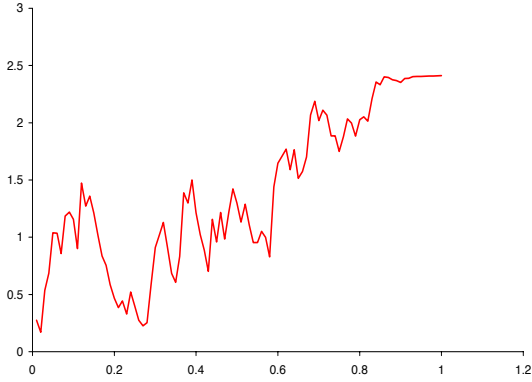


Figure 1: Hedging with  $\sigma_a$

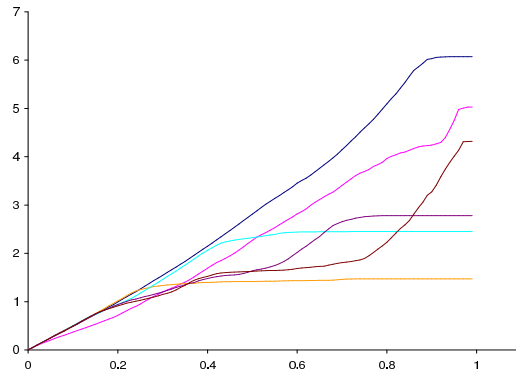


Figure 2: Hedging with  $\sigma_i$

## Advanced Notes

- Option pricing often assumes no discrepancy  $\sigma_a = \sigma_i \rightarrow \frac{1}{2}(\sigma_a^2 - \sigma_i^2) S^2 \Gamma dt = 0$ . Empirically this term is be positive but small because  $\sigma_i \geq \sigma_{Hist}$ .
- P&L vs.  $\Delta S$  parabolic chart (Lecture 2 on Volatility Smile by Emmanuel Derman) for positive convexity (e.g., call option) means that the larger step  $\Delta S$  the higher the profit from Gamma:  $P\&L = \frac{1}{2} \Gamma (\Delta S)^2$ , where direction given by the sign of  $\Delta$  does not matter, and  $(\frac{\Delta S}{S})^2 = \sigma^2 \Delta t$  – variance is proportional to squared returns.
- If  $\sigma_i$  the volatility that we fix at the time of entering a position then

$$\sigma_i^2 = \frac{\int_0^T \sigma_a^2 S^2 \Gamma dt}{\int_0^T S^2 \Gamma dt}$$

meaning that implied volatility is a Gamma-weighted average over actual volatility. This is consistent with representation of variance via the log contract with constant dollar Gamma.

- Vega P&L ‘sits’ within Delta P&L – we need volatility number to calculate Delta, therefore, how Delta changes is, in fact, important. Higher volatility gives a smooth Delta function, while low volatility gives a step-like Delta function and peaking Gamma. If volatility is not a constant, ‘the bastard Vega’ factor appears and other Greeks become uncertain.

Regarding the use of Itô lemma when expanding  $dV_i$ , when deriving **the Black-Sholes PDE** you have seen

$$dV = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} dS + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} dt$$

also seen

$$\left( \frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} \right) dt = r \left( V - S \frac{\partial V}{\partial S} \right) dt$$

where

$$dV = rVdt \quad \text{and so} \quad \Delta dS = \frac{\partial V}{\partial S} rSdt.$$



### Volatility information contained in ATM straddle

The straddle position is made up of a long call and a long put with the same strikes and expiries. Practitioners use the market prices of at-the-money straddles to deduce the at-the-money volatility.

The Black–Scholes value of a straddle is given below with  $C$  and  $P$  are the values of the call and the put respectively and we have used the put-call parity.

$$\begin{aligned} V_S &= C + P = C + C - Se^{-D(T-t)} + Ee^{-r(T-t)} \\ &= 2C - Se^{-D(T-t)} + Ee^{-r(T-t)} \end{aligned}$$

We can therefore deduce the price of a single call and hence the implied volatility

$$C = \frac{1}{2}(V_S + Se^{-D(T-t)} - Ee^{-r(T-t)})$$

Since the straddle is at the money we have  $S = E = S^*$  and  $t = t^*$ .

### Risk Reversal

The risk-reversal is a long call, with strike above the current spot, and a short put with a strike below the current spot. Both have the same expiry. Practitioners use the market price of the risk-reversal to deduce the volatility skew.

Assume that the strikes of the call and the put are a short distance  $\epsilon$  away from the current spot: the strike of the call is thus  $S^* + \epsilon$  and the strike of the put is  $S^* - \epsilon$ . With more informative notation,  $C(E, \sigma_{\text{imp}})$  means a call with strike  $E$  and implied volatility  $\sigma_{\text{imp}}$ , similarly for puts.

The Black–Scholes value of the risk-reversal in terms of the current spot and time is

$$\begin{aligned} V_{\text{RR}} &= C(S^* + \epsilon, \sigma_{\text{imp}}(S^* + \epsilon, T)) - P(S^* - \epsilon, \sigma_{\text{imp}}(S^* - \epsilon, T)) \\ &= C(S^* + \epsilon, \sigma_{\text{imp}}(S^* + \epsilon, T)) - C(S^* - \epsilon, \sigma_{\text{imp}}(S^* - \epsilon, T)) \\ &\quad + S^*e^{-D(T-t^*)} - (S^* - \epsilon)e^{-r(T-t^*)} \end{aligned}$$

If  $\epsilon$  is small, that is if the two strikes are close together we can expand and find

$$\begin{aligned} V_{\text{RR}} - S^*(e^{-D(T-t^*)} - e^{-r(T-t^*)}) &= \\ \epsilon \left( e^{-r(T-t^*)} + 2 \frac{\partial C}{\partial E}(S^*, b) + 2 \frac{\partial C}{\partial \sigma_{\text{imp}}}(S^*, b) a \right). \end{aligned}$$

where  $a(T)$  and  $b(T)$  parametrise the implied volatility  $\sigma\sqrt{T}$ .