

Options Implied Probability Distributions (OIPD)

- extracts risk-neutral probability distribution of an assets price at a specific future date from options prices

- outputs pdf and cdf of random variable S_t the price of the option @ expiration.

This is:

- A pricing measure
- consistent with no arbitrage
- Fully implied by traded options

This is NOT:

- a forecast
- a real-world probability
- belief distribution

Breeden - Litzenberger:

Given Pricing function $C(K)$

the risk neutral probability density function $f(K) = e^{rT} \frac{d^2 C(K)}{dK^2}$

Example: lets say that the value of a European call option is

$$C(K, T) = e^{-r(T-t)} \mathbb{E}[\max(S_T - K, 0)]$$

where T is expiry time
 S_T is price @ expiry

which is also represented as

$$C(K, T) = e^{-r(T-t)} \int_{-\infty}^{\infty} \max(S_T - K, 0) f(S_T) dS_T$$

pdf on random variable S_T

Now say we have a butterfly spread with a payout of ΔK . say it pays with probability p .

then, a reasonable non risk-seeking nor risk averse person would pay $\Delta K p$ for this option (remember expected value of Bernoulli distribution is p)

So if X is the amount we will pay, then $X = p \Delta K$

$$\Rightarrow p = \frac{X}{\Delta K}$$

strike	midline price
251	1.46
250	1.21
249	1.16

Butterfly spread buys 251, sells 2x250, buys 249

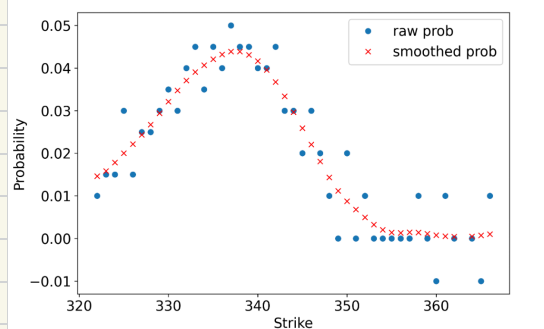
$$\Rightarrow X = 1.46 - 2(1.21) + 1.16 = 0.2$$

$$\Rightarrow p = \frac{X}{\Delta K} = \frac{0.2}{1} = 0.2 = 20\% \text{ chance of profit}$$

Repeat for every possible butterfly

↓ Smooth with Gaussian filter

↓ fit spline.



Moving from Discrete to Continuous Probabilities

Our old formula $P = \sum \frac{x}{\Delta K}$ $f(\Delta K) = \frac{x}{\Delta K}$ is a pmf.

We want a pdf. recall the differences:

$$\sum_{\Delta K} f(\Delta K) = 1$$

$$P(\Delta K) = f(\Delta K)$$

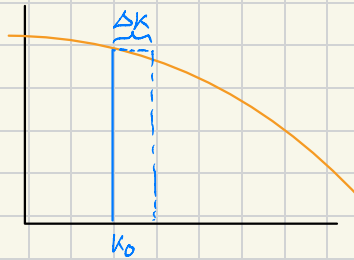
$$\int_{-\infty}^{\infty} f(\Delta K) = 1$$

$$P(A \leq \Delta K \leq B) = \int_A^B f(\Delta K)$$

So using a PDF, the probability the price at expiry S_T is within a band of width ΔK is

$$P(K_0 \leq S_T \leq K_0 + \Delta K) = \int_{K_0}^{K_0 + \Delta K} f(S_T) dS_T$$

$$\approx f(K_0) \Delta K$$



As $\Delta K \rightarrow 0$: $P(K_0 \leq S_T \leq K_0 + \Delta K)$ becomes $P(K_0 = S_T)$

$$P(S_T) = \lim_{\Delta K \rightarrow 0} f(S_T) \Delta K$$

$$\Rightarrow f(S_T) = \lim_{\Delta K \rightarrow 0} \frac{P(S_T)}{\Delta K}$$

$$f(S_T) = \lim_{\Delta K \rightarrow 0} \left(\frac{X/\Delta K}{\Delta K} \right) = \lim_{\Delta K \rightarrow 0} \left(\frac{X}{\Delta K^2} \right)$$

using butterfly formula \rightarrow

$$= \lim_{\Delta K \rightarrow 0} \left(\frac{C(S_T + \Delta K) - 2C(S_T) + C(S_T - \Delta K))}{\Delta K^2} \right)$$

$$= \frac{d^2 C}{dK^2}$$

Plugging in $C(K, T) = e^{-rT} \mathbb{E}[\max(S_T - K, 0)]$

$$= e^{-rT} \int_K^{\infty} (x - K) f(x) dx$$

We derived the Breeden-Litzenberger formula:

$$f(K) = \frac{d^2 C}{dK^2} e^{rT}$$

It may seem that this result is only applicable if everyone only buys butterfly spreads. Consider an alternative derivation:

$$C(K, T) = e^{-rT} \mathbb{E}[\max(S_T - K, 0)]$$

$$= e^{-rT} \int_0^{\infty} \max(S_T - K, 0) f(S_T) dS_T$$

$$= e^{-rT} \int_K^{\infty} (x - K) f(x) dx$$

$$= e^{-rT} \left(\int_K^{\infty} x f(x) dx - K \int_K^{\infty} f(x) dx \right)$$

$$\frac{dC}{dK} = e^{-rT} \left(\frac{d}{dK} \int_K^{\infty} x f(x) dx - \frac{d}{dK} K \int_K^{\infty} f(x) dx \right)$$

$$= e^{-rT} \left(-K f(K) - \left(\int_K^{\infty} f(x) dx - K f(K) \right) \right)$$

$$= e^{-rT} \left(\int_K^{\infty} f(x) dx \right)$$

since f is a PDF:

$$= e^{-rT} \left(\underbrace{\int_{-\infty}^K f(x) dx}_{\text{CDF!}} - 1 \right)$$

Derivative of CDF = PDF

$$\frac{d^2 C}{dK^2} = \frac{d}{dK} \left(e^{-rT} \int_K^{\infty} f(x) dx \right)$$

$$= e^{-rT} f(x)$$

$$\Rightarrow f(x) = \frac{d^2 C}{dK^2} e^{rT}$$