



# Session 8: Static Replication of European Payoffs

Tee Chyng Wen

QF620 Stochastic Modelling in Finance

# Behavior of Model Parameters – $\rho$

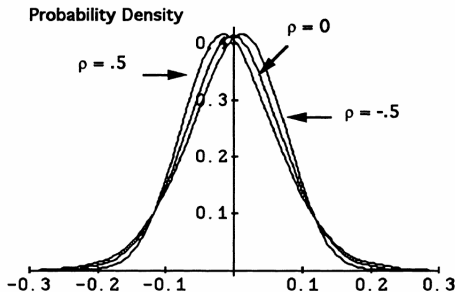
## Implication on Distribution

- The correlation parameter  $\rho$  is proportional to the skewness of stock returns.
- Intuitively, a negative correlation results in high volatility when the stock price drops, and this spreads the left tail of the probability density. The right tail is associated with low volatility and is not spread out.
- A negative correlation creates a fat left tail and a thin right tail in the stock return distribution.

## Implication on Pricing

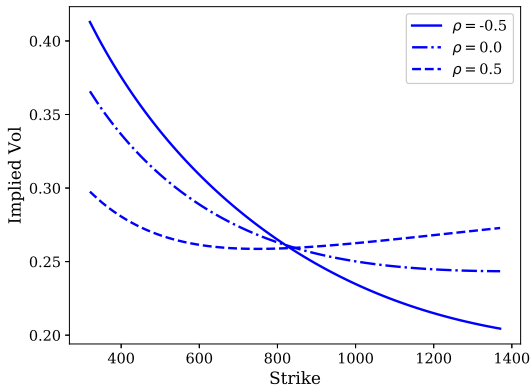
- This increases the prices of out-of-the-money puts and decreases the prices of out-of-the-money calls relative to the Black-Scholes model price.
- Intuitively, out-of-the-money put options benefit substantially from a fat left tail.
- A positive correlation will have completely opposite effects—it creates a fat right tail and a thin left tail.

## Behavior of Model Parameters – $\rho$



- ⇒ **Positive correlation** between stock and volatility is associated with **positive skew** in return distribution.
- ⇒ **Negative correlation** between stock and volatility is associated with **negative skew** in return distribution.

# Behavior of Model Parameters – $\rho$



Negative correlation increases the price of out-of-the-money put options and decreases the price of out-of-the-money call options.

# Behavior of Model Parameters – $\nu$

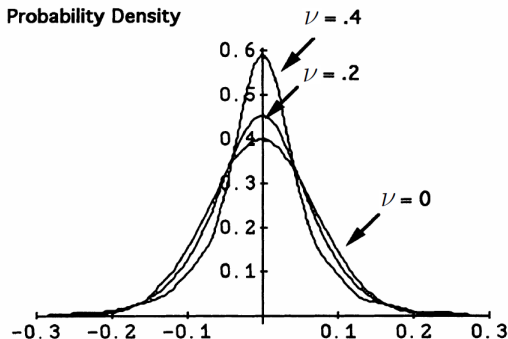
## Implication on Distribution

- When the volatility of volatility parameter is 0, we are back to a normal log-return distribution (if  $\beta = 0$ ).
- Otherwise, it increases the kurtosis of stock returns, creating two fat tails in both ends of the distribution.
- This has the effect of raising out-of-the-money puts and out-of-the-money call prices.

## Implication on Pricing

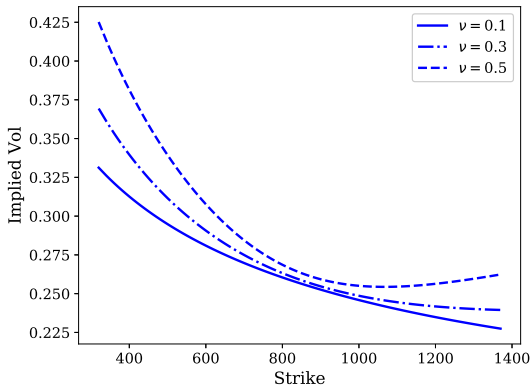
- If volatility is uncorrelated with stock return, then increasing the volatility of volatility only increases the kurtosis of spot return.
- In this case, random volatility is associated with increases in the prices of far-from-the-money options relative to near-the-money options.
- In contrast, the correlation of volatility with the spot return produces skewness.

## Behavior of Model Parameters – $\nu$



- ⇒ **Increasing volatility-of-volatility** has the effect of **increasing the kurtosis** of return.
- ⇒ When the volatility-of-volatility parameter is 0, volatility will be deterministic.

# Behavior of Model Parameters – $\nu$



Larger volatility-of-volatility  $\nu$  increases the price of out-of-the-money call and put options.

# What is the “model-free” framework?

In a **model-free** formulation, we let  $f(s)$  denote the risk-neutral probability density function of the stock price at time  $T$ , we can price a vanilla European call option maturing at time  $T$  as follows:

$$C(K) = e^{-rT} \mathbb{E}^*[(S_T - K)^+] = e^{-rT} \int_K^\infty (s - K) f(s) ds.$$

In earlier modelling approach, we will attempt to specify a model for the stock price process. A typical example is the Black-Scholes model, which will lead to:

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

We could also have used the Bachelier model, the displaced-diffusion model, or the SABR model.

⇒ Once a model is chosen, the risk-neutral density is also determined, by calibrating the model to market option data.



# What is the “model-free” framework?

Suppose we have sufficient liquid option quotes in the market, can we skip over the step of using a model to specify the stock price process, but instead **extract the risk-neutral density function** directly?

| Market Price | Model-Free Formula                              |
|--------------|---|
| $C(K_1)$     | $e^{-rT} \int_{K_1}^{\infty} (s - K_1) f(s) ds$ |
| $C(K_2)$     | $e^{-rT} \int_{K_2}^{\infty} (s - K_2) f(s) ds$ |
| $C(K_3)$     | $e^{-rT} \int_{K_3}^{\infty} (s - K_3) f(s) ds$ |
| $C(K_4)$     | $e^{-rT} \int_{K_4}^{\infty} (s - K_4) f(s) ds$ |
| $\vdots$     | $\vdots$  |

# Implied Risk-Neutral Density

- Black-Scholes model for European call and put options allow us to determine their prices by taking expectation of the option payoff on maturity, discount back to today.
- The model **assumes a lognormal process for the stock price** following

$$dS_t = rS_t dt + \sigma S_t dW_t^*,$$

under  $\mathbb{Q}^*$ , where the volatility  $\sigma$  is a model parameter that we need to determine.

- Since the vanilla option market is very liquid, we do not need to rely on any mathematical models to calculate the prices of options.
- Instead, the traded price of these options are published real-time by exchanges globally, and the process can now be reversed—given that an option traded at a particular price, what is the implied volatility that we should substitute into our Black-Scholes formula to give us this price, assuming that the underlying stock price is indeed following a lognormal process?

# Implied Risk-Neutral Density

- One option price allows us to determine one implied volatility for a particular **strike** and **maturity**.
- The market is constantly providing live information about option prices across a wide range of strikes for a given maturity.
- Given this information, we can now bring our analysis to the next level—instead of asking for just one single implied volatility to match one option price, we want to determine, for a given maturity, the **implied risk-neutral distribution**, that allows us to match the market volatility smile or skew.
- To this end, we need to apply **Leibniz's rule**:

$$I(x) = \int_{u(x)}^{v(x)} g(x, t) dt$$
$$\frac{dI(x)}{dx} = g(x, v(x)) \frac{dv}{dx} - g(x, u(x)) \frac{du}{dx} + \int_{u(x)}^{v(x)} \frac{\partial g(x, t)}{\partial x} dt$$

# Implied Risk-Neutral Density

This allows us to **extract the risk-neutral probability density function** from market-traded vanilla option prices.

Let  $f(s)$  denote the risk-neutral probability density, we can apply Leibniz's rule to obtain:

$$\begin{aligned}C(K) &= e^{-rT} \mathbb{E}[(S_T - K)^+] = e^{-rT} \int_K^\infty (s - K) f(s) ds \\ \frac{\partial C(K)}{\partial K} &= e^{-rT} \left[ \lim_{x \rightarrow \infty} (x - K) f(x) \frac{dx}{dK} - (K - K) f(K) \frac{dK}{dK} - \int_K^\infty f(s) ds \right] \\ &= -e^{-rT} \int_K^\infty f(s) ds \\ \frac{\partial^2 C(K)}{\partial K^2} &= -e^{-rT} \left[ \lim_{x \rightarrow \infty} f(x) \frac{dx}{dK} + f(K) \frac{dK}{dK} - \int_K^\infty \frac{\partial f(s)}{\partial K} ds \right] \\ &= e^{-rT} f(K).\end{aligned}$$

# Implied Risk-Neutral Density

We can also carry out the same procedure to the put options:

$$P(K) = e^{-rT} \mathbb{E}[(K - S_T)^+] = e^{-rT} \int_0^K (K - s) f(s) ds$$

These give us

$$\frac{\partial^2 C(K)}{\partial K^2} = e^{-rT} f(K) \quad \text{and} \quad \frac{\partial^2 P(K)}{\partial K^2} = e^{-rT} f(K).$$

This is the **Breedon-Litzenberger formula**, which showed in 1978 that the terminal distribution of the stock price implicit in the option prices, also known as the **implied distribution**, can be obtained by differentiating the call & put option prices twice with respect to the strike price.

Subsequently, **Carr and Madan** showed in 1998 that any European payoff can be replicated using a portfolio of cash, forward contracts, and European call & put options.

# Static Replication of European Payoff

To replicate any twice differentiable European payoff  $h(S_T)$ , we write

$$V_0 = e^{-rT} \mathbb{E}[h(S_T)] = e^{-rT} \int_0^\infty h(s) f(s) ds.$$

Let  $F = S_0 e^{rT}$ , we have

$$V_0 = e^{-rT} \mathbb{E}[h(S_T)] = \underbrace{\int_0^F h(K) \frac{\partial^2 P(K)}{\partial K^2} dK}_{(1)} + \underbrace{\int_F^\infty h(K) \frac{\partial^2 C(K)}{\partial K^2} dK}_{(2)}$$

Note that

- ① We have changed the dummy variable of the integral from  $s$  to  $K$ , as a reminder that the second-order derivatives of the call and put options are with respect to the strike.
- ② We are using liquid OTM and ATM options, i.e. low-strike puts and high-strike calls, to extract the risk-neutral density.

# Static Replication of European Payoff

Let us consider the call integral (2). Using integration-by-parts twice, we obtain

$$\begin{aligned}
 & \int_F^\infty h(K) \frac{\partial^2 C(K)}{\partial K^2} dK \\
 &= \left[ h(K) \frac{\partial C(K)}{\partial K} \right]_F^\infty - \int_F^\infty h'(K) \frac{\partial C(K)}{\partial K} dK \\
 &= \left[ \cancel{h(\infty) \frac{\partial C(\infty)}{\partial K}} \overset{0}{-} h(F) \frac{\partial C(F)}{\partial K} \right] - \left[ h'(K) C(K) \right]_F^\infty + \int_F^\infty h''(K) C(K) dK \\
 &= -h(F) \frac{\partial C(F)}{\partial K} - \left[ \cancel{h'(\infty) C(\infty)} \overset{0}{-} h'(F) C(F) \right] + \int_F^\infty h''(K) C(K) dK \\
 &= -h(F) \frac{\partial C(F)}{\partial K} + h'(F) C(F) + \int_F^\infty h''(K) C(K) dK.
 \end{aligned}$$

# Static Replication of European Payoff

Applying the same steps to the put integral (1), we can obtain

$$\int_0^F h(K) \frac{\partial^2 P(K)}{\partial K^2} dK = h(F) \frac{\partial P(F)}{\partial K} - h'(F) P(F) + \int_0^F h''(K) P(K) dK.$$

Combining both integrals (1) and (2), we have:

$$\begin{aligned} V_0 = h(F) \left[ -\frac{\partial C(F)}{\partial K} + \frac{\partial P(F)}{\partial K} \right] + h'(F) [C(F) - P(F)] \\ + \int_0^F h''(K) P(K) dK + \int_F^\infty h''(K) C(K) dK. \end{aligned}$$

This expression can be simplified further using **put-call parity**:

$$C(K) - P(K) = S_0 - Ke^{-rT}.$$



# Static Replication of European Payoff

Note that differentiating both sides of the put-call parity with respect to  $K$  yields:

$$\frac{\partial C(K)}{\partial K} - \frac{\partial P(K)}{\partial K} = -e^{-rT}.$$

Also, when  $K = F = S_0 e^{rT}$ , the call and put options are worth the same, so that:

$$C(F) - P(F) = S_0 - F e^{-rT} = 0.$$

Substituting both results, we arrive at the final static replication formula:

$$V_0 = e^{-rT} h(F) + \int_0^F h''(K) P(K) dK + \int_F^\infty h''(K) C(K) dK$$

**Reminder** Note that  $K$  in the integrals is a dummy variable — we use it to remind ourselves that the integrals are weighted across  $P(K)$  and  $C(K)$ , i.e. put and call options across a wide range of strikes.

# Static Replication of European Payoff

**Example** A financial contract pays  $aS_T^b$  on maturity date  $T$ , where  $a, b \in \mathbb{R}^+$  are positive real numbers. Use the static replication method to replicate this payoff using vanilla European call and put options.

**Solution** With  $h(S_T) = aS_T^b$ , we have

$$h'(S_T) = abS_T^{b-1}, \quad h''(S_T) = ab(b-1)S_T^{b-2}.$$

Hence, the payoff, which is twice differentiable, can be static replicated with a portfolio of options as follow:

$$\begin{aligned} V_0 = e^{-rT} aF^b &+ \int_0^F ab(b-1)K^{b-2}P(K) dK \\ &+ \int_F^\infty ab(b-1)K^{b-2}C(K) dK. \end{aligned}$$

# Static Replication of a Log Contract

**Example** Suppose we want to derive the valuation formula for a log contract paying  $\log \frac{S_T}{S_0}$  at maturity  $T$ , where  $S_t$  is the value of a stock.

- 1 Derive the valuation formula under Black-Scholes model.
- 2 Formulate the static replication portfolio using the Carr-Madan approach.

# Variance Swaps

**Variance swaps** are contracts which allow us to gain explicit volatility (and variance) exposure. This frees us from the need to worry about delta or gamma hedging if we were to use vanilla options to gain volatility exposure.

The payoff of a variance swap is given by

$$\text{Var Swap} = \text{Notional} \times (\sigma_R^2 - \sigma_K^2),$$

where  $\sigma_R^2$  is the **realized variance** of the stock and  $\sigma_K^2$  is the **strike variance**.

The realized variance  $\sigma_R^2$  is quantified as

$$\sigma_R^2 = \frac{252}{N} \sum_{i=1}^N \left( \log \left( \frac{S_{t_i}}{S_{t_{i-1}}} \right) \right)^2,$$

where  $i$  labels the value of the stock on each day and  $N$  is the total number of days in the contract. Variance swaps capture the realized variance of the underlying asset. It is an intuitive contract based on the definition of historical variance. The contract is often described in terms of the fair strike  $\sigma_K^2$ .

# Variance Swaps

To price a variance swap, we observe that the discrete sum over the log returns can be approximated by a continuous time integral

$$\sum_{i=1}^N \left[ \log \left( \frac{S_{t_i}}{S_{t_{i-1}}} \right) \right]^2 \approx \int_0^T \sigma_t^2 dt.$$

If we apply Itô's formula to a general stochastic differential equation

$$dS_t = rS_t dt + \sigma_t S_t dW_t^*,$$

we obtain

$$d \log S_t = \frac{dS_t}{S_t} - \frac{1}{2} \sigma_t^2 dt \quad \Rightarrow \quad \sigma_t^2 dt = 2 \left[ \frac{dS_t}{S_t} - d \log S_t \right].$$

Integrating both sides and then take expectation, we obtain

$$\begin{aligned} \int_0^T \sigma_t^2 dt &= 2 \int_0^T \frac{dS_t}{S_t} - 2 \log \left( \frac{S_T}{S_0} \right) \\ \mathbb{E}^* \left[ \int_0^T \sigma_t^2 dt \right] &= 2 \mathbb{E}^* \left[ \int_0^T \frac{dS_t}{S_t} \right] - 2 \mathbb{E}^* \left[ \log \left( \frac{S_T}{S_0} \right) \right]. \end{aligned}$$

# Variance Swaps

The first term on the RHS can be evaluated readily:

$$2\mathbb{E}^* \left[ \int_0^T \frac{dS_t}{S_t} \right] = 2\mathbb{E}^* \left[ \int_0^T \frac{rS_t dt + \sigma_t S_t dW_t^*}{S_t} \right] = 2rT.$$

The second term on the RHS is a static hedge of holding a log contract to expiry. It only depends on the initial stock price  $S_0$  and the final stock price  $S_T$ . This is perfectly suited for the static replication approach, and is the same problem we have solved previously for the log contract:

$$2\mathbb{E}^* \left[ \log \left( \frac{S_T}{S_0} \right) \right] = 2 \log \left( \frac{F}{S_0} \right) - 2e^{rT} \int_0^F \frac{P(K)}{K^2} dK - 2e^{rT} \int_F^\infty \frac{C(K)}{K^2} dK.$$

Since  $F = S_0 e^{rT}$ , this can be further simplified into

$$2\mathbb{E}^* \left[ \log \left( \frac{S_T}{S_0} \right) \right] = 2rT - 2e^{rT} \int_0^F \frac{P(K)}{K^2} dK - 2e^{rT} \int_F^\infty \frac{C(K)}{K^2} dK.$$

# Variance Swaps

Note that

- ⇒ The log contract can be replicated using a portfolio of European put and call options.
- ⇒ The weighting of the options is  $\frac{1}{K^2}$ . The portfolio contains all possible strikes.
- ⇒ The portfolio has more weight for downside options than upside options—indicating skew sensitivity.
- ⇒ The portfolio is asking us to place a lot more weight on low strike puts, relative to high strike calls.

Finally, we obtain

$$\mathbb{E} \left[ \int_0^T \sigma_t^2 dt \right] = 2e^{rT} \int_0^F \frac{P(K)}{K^2} dK + 2e^{rT} \int_F^\infty \frac{C(K)}{K^2} dK.$$

# VIX Index

The generalized formula used in the VIX calculation<sup>8</sup> is:

$$\sigma^2 = \frac{2}{T} \sum_i \frac{\Delta K_i}{K_i^2} e^{RT} Q(K_i) - \frac{1}{T} \left[ \frac{F}{K_0} - 1 \right]^2 \quad (1)$$

**WHERE...**

|              |   |
|--------------|---|
| $\sigma$ is  | $VIX/100 \Rightarrow VIX = \sigma \times 100$   |
| T            | Time to expiration  |
| F            | Forward index level derived from index option prices  |
| $K_0$        | First strike below the forward index level, F   |
| $K_i$        | Strike price of $i^{th}$ out-of-the-money option; a call if $K_i > K_0$ and a put if $K_i < K_0$ ; both put and call if $K_i = K_0$ . |
| $\Delta K_i$ | Interval between strike prices – half the difference between the strike on either side of $K_i$ :                                     |

$$\Delta K_i = \frac{K_{i+1} - K_{i-1}}{2}$$

(Note:  $\Delta K$  for the lowest strike is simply the difference between the lowest strike and the next higher strike. Likewise,  $\Delta K$  for the highest strike is the difference between the highest strike and the next lower strike.)

|          |  |
|----------|--|
| R        | Risk-free interest rate to expiration                                  |
| $Q(K_i)$ | The midpoint of the bid-ask spread for each option with strike $K_i$ . |