

# Options: Valorisation et Couverture dans le cadre Black-Scholes

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# Objectives

At the end of this lecture, you should understand:

- ▶ The assumptions and key concepts of the Black-Scholes model
- ▶ How to interpret de "delta" of an option
- ▶ How to set up a dynamic strategy to hedge the market risk of an option
- ▶ The risks associated with dynamic hedging, and the weaknesses of the Black-Scholes model,
- ▶ Why and how the Black-Scholes model is still used, despite its limitations

## The Black-Scholes Model Made Easy

- The Key Concepts of Black-Scholes

## Dynamic Hedging of Options

- Self-Financing Replicating Portfolio

- Sources of Hedging Error

## Implied Volatility

## Calculation of implied volatility

- Newton's method

- Jaeckel's method

## Vana-Volga Pricing and Hedging

## Smile Parametrisation

## Illustrations

## The Breeden-Litzenberger Formula

- Interpretation

# Historical Perspective

- ▶ Bachelier (1900) models stock prices as a random walk, and derives the first option pricing formula.
- ▶ James Boness (1964) derives almost the same option valuation formula as Black-Scholes, but uses the expected stock return as opposed to the risk-free rate
- ▶ Paul Samuelson (1965) also derives an option pricing formula similar to the Black-Scholes model, but it involves both the expected return of the stock and the expected return of the option.
- ▶ Ed Thorp (1967) reports to have hedged an option portfolio using Boness' formula, but replacing the expected return by the risk-free rate.

# The Black Scholes Model

A formal justification of the Black-Scholes model requires many assumptions:

- ▶ There is no arbitrage opportunity;
- ▶ It is possible to borrow and lend cash at a known constant risk-free interest rate  $r$ ;
- ▶ It is possible to buy and sell any amount of stock (this includes short selling);
- ▶ Transactions do not incur any costs;
- ▶ The stock price follows a log-normal distribution with constant and known drift  $\mu$  and volatility  $\sigma$ .

# The Black-Scholes Key Concept

”It is possible to create a hedged position, consisting of a long position in the stock and a short position in the option, whose value will not depend on the price of the stock.”

Fischer Black and Myron Scholes. “The Pricing of Options and Corporate Liabilities.”. In: *Journal of Political Economy* 81 (1973), pp. 637–654.

# The Black-Scholes Formula

$S_0$  Price of underlying asset

$T$  Option expiry

$K$  Strike

$r$  Interest rate

$\sigma$  Volatility

Call price:

$$C_0 = S_0 N(d_1) - Ke^{-rT} N(d_2)$$

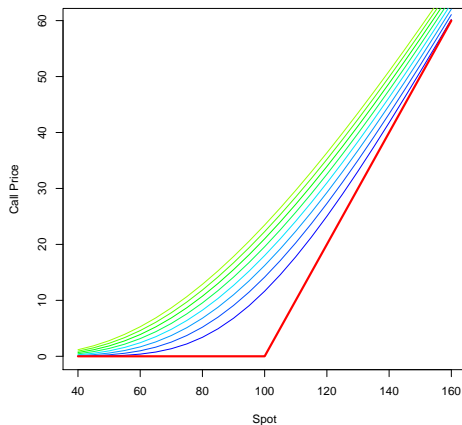
with:

$$d_1 = \frac{\ln(S_0/K) + (r + \sigma^2/2)T}{\sigma\sqrt{T}}$$

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

# Call Price as a function of Spot and $T$

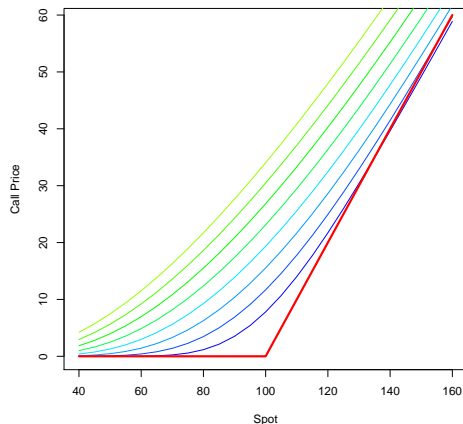
Strike = 100, Time to maturity 100 to 900 days.  $\sigma = .3$ .





# Call Price as a function of Spot and Volatility

Strike = 100, Time to maturity = 1 Yr,  $\sigma = .1, \dots, .9$ .



# The Replicating Portfolio

The Black-Scholes formula can be directly interpreted as the description of the replicating portfolio:

$$C = S_0 N(d_1) - e^{-rT} KN(d_2)$$

The replicating portfolio has:

- ▶  $N(d_1)$  stock
- ▶  $KN(d_2)$  € of nominal of a zero-coupon bond expiring at  $T$ .

## Review from Binomial Model

An option in a binomial model is equivalent to a portfolio:

- ▶ long  $\Delta$  units of stock
- ▶ funded by borrowing  $B \text{ €}$  at the risk-less rate

Such that, for a call worth  $C$ :

$$C = S_0 \Delta - B$$

The same principle applies with the Black-Scholes model:

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

# Black-Scholes Delta

The delta is the change in option price for a change of one e in the price of the underlying asset.

$$\text{Delta} = \frac{\text{Change in Option Value}}{\text{Change in Underlying Value}}$$

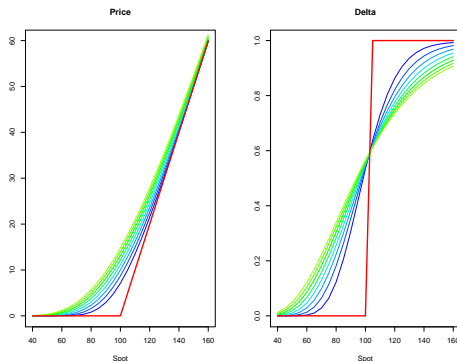
For a European call:

$$\Delta_C = N(d_1)$$

For a European put:

$$\Delta_p = N(d_1) - 1$$

# Price and Delta of a Call as a function of Maturity



# Construction of a hedge portfolio

Notation:

$C_t$  Value of derivative

$V_t$  Value of hedge portfolio

$B_t$  Amount borrowed/lent at the risk-free rate

$\Delta_t$  Delta of derivative

At  $t = 0$ , the derivative is sold at price  $C_0$  and the proceeds are used to purchase a hedge portfolio. The initial hedge is  $\Delta_0 S_0 - B_0$ , where  $B_0$  is computed from the accounting identity:

$$C_0 = \Delta_0 S_0 - B_0$$

## Example

A financial institution writes (sells) an at-the-money option on a stock worth €100. The option expires in two months, the hedge will be rebalanced every week (for illustration). Interest rate is 2% and volatility 30%.

Question: Compute the option price and the hedge portfolio.

# Initial Hedge Portfolio

- ▶ Call price (Black-Scholes):  $C_0 = 5.04$
- ▶ Delta:  $\Delta_0 = 0.5352$
- ▶ Amount borrowed:

$$\begin{aligned} B_0 &= C_0 - \Delta_0 S_0 \\ &= -48.48 \end{aligned}$$

Initial hedge portfolio			
Hedge Portfolio			
Call Price	Stock	Bond	Total
$C_0$	$\Delta_0 \times S_0$	$B_0$	
5.04	53.52	-48.48	5.04



## Rebalancing of hedge portfolio

The hedge must be adjusted periodically. At each step  $i$ , the decision rule is as follows:

1. Compute the value of the hedge portfolio formed at the previous time step:

$$V_i = -B_{i-1}e^{r\Delta t} + \Delta_{i-1}S_i$$

2. Compute the amount of stock to hold:

$$\Delta_i = \frac{\partial C_i}{\partial S_i}$$

3. The new hedge portfolio is  $\Delta_i S_i + B_i$ , with borrowing  $B_i$  determined by:

$$-B_i = V_i - \Delta_i S_i$$

At expiry of the derivative, the residual wealth is:

$$-C_T + \Delta_{T-1}S_T - B_{T-1}e^{r\Delta t}$$

# Hedge Effectiveness

The quality of a model is ultimately measured by the residual error:

$$\begin{aligned} E_T &= -B_{T-1}e^{r\Delta t} + \Delta_{T-1}S_T - C_T \\ &= V_T - C_T \end{aligned}$$

## Simulation 1: Call expiring out of the money

Week	stock price	$\Delta$	call	bond	hedge port.
1	100.00	0.54	5.05	-48.9	5.05
2	98.16	0.47	3.79	-42.0	4.05
3	90.05	0.18	0.90	-16.0	0.23
4	88.01	0.11	0.42	-9.8	-0.15
5	90.28	0.13	0.50	-11.6	0.09
6	94.67	0.25	1.02	-23.0	0.66
7	94.17	0.17	0.53	-15.5	0.53
8	95.65	0.16	0.34	-14.5	0.78
9	94.67	0.00	0.00	+1.6	0.62

Hedging discrepancy: 0.62 €.

## Example 2: Call expiring in the money

Week	stock price	$\Delta$	call	bond	hedge port.
1	100.00	0.54	5.05	-48.9	5.05
2	95.38	0.37	2.63	-32.7	2.55
3	93.58	0.29	1.73	-25.26	1.87
4	102.46	0.63	5.39	-60.1	4.45
5	101.23	0.58	4.22	-55.0	3.66
6	103.78	0.71	5.39	-68.6	5.12
7	103.34	0.72	4.56	-69.6	4.77
8	109.01	0.98	9.09	-98.0	8.82
9	103.94	1.00	3.94	-100.	4.06

Hedging discrepancy:  $4.06 - 3.94 = .12 \text{ €}$ .

# Large Scale Dynamic Hedging Simulation

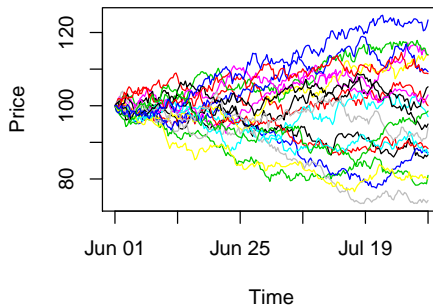
To test the effectiveness of delta-hedging with the Black-Scholes model:

1. Simulate price scenarios
2. Simulate the dynamic rebalancing of the hedge portfolio
3. For each path, observe the hedging error at expiry

# Simulated paths - log-normal process

$\sigma = 30\%$ ,  $T = 2$  months

Sample paths,  $\sigma: 30\%$



# First Simulation

Simulations in a “perfect Black-Scholes world”

- ▶ The volatility is known and constant
- ▶ the interest rate is constant
- ▶ No transaction costs

Delta-hedging simulation, maturity: 2 months,  
 $\sigma = .3, r = .02, K = 100, S_0 = 100$ . Option price: 5.05.  
200 time steps, 1000 simulations.

# Distribution of Hedging Error - ATM European Call

**distribution of wealth at expiry**  
**Vanilla IBM c @ 100 expiry: 2010-08-1**

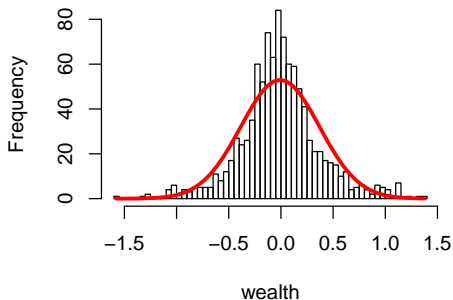


Figure: Distribution of hedging error  $\sigma = 0.4$



# Delta Hedging Error

Sources of hedging error:

1. Hedging frequency (Black-Scholes assumes continuous rebalancing);
2. Along one path, the volatility that is experienced may not be the theoretical volatility  $\sigma$ .

# Delta Hedging Error vs. Hedging Frequency

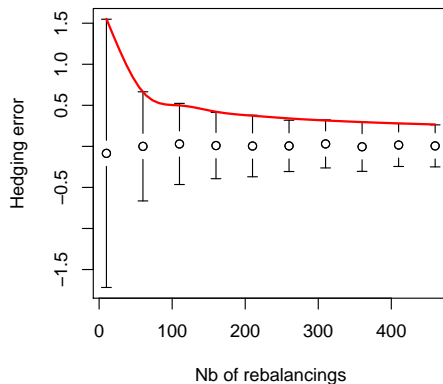


Figure: The variance of the hedging error is inversely related to the hedging frequency

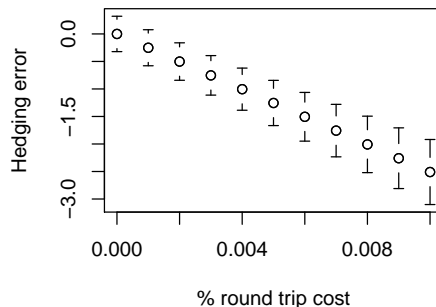
# Black-Scholes in Real Life

In real life, every assumption of the Black-Scholes model is invalidated:

- ▶ Interest rate is not constant
- ▶ Future volatility is unknown
- ▶ The distribution of  $S_T$  is not log-normal
- ▶ There are transaction costs and restrictions to shorting stocks

Let's measure the impact on hedging error.

# Delta Hedging Error vs. Transaction Cost



**Figure:** The hedging error is directly related to the magnitude of the transaction costs

# Hedging Error and Unknown Volatility

Consider:

- ▶ The pricing volatility, used to determine the initial value of the derivative
- ▶ The actual volatility, the one that is experienced during the live of the option.

You sell an option and dynamically hedge your risk by trading the delta-hedging portfolio.

- ▶ If the actual volatility is lower than the pricing (expected) volatility, you will on average make money.
- ▶ If the actual volatility is higher than the pricing (expected) volatility, you will on average lose money.

## More Greeks...

Hedging error is strongly dependent upon the curvature of the option price curve: the Gamma.

$$\Gamma = \frac{\partial^2 C}{\partial S^2}$$

$$\approx \frac{C(S_0 + h) - 2C(S_0) + C(S_0 - h)}{h^2}$$

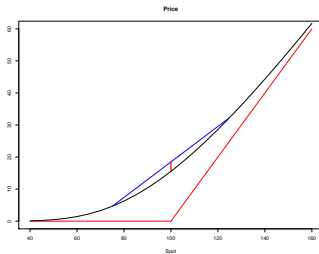


Figure: Gamma of a Call

## More Greeks...

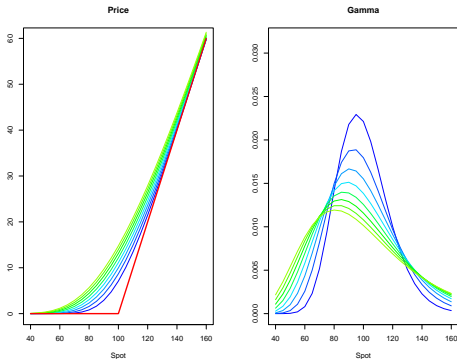


Figure: Price and Gamma of a Call as a function of maturity

# Hedging Error and Unknown Volatility

In the presence of unknown volatility, it can be shown that hedging error is related to:

1. The difference between the pricing (i.e. assumed) volatility and the effective volatility
2. The gamma of the option:
  - ▶ Hedging error is small when the gamma is small and keeps the same sign
  - ▶ Hedging error is large when the gamma is large and changes sign



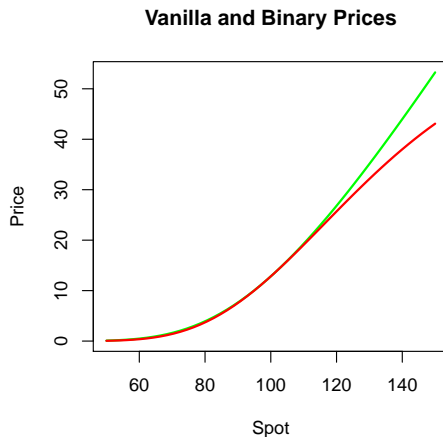
# Influence of Gamma on Hedging Error

We compare the hedging error of two simple derivatives expiring in 1 year:

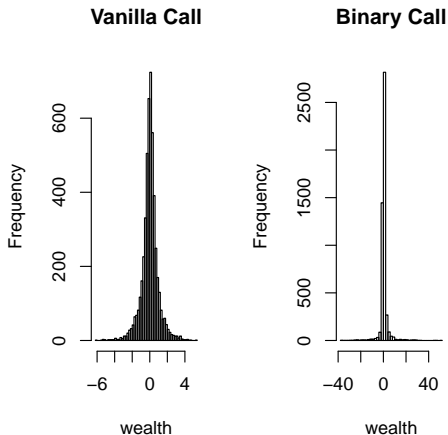
- ▶ a European call,  $K = 100$ ,
- ▶ a binary call struck at 126.65 e, that will pay 62.09 e if the option expires in the money.


The binary call is designed to have the same premium and same initial delta as the European option.

# PV of Vanilla and Binary Calls with same current price and delta



## Delta Hedging Errors: Vanilla vs. Binary Calls



In identical experimental conditions, the hedging error of the binary call is about **10 times larger** than the one of the European 

# Robustness of Black-Scholes

$$\epsilon_T = \frac{1}{2} \int_0^T [\Sigma^2 - \sigma_t^2] \frac{\partial^2 C}{\partial S^2} S^2 dt$$

N El-Karoui, M Jeanblanc-Picquè, and Steven E. Shreve.

“Robustness of the Black and Scholes Formula”. In:  
*Mathematical Finance* 8.2 (1998), pp. 93–126. ISSN: 1467-9965.  
DOI: 10.1111/1467-9965.00047

## Hedging Error: Conclusion

In a Black-Scholes world, hedging error is determined by four factors:

1. the hedging frequency
2. the transaction costs
3. the magnitude of the option gamma
4. the difference between  $\Sigma$  (the BS volatility used for pricing and hedging) and the volatility that is actually experienced,  $\sigma$ .

# Is Black-Scholes Still in Use?

- ▶ Black-Scholes in its pure form (delta hedging argument) has probably never been used in practice
- ▶ Complex derivatives have been managed with much more complex models (stochastic volatility, jumps, local volatility) since the mid 1990's.

# Is Black-Scholes Still in Use?

The Black-Scholes formula is widely used in vanilla option markets, but not in the way consistent with the theory:

- ▶ The Black-Scholes formula is used to operate a change of unit of measure for vanilla options: to convert prices into volatilities (it is easier to interpret volatility differences than price differences).
- ▶ Traders recognize that the main risk is unknown volatility / non normality of returns, and therefore hedge options with other options.

Espen Gaarder Haug and Nassim Nicholas Taleb. "Option Traders Use (very) Sophisticated Heuristics, Never the Black-Scholes-Merton Formula.". In: *Economic Behavior and Organization* 77 (2011).

# Is Black-Scholes Still in Use?

The Black-Scholes assumptions (normality of returns) are unfortunately still sometimes used by regulators to define risk measures.

1. Value at Risk, Pillar I of Basel II/III;
2. Solvency II risk models.



## The Implied Volatility Problem

- ▶ In the BS formula, we use  $\sigma$  as an input.
- ▶ In reality, exchanges quote options in price.
- ▶ The BS formula is used to convert an option price into the corresponding volatility.

Given the observed price  $C^*$  of a call, compute the volatility  $\sigma$  such that:

$$\begin{aligned} C^* &= C(S, K, T, r, \sigma) \\ &= f(\sigma) \end{aligned}$$

# Option data: Settlement prices of options on the WTI Feb09 futures contract

## NEW YORK MERCANTILE EXCHANGE NYMEX OPTIONS CONTRACT LISTING FOR 12/29/2008

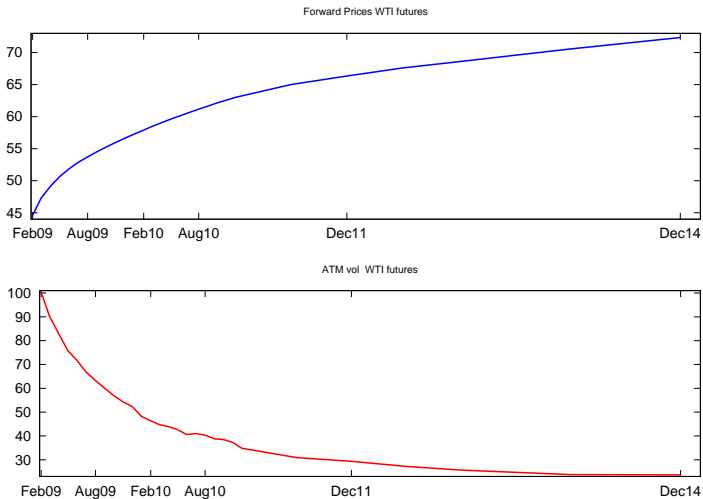
-----CONTRACT-----				TODAY'S	PREVIOUS	ESTIMATED
				SETTLE	SETTLE	VOLUME
LC	02 09	P	30.00	.53	.85	0
LC	02 09	P	35.00	1.58	2.28	0
LC	02 09	P	37.50	2.44	3.45	0
LC	02 09	C	40.00	3.65	2.61	10
LC	02 09	P	40.00	3.63	4.90	0
LC	02 09	P	42.00	4.78	6.23	0
LC	02 09	C	42.50	2.61	1.80	0
LC	02 09	C	43.00	2.43	1.66	0
LC	02 09	P	43.00	5.41	6.95	100

# Option data: Options on the S&P 500 Index

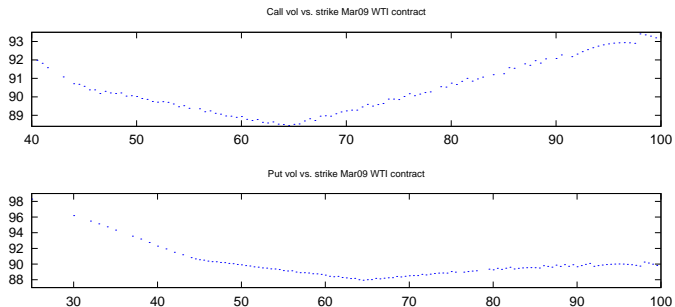
(Source:CBOE)

SPX (SP 500 INDEX)	1290.59	+7.24						
Jan 24 2011 @ 14:03 ET								
Calls	Last Sale	Net	Bid	Ask	Vol	Open	Int	...
11 Jan 1075.00 (SPXW1128A1075-E)	0.0	0.0	215.30	217.00	0	0		...
11 Jan 1100.00 (SPXW1128A1100-E)	0.0	0.0	190.60	191.80	0	0		...

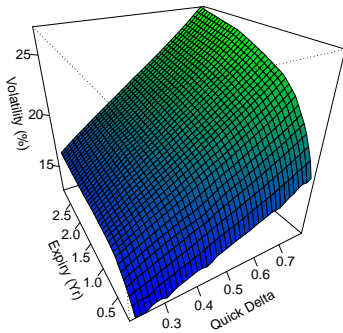
# ATM Volatility



# WTI Option Implied Vol



# Implied Volatility of S&P 500 Index Options (24-jan-2011)



# The Implied Volatility

Option traders study the implied volatility rather than the option prices.

Why?

- ▶ A measure of value, irrespective of strike and maturity
- ▶ The BS model assumes a constant volatility: deviations from that assumption are useful to study

# The Implied Volatility Problem

Given the observed price  $C^*$  of a call, compute the volatility  $\sigma$  such that:

$$\begin{aligned} C^* &= C(S, K, T, r, \sigma) \\ &= f(\sigma) \end{aligned}$$



# The Implied Volatility Problem

There is a change in convexity in  $C = f(\sigma)$ :

$$\frac{\partial C}{\partial \sigma} = S n(d_1) \sqrt{T}$$

$$\frac{\partial^2 C}{\partial \sigma^2} = S \sqrt{T} n(d_1) \frac{1}{\sigma} \left[ \frac{1}{\sigma^2 T} \ln\left(\frac{F}{K}\right)^2 - \frac{1}{4} \sigma^2 T \right]$$

with  $F = S e^{rT}$ .

Thus,  $C(\sigma)$  is convex on the interval  $(0, \sqrt{\frac{2|\ln(F/K)|}{T}}]$ , and concave otherwise.

## Convergence of Newton's Method

To ensure convergence of Newton's method, one must carefully choose the initial point.

### Theorem

*Let  $f$  be defined on the interval  $[a, b]$  and assume that:*

1.  $f(x^*) = 0$  for some  $x^* \in [a, b]$
2.  $f'(x) > 0$
3.  $f''(x) \geq 0$

*Then Newton's method converges monotonically from  $x_0 = b$ . If*

1.  $f(x^*) = 0$  for some  $x^* \in [a, b]$
2.  $f'(x) > 0$
3.  $f''(x) \leq 0$

*Then Newton's method converges monotonically from  $x_0 = a$ .*

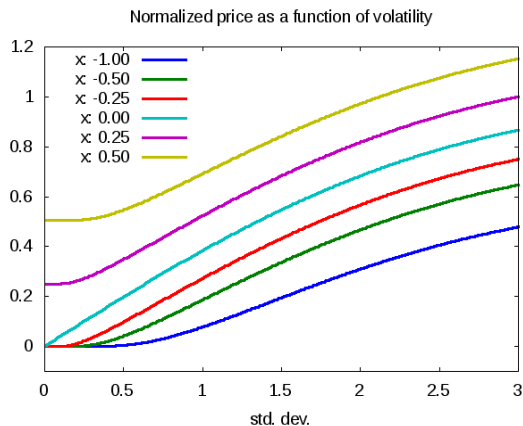
# Convergence of Newton's Method

Consider now Newton's method started at

$$\sigma_0 = \sqrt{\frac{2|\ln(F/K)|}{T}}$$

- ▶ If  $f(\sigma_0) > 0$ , we are in case I of theorem.
- ▶ If  $f(\sigma_0) < 0$  we are in case II.

# Normalized Call price as a function of $\sigma$



# Implied Volatility by Newton's Method

The following algorithm generates a monotonic series  $(\sigma_n)$ :

1. Set  $\sigma_0 = \sqrt{\frac{2|\ln(F/K)|}{T}}$
2. While  $|C(\sigma_n) - C^*| > \epsilon$ :

2.1 Let

$$\sigma_{n+1} = \sigma_n + \frac{C^* - C(\sigma_n)}{\frac{\partial C}{\partial \sigma}}$$

2.2  $n \leftarrow n + 1$

## Jackel's Method

Given an option price  $p$ , we must solve for  $\sigma$

$$p = \delta \theta \left[ F \Phi \left( \theta \left[ \frac{\ln(F/K)}{\sigma} + \frac{\sigma}{2} \right] \right) - K \Phi \left( \theta \left[ \frac{\ln(F/K)}{\sigma} - \frac{\sigma}{2} \right] \right) \right]$$

where:

$\delta$  discount factor

$\theta$  1 for call, -1 for put

$F$  Forward price:  $Se^{(r-d)T}$

$\sigma$   $\sigma\sqrt{T}$

## Jackel's Method

Set:

$$\begin{aligned}x &= \ln(F/K) \\ b &= \frac{p}{\delta \sqrt{FK}}\end{aligned}$$

The Black-Scholes equation becomes:

$$b = \theta \left[ e^{x/2} \Phi \left( \theta \left[ \frac{x}{\sigma} + \frac{\sigma}{2} \right] \right) - e^{-x/2} \Phi \left( \theta \left[ \frac{x}{\sigma} - \frac{\sigma}{2} \right] \right) \right]$$

The normalized price function changes convexity at  $\sigma_c = \sqrt{|x|}$

# Jaeckel's Method

Further defines

$$f(\sigma) = \begin{cases} \ln\left(\frac{b-i}{\bar{b}-i}\right) & \text{if } \bar{b} < b_c \\ b - \bar{b} & \text{otherwise} \end{cases}$$

with:

$b_c$   $b(x, \sigma_c, \theta)$

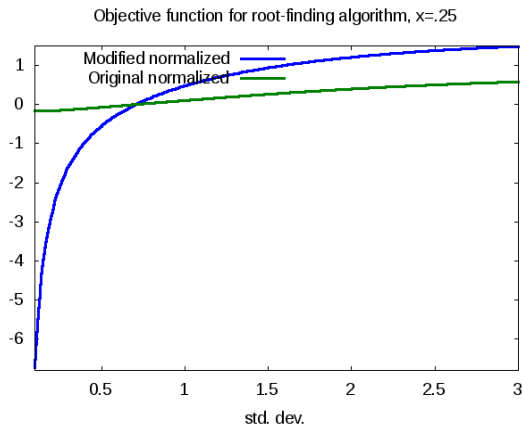
$i$  normalized intrinsic value:  $1_{\theta x > 0} \theta (e^{x/2} - e^{-x/2})$

$\bar{b}$  target normalized price

The function  $f(\sigma)$  is now monotonously concave.



# Transformed objective function



- ▶ Vega:  $\nu = \frac{\partial O}{\partial \sigma}$
- ▶ Vanna:  $\frac{\partial \nu}{\partial S}$
- ▶ Volga:  $\frac{\partial \nu}{\partial \sigma}$

# Vanna-Volga Dynamic Hedging

- ▶ A practical method for pricing and hedging derivatives, taking into account an uncertain volatility.
- ▶ Popular for Foreign Exchange derivatives
- ▶ Relates the price of a complex derivative to the known price of simpler, liquid instruments.

## Vanna-Volga Dynamic Hedging

An extended version of Ito's lemma with random volatility:

$$\begin{aligned} dO(t, K) = & \frac{\partial O}{\partial t} dt + \frac{\partial O}{\partial S} dS_t + \frac{\partial O}{\partial \sigma} d\sigma_t \\ & + \frac{1}{2} \frac{\partial^2 O}{\partial S^2} (dS_t)^2 + \frac{1}{2} \frac{\partial^2 O}{\partial \sigma^2} (d\sigma_t)^2 + \frac{\partial^2 O}{\partial S \partial \sigma} dS_t d\sigma_t \end{aligned}$$

# Vanna-Volga Dynamic Hedging

$$\begin{aligned}
 dO(t, K) - \Delta_t dS_t - \sum_{i=1}^3 x_i dC_i(t, K_i) = \\
 \left[ \frac{\partial O}{\partial t} - \sum_i x_i \frac{\partial C_i}{\partial t} \right] dt + \left[ \frac{\partial O}{\partial S} - \Delta_t - \sum_i x_i \frac{\partial C_i}{\partial S} \right] dS_t \\
 + \left[ \frac{\partial O}{\partial \sigma} - \sum_i x_i \frac{\partial C_i}{\partial \sigma} \right] d\sigma_t + \left[ \frac{\partial^2 O}{\partial S^2} - \sum_i x_i \frac{\partial^2 C_i}{\partial S^2} \right] (dS_t)^2 \\
 + \left[ \frac{\partial^2 O}{\partial \sigma^2} - \sum_i x_i \frac{\partial^2 C_i}{\partial \sigma^2} \right] (d\sigma_t)^2 + \left[ \frac{\partial^2 O}{\partial S \partial \sigma} - \sum_i x_i \frac{\partial^2 C_i}{\partial S \partial \sigma} \right] dS_t d\sigma_t
 \end{aligned}$$

# Vanna-Volga Hedging

$$dO(t, K) - \Delta_t dS_t - \sum_{i=1}^3 x_i dC_i(t, K_i) =$$
$$r \left[ O(t, K) - \sum_i x_i C_i(t, K_i) \right] dt$$

## Vanna-Volga Dynamic Hedging

The weights  $x_i$  are obtained by solving the system of linear equations:

$$\frac{\partial O}{\partial \sigma} = \sum_i x_i \frac{\partial C_i}{\partial \sigma}$$

$$\frac{\partial^2 O}{\partial \sigma^2} = \sum_i x_i \frac{\partial^2 C_i}{\partial \sigma^2}$$

$$\frac{\partial^2 O}{\partial S \partial \sigma} = \sum_i x_i \frac{\partial^2 C_i}{\partial S \partial \sigma}$$

or,

$$b = Ax$$

## Vanna-Volga Dynamic Hedging

Choice of hedge instruments

- ▶ An at-the-money straddle:

$$C_1 = C(S) + P(S)$$

- ▶ A “risk reversal”, traditionally defined as

$$C_2 = P(K_1) - C(K_2)$$

with  $K_1$  and  $K_2$  chosen so that the options have a delta of .25.

- ▶ A “butterfly”, defined as

$$C_3 = \beta(P(K_1) + C(K_2)) - (P(S) + C(S))$$

with  $\beta$  determined to set the vega of the butterfly to 0.



## Vanna-Volga Dynamic Hedging: Calculation Steps

- compute the risk indicators for the option to be priced:

$$b = \begin{pmatrix} \frac{\partial O}{\partial \sigma} \\ \frac{\partial^2 O}{\partial \sigma^2} \\ \frac{\partial^2 O}{\partial \sigma \partial S} \end{pmatrix} \quad (1)$$

- compute A matrix

$$A = \begin{pmatrix} \frac{\partial C_1}{\partial \sigma} & \cdots & \frac{\partial C_3}{\partial \sigma} \\ \frac{\partial^2 C_1}{\partial \sigma^2} & \cdots & \frac{\partial^2 C_3}{\partial \sigma^2} \\ \frac{\partial^2 C_1}{\partial \sigma \partial S} & \cdots & \frac{\partial^2 C_3}{\partial \sigma \partial S} \end{pmatrix} \quad (2)$$

## Vanna-Volga Dynamic Hedging: Calculation Steps

- solve for  $x$ :

$$b = Ax$$

- the corrected price for  $O$  is:

$$O^M(t, K) = O^{BS}(t, K) + \sum_{i=2}^3 x_i (C_i^M(t) - C_i^{BS}(t)) \quad (3)$$

where  $C_i^M(t)$  is the market price and  $C_i^{BS}(t)$  the Black-Scholes price (i.e. with flat volatility).

# Vanna-Volga Dynamic Hedging

Neglecting the off diagonal terms in  $A$ , a simplified procedure is to estimate  $x_i$  by:

$$x_2 = \frac{\frac{\partial^2 O}{\partial \sigma^2}}{\frac{\partial^2 C_2}{\partial \sigma^2}}$$

$$x_3 = \frac{\frac{\partial^2 O}{\partial \sigma \partial S}}{\frac{\partial^2 C_3}{\partial \sigma \partial S}}$$

In practice, the weights  $x_i$  are scaled to better fit market prices.

# Smile Paramétrisation

Risk reversals and butterflies provide a simple paramétrisation of the smile:

- ▶ the risk reversal measures asymmetry
- ▶ the butterfly measures the convexity

# Volatility Interpolation

Given the ATM volatility and at two other strikes, determine the volatility at an arbitrary strike  $K$ .

# Vana-Volga Smile Interpolation

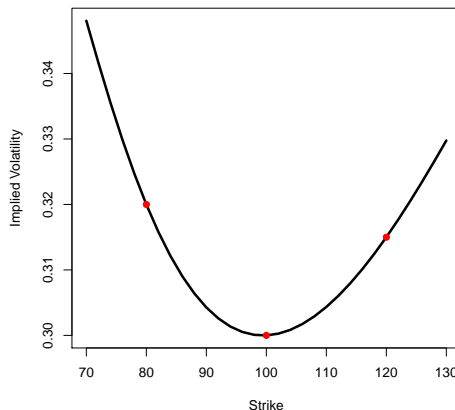


Figure: Interpolated Volatility Curve with Vanna-Volga algorithm

# Vanna-Volga Pricing of a Binary Option

Consider a one-year binary call, struck at the money. Assume that the smile is quadratic.

Use the traditional benchmark instruments of the FX market:

- ▶ Straddle,
- ▶ Risk-reversal
- ▶ Butterfly

## Vanna-Volga Pricing of a Binary Option

### Steps:

- ▶ Compute the strikes corresponding to a  $25\Delta$  call and put,
- ▶ Compute the value of each benchmark instrument. The butterfly must be vega-neutral,
- ▶ Compute the risk indicators (vega, vanna, volga) for the binary option,
- ▶ ... and for the benchmark instruments,
- ▶ Compute the smile cost of each benchmark: the price with the smile effect less price at the ATM volatility.
- ▶ Compute the price correction for the binary option.



# Pricing a Binary Option

In summary, we get:

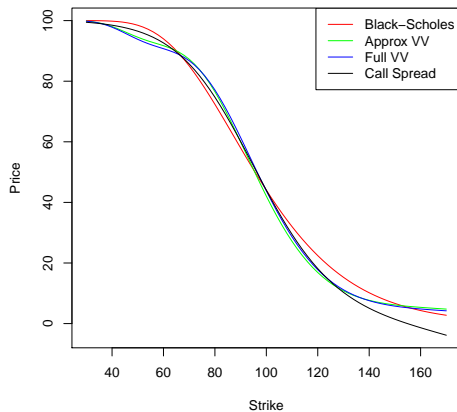
- ▶ Black-Scholes price: 32
- ▶ With approximate Vanna-Volga correction:  
 $32 + -4.86 = 27.14$
- ▶ With accurate Vanna-Volga correction:  $32 + -3.39 = 28.61$
- ▶ the approximation by a call spread is: 29.44

# Pricing Comparison

comparison of binary option value for a range of strikes, computed with four methods:

1. the regular Black-Scholes method, assuming a flat volatility
2. the Black-Scholes price with approximate Volga-Vanna correction
3. same as above, but with an accurate calculation of the Vanna-Volga correction
4. the value of a call spread centered at the strike

# Vana-Volga Pricing of a Binary Option



# The Breeden-Litzenberger formula

Risk-neutral density of the underlying asset at maturity  $T$  as a function of derivative prices:

$$p_T(K) = e^{rT} \frac{\partial^2 C(S, K, T)}{\partial K^2} \quad (4)$$

where  $C(S, K, T)$  is the price of a call of strike  $K$ , maturity  $T$ , when the current spot is  $S$ .

## The Breeden-Litzenberger formula

By definition of the risk-neutral probability,

$$C(S, K, T) = e^{-rT} \int_K^{\infty} (S_T - K) p(S_T) dS_T \quad (5)$$

Applying Leibniz's Rule to get:

$$\frac{\partial C(S, K, T)}{\partial K} = -e^{-rT} \int_K^{\infty} p(S_T) dS_T$$

Let  $F(K)$  the cumulative density function of  $S_T$ ,

$$\begin{aligned} e^{rT} \frac{\partial C(S, K, T)}{\partial K} &= - \int_K^{\infty} p(S_T) dS_T \\ &= F(K) - 1 \end{aligned}$$

# The Breeden-Litzenberger formula

Differentiate again with respect to  $K$  to get:

$$\frac{\partial^2 C}{\partial K^2} e^{rT} = p(K) \quad (6)$$

## Local Vol: Interpretation

Consider a butterfly spread centered at  $K$ , and scaled to yield a maximum payoff of 1. Let  $\phi(S_T)$  be the payoff function. The value of the butterfly is:

$$\begin{aligned} V &= \frac{1}{\Delta K} [C(K + \Delta K) - 2C(K) + C(K - \Delta K)] \\ &= e^{-rT} \int_0^\infty \phi(S) p(S) dS \end{aligned}$$

## Local Vol: Interpretation

In the interval  $[K - \Delta K, K + \Delta K]$ , approximate  $p(S)$  by the constant  $p(K)$  to get:

$$\begin{aligned} V &= e^{-rT} p(K) \int_0^\infty \phi(S) dS \\ &= e^{-rT} p(K) \Delta K \end{aligned}$$

Finally, use the definition of the derivative:

$$\lim_{\Delta K \rightarrow 0} \frac{1}{\Delta K^2} [C(K + \Delta K) - 2C(K) + C(K - \Delta K)] = \frac{\partial^2 C(K)}{\partial K^2} \quad (7)$$

to get:

$$p_T(K) = e^{rT} \frac{\partial^2 C(K)}{\partial K^2} \quad (8)$$



Analytical expression for the density of  $S_T$ 

$$p(K) = n(d_2) \left\{ \frac{1}{K\sigma\sqrt{T}} + \frac{\partial\sigma}{\partial K} \frac{2d_1}{\sigma} + \left( \frac{\partial\sigma}{\partial K} \right)^2 \frac{\sqrt{T} K d_1 d_2}{\sigma} + \frac{\partial^2\sigma}{\partial K^2} K\sqrt{T} \right\} \quad (9)$$

## Illustration: Shimko's Model

Fit a quadratic model to the implied volatility, in order to get analytical expressions for  $\frac{\partial \sigma}{\partial K}$ ,  $\frac{\partial^2 \sigma}{\partial K^2}$ :

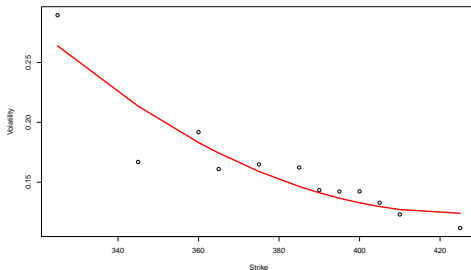
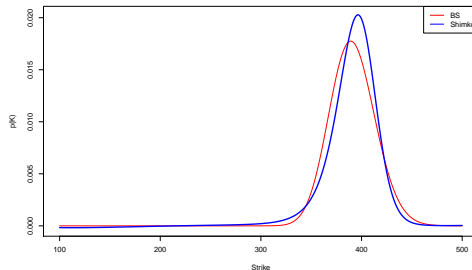


Figure: Quadratic Volatility Model

## Illustration: Shimko's Model

The density implied from the quadratic volatility smile clearly exhibit “fat tails”.



**Figure:** Density of  $S_T$ , with constant volatility and quadratic model for implied volatility.

## Consequence for Option Pricing

To illustrate the importance of correctly accounting for the volatility smile, we now consider a digital option maturing at the same time as our European options. We want to price this option in a way that is consistent with the observed volatility smile.

A naive approach would be to look up the Black-Scholes volatility corresponding to the strike, and price the digital option accordingly. The price of a digital cash-or-nothing call is given by:

$$C = e^{-rT} \Phi(d_2) \quad (10)$$

with:

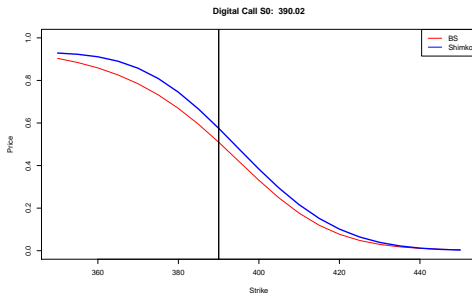
$$\begin{aligned} \Phi(x) &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}x^2} \\ d_2 &= \frac{\ln \frac{S}{K} + (r - d - \frac{\sigma^2}{2})T}{\sigma\sqrt{T}} \end{aligned}$$

## Consequence for Option Pricing

However, since we know the density of  $S_T$ , we can directly compute the expected discounted value of the digital payoff:

$$C = e^{-rT} \int_K^{\infty} p(S_T) dS_T$$

## Digital Price with and without Smile



**Figure:** Comparison of prices of a digital option, using a log-normal density for  $S_T$ , and using the density implied by the volatility smile fitted to a quadratic function.

## WTI Smile

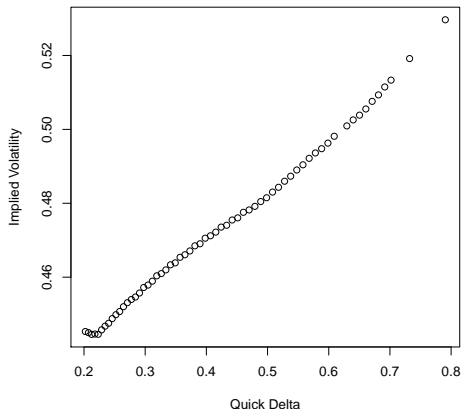


Figure: Implied volatility of WTI NYMEX options on the December 2009 Futures, observed on April 21, 2009

# WTI Smile

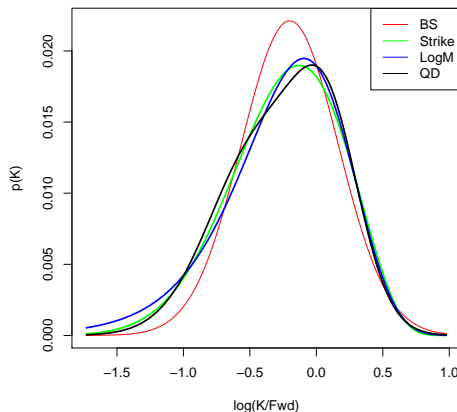


Figure: Implied density of  $F_T$ , the December 2009 WTI Nymex Futures contract. Calculation is performed by finite difference with implied



# References

- [1] Fischer Black and Myron Scholes. “The Pricing of Options and Corporate Liabilities.”. In: *Journal of Political Economy* 81 (1973), pp. 637–654.
- [2] Espen Gaarder Haug and Nassim Nicholas Taleb. “Option Traders Use (very) Sophisticated Heuristics, Never the Black-Scholes-Merton Formula.”. In: *Economic Behavior and Organization* 77 (2011).
- [3] N El-Karoui, M Jeanblanc-Picquè, and Steven E. Shreve. “Robustness of the Black and Scholes Formula”. In: *Mathematical Finance* 8.2 (1998), pp. 93–126. ISSN: 1467-9965. DOI: 10.1111/1467-9965.00047.