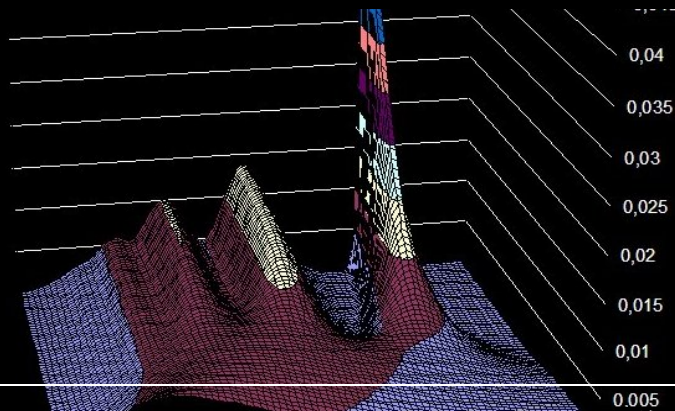


Pricing and Hedging without Fast Fourier Transform

A revisited, reliable and stable Fourier Transform Method for Affine Jump Diffusion Models



Marcello Minenna - Paolo Verzella
Quant Congress Europe – Oct. 31, 2005 - London



Syllabus of the presentation

- Review of Fourier Methods in Option Pricing
- Calibration and Performance
- Greek Behavior of New FT-Q



2



Syllabus of the presentation

- **Review of Fourier Methods in Option Pricing**
- Calibration and Performance
- Greek Behavior of New FT-Q

Review of Fourier Methods in Option Pricing – theory

European Call Maturity T Terminal Spot Price S_T

In AJD models Call Price can be expressed in a form close to the canonical Black – Scholes - Merton style

$$C_t = S_t P_1(\Theta) - Ke^{-rt} P_2(\Theta)$$

where

$$P_1(\Theta), P_2(\Theta) = \Pr(\ln S_T \geq \ln[K])$$

under different martingale measures



3



4



$P_1(\Theta), P_2(\Theta) = \Pr(\ln S_T \geq \ln[K])$
under different martingale measures



determined by using the Levy's inversion formula, i.e.:

$$\Pr(\ln S_T \geq \ln[K]) = \frac{1}{2} + \frac{1}{\pi} \int_0^{\infty} \operatorname{Re} \left[\frac{e^{-i\phi \ln[K]} \tilde{f}_j(\phi)}{i\phi} \right] d\phi$$

$$\Pr(\ln S_T \geq \ln[K]) = \frac{1}{2} + \frac{1}{\pi} \int_0^{\infty} \operatorname{Re} \left[\frac{e^{-i\phi \ln[K]} \tilde{f}_j(\phi)}{i\phi} \right] d\phi$$



a close formula for the Characteristic Function of the log – terminal price, i.e.:

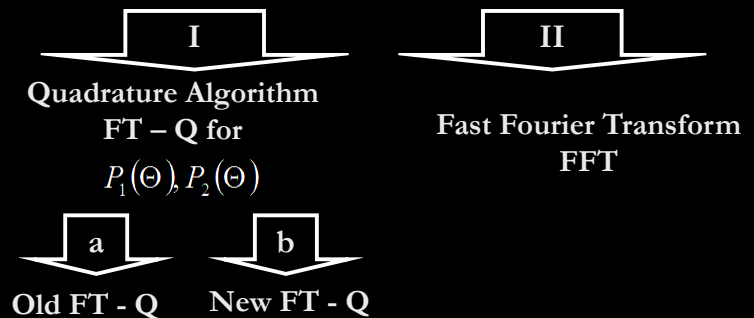
$$\tilde{f}_T(\phi) = E[e^{i\phi \ln S_T}]$$

$$\tilde{f}_T(\phi) = E[e^{i\phi \ln S_T}]$$



a closed formula for AJD models

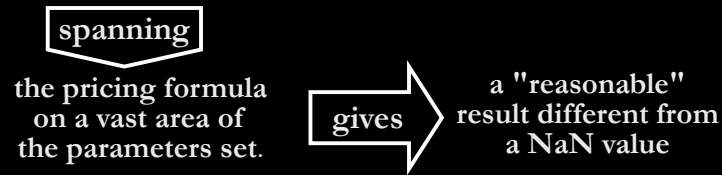
How to compute: C_i



Algorithms Valuation Criteria

STABILITY

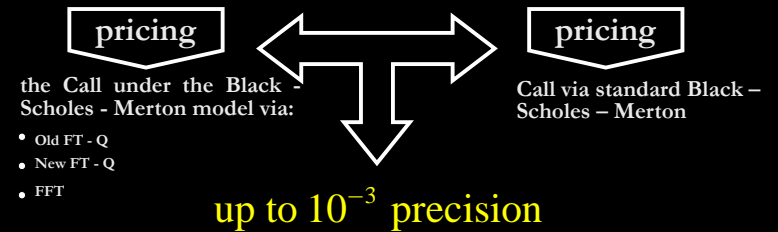
The algorithm is defined stable **if and only if**
it "closes" the quadrature scheme



Algorithms Valuation Criteria

ACCURACY

The algorithm is defined accurate **if and only if**



Algorithms Valuation Criteria

SPEED

The algorithm is defined fast **with respect to**
the results of the **FFT algorithm**



High Order Newton Cotes Algorithm

Up to 8th

$$C_t = S_t P_1(\theta) - Ke^{-rt} P_2(\theta)$$



Pros (+)

ACCURACY

Cons (-)

STABILITY
SPEED



In order to overcome the cited problems of Old FT – Q:



In order to overcome the cited problems of Old FT – Q:

- Gauss - Lobatto Quadrature Algorithm
- Re-adjustment of $\tilde{f}_T(\phi) = E[e^{i\phi \ln S_T}]$



$$C_t = S_t P_1(\Theta) - Ke^{-r\tau} P_2(\Theta)$$



Pros (+)

STABILITY
ACCURACY

Cons (-)

SPEED



Cooley - Tukey algorithm

Applied to the equivalent formula via a recombinant FFT parameters

$$C_t = \frac{e^{-\alpha \ln K}}{\pi} \int_0^\infty e^{-i\phi \ln K} \tilde{f}_j(\phi) d\phi \quad \text{for ATM}$$



Pros (+)

SPEED
FASTER
(up to 20 times the quadrature algorithms)

Cons (-)

STABILITY
* The formula must be changed arbitrarily according to Option moneyness

ACCURACY
** the recombinant FFT parameters must be changed according to the choice of the pricing models

Syllabus of the presentation

- Review of Fourier Methods in Option Pricing
- **Calibration and Performance**
- Greek Behaviour of New FT-Q

The Calibration Procedure and Performance

$$SSE_t = \min_{v(t), \Phi} \sum_{n=1}^N [C_{Market}(S_t) - C_{AJD}(S_t)]^2$$



Quadrature Algorithm
FT - Q



Fast Fourier Transform
FFT



Old FT - Q



New FT - Q

The Calibration Procedure and Performance

$$SSE_t = \min_{v(t), \Phi} \sum_{n=1}^N [C_{Market}(S_t) - C_{AJD}(S_t)]^2 \xleftrightarrow{\text{through}} \text{Quadrature Algorithm Old FT - Q}$$

Pros (+)

Cons (-)

STABILITY
ACCURACY
SPEED

The Calibration Procedure and Performance

$$SSE_t = \min_{v(t), \Phi} \sum_{n=1}^N [C_{Market}(S_t) - C_{AJD}(S_t)]^2 \xleftrightarrow{\text{through}} \text{Fast Fourier Trasform FFT}$$

Pros (+)

Cons (-)

SPEED

STABILITY *
ACCURACY **

The Calibration Procedure and Performance

$$SSE_t = \min_{v(t), \Phi} \sum_{n=1}^N [C_{Market}(S_t) - C_{AJD}(S_t)]^2 \xleftrightarrow{\text{through}} \text{Quadrature Algorithm New FT - Q}$$

Pros (+)

Cons (-)

STABILITY
ACCURACY
SPEED

The Calibration Procedure and Performance

By keeping in mind that only New FT-Q is stable and accurate, some figures on speed

Original Option Pricing Formulas are used

FFT	Heston Model	Merton Model	BCC Model
	7.26 sec.	10.54 sec.	18.33 sec.
NEW FT - Q	Heston Model	Merton Model	BCC Model
	55.12 sec.	66.48 sec.	110.39 sec.
OLD FT - Q	Heston Model	Merton Model	BCC Model
	390.41 sec.	454.76 sec.	722.1 sec.

By now, the speed of Fourier Trasform method is closer than ever to the FFT calibration time

Calibration Performances using
Option Readjusted Pricing Formulas

where available

	Heston Model	Merton Model	BCC Model
FFT	7.24 sec.	10.54 sec.	18.32 sec.
NEW FT - Q	23.13 sec.	66.48 sec.	48.7 sec.
OLD FT - Q	331.6 sec.	454.76 sec.	688.5 sec.

- Review of Fourier Methods in Option Pricing
- Calibration Procedure and Performance
- **Greek Behaviour of New FT-Q**

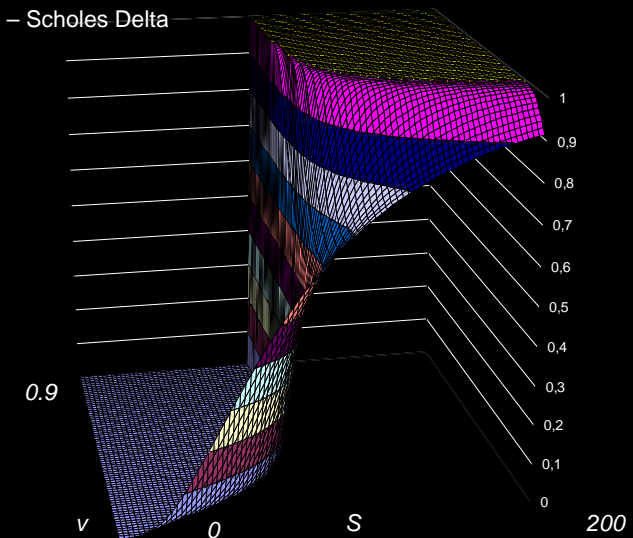
Greek behaviour of new FT-Q

An impressive methodology to test Stability of the New FT – Quadrature algorithm is to compute Greeks

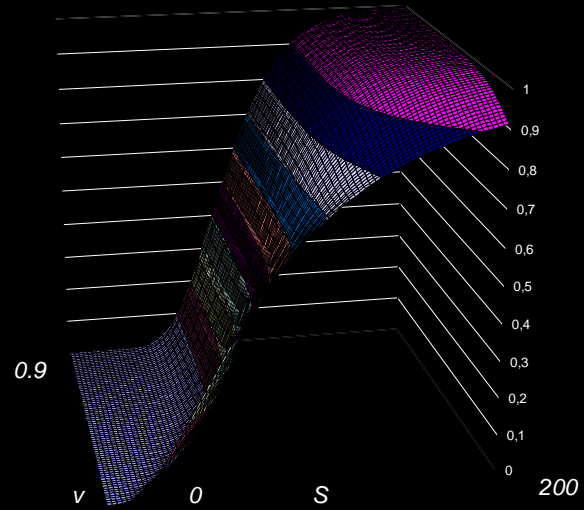
Infact, in an AJD setting the Greeks are available in closed form

So, an extended spanning of the AJD Greeks on the parameters set is useful to assess models and test Stability

Black – Scholes Delta



Heston Delta



Lambda = -2

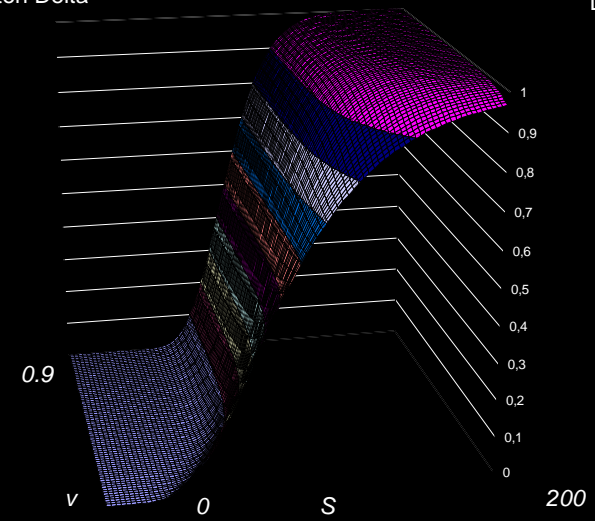
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Delta



Lambda = 2 ↑

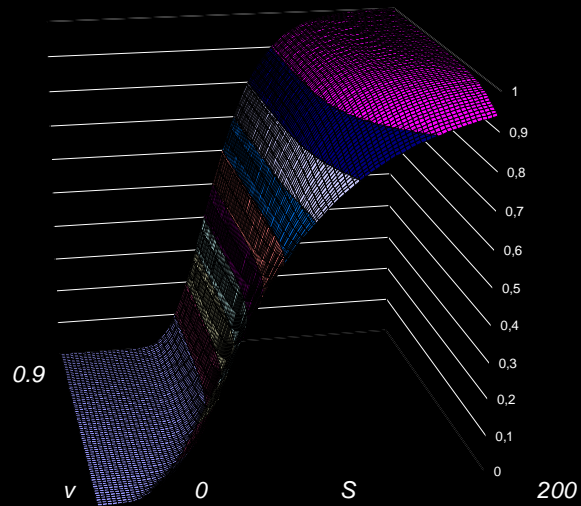
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Delta



Rho = -1

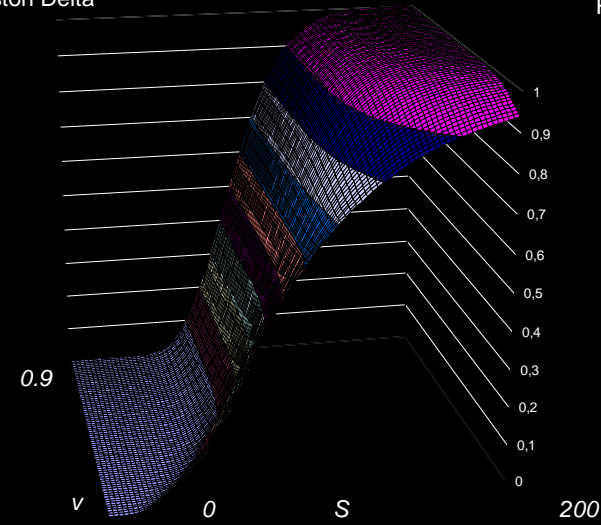
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Lambda = 0

Heston Delta



Rho = 1

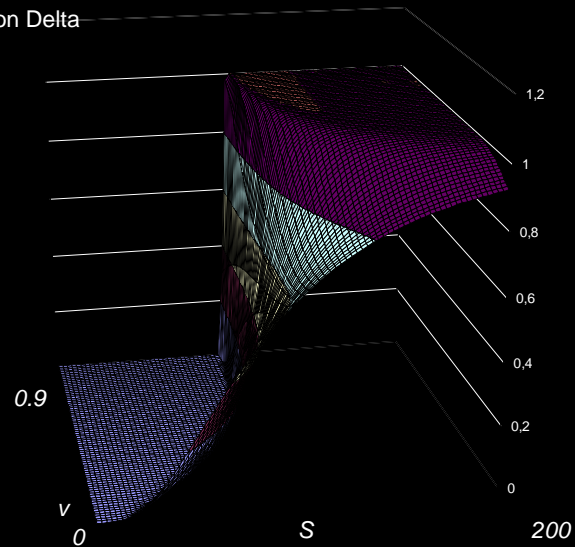
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Lambda = 0

Merton Delta

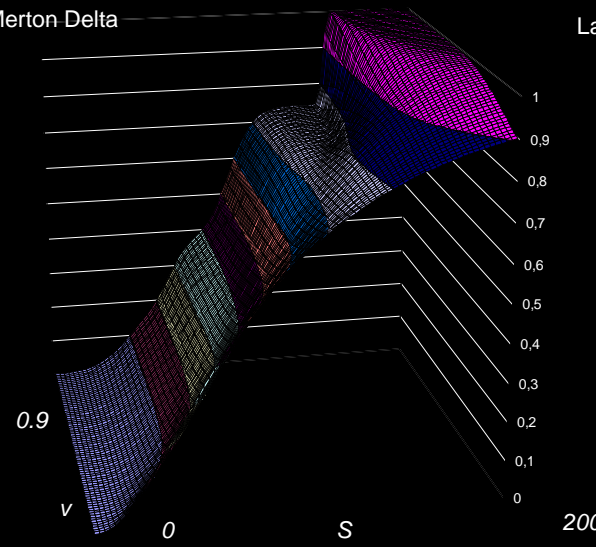


LambdaJ = 0

EtaJ = 0

MuJ = 0

Merton Delta

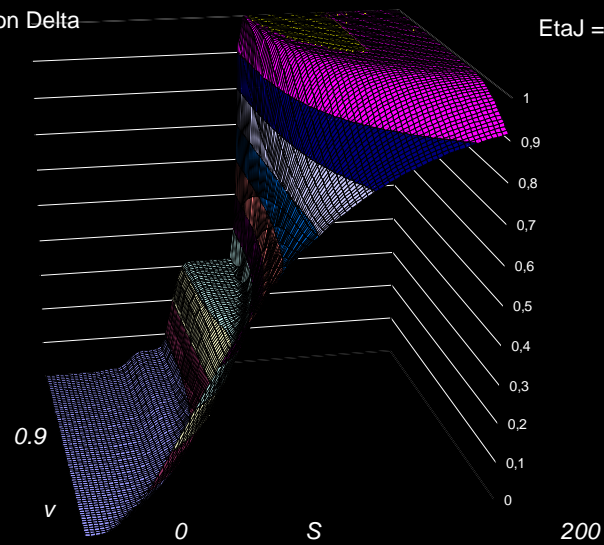


LambdaJ = 1.8 ↑

EtaJ = 0.1

MuJ = 0.5

Merton Delta

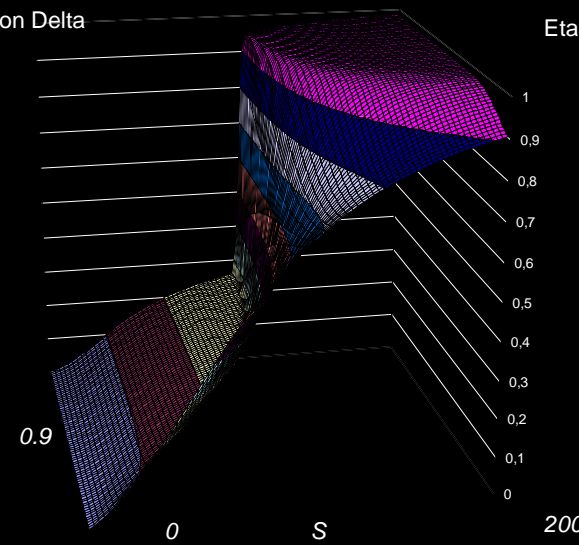


EtaJ = 0.1

LambdaJ = 0.5

MuJ = 0.5

Merton Delta

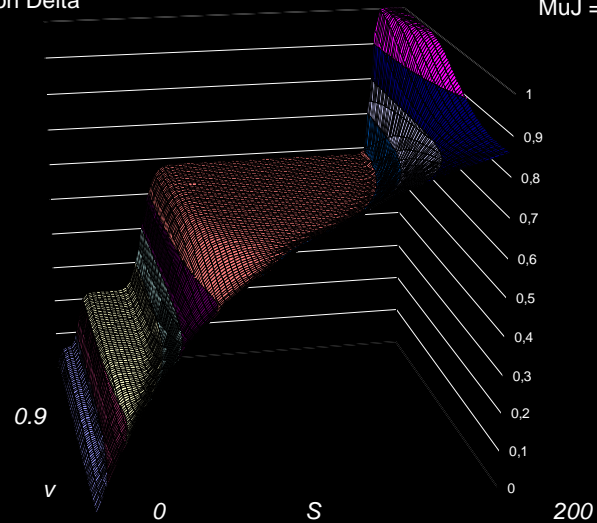


EtaJ = 0.75 ↑

LambdaJ = 0.5

MuJ = 0.5

Merton Delta



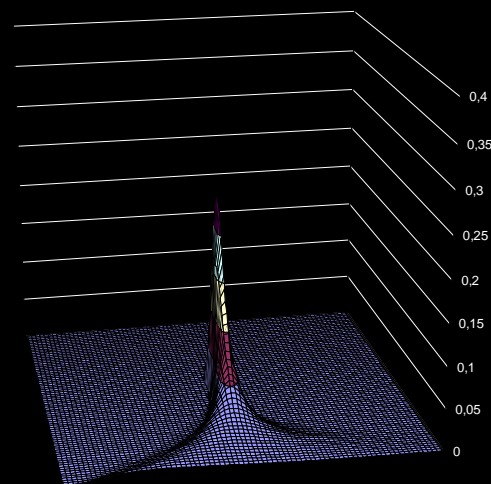
$\mu J = 2.5$



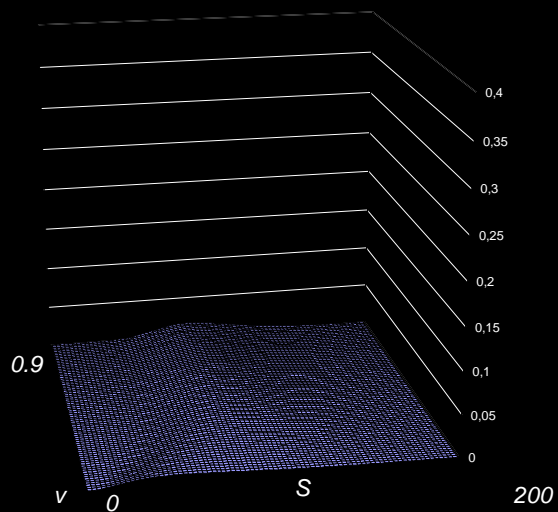
$\lambda J = 0.5$

$\eta = 0.1$

Black – Scholes Gamma



Heston Gamma



$\lambda = -2$

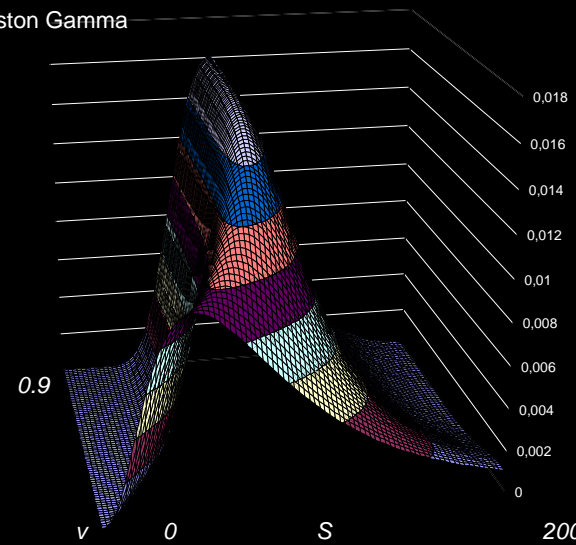
$\kappa V = 2$

$\theta V = 0.3$

$\eta V = 0.1$

$\rho = 0$

Heston Gamma



$\lambda = -2$

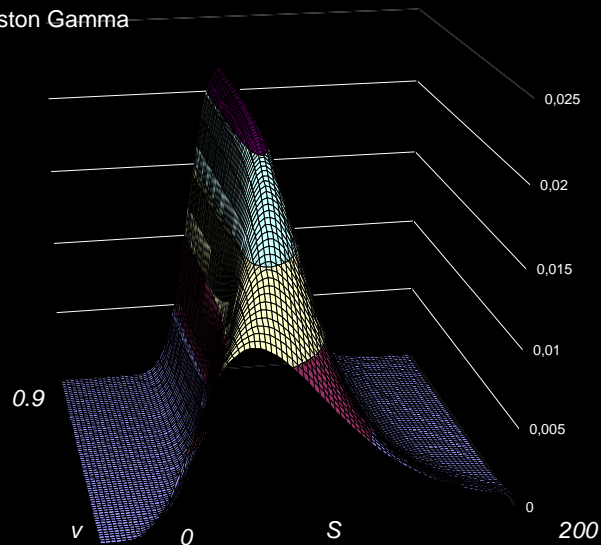
$\kappa V = 2$

$\theta V = 0.3$

$\eta V = 0.1$

$\rho = 0$

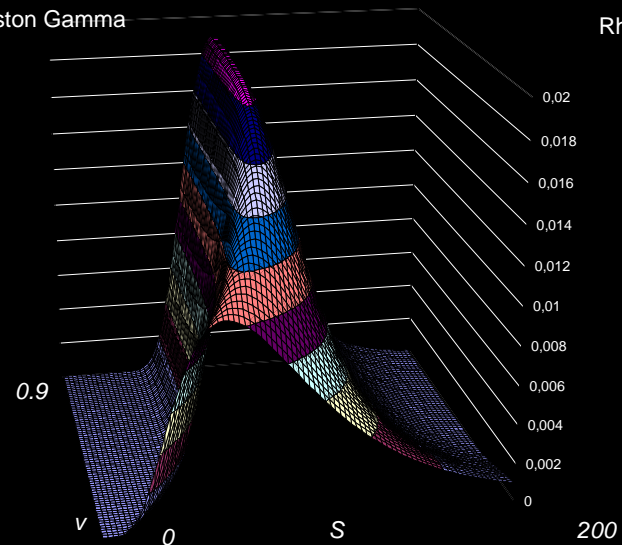
Heston Gamma



Lambda = 2 ↑

CappaV = 2
 ThetaV = 0.3
 EtaV = 0.1
 Rho = 0

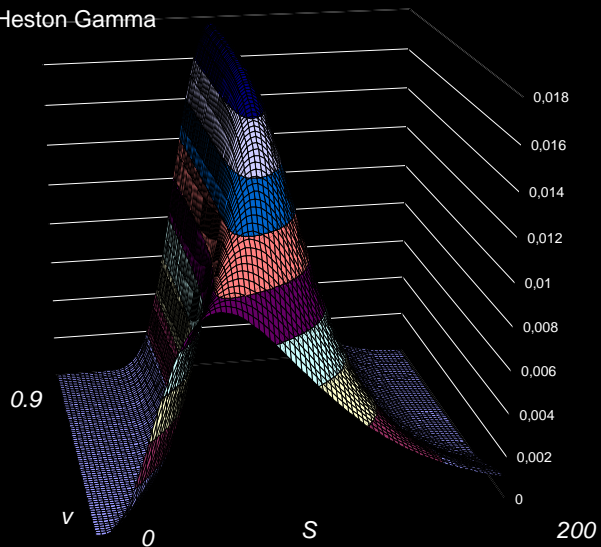
Heston Gamma



Rho = -1

CappaV = 2
 ThetaV = 0.3
 EtaV = 0.1
 Lambda = 0

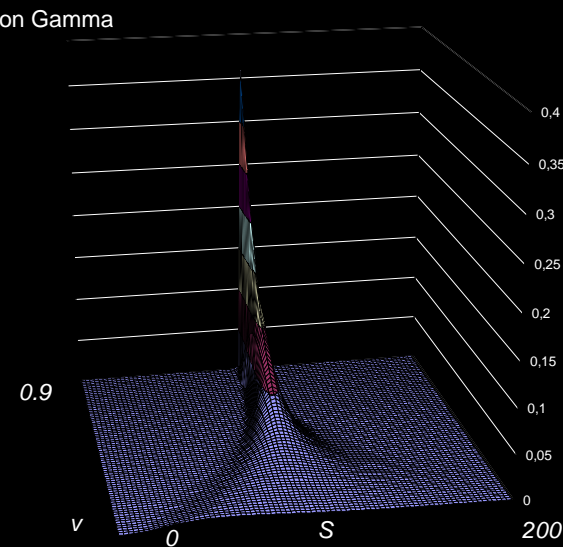
Heston Gamma



Rho = 1

CappaV = 2
 ThetaV = 0.3
 EtaV = 0.1
 Lambda = 0

Merton Gamma

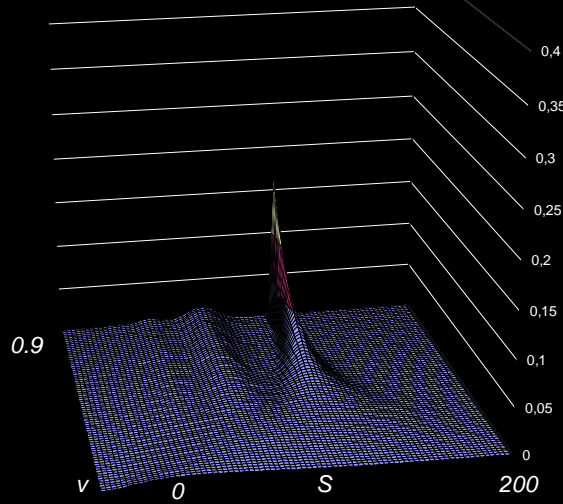


LambdaJ = 0

EtaJ = 0
 MuJ = 0

Merton Gamma

LambdaJ = 0.9

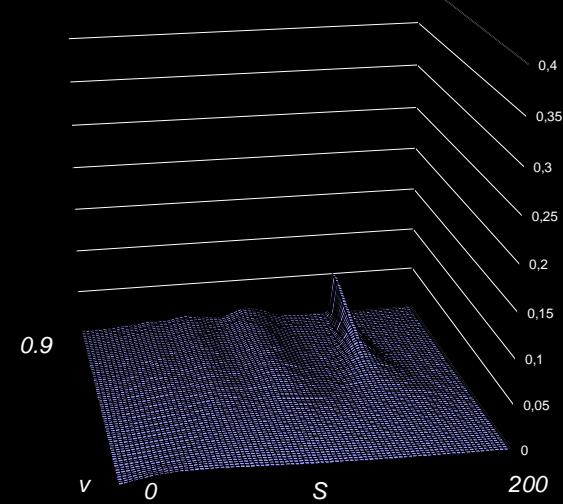


EtaJ = 0.1

MuJ = 0.5

Merton Gamma

LambdaJ = 1.8

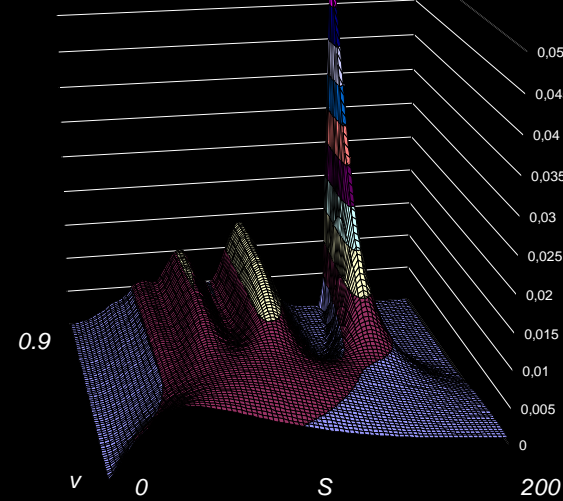


EtaJ = 0.1

MuJ = 0.5

Merton Gamma

LambdaJ = 1.8

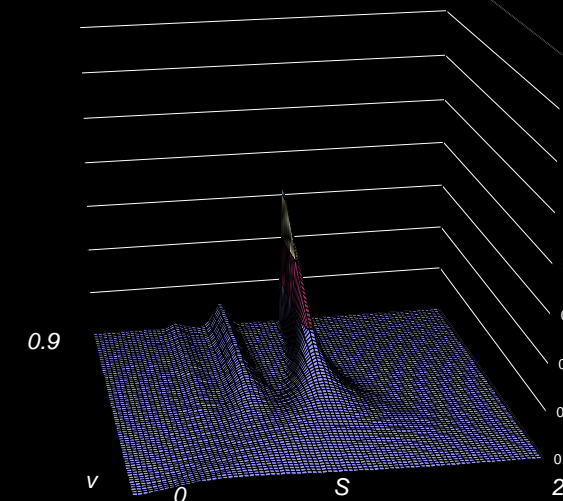


EtaJ = 0.1

MuJ = 0.5

Merton Gamma

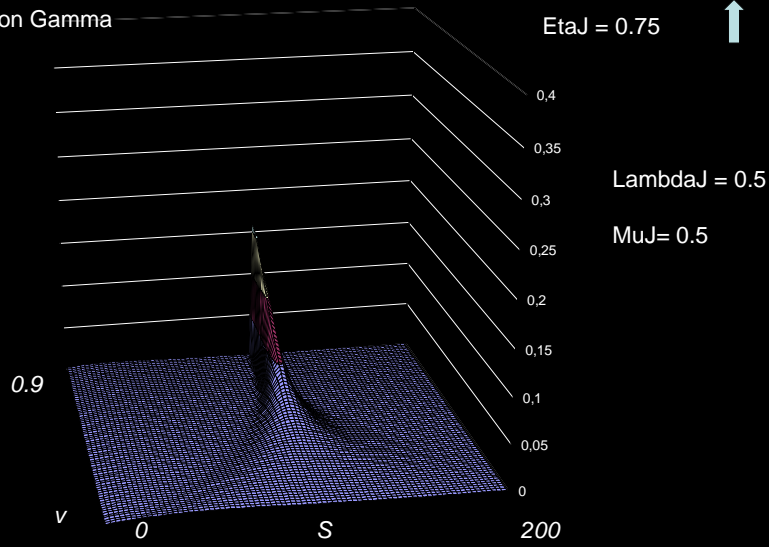
EtaJ = 0.1



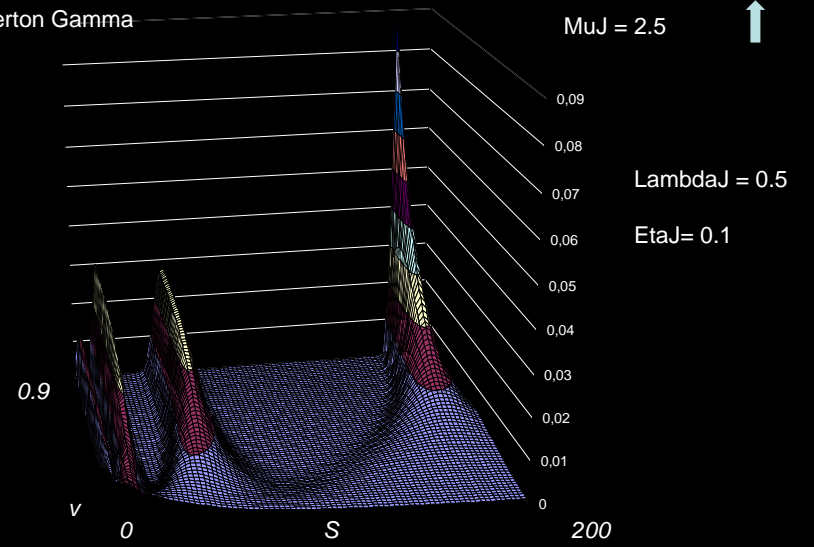
LambdaJ = 0.5

MuJ = 0.5

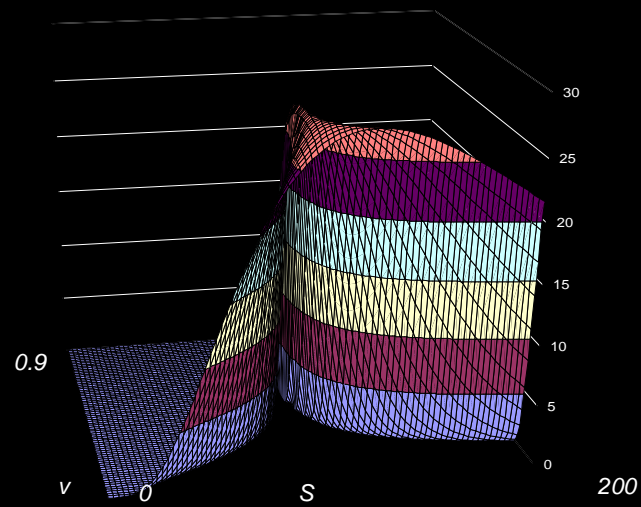
Merton Gamma



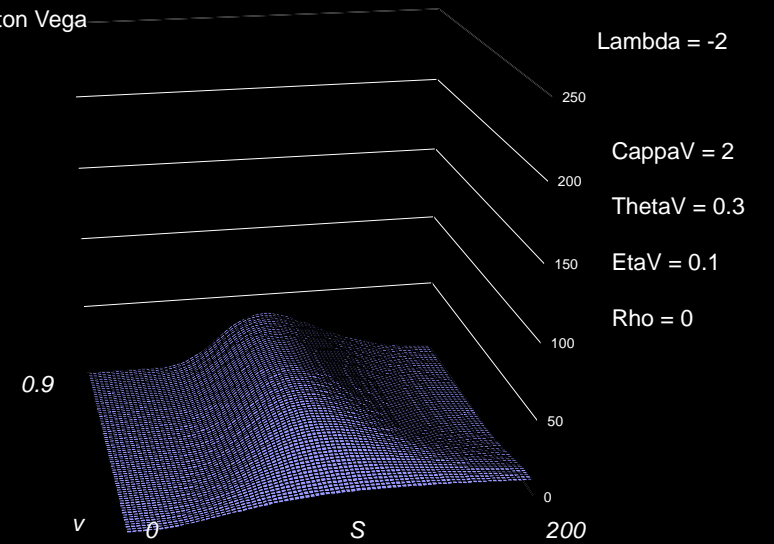
Merton Gamma



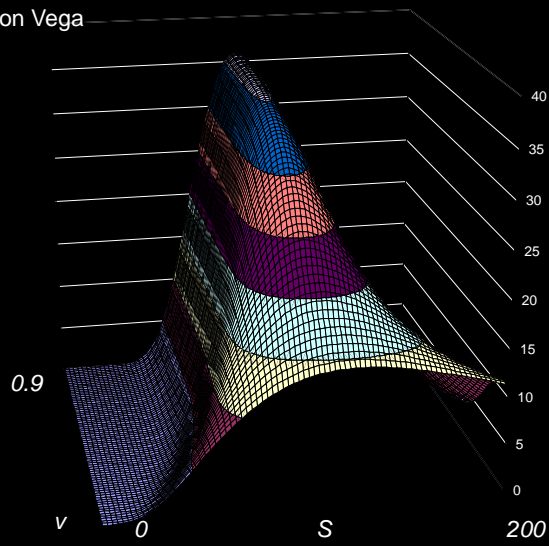
Black – Scholes Vega



Heston Vega



Heston Vega



Lambda = -2

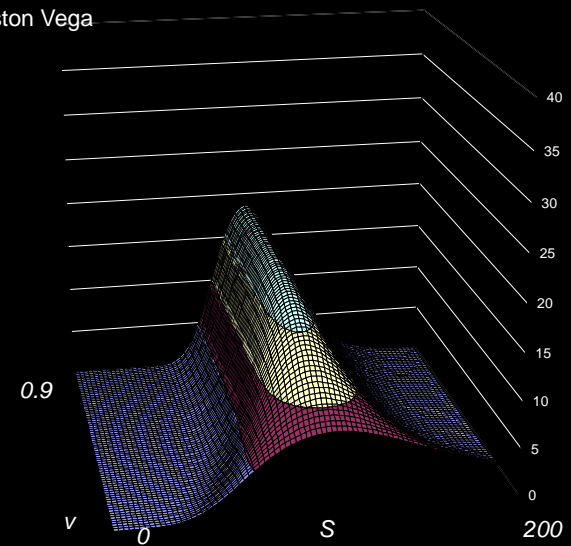
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Vega



Lambda = 2 ↑

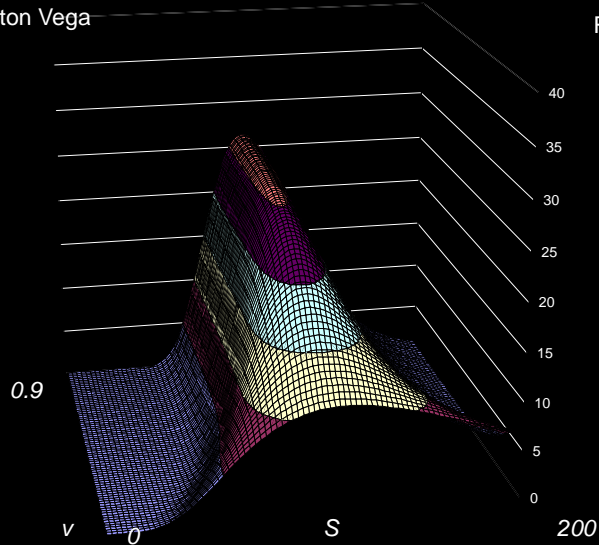
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Vega



Rho = -1

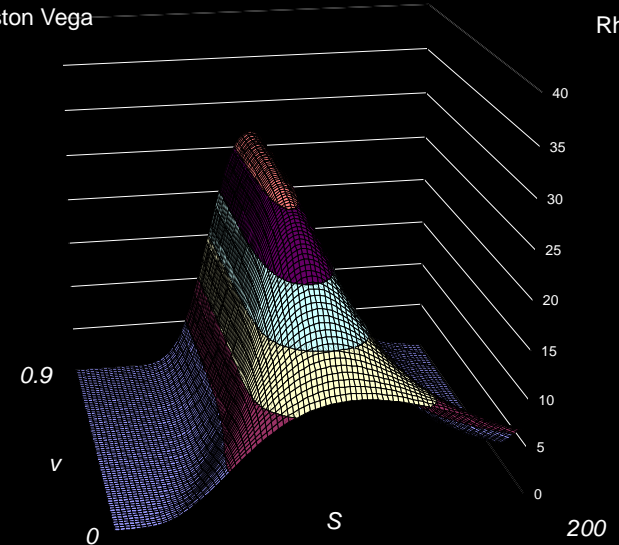
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Lambda = 0

Heston Vega



Rho = 1

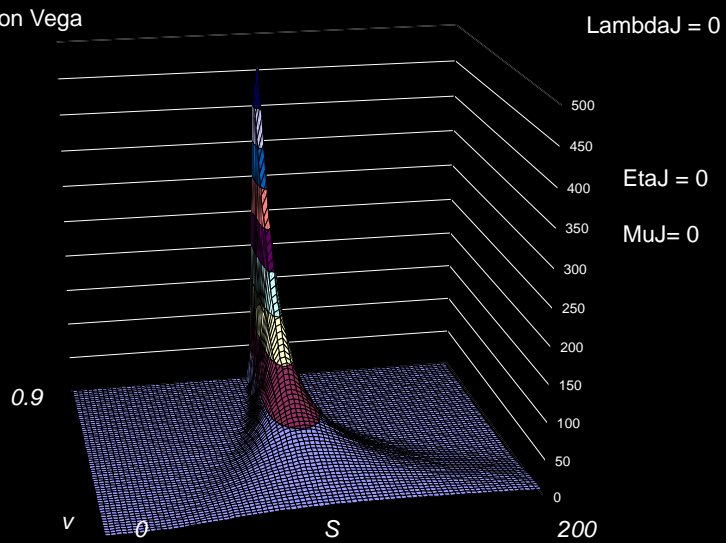
CappaV = 2

ThetaV = 0.3

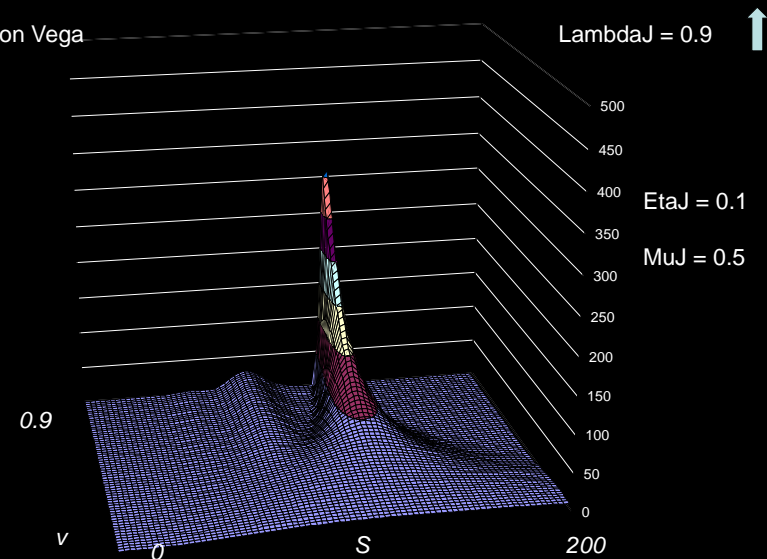
EtaV = 0.1

Lambda = 0

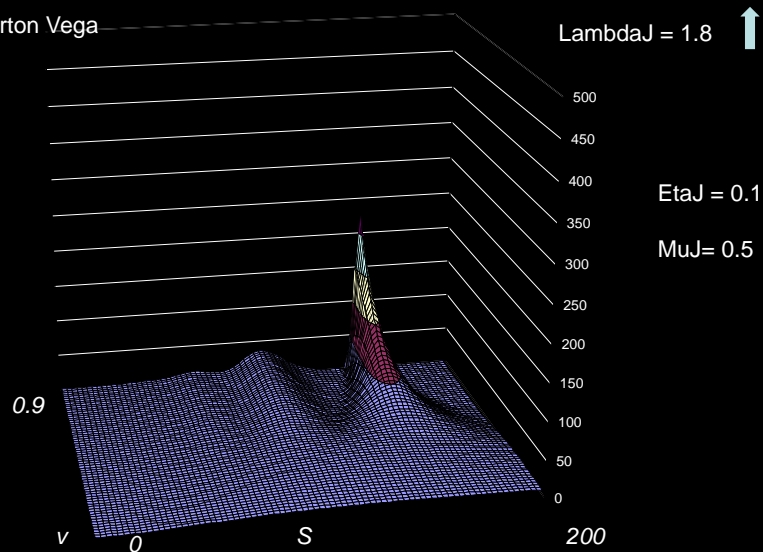
Merton Vega



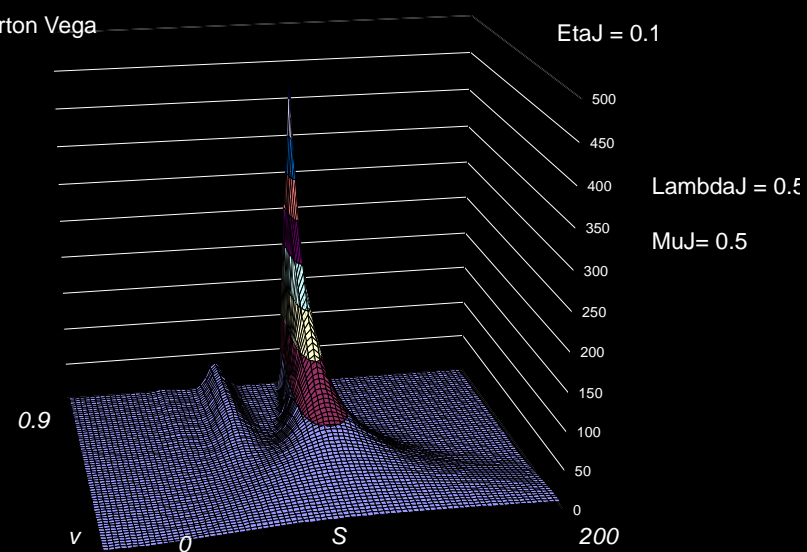
Merton Vega



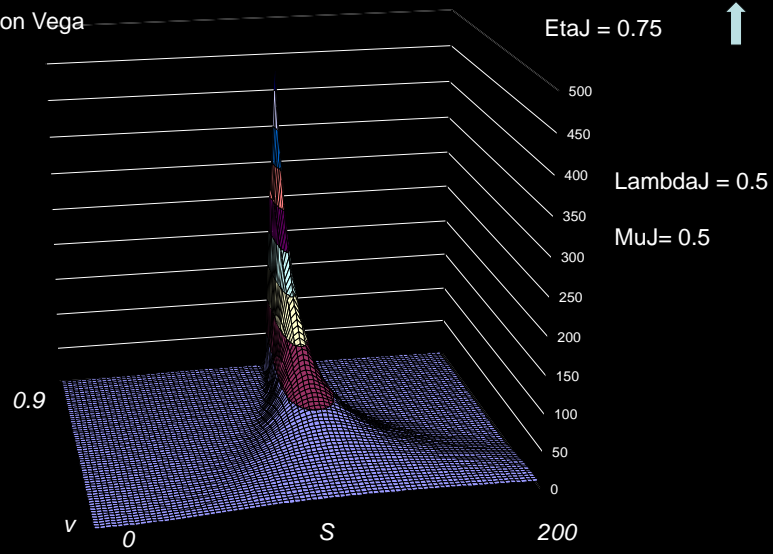
Merton Vega



Merton Vega



Merton Vega



Merton Vega

