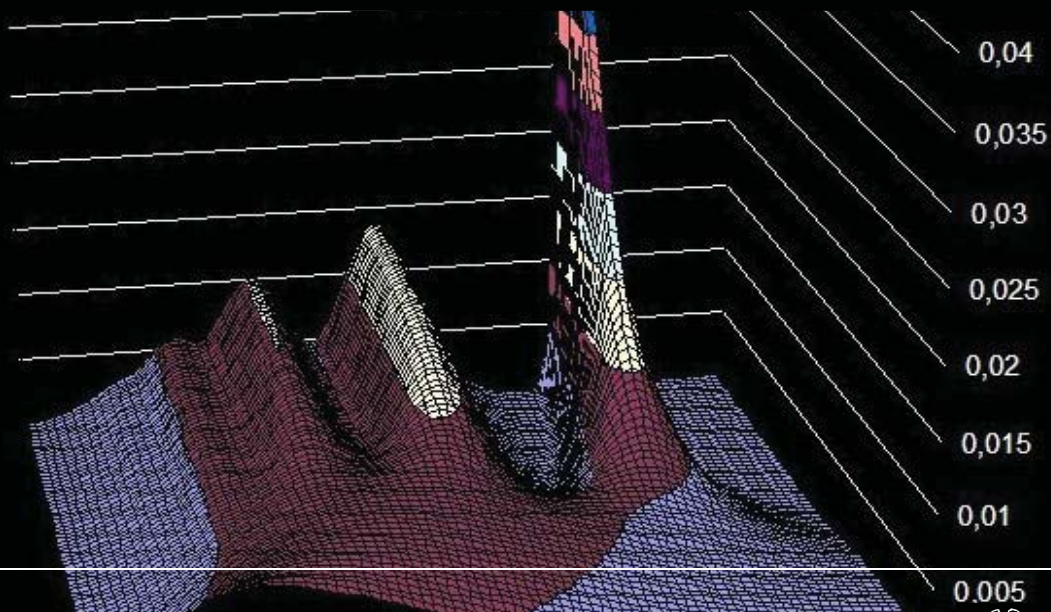


Towards a feasible calibration of Affine Jump Diffusion Models

Greeks Behavior and Inverse Fourier Transform Stability



Marcello Minenna - Paolo Verzella

XXIX AMASES Conference – Sept. 15, 2005 - Palermo



CONSOB

Syllabus of the presentation

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



CONSOB

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

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

where

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

under different martingale measures

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

can be

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

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

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

requires

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

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

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



a closed formula for AJD models

How to compute: $P_1(\Theta), P_2(\Theta)$



Quadrature Algorithm
FT - Q



Old FT - Q



New FT - Q



Fast Fourier Transform
FFT



Newton Rapson Algorithm



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



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



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

$$P_1(\Theta), P_2(\Theta)$$



Quadrature Algorithm
New FT - Q

Pros (+)

STABILITY

ACCURACY

Cons (-)

SPEED

$$P_1(\Theta), P_2(\Theta)$$



Fast Fourier Transform
FFT

Simpson algorithm

Applied to the equivalent formula via a recombining FFT parameters



$$C_t = \frac{e^{-\alpha k}}{\pi} \int_0^\infty e^{-i\phi \ln[K]} \tilde{f}_j(\phi) d\phi$$

$$P_1(\Theta), P_2(\Theta)$$



Fast Fourier Transform
FFT

Pros (+)

SPEED

FASTER

(up to 20 times the quadrature algorithms)

Cons (-)

STABILITY

EXPLOSION

for OTM options

ACCURACY

ARBITRARY

choice of the recombinant FFT parameters

Syllabus of the presentation

- Review of Fourier Methods in Option Pricing
- **Calibration Procedure 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 Trasform
FFT



Old FT - Q



New FT - Q

The Calibration Procedure and Performance

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

Pros (+)

Cons (-)

STABILITY
ACCURACY
SPEED

The Calibration Procedure and Performance

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

through

Fast Fourier Trasform
FFT

Pros (+)

SPEED

Cons (-)

STABILITY
ACCURACY



The Calibration Procedure and Performance

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

through

Quadrature Algorithm
New FT - Q

Pros (+)

STABILITY
ACCURACY
SPEED

Cons (-)



The Calibration Procedure and Performance

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

FFT	Heston Model	Merton Model	BCC Model
	7.2 sec.	10.6 sec.	18.4 sec.
NEW FT - Q	Heston Model	Merton Model	BCC Model
	23.1 sec.	29.5 sec.	48.7 sec.
OLD FT – Q	Heston Model	Merton Model	BCC Model
	331.6 sec.	410.9 sec.	688.5 sec.

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

Syllabus of the presentation

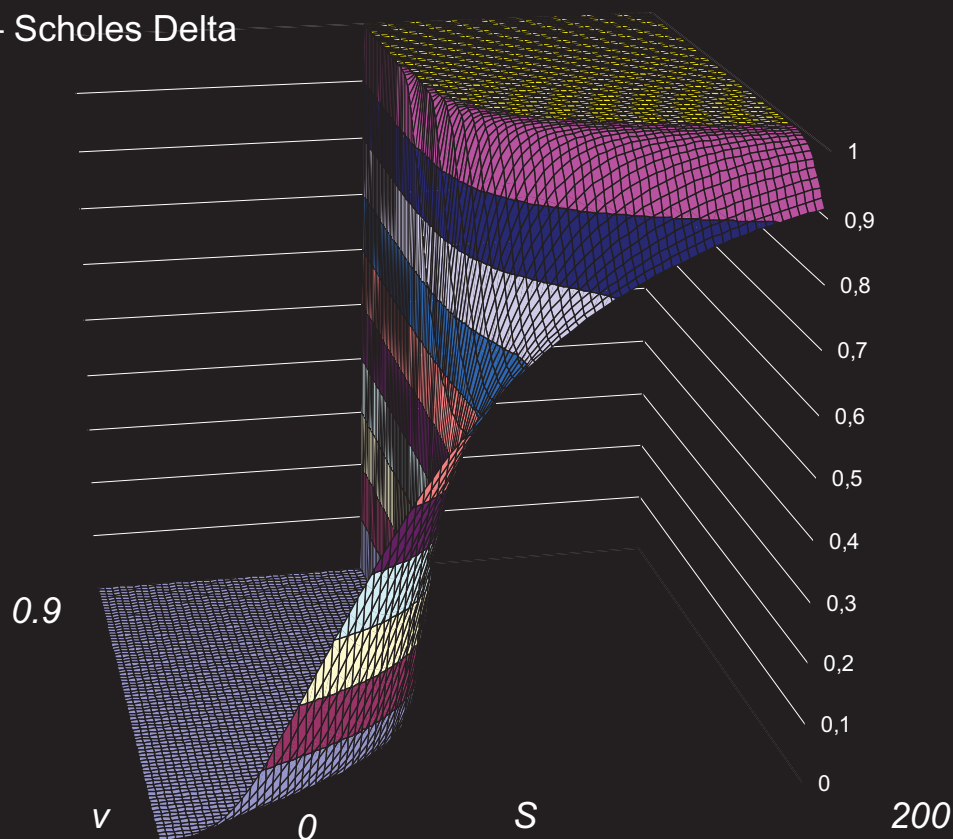
- Review of Fourier Methods in Option Pricing
- Calibration Procedure and Performance
- **Greek Behaviour of New FT-Q**

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

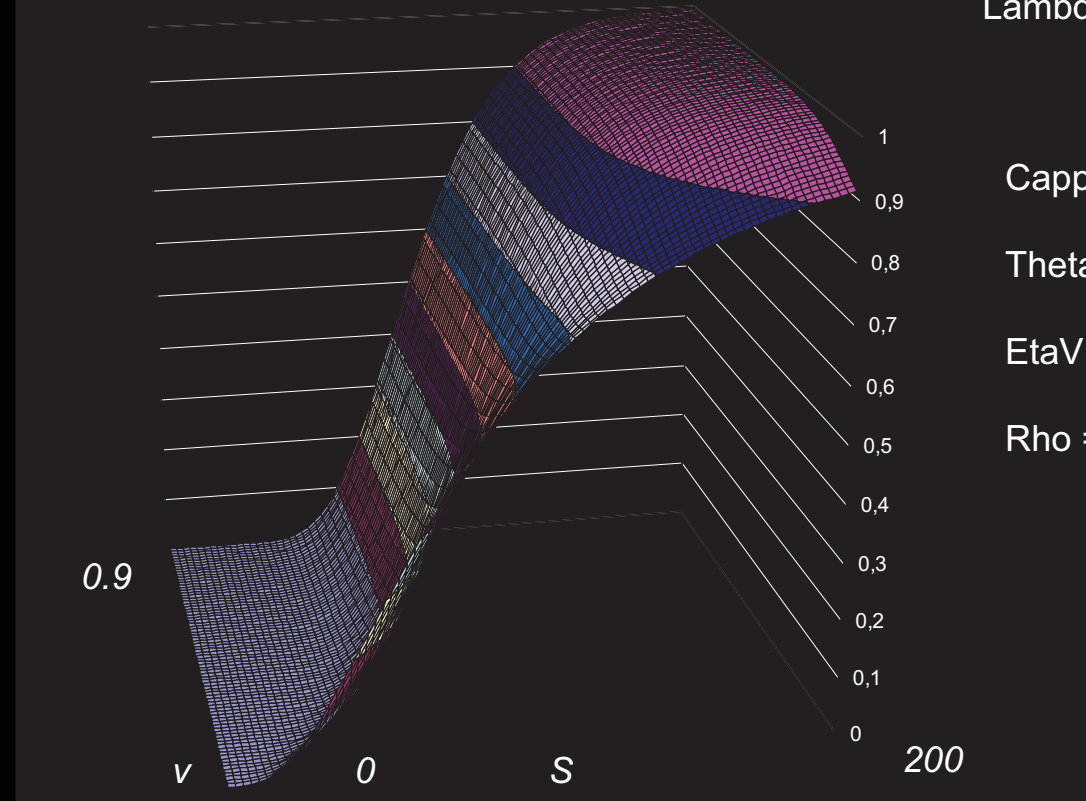
Infact, in an AJD setting the Greeks are available in closed form

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

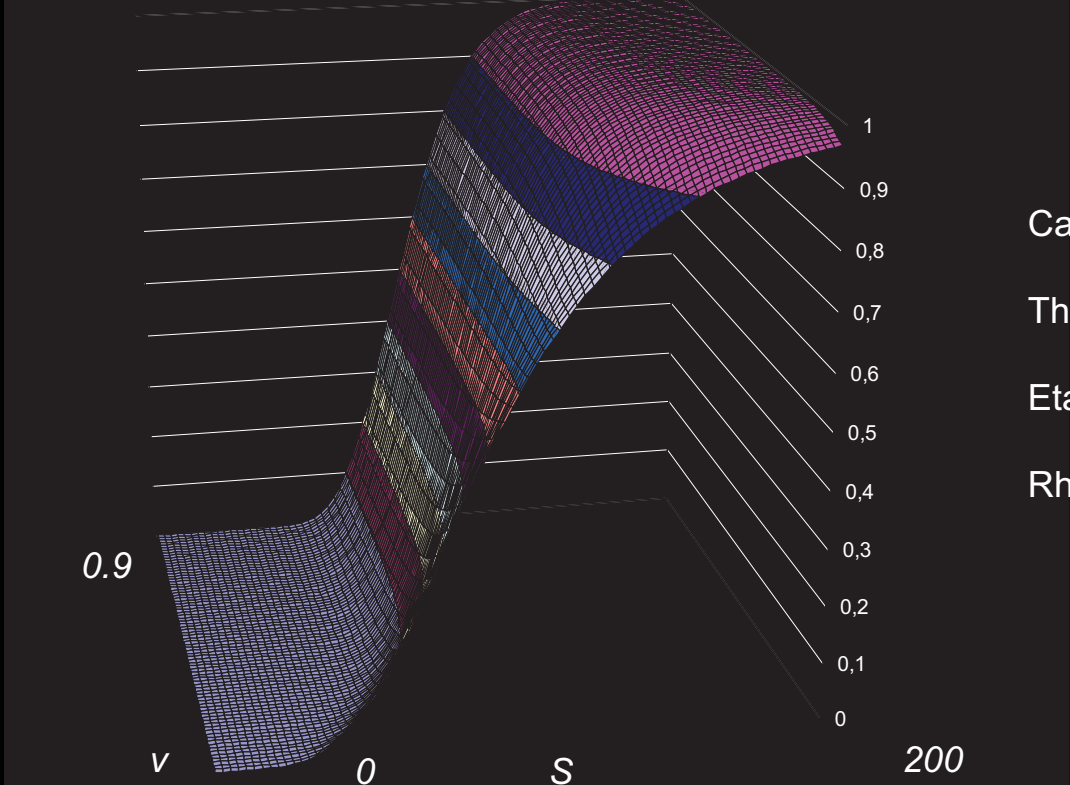
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Delta



Lambda = 2

CappaV = 2

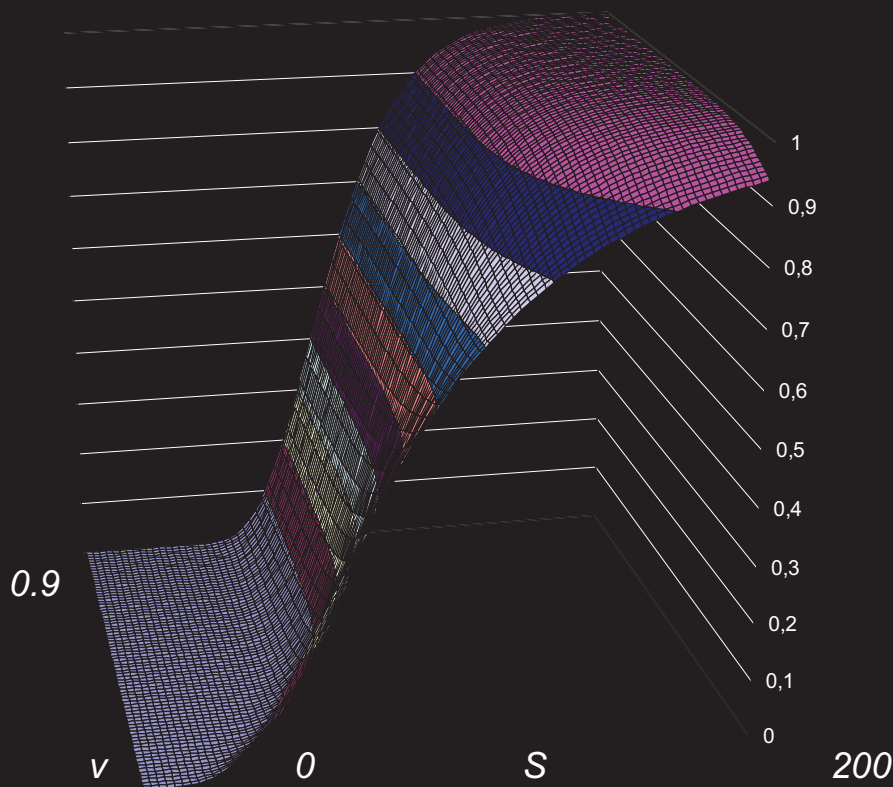
ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Delta

$\rho = -1$



$\kappa_V = 2$

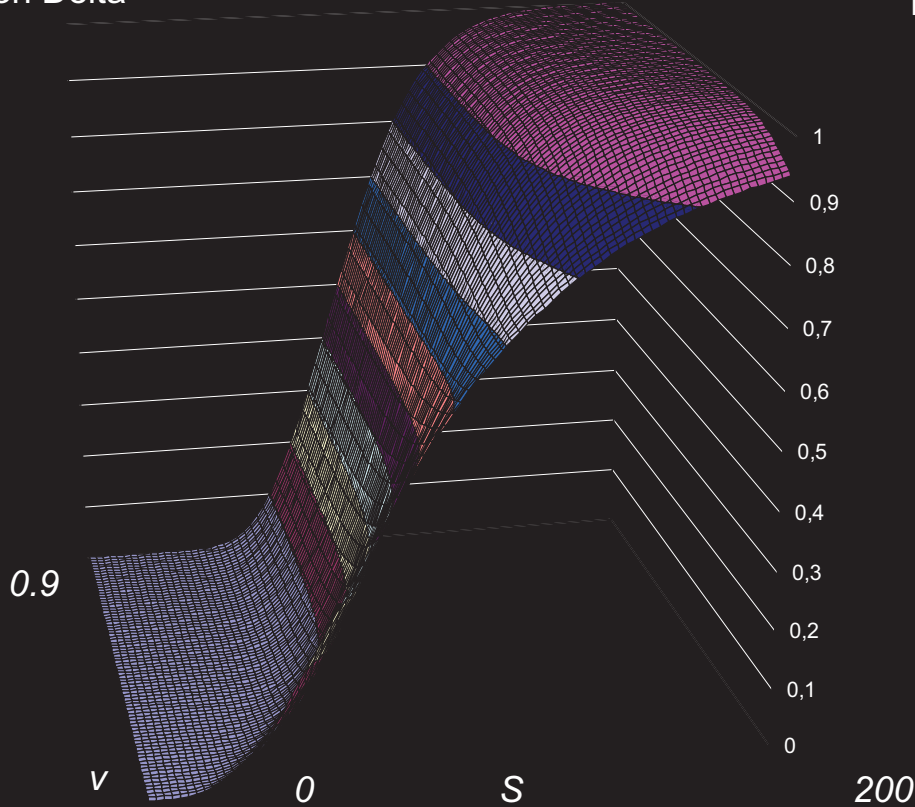
$\theta_V = 0.3$

$\eta_V = 0.1$

$\lambda = 0$

Heston Delta

$\rho = 1$



$\kappa_V = 2$

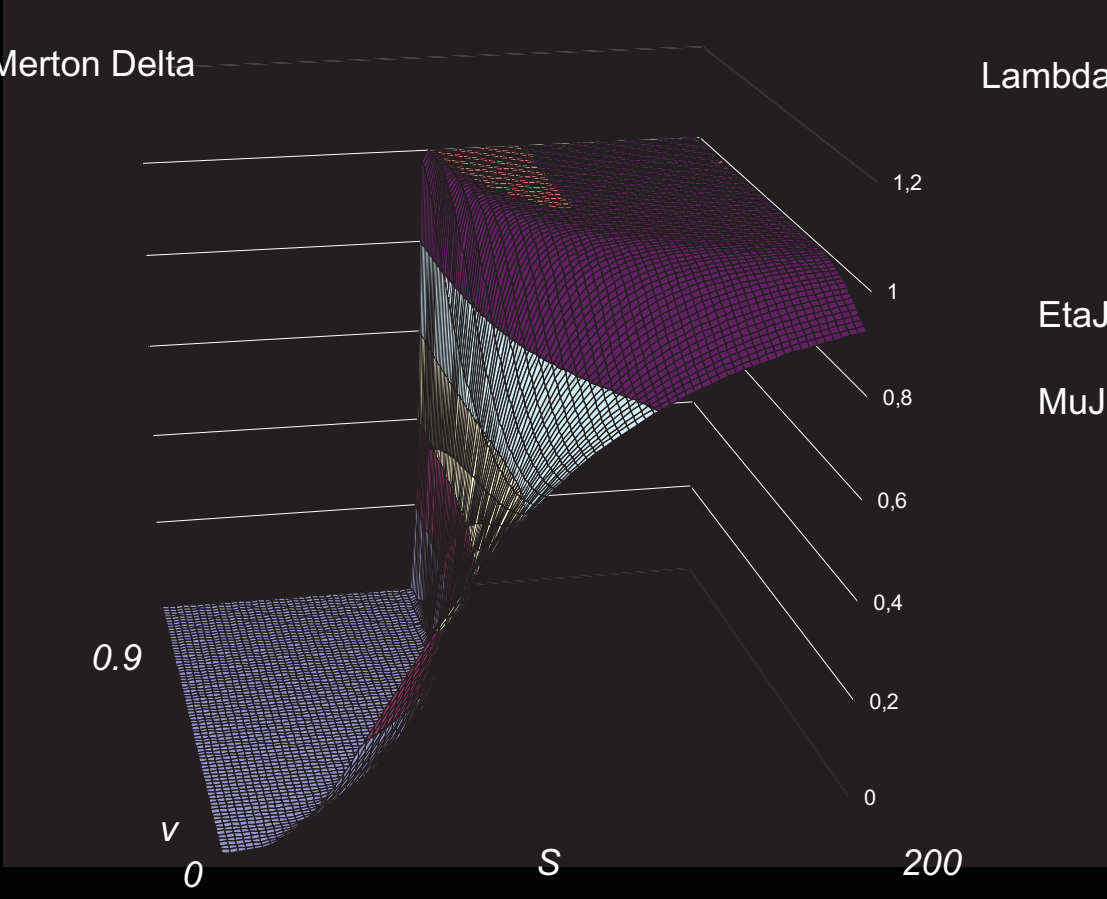
$\theta_V = 0.3$

$\eta_V = 0.1$

$\lambda = 0$

Merton Delta

$\text{LambdaJ} = 0$

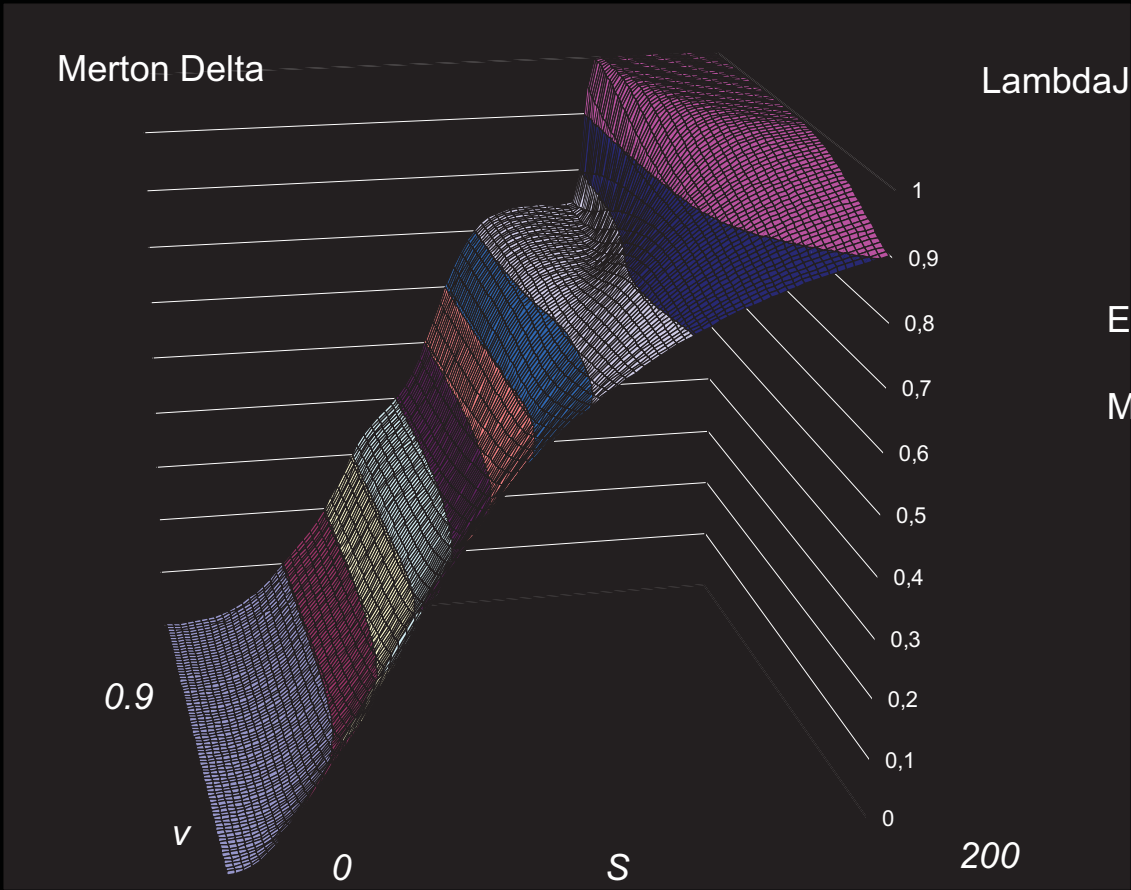


$\text{EtaJ} = 0$

$\text{MuJ} = 0$

Merton Delta

$\text{LambdaJ} = 1.8$



$\text{EtaJ} = 0.1$

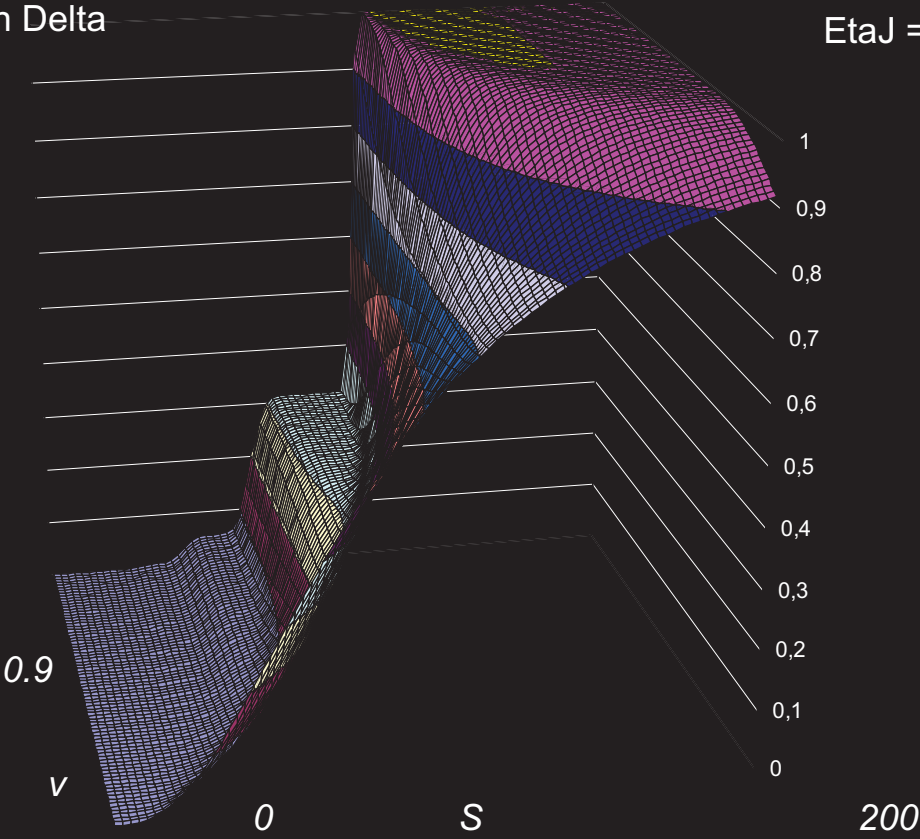
$\text{MuJ} = 0.5$

Merton Delta

EtaJ = 0.1

LambdaJ = 0.5

MuJ= 0.5

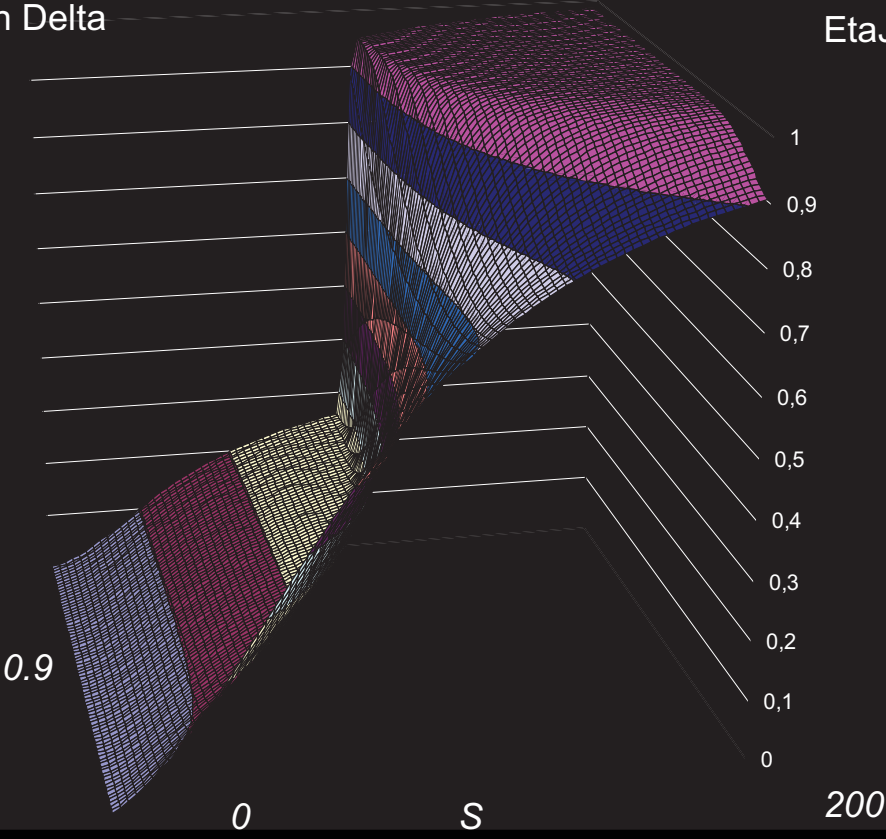


Merton Delta

EtaJ = 0.75

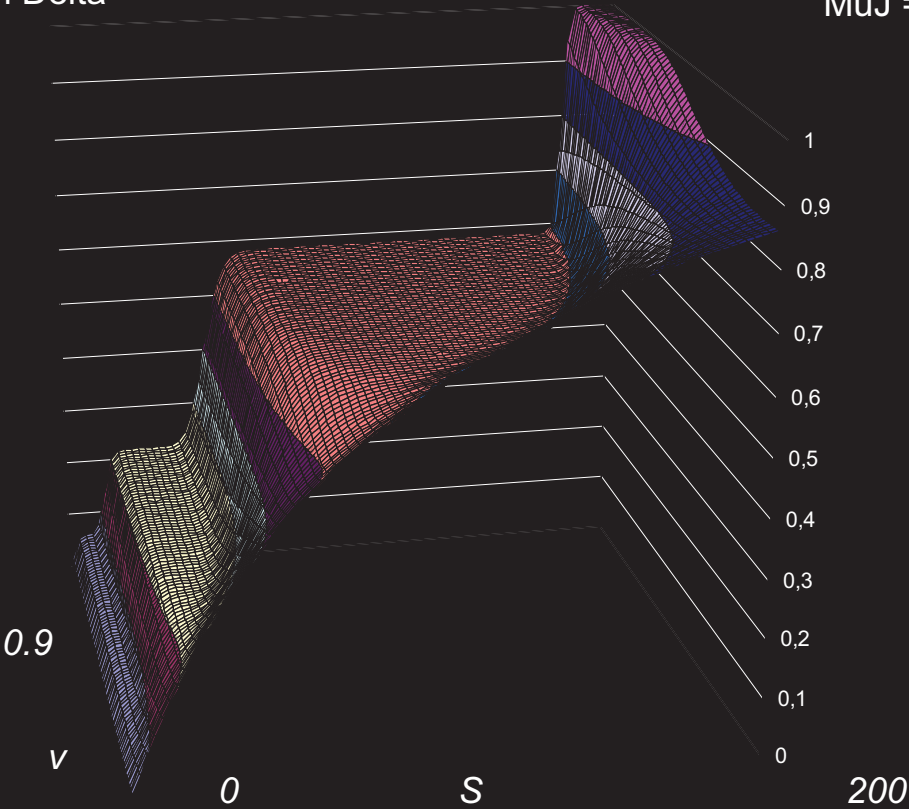
LambdaJ = 0.5

MuJ= 0.5



Merton Delta

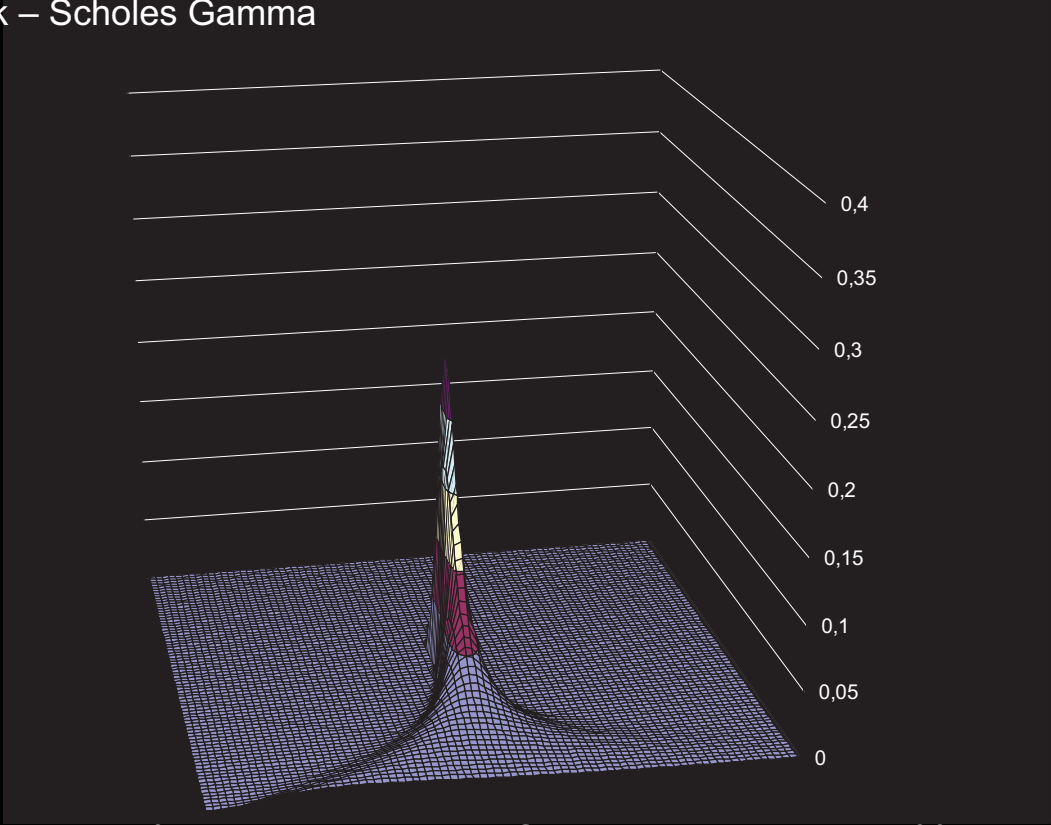
MuJ = 2.5



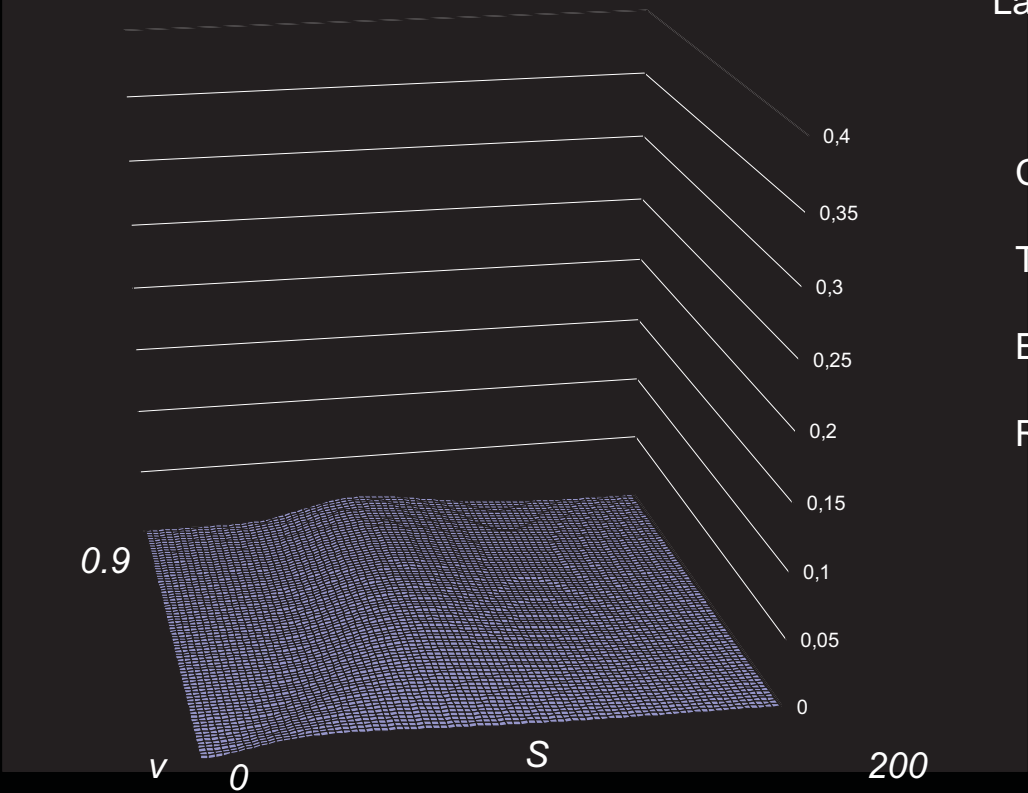
LambdaJ = 0.5

Eta= 0.1

Black – Scholes Gamma



Heston Gamma



$\Lambda = -2$

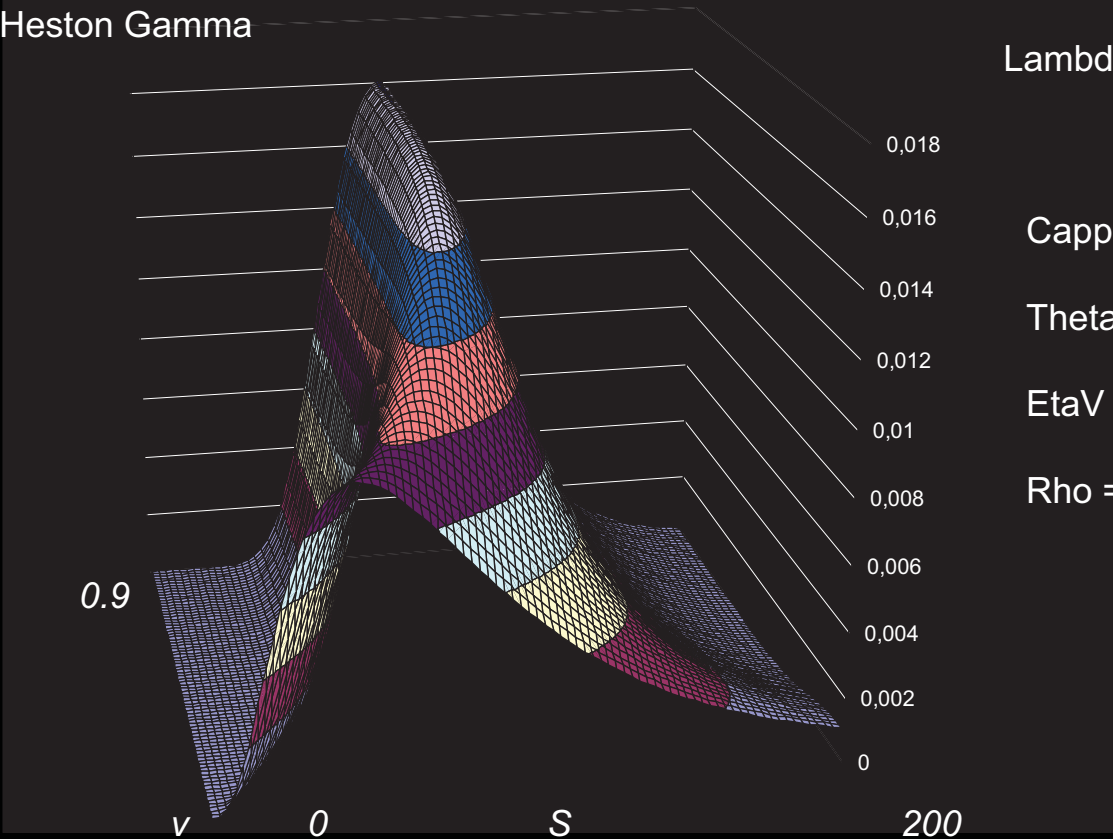
$\kappa_V = 2$

$\theta_V = 0.3$

$\eta_V = 0.1$

$\rho = 0$

Heston Gamma



$\Lambda = -2$

$\kappa_V = 2$

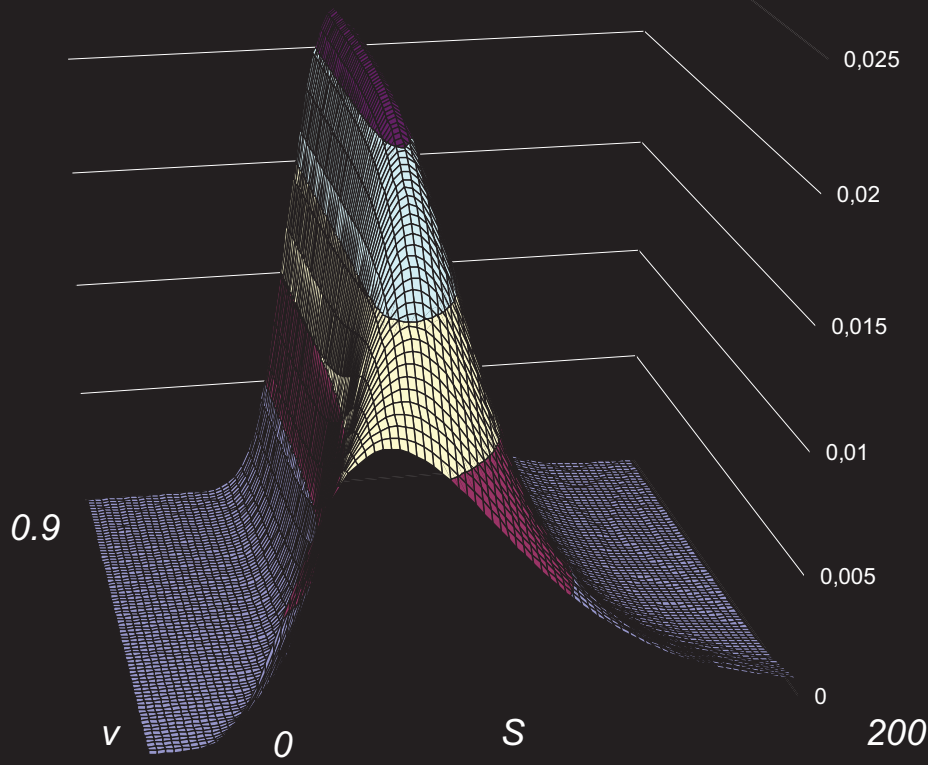
$\theta_V = 0.3$

$\eta_V = 0.1$

$\rho = 0$

Heston Gamma

Lambda = 2 



CappaV = 2

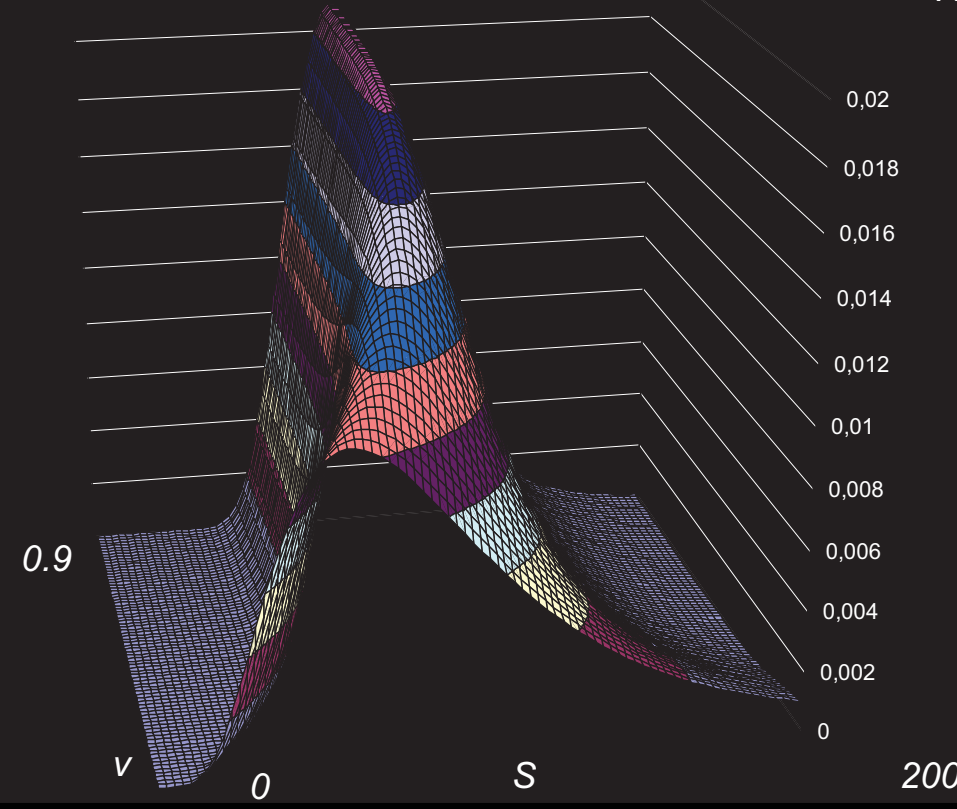
ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Gamma

Rho = -1



CappaV = 2

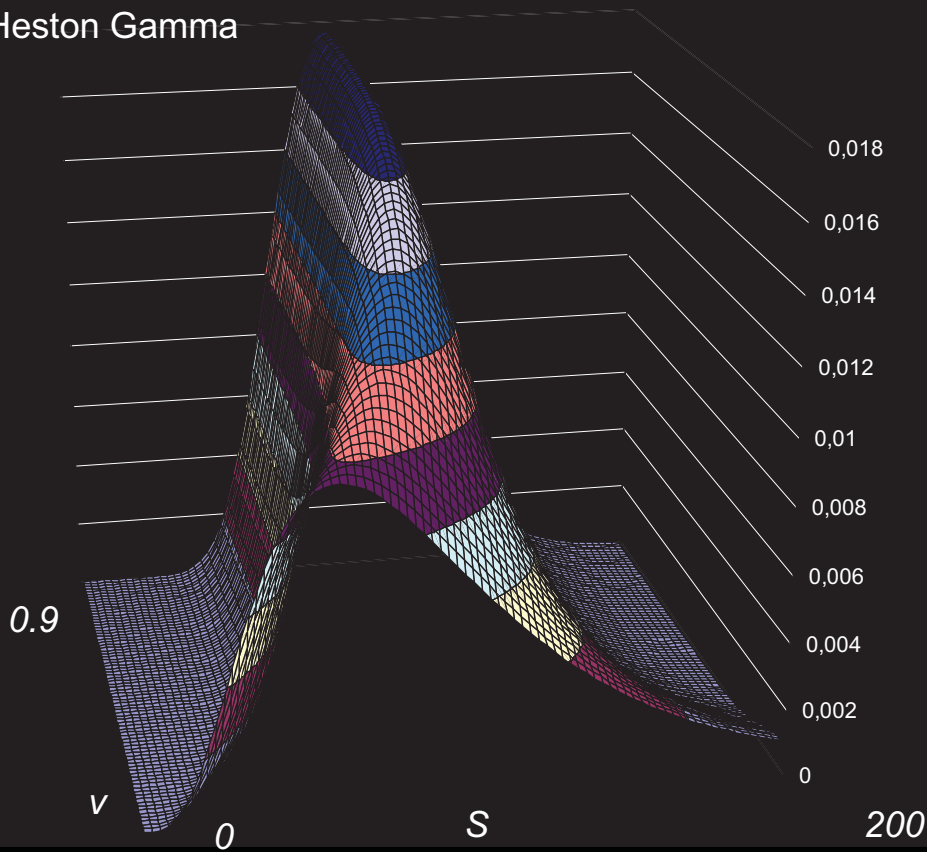
ThetaV = 0.3

EtaV = 0.1

Lambda = 0

Heston Gamma

$\rho = 1$



$\text{CappaV} = 2$

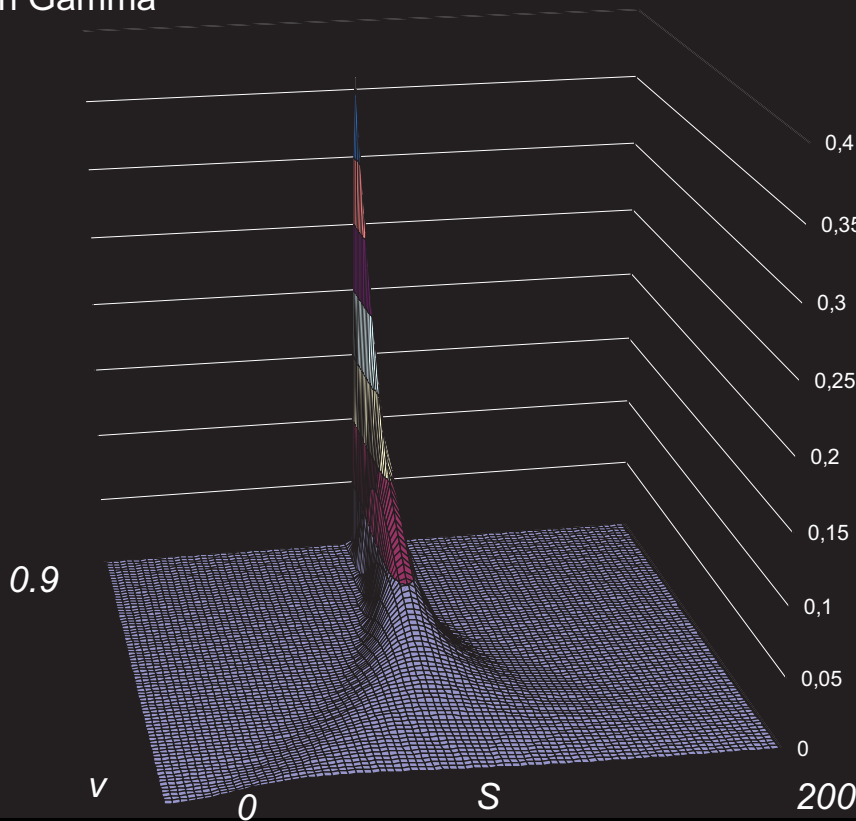
$\text{ThetaV} = 0.3$

$\text{EtaV} = 0.1$

$\text{Lambda} = 0$

Merton Gamma

$\text{LambdaJ} = 0$

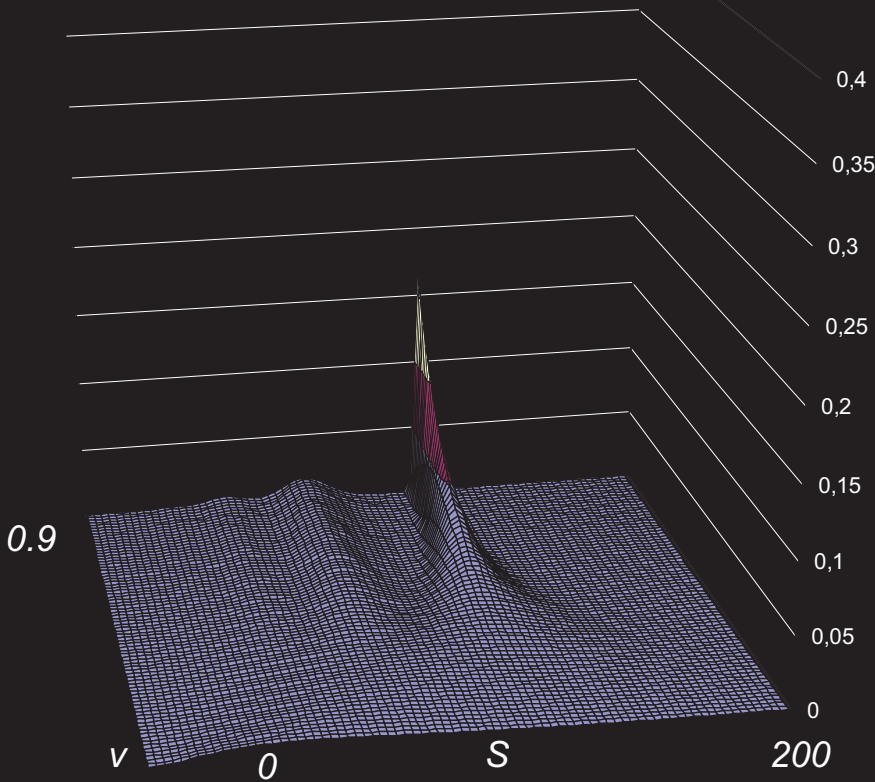


$\text{EtaJ} = 0$

$\text{MuJ} = 0$

Merton Gamma

LambdaJ = 0.9



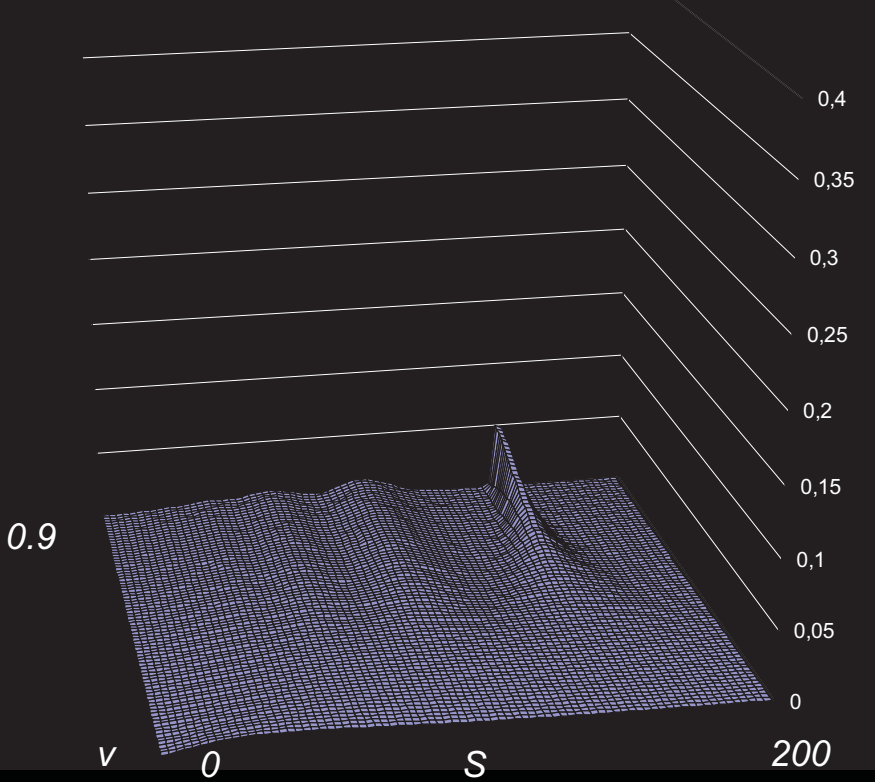
EtaJ = 0.1

MuJ = 0.5



Merton Gamma

LambdaJ = 1.8



EtaJ = 0.1

MuJ = 0.5

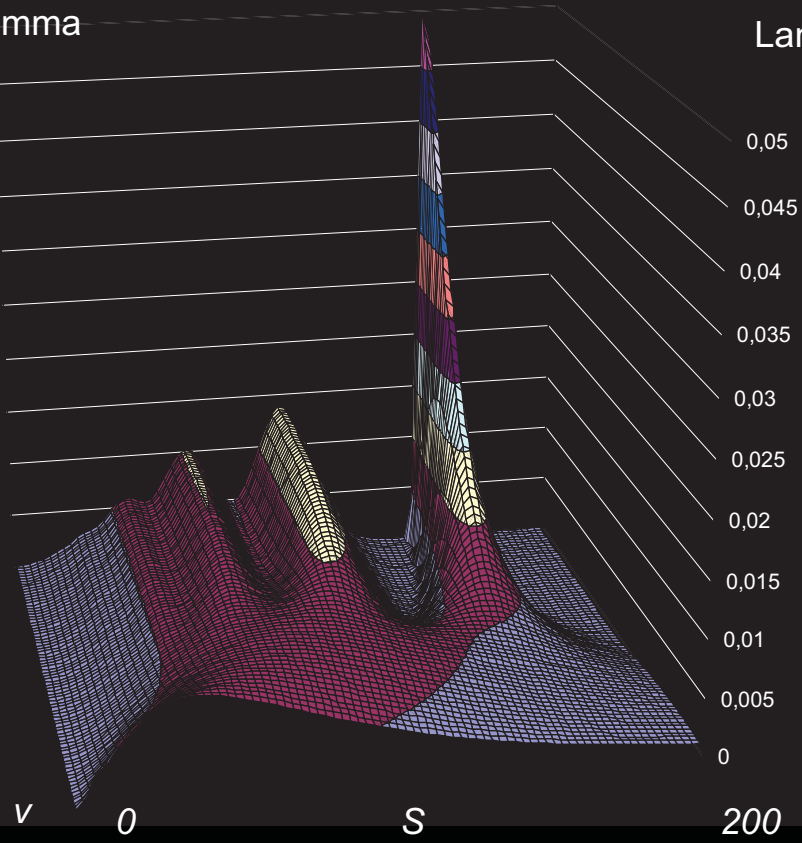


Merton Gamma

LambdaJ = 1.8



0.9



EtaJ = 0.1

MuJ = 0.5



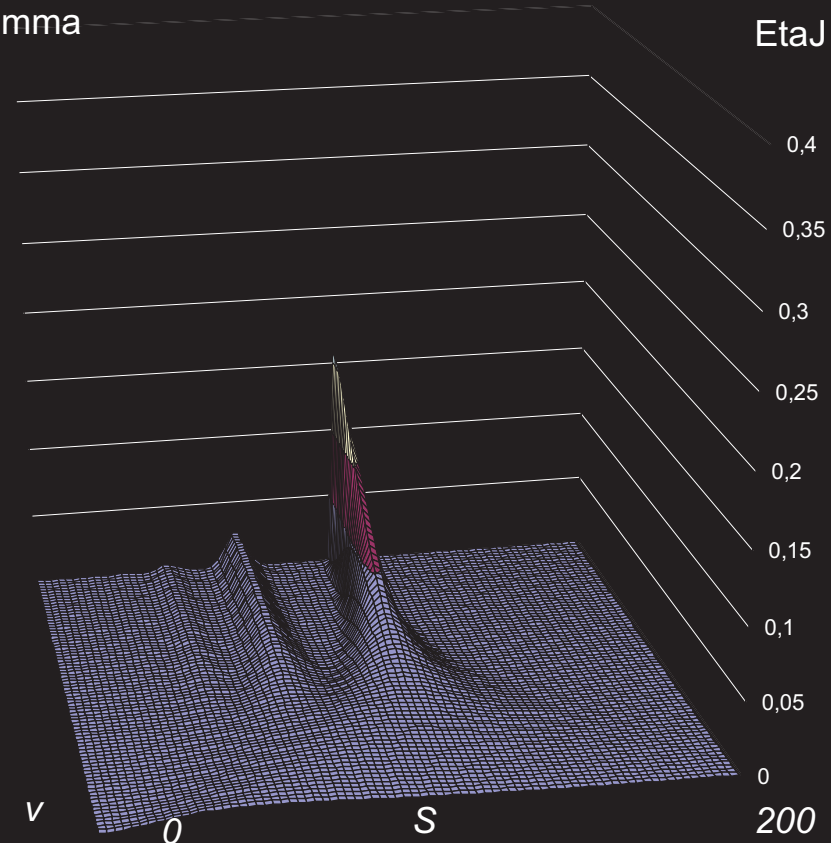
Merton Gamma

EtaJ = 0.1

LambdaJ = 0.5

MuJ = 0.5

0.9



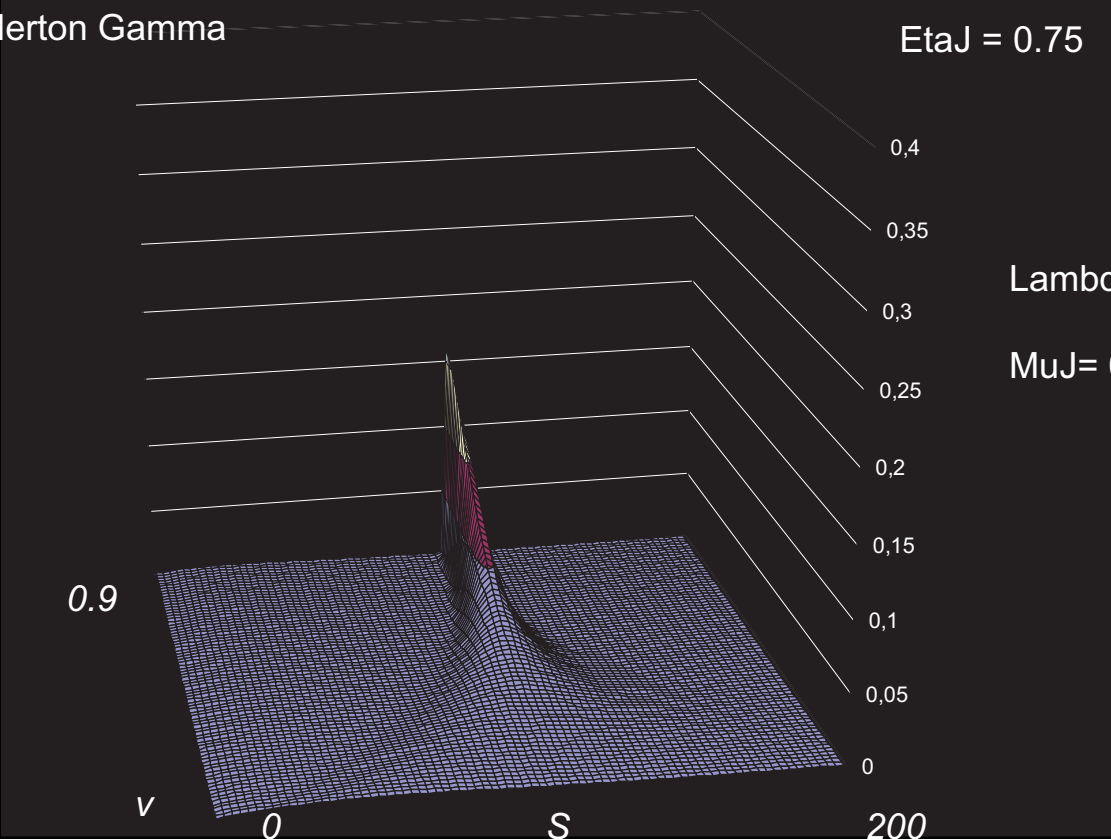
Merton Gamma

$\text{EtaJ} = 0.75$



$\text{LambdaJ} = 0.5$

$\text{MuJ} = 0.5$



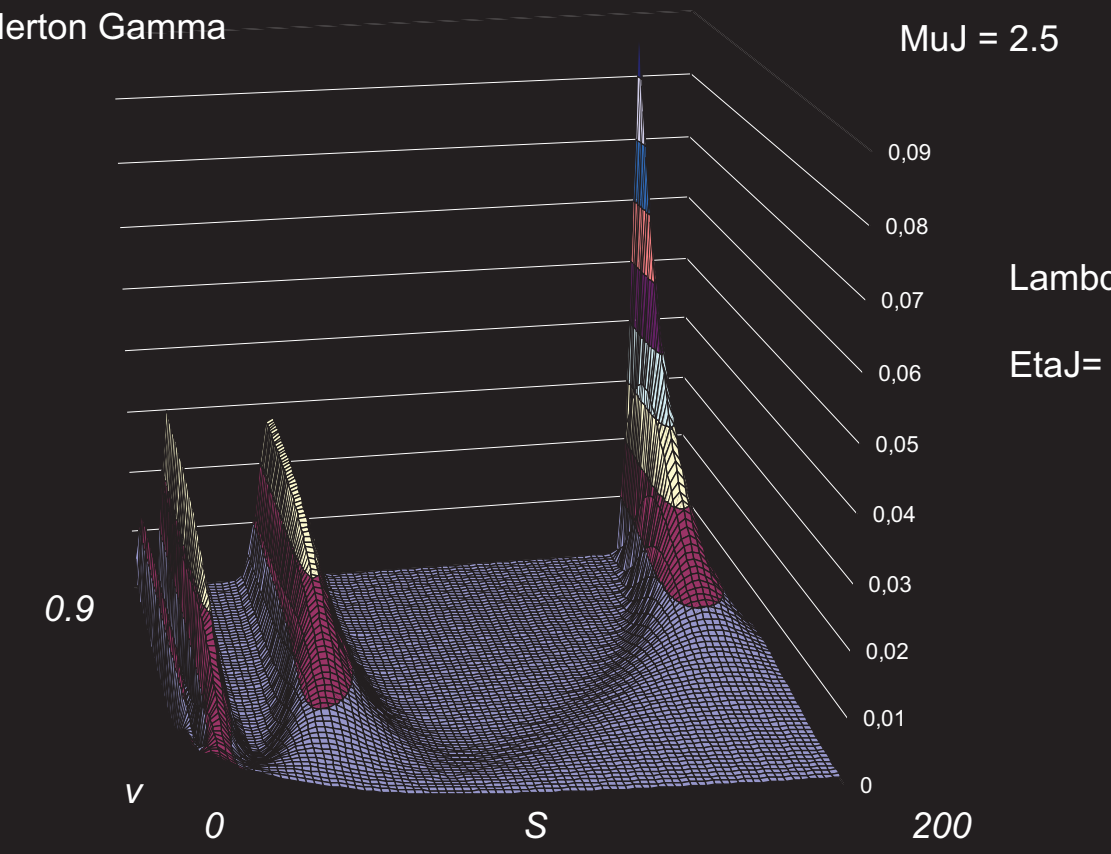
Merton Gamma

$\text{MuJ} = 2.5$

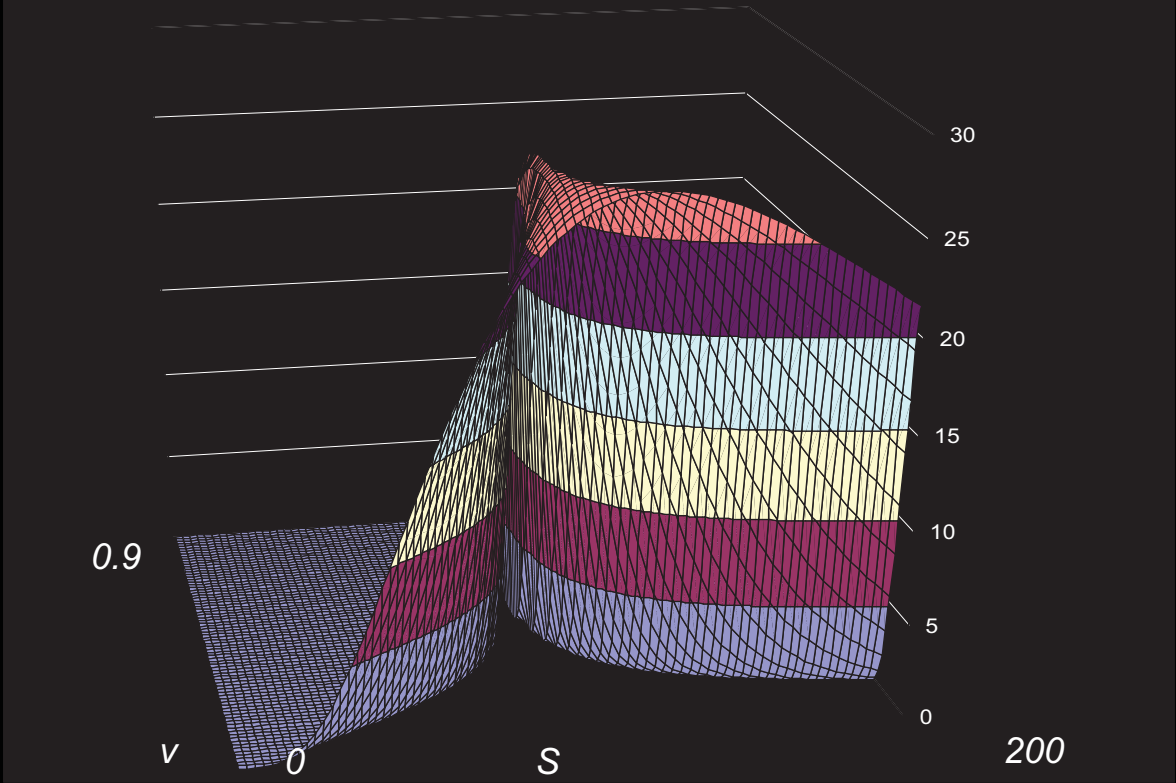


$\text{LambdaJ} = 0.5$

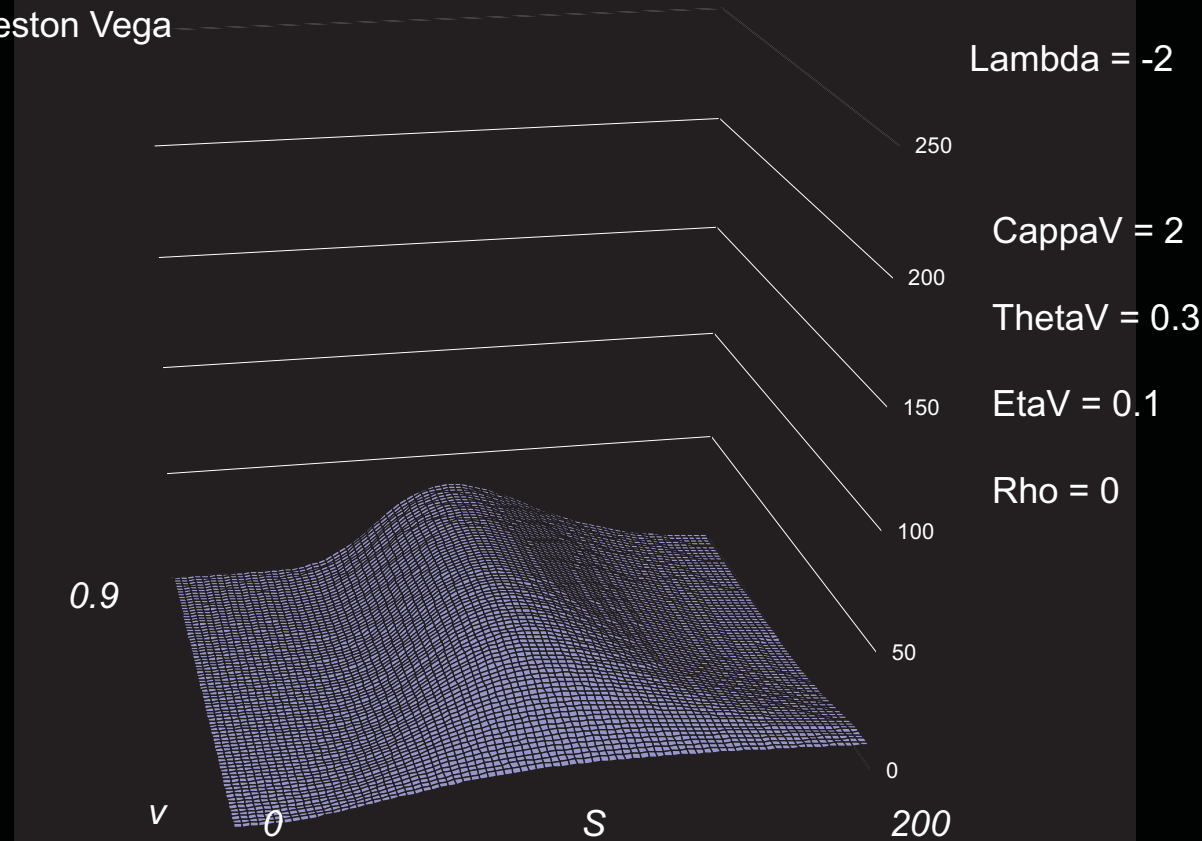
$\text{EtaJ} = 0.1$



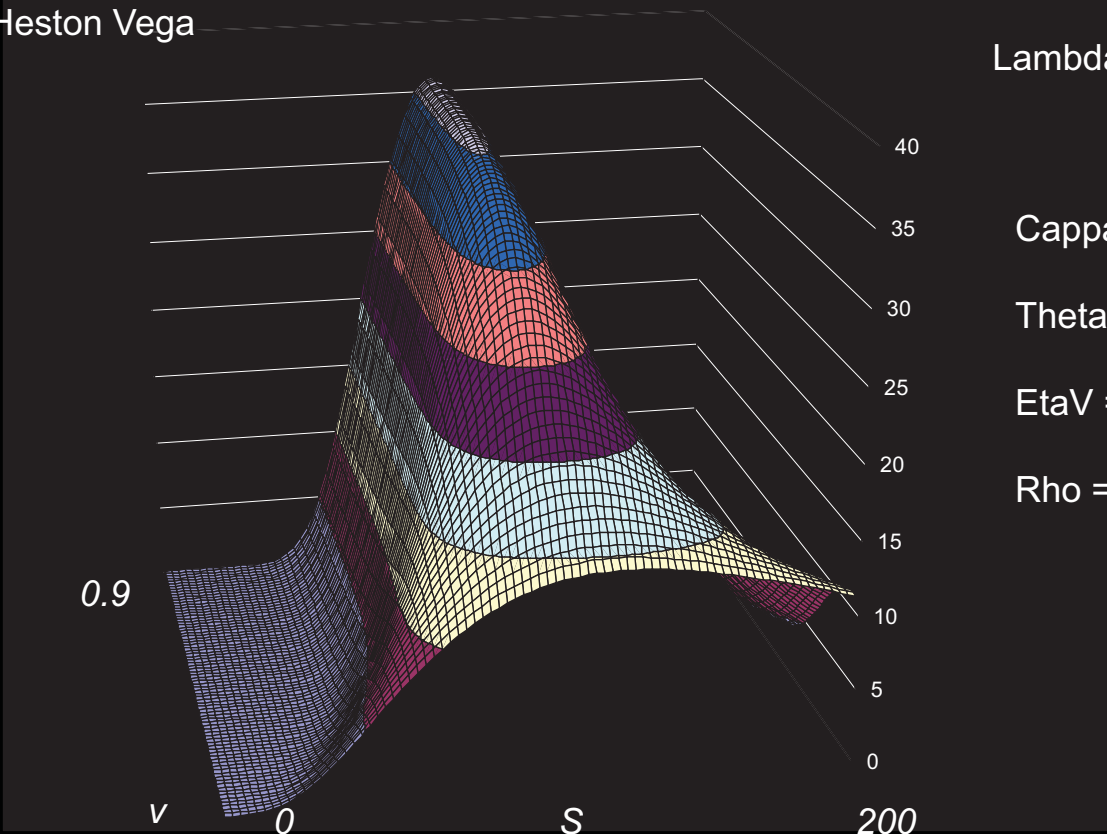
Black – Scholes Vega



Heston Vega



Heston Vega



Lambda = -2

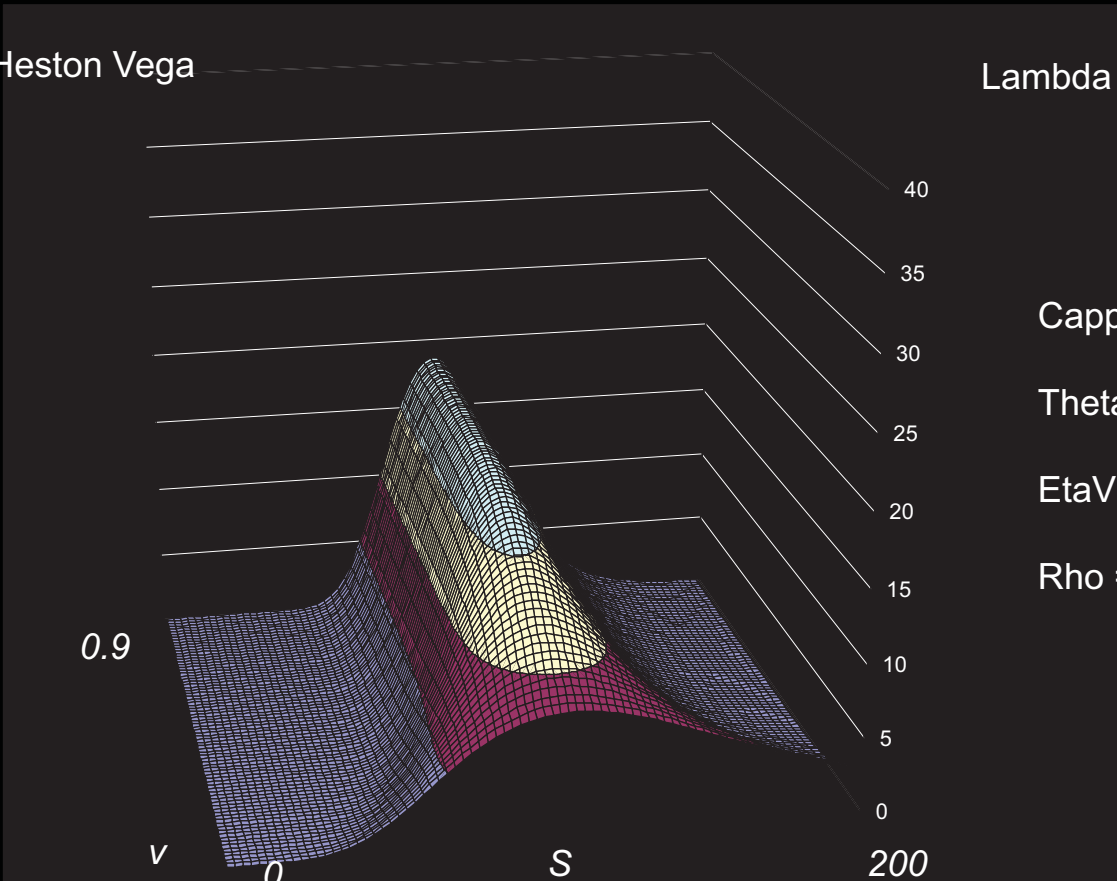
CappaV = 2

ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Vega



Lambda = 2



CappaV = 2

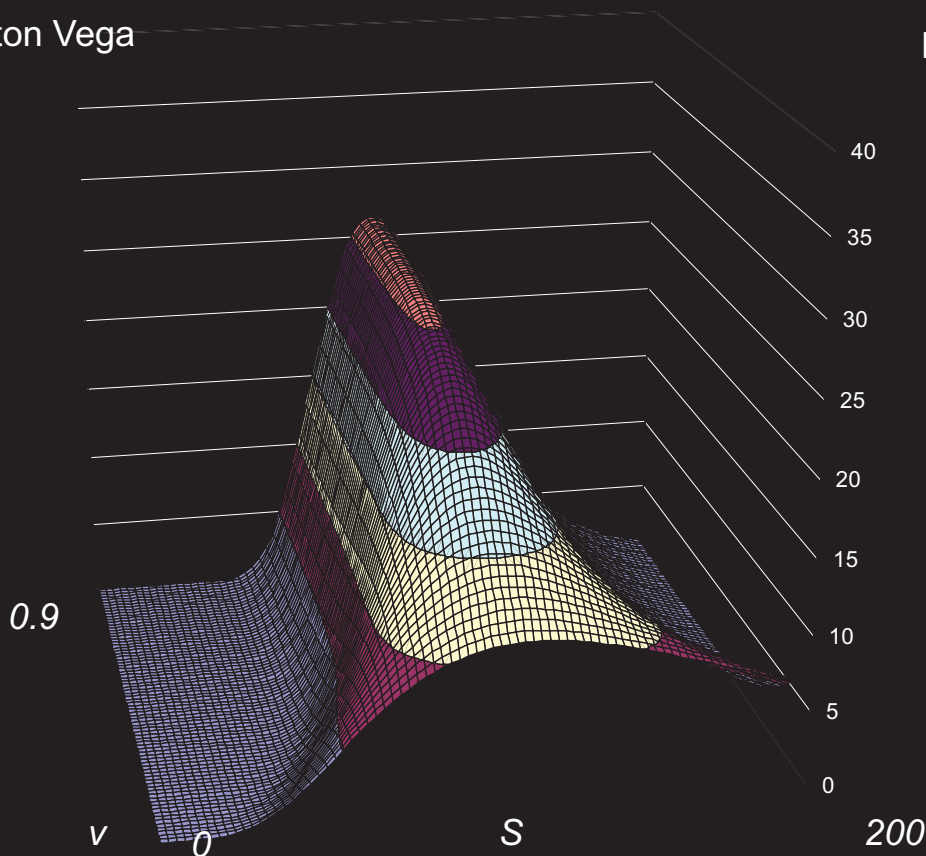
ThetaV = 0.3

EtaV = 0.1

Rho = 0

Heston Vega

Rho = -1



CappaV = 2

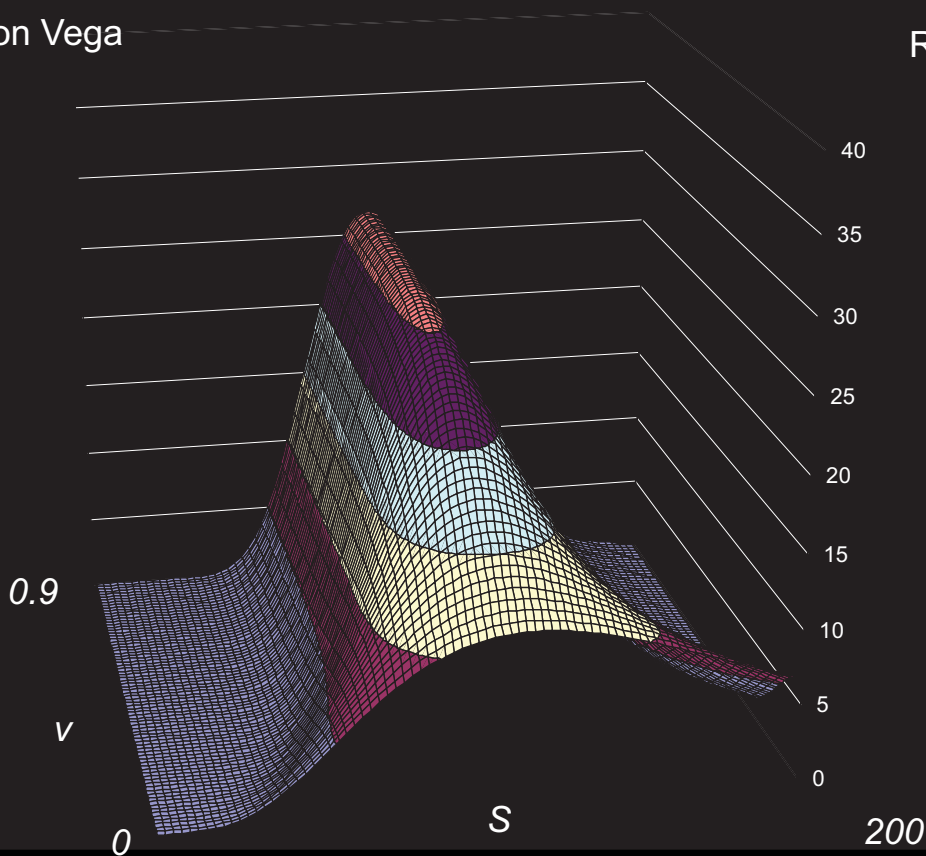
ThetaV = 0.3

EtaV = 0.1

Lambda = 0

Heston Vega

Rho = 1



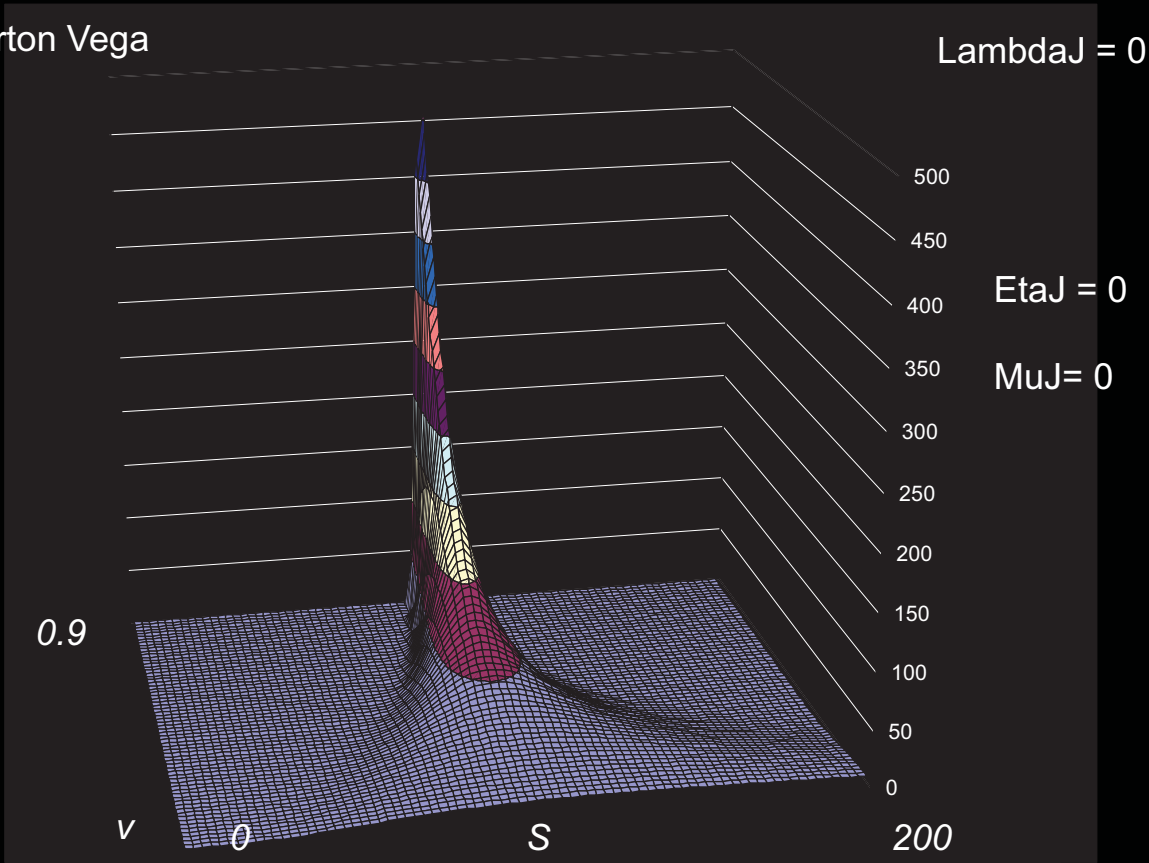
CappaV = 2

ThetaV = 0.3

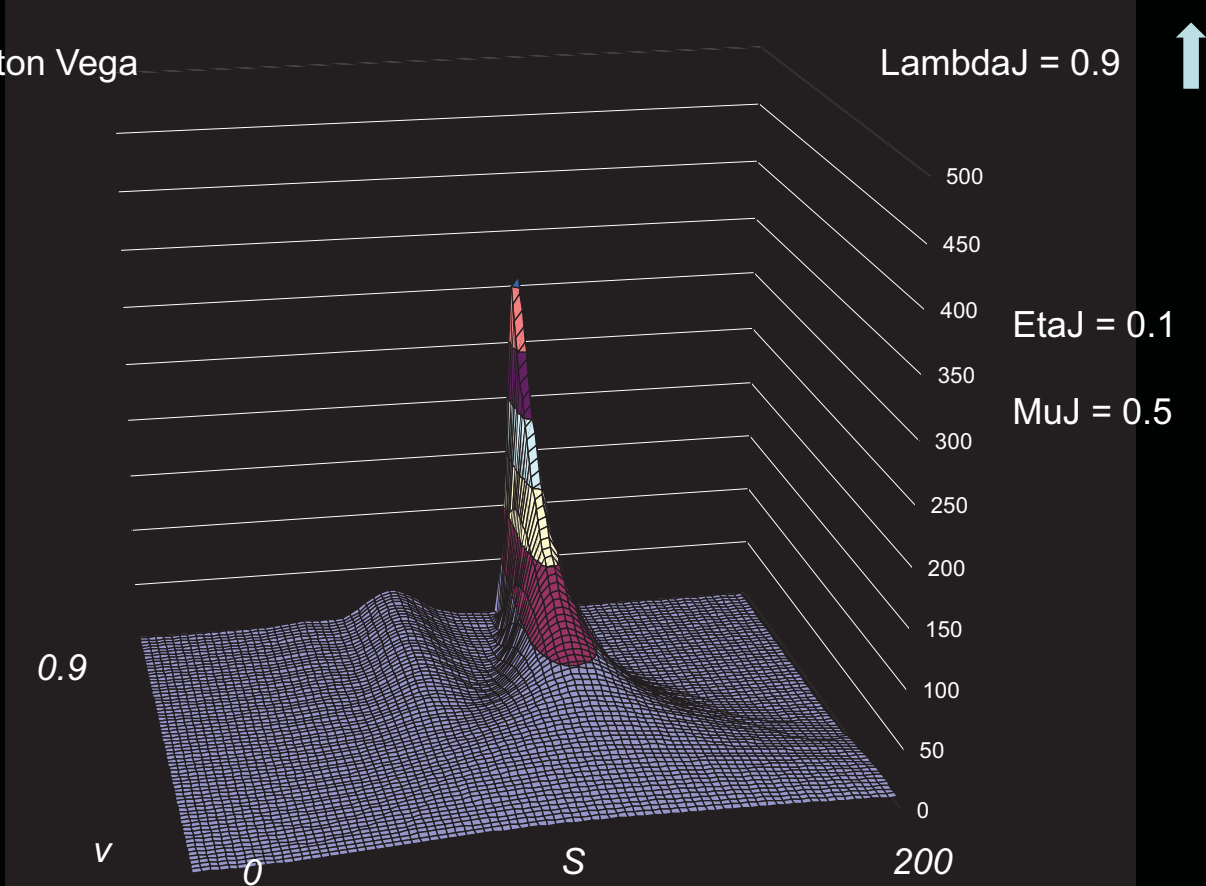
EtaV = 0.1

Lambda = 0

Merton Vega



Merton Vega



Merton Vega

$\Lambda J = 1.8$



$\eta J = 0.1$

$\mu J = 0.5$

0.9

v

0

S

200

500

450

400

350

300

250

200

150

100

50

0



CONSOB

Merton Vega

$\eta J = 0.1$

$\Lambda J = 0.5$

$\mu J = 0.5$

0.9

v

0

S

200

500

450

400

350

300

250

200

150

100

50

0



CONSOB

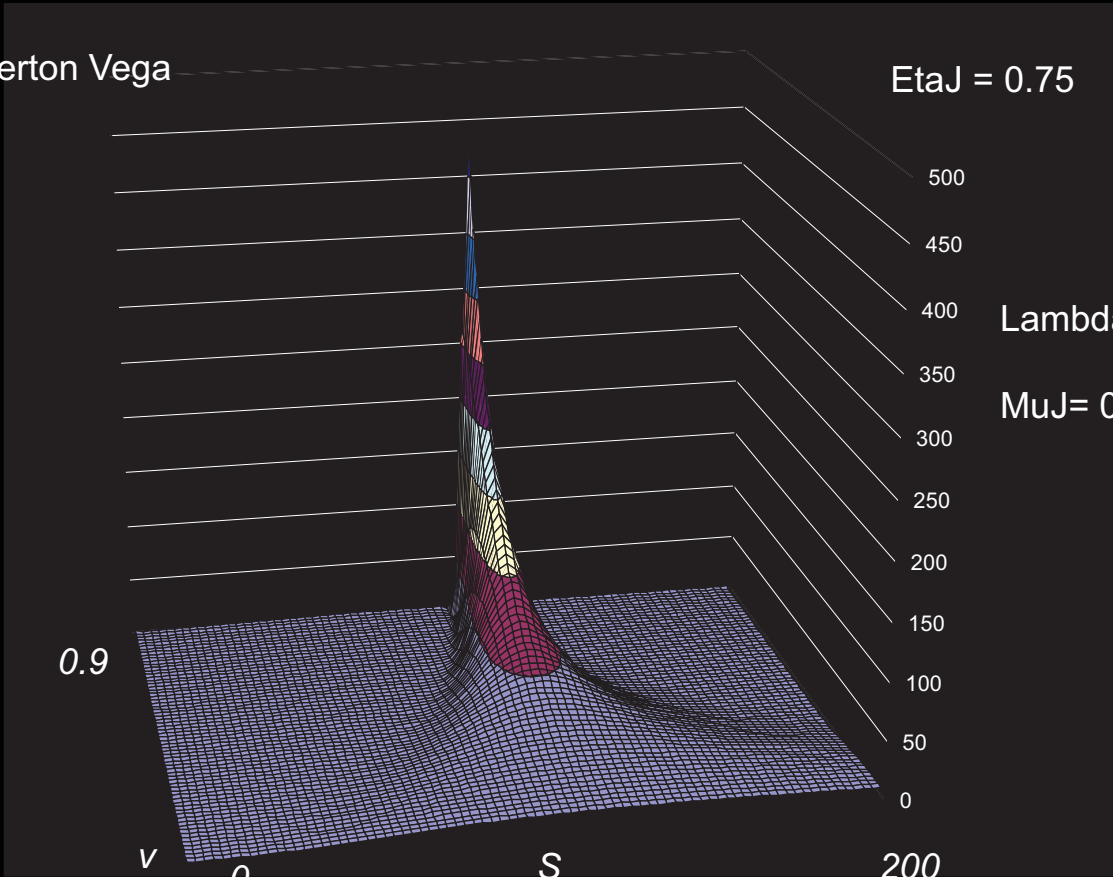
Merton Vega

$\text{EtaJ} = 0.75$



$\text{LambdaJ} = 0.5$

$\text{MuJ} = 0.5$



Merton Vega

$\text{MuJ} = 2.5$



$\text{LambdaJ} = 0.5$

$\text{EtaJ} = 0.5$

