

Rational Shapes of the Volatility Surface

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References

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Goals

- Derive arbitrage bounds on the slope and curvature of volatility skews.
- Investigate the strike and time behavior of these bounds.
- Specialize to stochastic volatility and jumps.
- Draw implications for parameterization of the volatility surface.

Slope Constraints

- No arbitrage implies that call spreads and put spreads must be non-negative. *i.e.*

$$\frac{\partial C}{\partial K} \leq 0 \text{ and } \frac{\partial P}{\partial K} \geq 0$$

- In fact, we can tighten this to

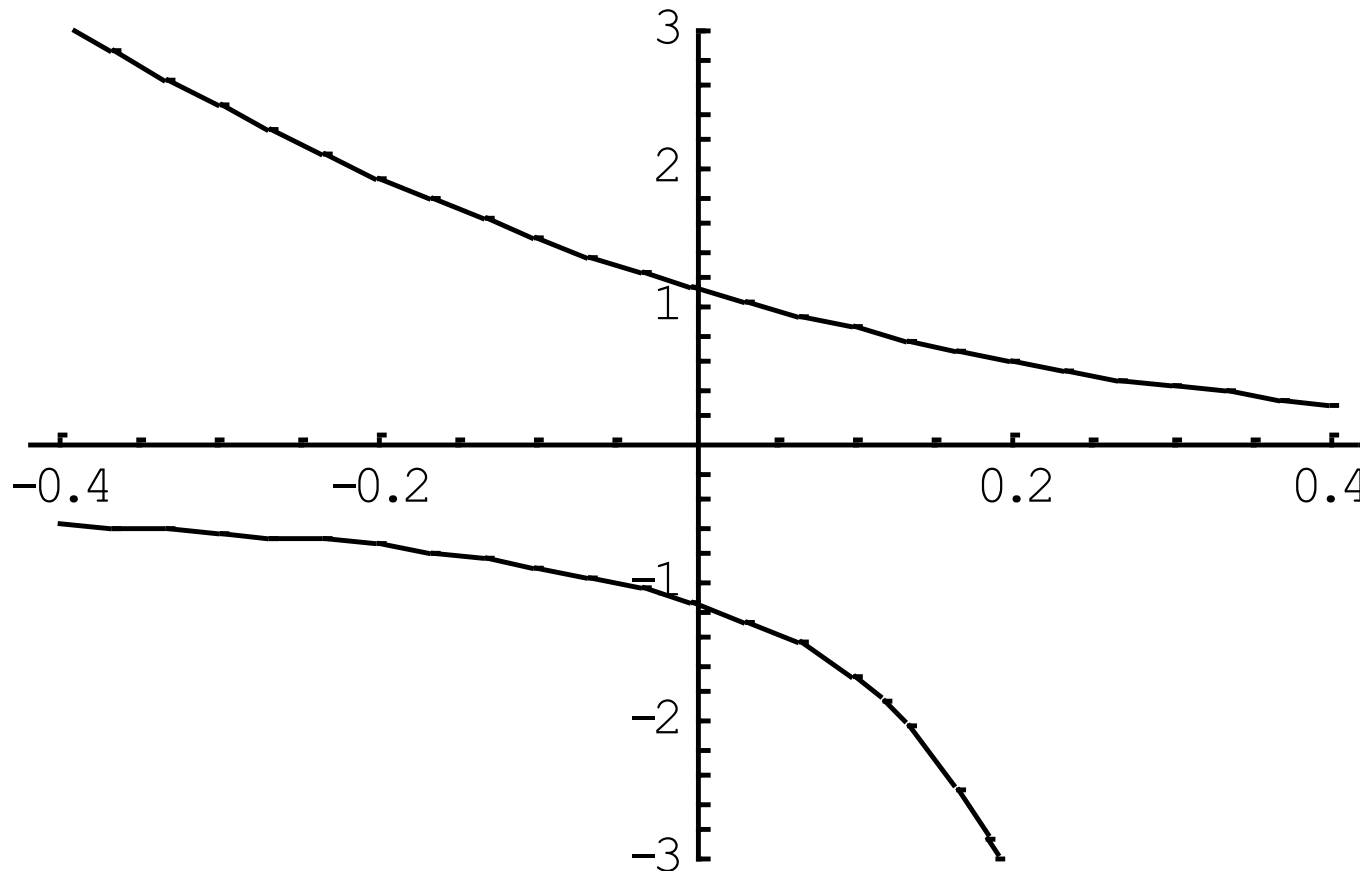
$$\frac{\partial C}{\partial K} \leq 0 \text{ and } \frac{\partial P}{\partial K} \geq 0$$

- Translate these equations into conditions on the implied total volatility $\sigma[y]$ as a function of $y = \ln(K / F)$.
- In conventional notation, we get

$$\sigma'[y] \leq \sqrt{2\pi} \exp\left(-\frac{d_2^2}{2\sigma^2}\right)$$

$$\sigma'[y] \geq -\sqrt{2\pi} \exp\left(-\frac{d_1^2}{2\sigma^2}\right)$$

- Assuming $\sigma[y] = 0.25 - 0.3y$ we can plot these bounds on the slope as functions of y .



- Note that we have plotted bounds on the slope of *total* implied volatility as a function of y . This means that the bounds on the slope of BS implied volatility get tighter as time to expiration increases by $1/\sqrt{T}$.

Convexity Constraints

- No arbitrage implies that call and puts must have positive convexity. *i.e.*

$$\frac{\partial^2 C}{\partial K^2} \geq 0 \text{ and } \frac{\partial^2 P}{\partial K^2} \geq 0$$

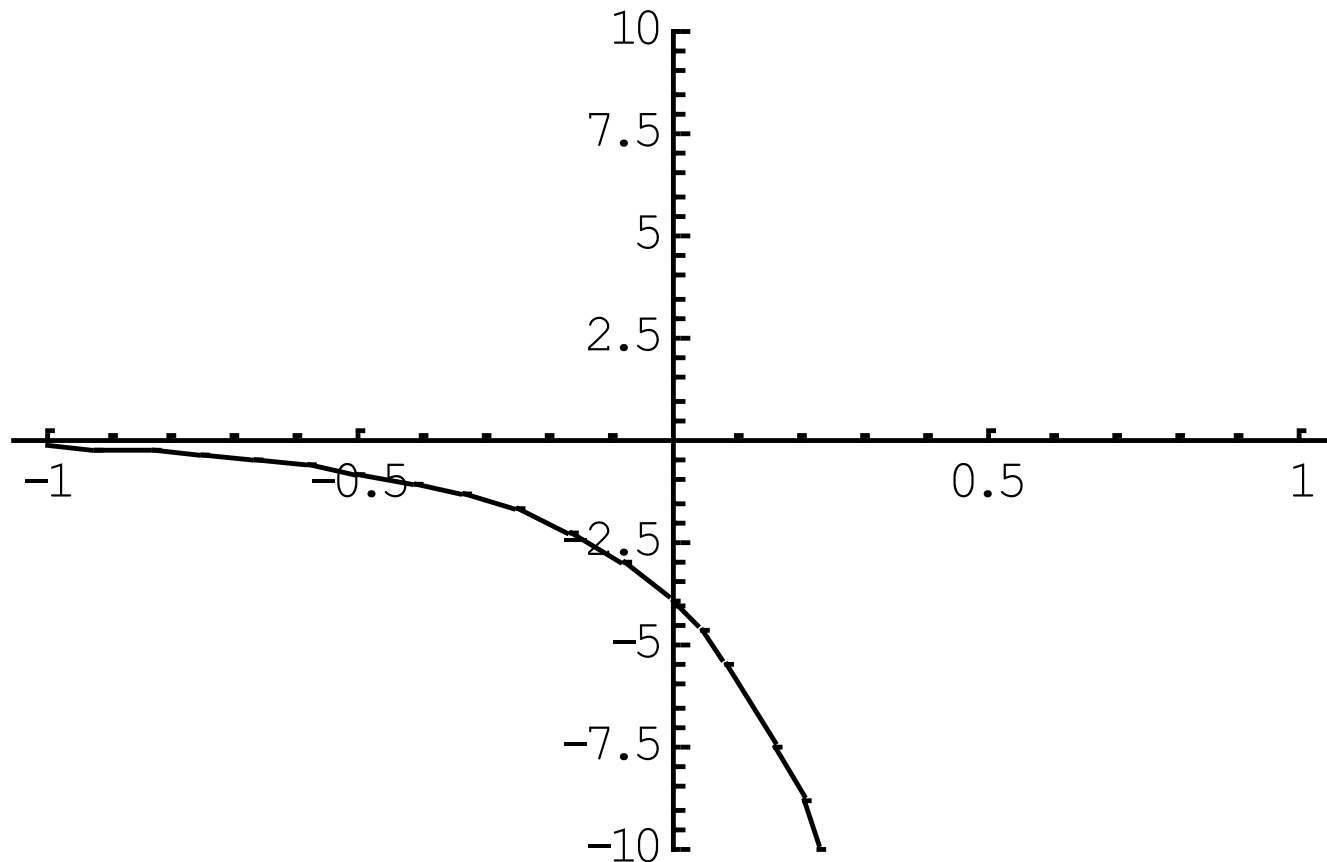
- Translating these into our variables gives

$$\frac{\partial^2 C}{\partial y^2} \geq \frac{\partial C}{\partial y}$$

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- Again, assuming $\sigma[y] = 0.25$ and $\sigma'[y] = -0.3$
we can plot this lower bound on the convexity
as a function of .



Implication for Variance Skew

- Putting together the vertical spread and convexity conditions, it may be shown that implied variance may not grow faster than linearly with the log-strike.
- Formally,

$$\frac{v[y]}{y} \equiv \frac{\sigma_{BS}^2[y]}{y} \rightarrow \text{some constant } A \text{ as } |y| \rightarrow \infty$$

Local Volatility

- Local volatility $\sigma(K, T)$ is given by

$$\frac{\sigma^2(K, T)}{2} = \frac{\frac{\partial C}{\partial T}}{K^2 \frac{\partial^2 C}{\partial K^2}}$$

- Local variances are non-negative iff arbitrage constraints are satisfied.

Time Behavior of the Skew

- Since in practice, we are interested in the lower bound on the slope for most stocks, let's investigate the time behavior of this lower bound.
- Recall that the lower bound on the slope can be expressed as

$$-\sqrt{2\pi} \exp\left(-d_1^2/2\right) \sigma \sqrt{T} \phi(d_1)$$

- For small times, $d_1 \approx 0$ and $N(d_1) \approx \frac{1}{2}$

so

$$\sigma'[0] \geq -\sqrt{\frac{\pi}{2}}$$

Reinstating explicit dependence on T , we get

$$\sigma_{BS}'[0] \geq -\sqrt{\frac{\pi}{2T}}$$

That is, \sqrt{T} for small T .

- Also,

$$d_1 = \frac{\sigma[0]}{2} \rightarrow \infty \text{ as } t \rightarrow \infty$$

- Then, the lower bound on the slope

$$\sigma'[0] \geq -\sqrt{2\pi} \exp(-d_1^2/2) \sigma[0]$$

$$\approx -\frac{1}{d_1} = -\frac{2}{\sigma[0]}$$

- Making the time-dependence of $\sigma[0]$ explicit,

$$\sigma_{BS}'[0] \geq -\frac{1}{T} \frac{2}{\sigma_{BS}[0]} \text{ as } T \rightarrow \infty$$

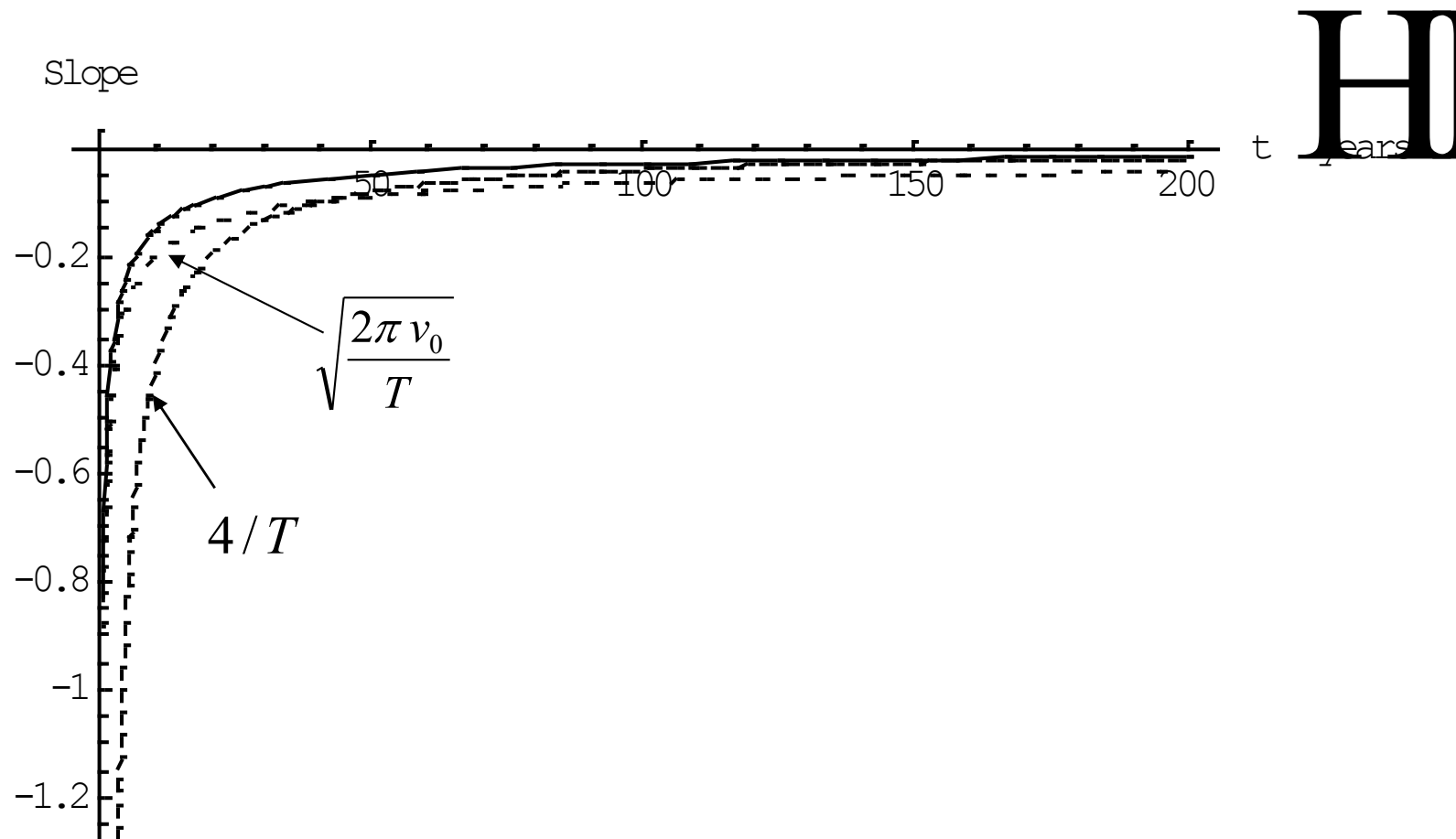
- In particular, the time dependence of the at-the-money skew cannot be

$$\sigma_{BS}'[0] \approx -\frac{1}{\sqrt{T}}$$

because for any choice of positive constants a, b

$$\exists T \text{ large enough s.t. } -\frac{a}{\sqrt{T}} < -\frac{b}{T}$$

- Assuming $\sigma_{BS}[0] = 0.25$, we can plot the variance slope lower bound as a function of time.

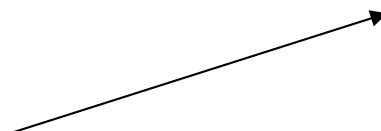


A Practical Example of Arbitrage

- We suppose that the ATMF 1 year volatility and skew are 25% and 11% per 10% respectively. Suppose that we extrapolate the vol skew using a $1/\sqrt{T}$ rule.
- Now, buy 99 puts struck at 101 and sell 101 puts struck at 99. What is the value of this portfolio as a function of time to expiration?

Current Market	100.00	100.00	100.00	100.00
Dividends (cts. yield or schedule)	0.00%	0.00%	0.00%	0.00%
Strike	101.00	99.00	101.00	99.00
Start Date (date on which strike is set)	03-Apr-98	03-Apr-98	03-Apr-98	03-Apr-98
Shares = s, Notional = n	s	s	s	s
Expiration Date	03-Apr-99	03-Apr-99	03-Apr-02	03-Apr-02
Stock Rate (sa/365 rate or curve)	0.000%	0.000%	0.000%	0.000%
Pay Rate (sa/365 rate or curve)	0.000%	0.000%	0.000%	0.000%
Volatility (number or curve)	23.90%	26.10%	24.45%	25.55%
Call =c, Put= p	p	p	p	p
Option Price	10.07	9.84	19.92	19.58
Delta	-0.4690	-0.4329	-0.4113	-0.3916
Gamma (per 1%)	0.0166	0.0151	0.0080	0.0075
Vega per 1% vol	0.3976	0.3932	0.7774	0.7675
Theta per day	-0.0130	-0.0141	-0.0065	-0.0067
Position	99	-101	99	-101
Value	996.72	(993.70)	1,972.34	(1,977.18)
Portfolio Value	3.02		(4.83)	

Arbitrage!



With more reasonable parameters, it takes a long time to generate an arbitrage though....

Current Market	100.00	100.00	100.00	100.00
Dividends (cts. yield or schedule)	0.00%	0.00%	0.00%	0.00%
Strike	101.00	99.00	101.00	99.00
Start Date (date on which strike is set)	03-Apr-98	03-Apr-98	03-Apr-98	03-Apr-98
Shares = s, Notional = n	s	s	s	s
Expiration Date	03-Apr-99	03-Apr-99	03-Apr-48	03-Apr-48
Stock Rate (sa/365 rate or curve)	0.000%	0.000%	0.000%	0.000%
Pay Rate (sa/365 rate or curve)	0.000%	0.000%	0.000%	0.000%
Volatility (number or curve)	24.70%	25.30%	24.96%	25.04%
Call =c, Put= p	p	p	p	p
Option Price	10.39	9.52	63.07	61.61
Delta	-0.4668	-0.4340	-0.1902	-0.1864
Gamma (per 1%)	0.0161	0.0156	0.0015	0.0015
Vega per 1% vol	0.3975	0.3934	1.8909	1.8670
Theta per day	-0.0135	-0.0136	-0.0013	-0.0013
Position	99	-101	99	-101
Value	1,028.21	(961.92)	6,244.14	(6,222.68)
Portfolio Value	66.30	No arbitrage!	21.46	



So Far....

- We have derived arbitrage constraints on the slope and convexity of the volatility skew.
- We have demonstrated that the $1 / \sqrt{T}$ rule for extrapolating the skew is inconsistent with no arbitrage. Time dependence must be at most $1 / T$ for large T

Stochastic Volatility

- Consider the following special case of the Heston model:

$$dx = \mu dt + \sqrt{v} dZ$$

$$dv = -\lambda(v - \bar{v})dt - \eta\sqrt{v} dZ$$

- In this model, it can be shown that

$$\left. \frac{\partial v_{BS}}{\partial y} \right|_{y=0} \approx -\eta \frac{1}{\lambda T} \left\{ 1 - \frac{1 - e^{-\lambda T}}{\lambda T} \right\}$$

- For a general stochastic volatility theory of the form:

$$dx = \mu dt + \sqrt{v} dZ_1$$

$$dv = -\lambda(v - \bar{v})dt - \eta \beta(v) \sqrt{v} dZ_2$$

with

$$\langle dZ_1, dZ_2 \rangle = \rho dt$$

we claim that (very roughly)

$$\left. \frac{\partial v_{BS}}{\partial y} \right|_{y=0} \approx \rho \eta \beta(v) \frac{1}{\lambda T} \left\{ 1 - \frac{1 - e^{-\lambda T}}{\lambda T} \right\}$$

- Then, for very short expirations, we get

$$\left. \frac{\partial v_{BS}}{\partial y} \right|_{y=0} \approx \frac{\rho \eta \beta(v)}{2}$$

- a result originally derived by Roger Lee and for very long expirations, we get

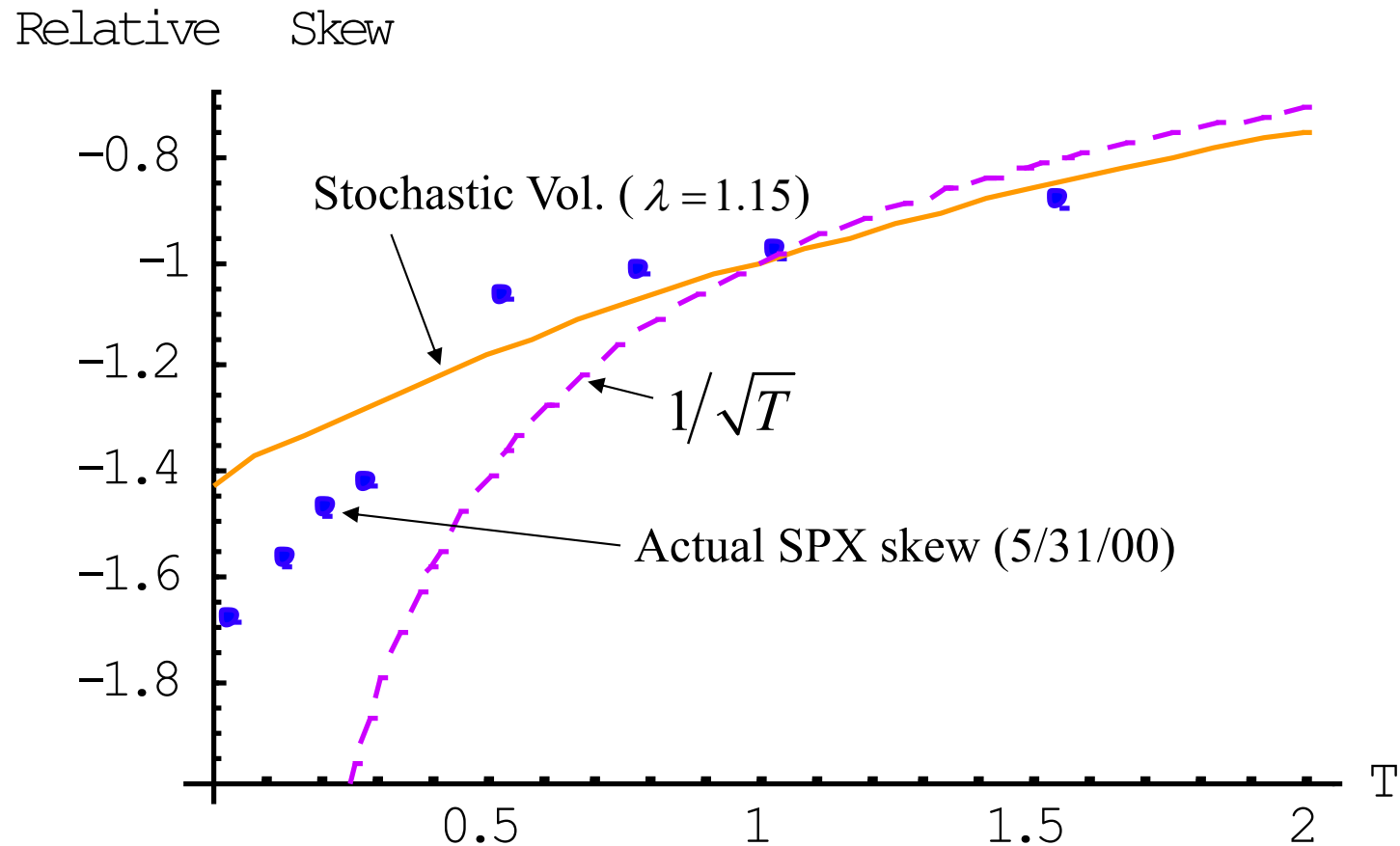
$$\left. \frac{\partial v_{BS}}{\partial y} \right|_{y=0} \approx \frac{\rho \eta \beta(v)}{\lambda T}$$

- Both of these results are consistent with the arbitrage bounds.

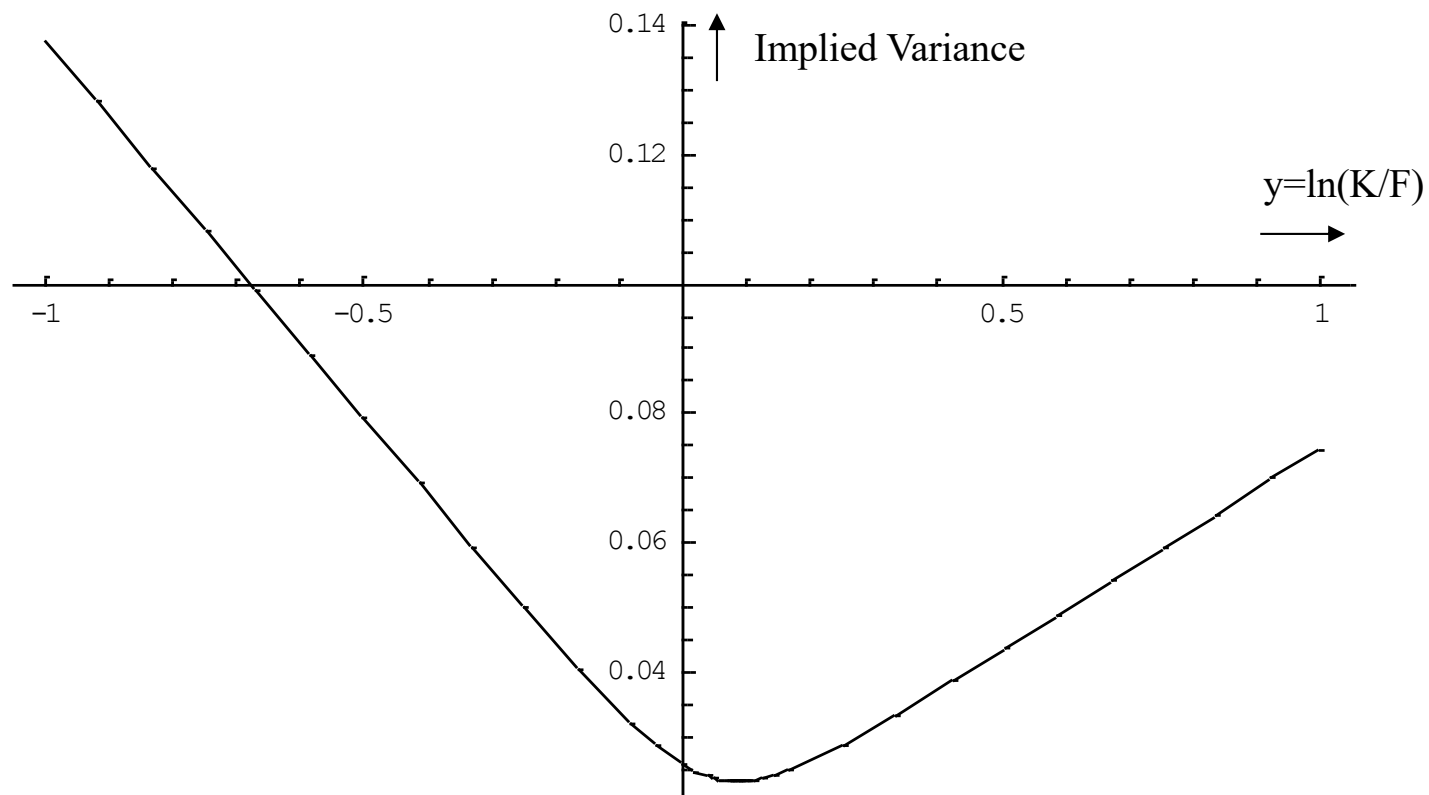
Doesn't This Contradict \sqrt{T} ?

- Market practitioners' rule of thumb is that the skew decays as $1/\sqrt{T}$.
- Using $\lambda = 1.15$ (from Bakshi, Cao and Chen), we get the following graph for the relative size of the at-the-money variance skew:

ATM Skew as a Function of T



Heston Implied Variance



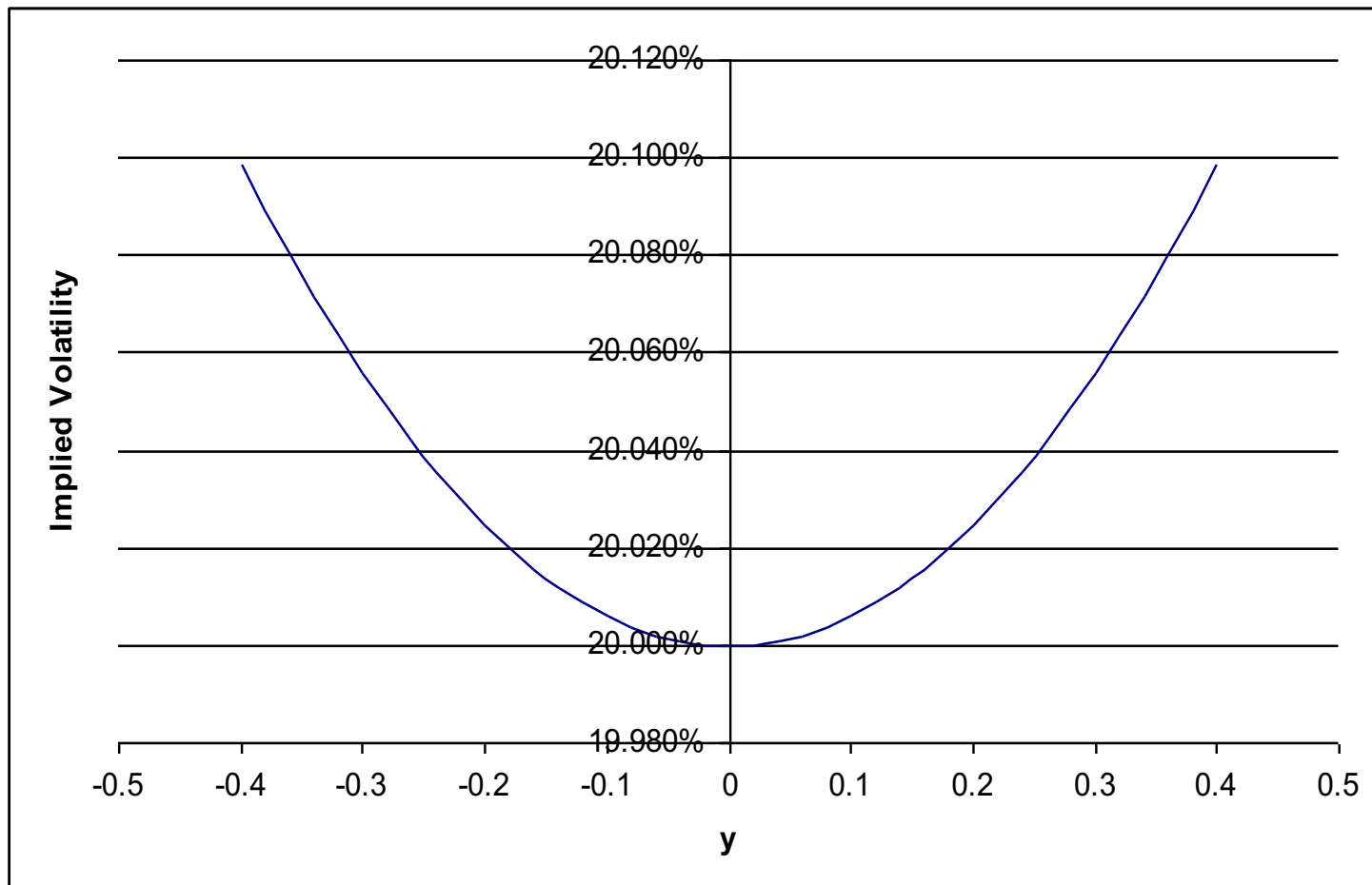
Parameters: $\nu = 0.04$, $\bar{\nu} = 0.04$, $\lambda = 1.15$, $\rho = -0.39$, $\eta = 0.64$

from Bakshi, Cao and Chen.

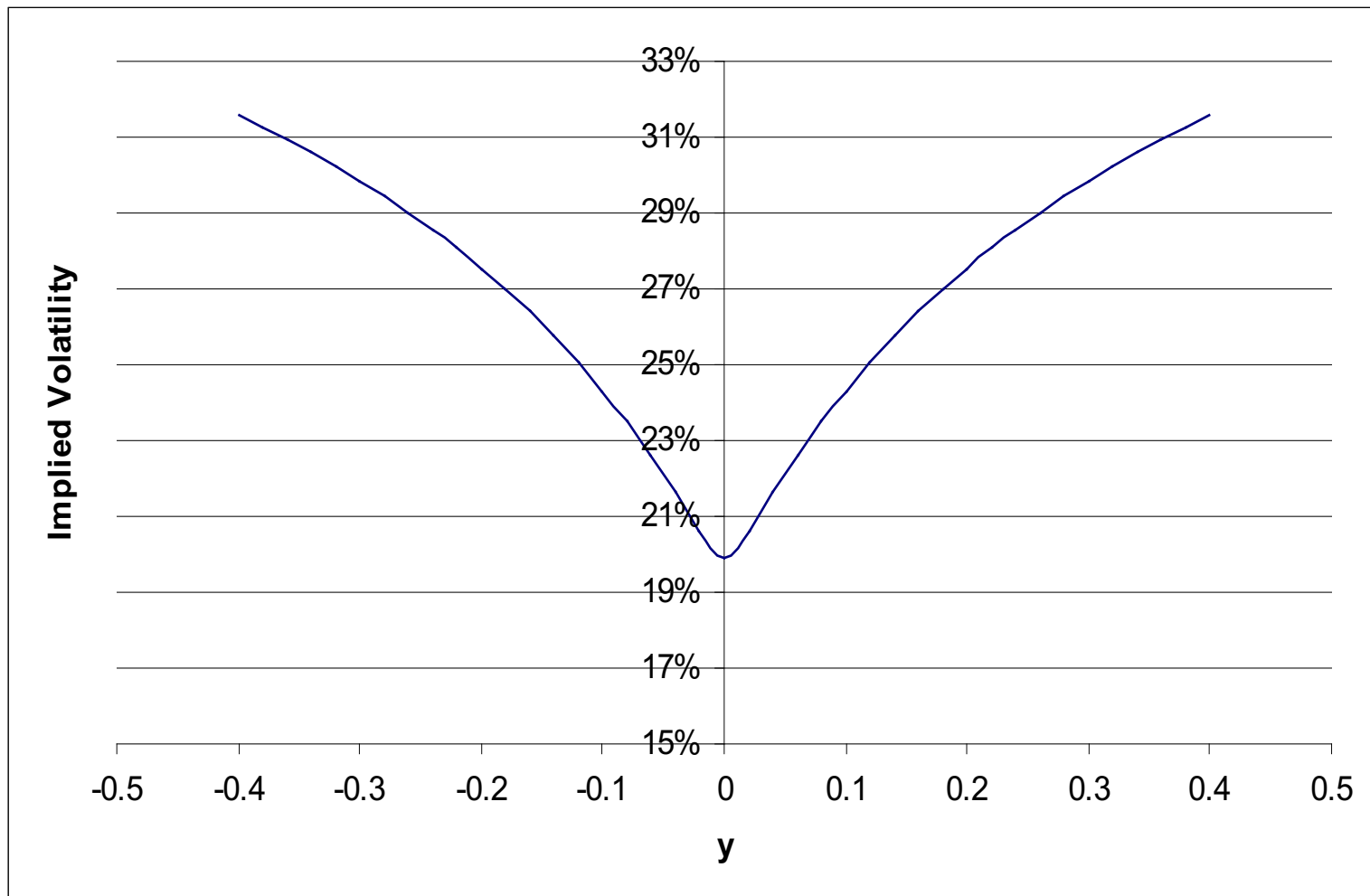
A Simple Regime Switching Model

- To get intuition for the impact of volatility convexity, we suppose that realised volatility over the life of a one year option can take one of two values each with probability $1/2$. The average of these volatilities is 20%.
- The price of an option is just the average option price over the two scenarios.
- We graph the implied volatilities of the resulting option prices.

High Vol: 21%; Low Vol: 19%



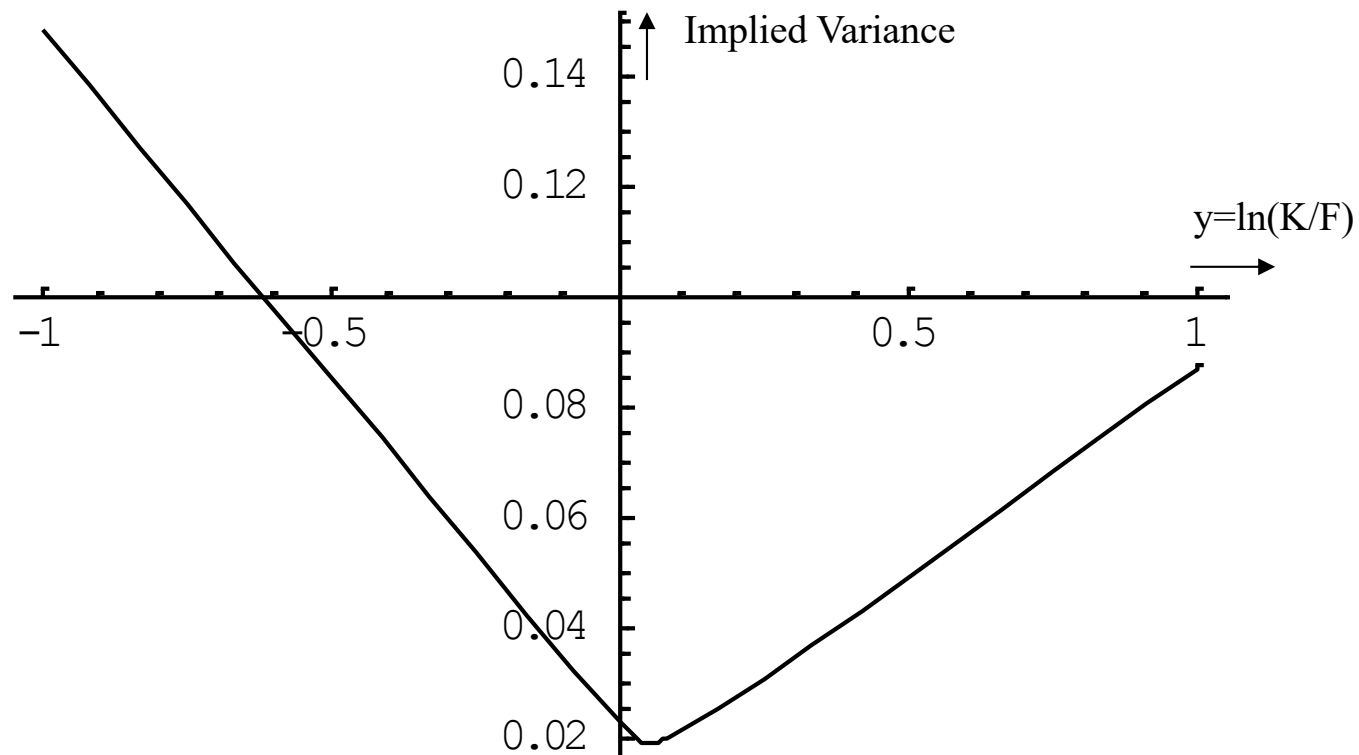
High Vol: 39%; Low Vol: 1%



Intuition

- As $|y| \rightarrow \infty$, implied volatility tends to the highest volatility.
- If volatility is unbounded, implied volatility must also be unbounded.
- From a trader's perspective, the more out-of-the-money (OTM) an option is, the more volatility convexity it has. Provided volatility is unbounded, more OTM options must command higher implied volatility.

Asymmetric Variance Gamma Implied Variance

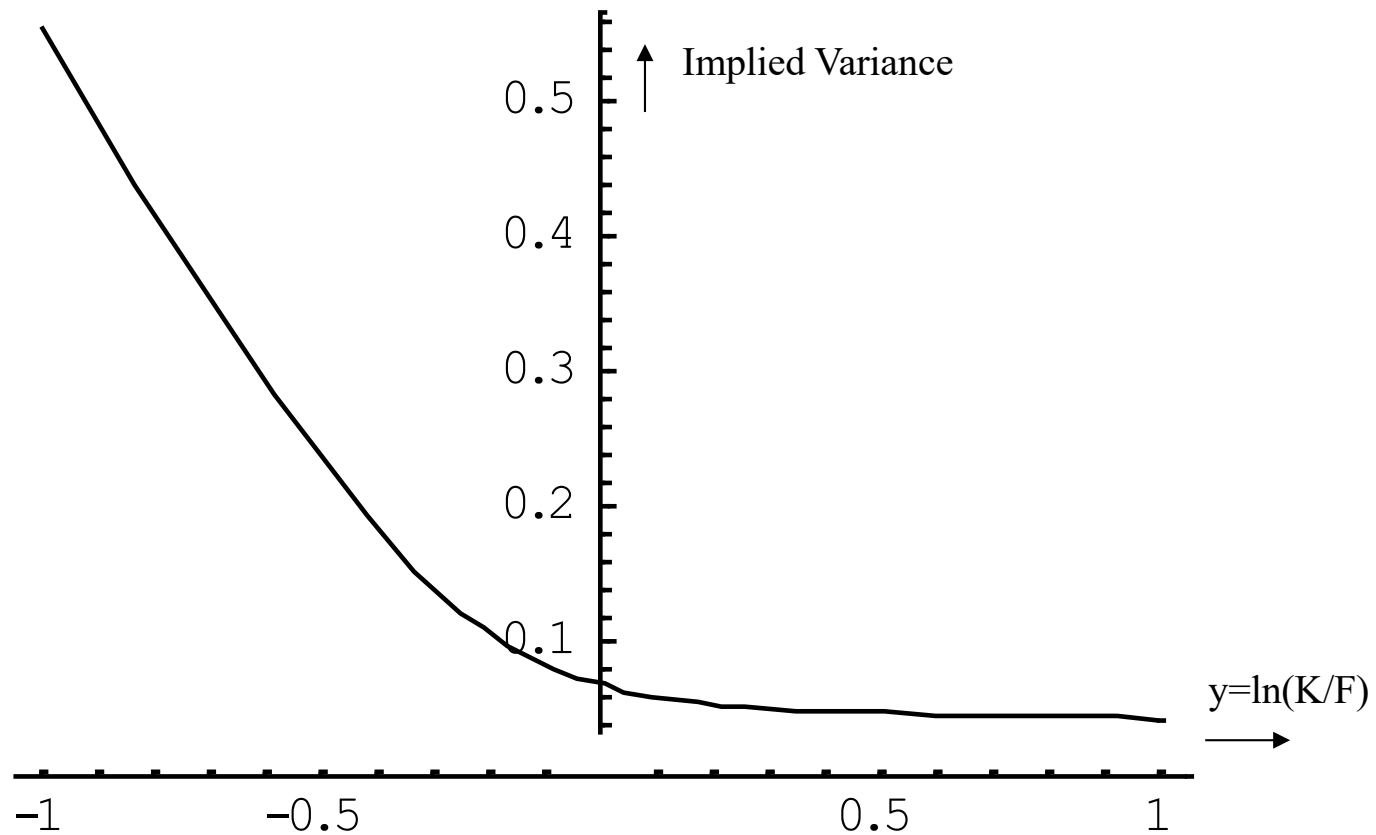


Parameters: $\bar{w} = 0.04, \nu = 0.1, \theta = -1.5, \rho = -0.4$

Jump Diffusion

- Consider the simplest form of Merton's jump-diffusion model with a constant probability λ of a jump to ruin.
- Call options are valued in this model using the Black-Scholes formula with a shifted forward price.
- We graph 1 year implied variance as a function of log-strike with $\nu = 0.04$, $\lambda = 0.05$:

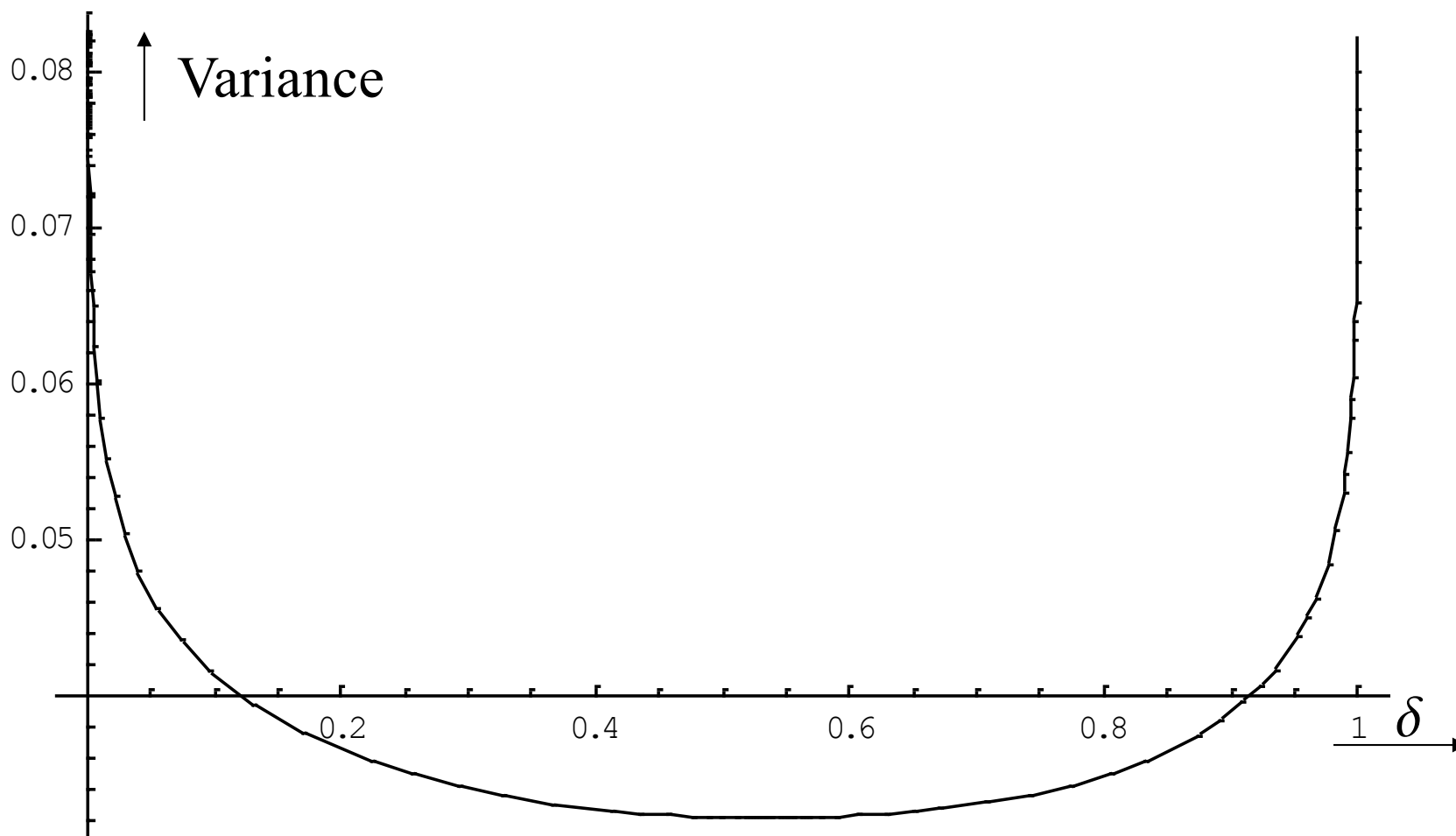
Jump-to-Ruin Model



Parameters: $\bar{v} = 0.04, \lambda = 0.05$

- So, even in jump-diffusion, v is linear in y as $|y| \rightarrow \infty$.
- In fact, we can show that for many economically reasonable stochastic-volatility-plus-jump models, implied BS variance must be asymptotically linear in the log-strike y as $|y| \rightarrow \infty$.
- This means that it does not make sense to plot implied BS variance against delta. As an example, consider the following graph of v vs. δ in the Heston model:

Variance vs δ in the Heston Model



Implications for Parameterization of the Volatility Surface

- Implied BS variance v must be parameterized in terms of the log-strike y (vs delta doesn't work).
- v is asymptotically linear in y as $|y| \rightarrow \infty$

- $\left. \frac{\partial v}{\partial y} \right|_{y=0}$ decays as $\frac{1}{T}$ as $T \rightarrow \infty$

- $\left. \frac{\partial v}{\partial y} \right|_{y=0}$ tends to a constant as $T \rightarrow 0$