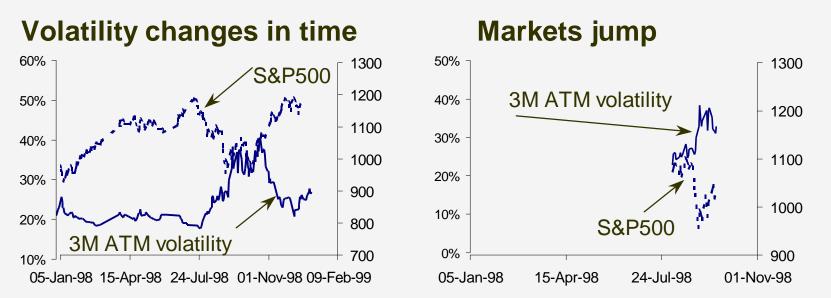
## **Perturbative Analysis of Volatility Smiles**

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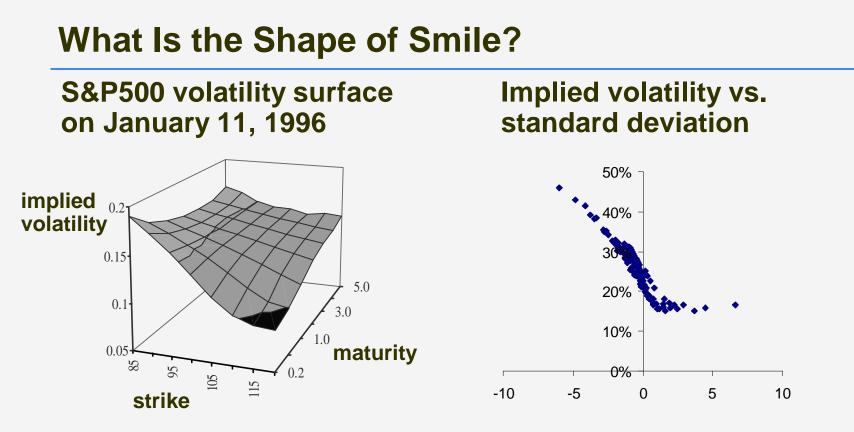
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## **Main Risks in Options Markets**



- Index volatility is mean-reverting
- It is negatively correlated with the price
- A jump in price often entails a volatility jump

#### Most models ignore at least one of these risks

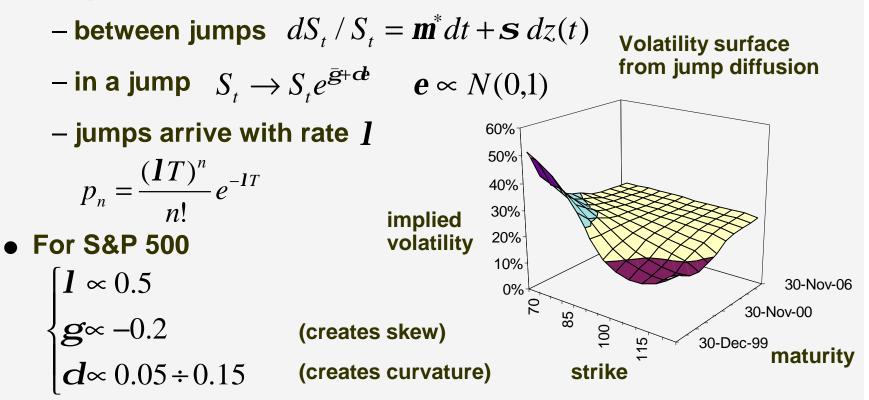


- Implied volatility decreases with strike price
- The skew slope is the greatest for short maturities

What underlying processes produce such skews?

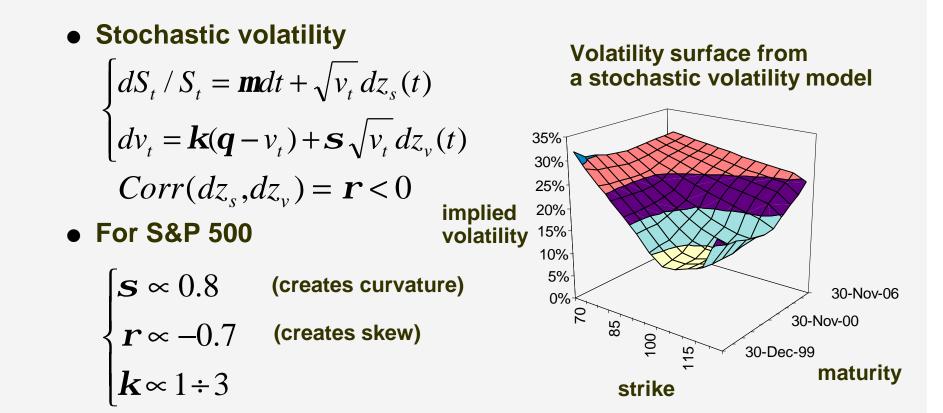
#### Is the Skew Due to Jumps?

Jump Diffusion model



The jump diffusion model works well for short maturities

#### What Happens at Longer Maturities?



#### Stochastic volatility models work well for long maturities

# How to Combine Stochastic Volatility and Jump Diffusion ?

- between jumps  $\begin{cases} dS/S = \mathbf{m}dt + \sqrt{v}dz_1 \\ dv = \mathbf{k}(\mathbf{q} v)dt + \mathbf{s}\sqrt{v}dz_2 \end{cases}$  Corr(dz<sub>1</sub>, dz<sub>2</sub>) = **r**
- market crashes form a Poisson process with rate  $\boldsymbol{l}$  $\begin{cases} \log S \to \log S + \boldsymbol{g} + \boldsymbol{d} \boldsymbol{e} & \boldsymbol{e} \propto N(0,1) \\ v \to v + \boldsymbol{g} \end{pmatrix}$
- the option price obeys the equation

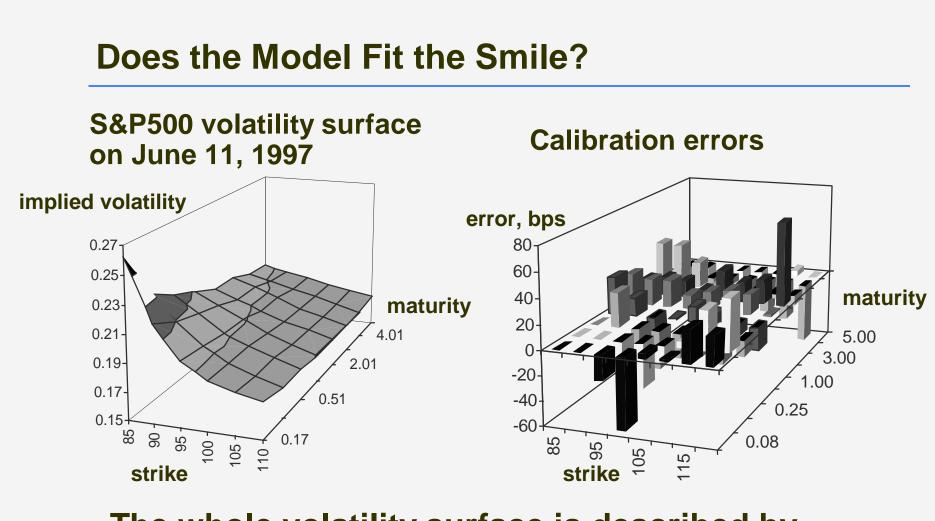
$$\frac{\partial f}{\partial t} + \mathbf{m}^* S \frac{\partial f}{\partial S} + \mathbf{k}(\mathbf{q} - v) \frac{\partial f}{\partial v} + \frac{1}{2} v \left\{ S^2 \frac{\partial^2 f}{\partial S^2} + \mathbf{s}^2 \frac{\partial^2 f}{\partial v^2} + 2 \mathbf{rs} S \frac{\partial^2 f}{\partial S \partial v} \right\}$$
$$+ \mathbf{I} E^* \left[ f (Se^{\mathbf{g} + \mathbf{d} \mathbf{e}}, v + \mathbf{g}) - f (S, v) \right] = rf$$

**European option prices can be computed analytically** 

#### What Is the Distribution of Stock Prices?

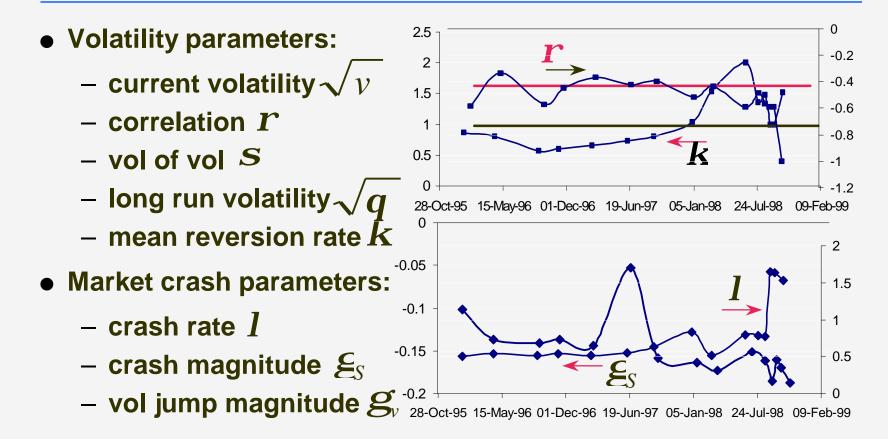
- Call prices equal  $C = S P_1 K e^{-rT} P_0$
- Find the characteristic functional  $f(t, f) = E^* \left[ e^{if \ln(S/F)} \right] = \text{Fourier Transform of } P'_0$ • Use the affine ansatz  $\hat{P}_n = e^{C(T-t, j) + D(T-t, j) \vee}$  to derive  $\begin{cases} C(t, j) = C_H(t, j) + It \left[ e^{ijg - j^2 d^2/2} I(t) - 1 \right] & p_{\pm} = \frac{\mathcal{E}_{\nu}}{s^2} (b - rsj i \pm d) \\ D(t, j) = D_H(t, j) & I(t) = \frac{1}{t} \int_0^t e^{g D(t, j)} dt = -\frac{2g}{p_+ p_-} \int_0^{-g D(t, j)} \frac{e^{-z} dz}{(1 + z/p_+)(1 + z/p_-)} \end{cases}$

This model accounts for the main risks of options markets



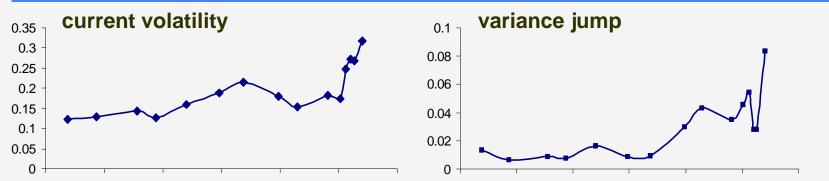
The whole volatility surface is described by one set of constant parameters

#### **Are Smile Parameters Stable Over Time?**

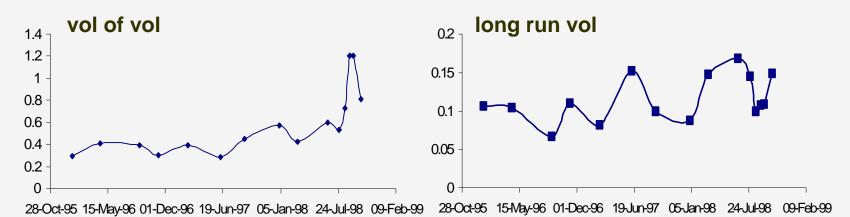


# Mean reversion, correlation and crash size are constant

#### **Patterns in Stochastic Volatility Parameters**



28-Oct-95 15-May-96 01-Dec-96 19-Jun-97 05-Jan-98 24-Jul-98 09-Feb-99 28-Oct-95 15-May-96 01-Dec-96 19-Jun-97 05-Jan-98 24-Jul-98 09-Feb-99



## Long run diffusion volatility is relatively stable

#### What Is the Intuition?

- How does each source of risk affect the smile slope
  - at long maturities
  - at short maturities
- What is its effect on
  - ATM volatility
  - smile curvature
- For many models, the "weak smile expansion" is a good guide.
- However, the natural expansion is for the characteristic functional, not the implied volatilities.

#### How to construct the weak smile expansion?

#### Linking Characteristic Functionals to Implied Volatilities

• The characteristic functional

The probability distribution

$$F_t(\mathbf{h}) = \int_0^\infty p(K) e^{i\mathbf{h}\ln(K/F)} dK$$
$$p(K) = e^{rT} \frac{\partial^2 C}{\partial K^2}$$

• Introduce implied standard deviation  $\mathbf{j} = \mathbf{\bar{s}}(K,T)\sqrt{T}$ Parametrize  $\mathbf{j} = \mathbf{j}(z)$ where  $z \equiv d_2 = \frac{\ln(F/K)}{\mathbf{j}} - \frac{\mathbf{j}}{2} = \ln(M/K), \qquad M = Fe^{-\mathbf{j}^2/2}$ • Then  $p(K) dK = N'(z) dz \left\{ -1 + \frac{z}{\mathbf{j}/\mathbf{j} + z + \mathbf{j}} - \frac{\partial}{\partial z} \left( \frac{1}{\mathbf{j}/\mathbf{j} + z + \mathbf{j}} \right) \right\} \qquad \mathbf{j} \equiv \frac{\partial \mathbf{j}}{\partial z}$ 

#### Linking Characteristic Functionals to Implied Volatilities

• Changing the integration variable to *z* and integrating by parts

$$F(\mathbf{h}) = \int_{-\infty}^{+\infty} dz \, N'(z) \, e^{-i\mathbf{h}\mathbf{j}\left(\frac{1}{2}\mathbf{j}+z\right)} (1+i\mathbf{h}\mathbf{j})$$

• In terms of  $w \equiv z + ihj(z)$ 

$$F(\mathbf{h}) = \int_{-\infty}^{+\infty} dw \, N'(w) \, e^{-\frac{1}{2}\mathbf{h}(\mathbf{h}+i)\mathbf{j}^{2}(w)}$$

$$w = \frac{\ln(F/K)}{j(w)} + \left(ih - \frac{1}{2}\right)j(w)$$

# $F(\mathbf{h})$ is related to the analytic continuation of $\mathbf{j}$

#### What Is the First Order Perturbation?

- Assume  $\mathbf{j}^{2}(w) \cong \mathbf{j}_{0}^{2} + \mathbf{y}_{1}(\mathbf{h}, w)$ with  $\mathbf{j}_{0}$  independent of w. • Then  $F(\mathbf{h}) = e^{-\frac{1}{2}\mathbf{j}_{0}^{2}\mathbf{h}(\mathbf{h}+i)} \{1 + F_{1}(\mathbf{h})\}$   $\int_{-\infty}^{+\infty} dw N'(w)\mathbf{y}_{1}(\mathbf{h}, w) = -\frac{2}{\mathbf{h}(\mathbf{h}+i)} F_{1}(\mathbf{h})$  $w = -\frac{x}{\mathbf{j}_{0}} + \mathbf{j}_{0}\left(i\mathbf{h}-\frac{1}{2}\right) + O(\mathbf{y}_{1})$
- As a result

$$\mathbf{y}_{1}(x) = -\frac{2\mathbf{j}_{0}}{\sqrt{2\mathbf{p}}} e^{\frac{x^{2}}{2\mathbf{j}_{0}^{2}}} \int_{-\infty}^{+\infty} \frac{F_{1}(\mathbf{h}^{*}) d\mathbf{h}^{*}}{\mathbf{h}^{*2} + \frac{1}{4}} e^{-\frac{1}{2}\mathbf{j}_{0}^{2}\mathbf{h}^{*2} - ix\mathbf{h}^{*}} \qquad \mathbf{h}^{*} = \mathbf{h} + i/2$$

The smile slope is a simple integral of  $F_1(\mathbf{h}^*)$ 

#### What Is the Effect of Price Jumps?

• In the Merton model

The ATM smile slope

$$\frac{\partial \mathbf{y}_{1}}{\partial x}\Big|_{x=0} = 2\mathbf{I}T(e^{\mathbf{g}}-1) \qquad \Rightarrow \quad \frac{d\overline{\mathbf{s}}}{dx}\Big|_{x=0} = \frac{\mathbf{I}(e^{\mathbf{g}}-1)}{\mathbf{s}_{0}}$$

- stable calibration of expected loss  $I(e^g-1)$ 

– more noise in l and  $\underline{\beta}$ 

• The smile curvature  $(\boldsymbol{\xi} = 0)$ 

$$\frac{\partial^2 \mathbf{y}_1}{\partial x^2} \bigg|_{x=0} = \frac{\mathbf{I} T \mathbf{d}^4}{4 \mathbf{j}_0^2} \qquad \mathbf{\overline{S}} \approx \mathbf{S}_0 + \frac{\mathbf{I}}{16} \frac{\mathbf{d}^4 x^2}{\mathbf{S}_0^3 T}$$

at small *d*, very straight skews
strong dependence on *d*when *d* is large

#### What Is the Effect of Stochastic Volatility?

- Black-Scholes variance  $\mathbf{j}_{0}^{2} \equiv \mathbf{q}T + \frac{\mathbf{v} \mathbf{q}}{\mathbf{k}}(1 e^{-\mathbf{k}T})$
- As  $T \rightarrow 0$ ,  $y'_{1}(x) = \frac{1}{2} rsT$ Hence the smile slope (in stdev space)

$$\frac{1}{\overline{\boldsymbol{s}}_0}\frac{d\overline{\boldsymbol{s}}}{dz} = \sqrt{T}\,\frac{d\overline{\boldsymbol{s}}}{dx} = \frac{\boldsymbol{rs}}{4\sqrt{v_0}}\,\sqrt{T} \approx 0.16$$

• As 
$$T \to \infty$$
,  $y'_{1}(x) = \frac{\mathbf{rs}}{\mathbf{k}}$   
 $\frac{1}{\mathbf{s}_{0}} \frac{d\mathbf{s}}{dz} = \frac{\mathbf{rs}}{2\mathbf{k}\sqrt{qT}} \propto 0.06$ 

- calibration of *rs* more stable

long run skew often too flat

#### The long run Heston smile is often too flat

#### How Jumps in Volatility Change the Picture?

• If  $\boldsymbol{\xi} = \boldsymbol{d} = 0$ , only the change in volatility level

$$\mathbf{y}_{1}(x) = -(\mathbf{I}T)(\mathbf{g}T) \frac{1 - \mathbf{k}T - e^{-\mathbf{k}T}}{(\mathbf{k}T)^{2}} \qquad \qquad \rightarrow \frac{1}{2}(\mathbf{I}T)(\mathbf{g}T) \qquad \text{as } T \to 0$$
$$\rightarrow \frac{\mathbf{I}g}{\mathbf{k}}T \qquad \qquad \text{as } T \to \infty$$

• Interaction of vol jumps with price jumps

$$y_{1}(x) = \frac{(lT)(gT)}{2} \frac{g}{j_{0}^{2}} \frac{1-kT-e^{-kT}}{(kT)^{2}}$$

$$-as T \rightarrow 0, \quad \overline{s}'_{x} = \frac{lg}{s_{0}} \rightarrow (1+a_{0}) \frac{lg}{s_{0}} \qquad a_{0} = \frac{g}{4s_{0}^{2}}$$
for the jump from 15% to 35% 
$$g \approx 0.10 \Rightarrow a_{0} \approx 1.0$$

$$-as T \rightarrow \infty, \quad y_{1}'(x) = \frac{lgg}{kq} \quad vs. \quad \frac{rs}{k} \qquad \Rightarrow a_{\infty} = \frac{lgg}{rsq} \approx 0.9$$
Volatility jumps significantly affect the skew

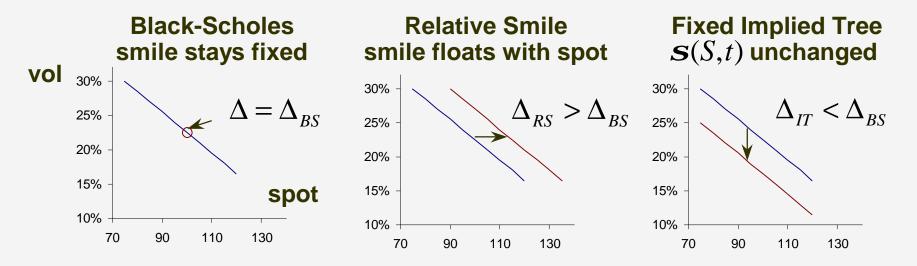
#### What Is the Delta?

• When the spot moves, the smile can move too

hus 
$$\Delta = \frac{dC}{dS} = \frac{\partial C}{\partial S} + \frac{\partial C}{\partial \overline{S}} \frac{d\overline{S}}{dS}$$

• Three regimes

T



# Stochastic volatility and jump diffusion yield relative smiles

#### How to Minimize the P(L) Variance?

• Given d and d,  $dC = \Delta dS + \Lambda dr$ 

Hedge with y shares:

P/L = dC - y dS

• In a stochastic volatility model  $Var(P/L) = (\Delta - y)^2 Var(ds) + 2(\Delta - y) \Lambda Cov(ds, dr) + \Lambda^2 Var(dr)$ Minimize with respect to y:  $y = \Delta + rs\Lambda/S < \Delta$ • Since  $\Delta = \Delta_{BS} - \Lambda \overline{s}^2(x)'/S$  $y = \Delta_{BS} + rs\Lambda/2S$ 

Optimal "risk management" delta <  $\Delta_{BS}$ 

#### What Is the Meaning of the Implied Tree?

- Imagine the world is described by a stochastic volatility model, but we hedge with the implied tree model
- Then the smile slope  $\frac{d\overline{s}}{dx} = \frac{rs}{4\overline{s}}$
- When we move the spot, keeping the implied tree fixed,

$$\frac{d\mathbf{\bar{s}}^{2}(x,T)}{dt} = \frac{1}{T} \int_{0}^{T} dt \ E_{BB} \left[ \frac{\partial \mathbf{\bar{s}}^{2}(\mathbf{x},t)}{\partial \mathbf{x}} \right|_{\mathbf{x}=\mathbf{x}_{BB}(t)} \right]$$

Thus

$$\Delta_{IT} = \Delta_{BS} + \mathbf{rs} \Lambda / 2S = y$$

# Implied tree delta mimicks the risk management delta

#### **Summary and Overview**

- Stochastic volatility and market jumps produce a skewed surface of implied volatilities
- The effect of volatility jumps on the skew is highly significant
- Perturbative expansions are a useful tool for understanding the smile
- The optimal delta depends on the dynamics of volatility