

Fixed Income Derivatives: Managing Smile Risk

Patrick S. Hagan

pathagan1954@yahoo.com

Martingales

- Martingales

$M(t) = M(t, *)$ = random process

M can depend on anything known by time t

- Given a probability measure Q , $M(t, *)$ is a Martingale iff

$$M(t, *) = E\{M(T, *) \mid t\} \quad \text{for all } T > t$$

↳ given all info known by time t

"for all times $T > t$, the average of $M(T)$ is $M(t)$ "



- Martingale representation theorem: If $M(t)$ is a Martingale then

$$dM = A(t, *) dW(t)$$

for some $A(t, *)$, where dW is Brownian motion under Q

Arbitrage free pricing

- Choose any positive, tradable security that has no cash flow as the *numeraire*. Then there is a probability measure Q such that $V(t, \cdot)/N(t, \cdot)$ is a Martingale for all tradable securities. Here $V(t, \cdot)$ is the security price and $N(t, \cdot)$ is the value of the numeraire:

$$\frac{V(t)}{N(t)} = E \left\{ \frac{V(T, \cdot)}{N(T, \cdot)} \mid t \right\} \quad \underline{\underline{\text{all } T > t}} \quad \begin{matrix} \text{no} \\ \text{cash flow} \end{matrix}$$

//

$$\frac{V(t)}{N(t)} = E \left\{ \left(\frac{V(T, \cdot)}{N(T, \cdot)} + \int_t^T \frac{C(t', \cdot)}{N(t', \cdot)} dt' \right) \mid t \right\} \quad \underline{\underline{\text{all } T > t}} \quad \begin{matrix} \text{cash flow} \\ C(t, \cdot) \end{matrix}$$

- complete markets: Harrison & Pliska et al – *hedging arguments* (BS)
- incomplete markets: *law of one price*

- $N(t) = e^{\int_0^t r(t') dt'}$ money market numeraire

$$V(t) = N(t) E \left\{ \frac{V(T)}{N(T)} \mid t \right\} = E \left\{ e^{-\int_t^T r(t') dt'} V(T) \mid t \right\} \quad \underline{\underline{\text{all } T > t}} \quad \begin{matrix} \text{risk neutral measure} \end{matrix}$$

- $N(t) = Z(t, T_{set})$ zero coupon bond, maturity T_{set}

$$V(t) = Z(t, \cancel{T_{set}}) E \left\{ \frac{V(T)}{Z(T, T_{set})} \mid t \right\} \quad \underline{\underline{\text{all } T > t}} \quad \begin{matrix} \text{forward measure} \\ \text{for } T_{set} \end{matrix}$$

Swaptions

- $$V_{sw}(t) = R_f \sum_{j=1}^n \alpha_j Z(t, t_j) + Z(t, t_n) - Z(t, t_0) = [R_f - R_s(t)] L(t)$$

receiver
 payer

$$R_s(t) = \frac{Z(t, t_0) - Z(t, t_n)}{L(t)}; \quad L(t) = \sum_{j=1}^n \alpha_j Z(t, t_j)$$

- $$V_{swptn}(t_{ex}) = [R_f - R_s(t_{ex})]^+ L(t_{ex}) \quad \text{at } t_{ex}$$

payoff

- Choose $N(t) = L(t) = \sum_{j=1}^n \alpha_j Z(t, t_j)$

$$\frac{V_{swptn}(0)}{L(0)} = E \left\{ \frac{V_{swptn}(T)}{L(T)} \mid 0 \right\} \quad \text{for any } T > 0$$

use $T = t_{ex}$

$$V_{swptn}(0) = L(0) E \left\{ [R_f - \underline{R_s(t_{ex})}]^+ \mid 0 \right\}$$

- $R_s(t)$ is a Martingale:

$$dR_s = A(t, *) dW \stackrel{\text{theory}}{=} \sigma R_s dW$$

$$L(0) = \sum_j \alpha_j D(t_j); \quad R_s^0 = \frac{D(t_0) - D(t_n)}{\sum_j \alpha_j D(t_j)}$$

Backwards Kolmogorov equation

- If $U(t, \mathbf{x}) = E\{F(\mathbf{X}(T) \mid \mathbf{X}(\frac{T}{t}) = \mathbf{x}\}$

where

$$dX_j = \mu_j(t, \mathbf{X})dt + a_j(t, \mathbf{X})dW_j \quad j = 1, 2, \dots$$

with

$$dW_j dW_k = \rho_{jk} dt$$

Then $U(t, \mathbf{x})$ is the solution of

$$\frac{\partial U}{\partial t} + \sum_j \mu_j(t, \mathbf{x}) \frac{\partial U}{\partial x_j} + \frac{1}{2} \sum_{jk} \rho_{jk} a_j(t, \mathbf{x}) a_k(t, \mathbf{x}) \frac{\partial^2 U}{\partial x_j \partial x_k} = 0$$

for $t < T$, with

$$U(t, \mathbf{x}) = F(\mathbf{x}) \quad \text{at } t = T$$

- $\tilde{V}(t, R) = E\{[R_f - R_s(t_{ex})]^+ \mid R_s(\frac{T}{t}) = R\}$

$$dR_s = \sigma R_s dW$$

So

$$\begin{cases} \frac{\partial \tilde{V}}{\partial t} + \frac{1}{2} \sigma^2 R^2 \frac{\partial^2 V}{\partial R^2} = 0 & \text{for } t < t_{ex} \\ \tilde{V} = [R_f - R]^+ & \text{at } t = t_{ex} \end{cases}$$

or

$$\tilde{V}(0, R_s^0) = R_f N(d_1) - R_s^0 N(d_2)$$

$$d_{1,2} = \frac{\log R_f / R_s^0 \pm \frac{1}{2} \sigma^2 t_{ex}}{\sigma t_{ex}^{1/2}}$$

Black's formula

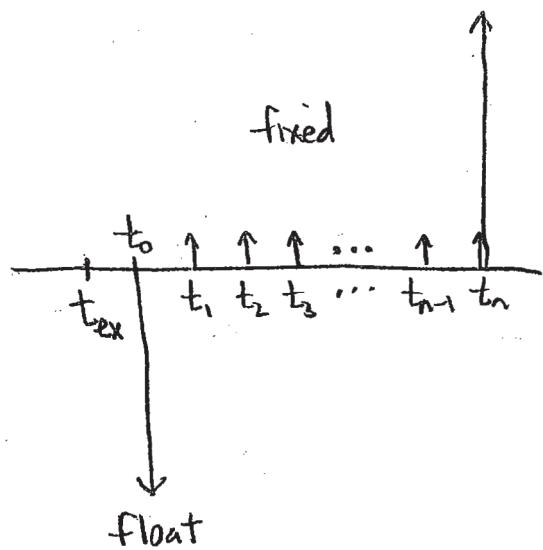
Swaption prices

$$V_{swptn}(0) = L_0 [R_f N(d_1) - R_s^0 N(d_2)]$$

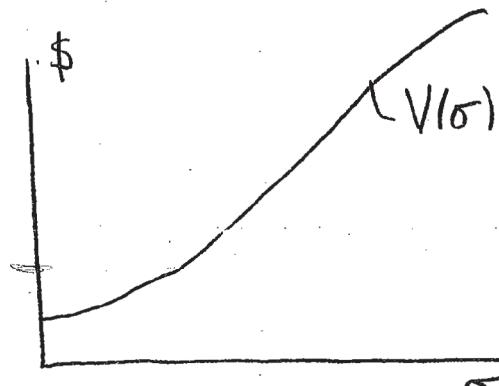
$$d_{1,2} = \frac{\log R_f / R_s^0 \pm \frac{1}{2} \sigma^2 t_{ex}}{\sigma t_{ex}^{1/2}}$$

$$L_0 = \sum_j \alpha_j D(t_j); \quad R_s^0 = \frac{D(t_0) - D(t_n)}{\sum_j \alpha_j D(t_j)}$$

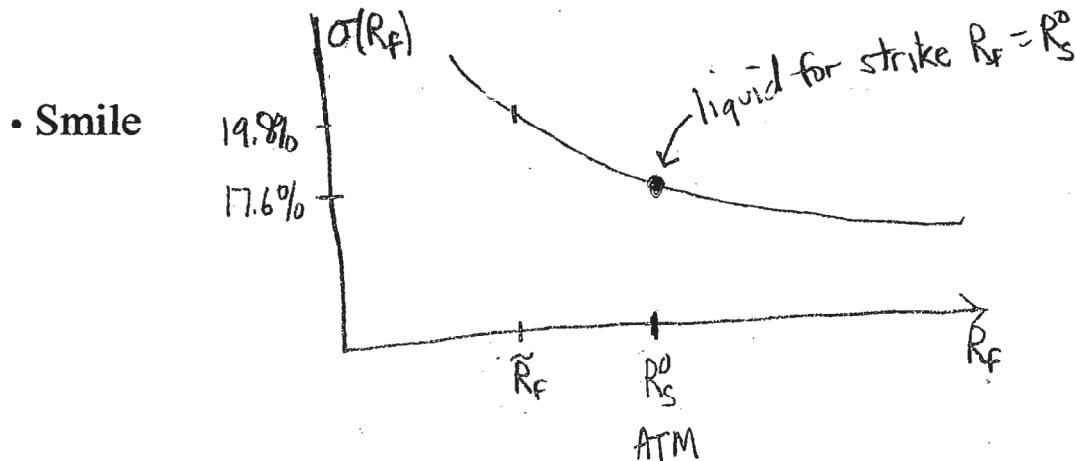
- Payer: $R_f \leq R_s^0$



- Market price quoted as σ
“3 into 7 receiver @ 8%”
“17.6% vol”

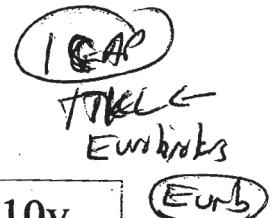


implied vol



Managing vol books

Prob Yonae



- Swaption/caplet vol matrix

$t_{ex} \setminus \text{tenor}$	3m	1y	2y	3y	...	10y
1m	0.182	0.180	0.17	0.176	...	0.160
3m	0.177	0.175	0.174	0.170	...	0.157
6m	0.173	0.172	0.174	0.165	...	0.155
1y	0.171	0.170	0.170	0.167	...	0.150
:	:	:	:	:	...	:
10y	0.164	0.161	0.160	0.158	...	0.142

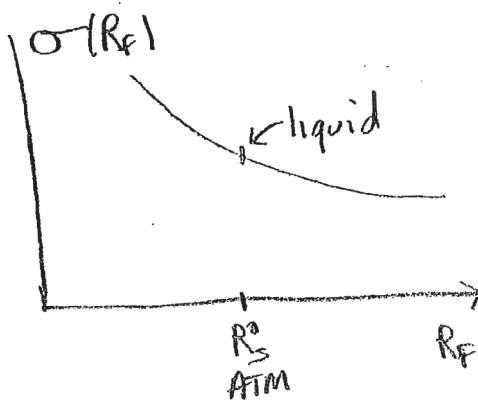
- vols are for ATM options ($R_f = R_s^0$) liquid
- broker quotes, options on EDF's
- interpolate for off-grid swaptions

- Find V_{book}

- bump vol, find $\Delta V_{book} \rightarrow$ buy/sell swaption \rightarrow vega neutral
- bump YC instrument (e.g., EDF), find $\Delta V_{book} \rightarrow$ buy/sell instrument

\rightarrow delta neutral

- off market ($R_f \neq R_s^0$)?

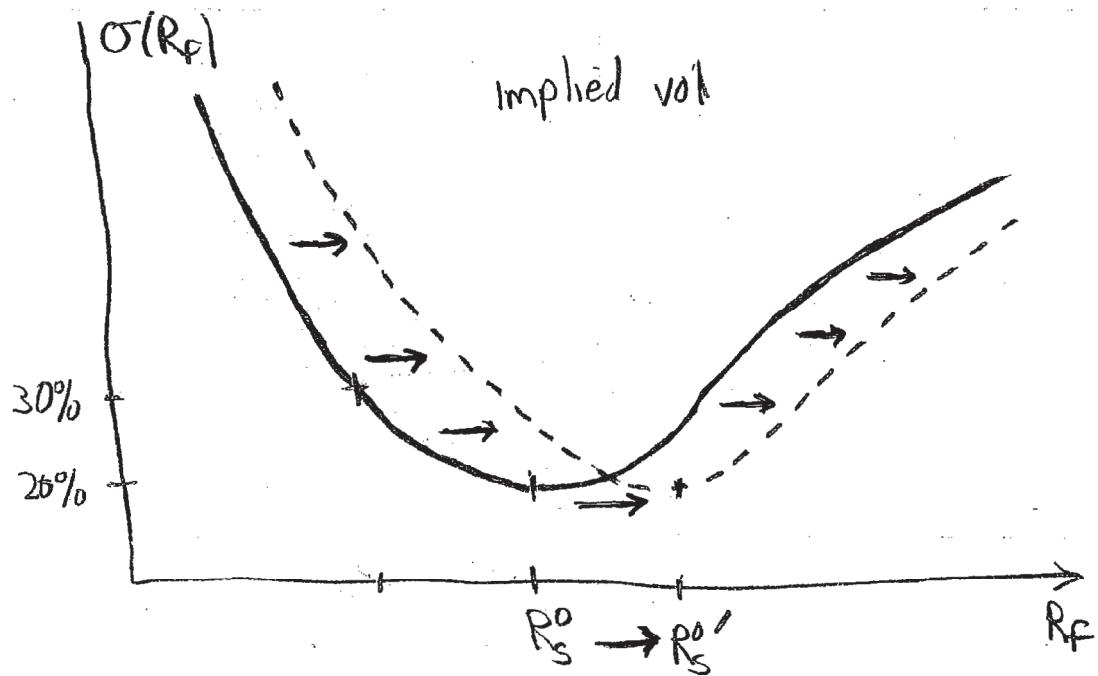
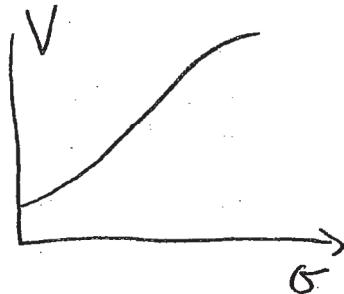


VolMat ID	USDVM.001	10/25/2001	USD	10/25/2001					
Term Tenor	3M	1Y	2Y	3Y	4Y	5Y	7Y	10Y	15Y
1M	5.25	12.25	13.50	14.13	14.50	14.75	14.50	14.25	13.25
3M	7.55	13.00	14.13	14.38	14.63	14.75	14.63	14.50	13.50
6M	11.44	14.25	14.88	15.00	15.13	15.25	15.00	14.75	13.50
9M	15.76	15.50	15.63	15.56	15.50	15.50	15.31	15.13	13.75
1Y	16.20	16.75	16.38	16.13	15.88	15.75	15.63	15.50	14.00
18M	16.78	17.25	16.75	16.56	16.31	16.13	15.88	15.63	14.13
2Y	19.25	17.75	17.13	17.00	16.75	16.50	16.13	15.75	14.25
3Y	19.25	17.75	17.13	17.00	16.75	16.50	16.13	15.63	14.00
4Y	19.00	17.50	16.88	16.63	16.38	16.13	15.63	15.13	13.63
5Y	18.50	17.38	16.50	16.13	15.88	15.50	15.13	14.63	13.13
7Y	16.25	15.38	14.88	14.38	14.13	13.88	13.50	13.25	11.88
10Y	14.00	13.50	13.00	12.50	12.25	12.00	11.38	11.00	9.88
15Y	11.00	11.00	10.75	10.63	10.50	10.25	9.88	9.38	8.75

Implied vols

- $V(R_s^0) = L_0[R_f N(d_1) - R_0 N(d_2)]$

$$d_{1,2} = \frac{\log R_f/R_s^0 + \frac{1}{2}\sigma^2 t_{ex}}{\sigma t_{ex}^{1/2}}$$



- Different model at each strike R_f

$\text{vega} = \frac{\partial}{\partial \sigma} V(0, R_s^0, R_f)$ hedge all vega with 1 ATM swaption expensive

$\Delta = \frac{\partial}{\partial R_s^0} V(0, R_s^0, R_f)$ hedge all Δ with 1 ATM swaption cheap

$R_s^0 \rightarrow R_s^0 + \delta, \quad \sigma \rightarrow \sigma + ?$ systematic change in σ with R_s^0

- Need single model for all strikes

Local vol models (Dupire/Derman-Kani)

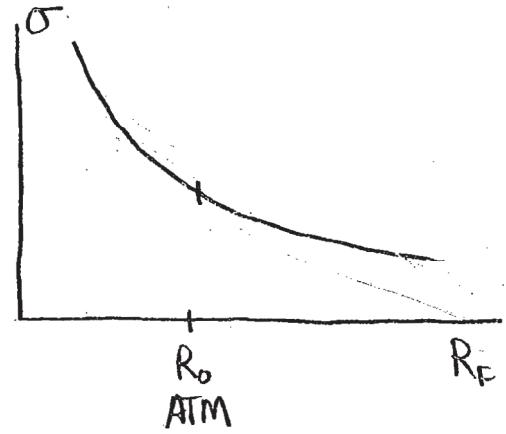
- $V(R_s^0, R_f) = L_0 E \{ [R_f - R_s(t_{ex})]^+ \mid R_s(0) = R_s^0 \}$

$$dR_s = C(t, *) dW \stackrel{\text{theory}}{\underset{\text{model}}{\leftarrow\rightarrow}} A(t, R_s) R_s dW = A(R_s) R_s dW$$

new model

- given $A(R_s)$, solve for $V(R_s^0, R_f)$
- find implied vol $\sigma_{imp}(R_f)$ from $V(R_s^0, R_f)$
- vary $A(R_s)$ until $\sigma_{imp}(R_f)$ matches market for all R_f

single self-consistent model for all strikes R_f



- Singular perturbation analysis:

$$V(R_s^0, R_f) = L_0 [R_f N(d_1) - R_s^0 N(d_2)]$$

$$d_{1,2} = \frac{\log R_f / R_s^0 + \frac{1}{2} \sigma^2 t_{ex}}{\sigma t_{ex}^{1/2}}$$

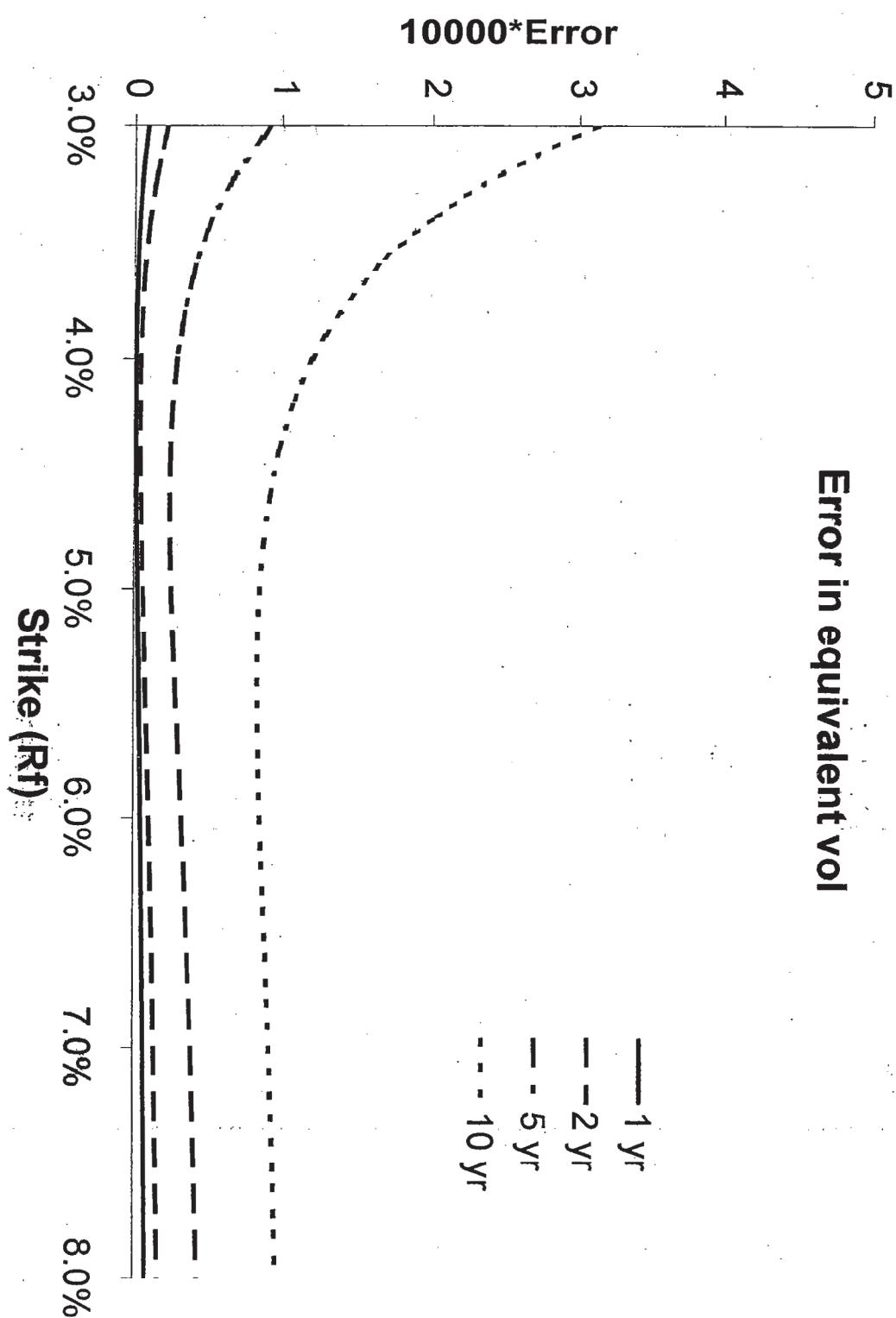
with

$$\sigma_{imp}(R_s^0, R_f) = A(R_{av}) \left\{ 1 + \frac{1}{24} (\gamma_2 - 2\gamma_1^2 - 2\gamma_1) \frac{(R_s^0 - R_f)^2}{R_{av}^2} + \frac{1}{24} (2\gamma_2 - \gamma_1^2 + 2\gamma_1) A^2 (R_{av}) t_{ex} + \dots \right\}$$

\leftarrow small: 1% or less

$$R_{av} = \frac{1}{2} (R_f + R_s^0); \quad \gamma_1 = \frac{R_{av} A'(R_{av})}{A(R_{av})}; \quad \gamma_2 = \frac{R_{av}^2 A''(R_{av})}{A(R_{av})}$$

Error in equivalent vol



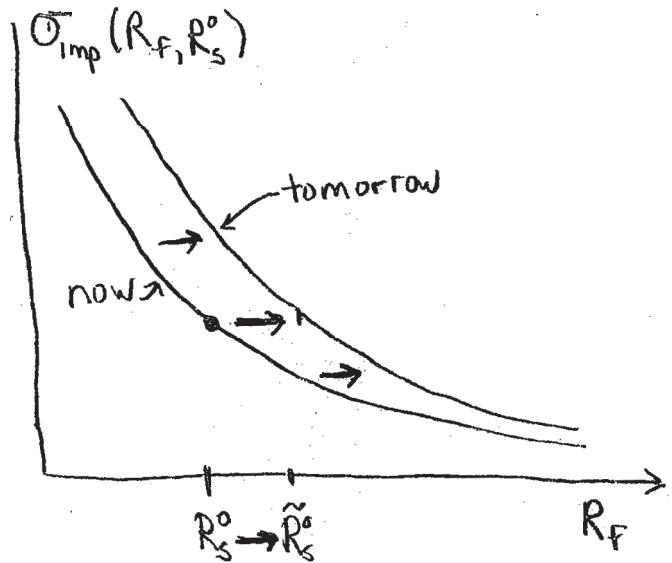
units: 1 on vertical axis \Rightarrow 20.00% \rightarrow 20.01%

Reprise

$$dR_s = aA(R_s)R_sdW$$

$$V(R_s^0, R_f) = BS(t_{ex}, R_s^0, R_f, \sigma_{imp})$$

$$\sigma_{imp}(R_s^0, R_f) = aA(R_{av})\{1 + \dots\}$$



- calibration

$$\sigma_{market}(R_f) = aA(\frac{1}{2}(R_s^0 + R_f)\{1 + \dots\}$$

so

$$aA(R_s) = \sigma_{market}(2R_f - R_s^0)\{1 - \dots\}$$

iterate

- hedging

$$\frac{\partial V}{\partial R_s^0} = \frac{\partial BS}{\partial R_s^0} + \frac{\partial BS}{\partial \sigma_{imp}} \frac{\partial \sigma_{imp}(R_s^0, R_f)}{\partial R_s^0}$$

$$\begin{array}{lll} \text{new } \Delta & \text{old } \Delta & \text{old vega} \cdot \text{systematic shift} \\ \hline \end{array}$$

new term

for each R_f

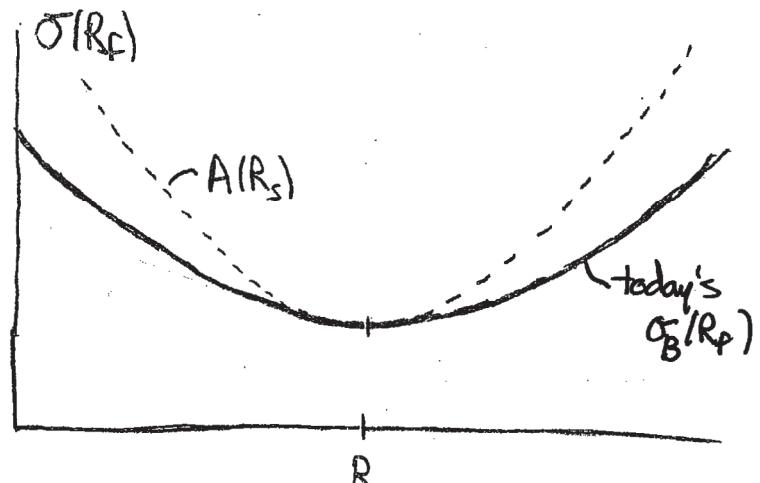
$$\frac{\partial V}{\partial a} = \frac{\partial BS}{\partial \sigma_{imp}} \frac{\partial \sigma_{imp}(R_s^0, R_f)}{\partial a}$$

$$\begin{array}{lll} \text{new } \text{vega} & \text{old } \text{vega} & \text{scaling factor} \\ \hline \end{array}$$

for each R_f

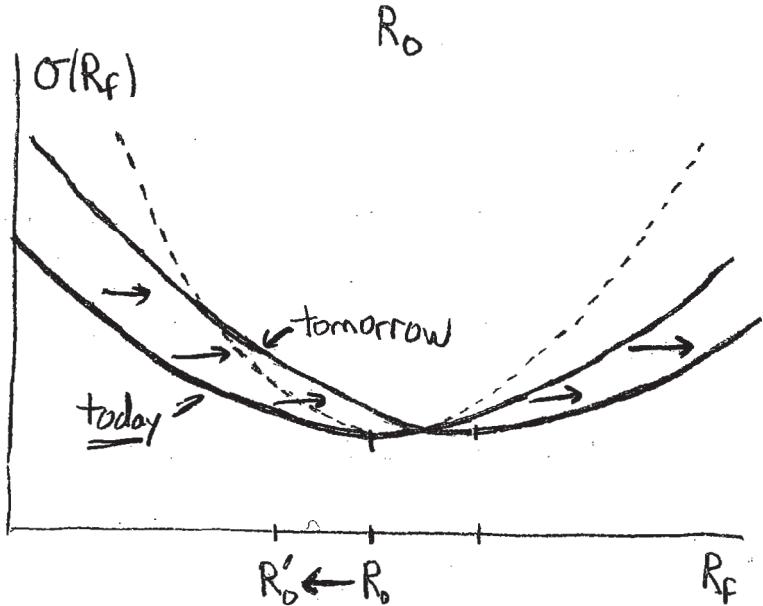
Does it work

$dR_s = A(R_s)R_sdW$
pick $A(R_s)$ to match
today's smile



tomorrow predict:

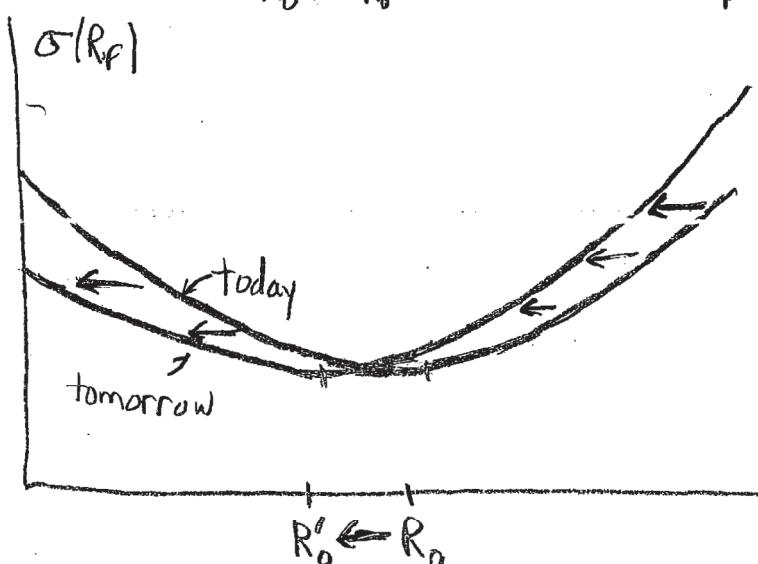
if $R_s^0 \rightarrow R_s^0 + \delta$ then
the smile shifts the
other way by δ



experience says

Only reasonable models are
 $A(R_s) = a(t)R_s^\beta$

scaling



Stochastic $\alpha\beta\rho$ model

- Fundamental theory

$$V = L \mathbb{E}\{[R_s(t_{ex}) - R_f]^+ | R_s(0) = R_s^0\}$$

$$dR_s = A(*)dW$$

- 1 factor Markovian models failed

- non-Markovian models: too messy
- use 2 factor Markovian model
- second factor: make vol stochastic
account for vega risk

- stochastic $\alpha\beta\rho$ model

$$dR_s = aR_s^\beta dW_1$$

$$da = vadW_2$$

$$dW_1 dW_2 = \rho dt$$

- first factor: CEV model for rates

$$\beta = 0? \quad \beta = \frac{1}{2}? \quad \beta = 1?$$

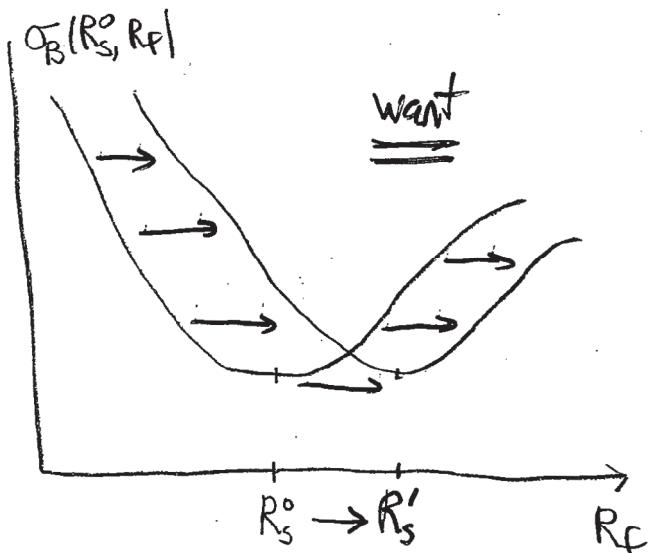
- stochastic vol produces smile (*adverse selection*)

- correlate stochastic vol and rates
use correlation to set skew

SABR model

- SABR

$$\begin{aligned} dR_s &= \alpha R_s^\beta dW_1 \\ d\alpha &= \nu \alpha dW_2 \\ dW_1 dW_2 &= \rho dt \end{aligned}$$



- option prices: Black's formula with

$$\sigma_B \approx \frac{\alpha}{(R_s^0)^{1-\beta}} \left\{ 1 - \underbrace{(1-\beta-\rho\lambda)}_{\substack{\text{ATM vol} \\ \text{beta skew}}} \frac{R_f - R_s^0}{2R_s^0} + \underbrace{(2-3\rho^2)\lambda^2}_{\substack{\text{Vanna skew}}} \frac{(R_f - R_s^0)^2}{12(R_s^0)^2} + \dots \right.$$

volga smile

$$\lambda = \frac{\nu(R_s^0)^{1-\beta}}{\alpha} = \frac{\text{vol of vol}}{\text{ATM vol}}$$

- implied vol is very rough ...

Equivalent vol for the SABR model

- Accurate formula for all strikes R_f :

$$V(R_s^0, R_f) = BS(R_s^0, R_f, \sigma_B)$$

with

$$\sigma_B(R_s^0, R_f) = \frac{\alpha}{(R_s^0 R_f)^{(1-\beta)/2} \left\{ 1 + \frac{(1-\beta)^2}{24} \log^2 R_s^0 / R_f \right\}} \cdot \frac{z}{x(z)} \cdot$$

$$\left\{ 1 + [(1 - \beta)^2 + 6\rho\beta\lambda + (2 - 3\rho^2)\lambda^2] \frac{\alpha^2 t_{ex}}{24(R_s^0 R_f)^{1-\beta}} + \dots \right.$$

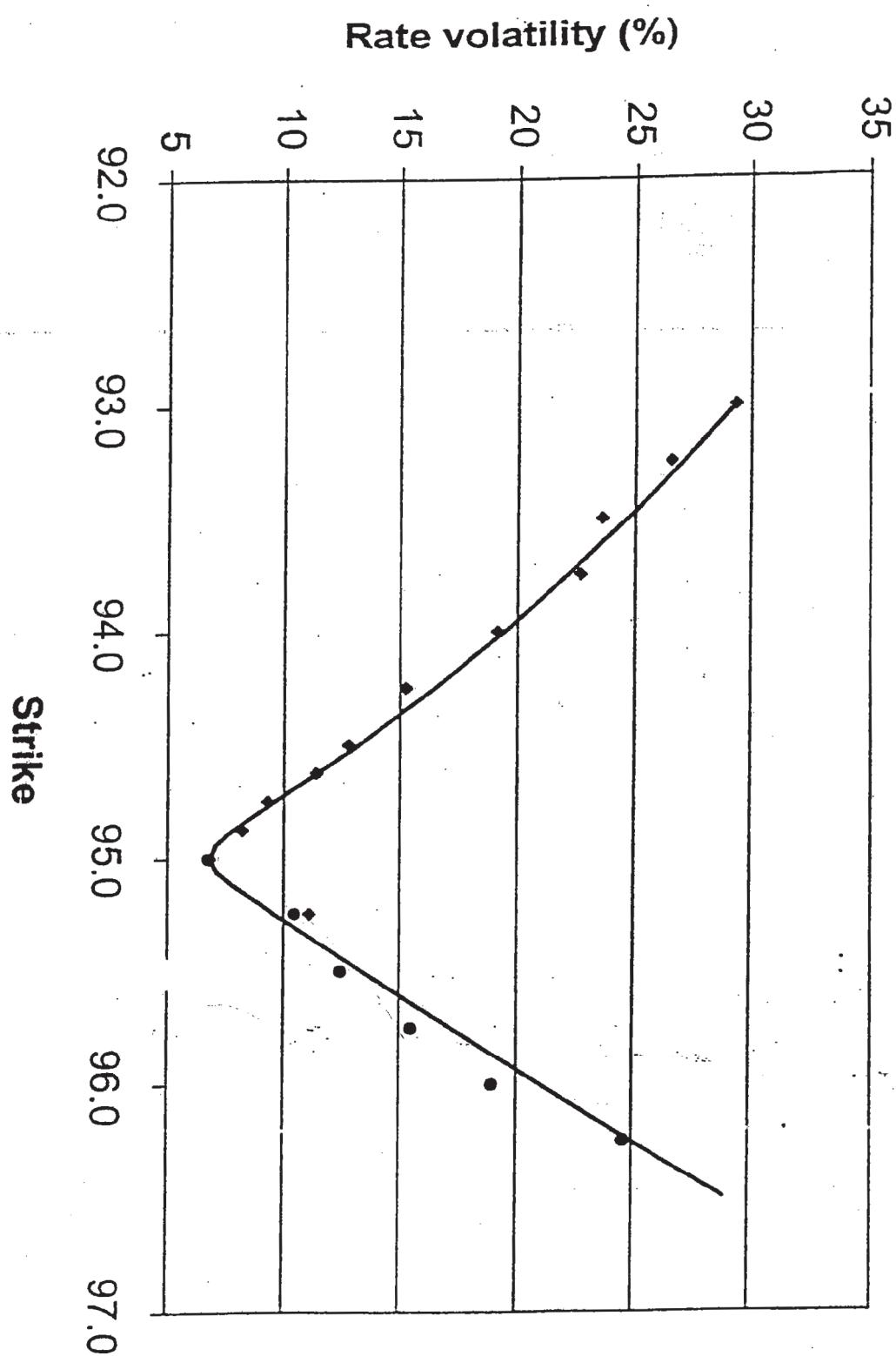
$$\lambda = \frac{\nu}{\alpha} (R_s^0 R_f)^{(1-\beta)/2} = \frac{\text{"vol of vol"}}{\text{"rate vol"}}$$

$$z = \frac{\nu}{\alpha} (R_s^0 R_f)^{(1-\beta)/2} \log R_s^0 / R_f$$

$$x(z) = \log \frac{z - \rho + \sqrt{1 - 2\rho z + z^2}}{1 - \rho} \Leftrightarrow z = \sinh x - \rho(\cosh x - 1)$$

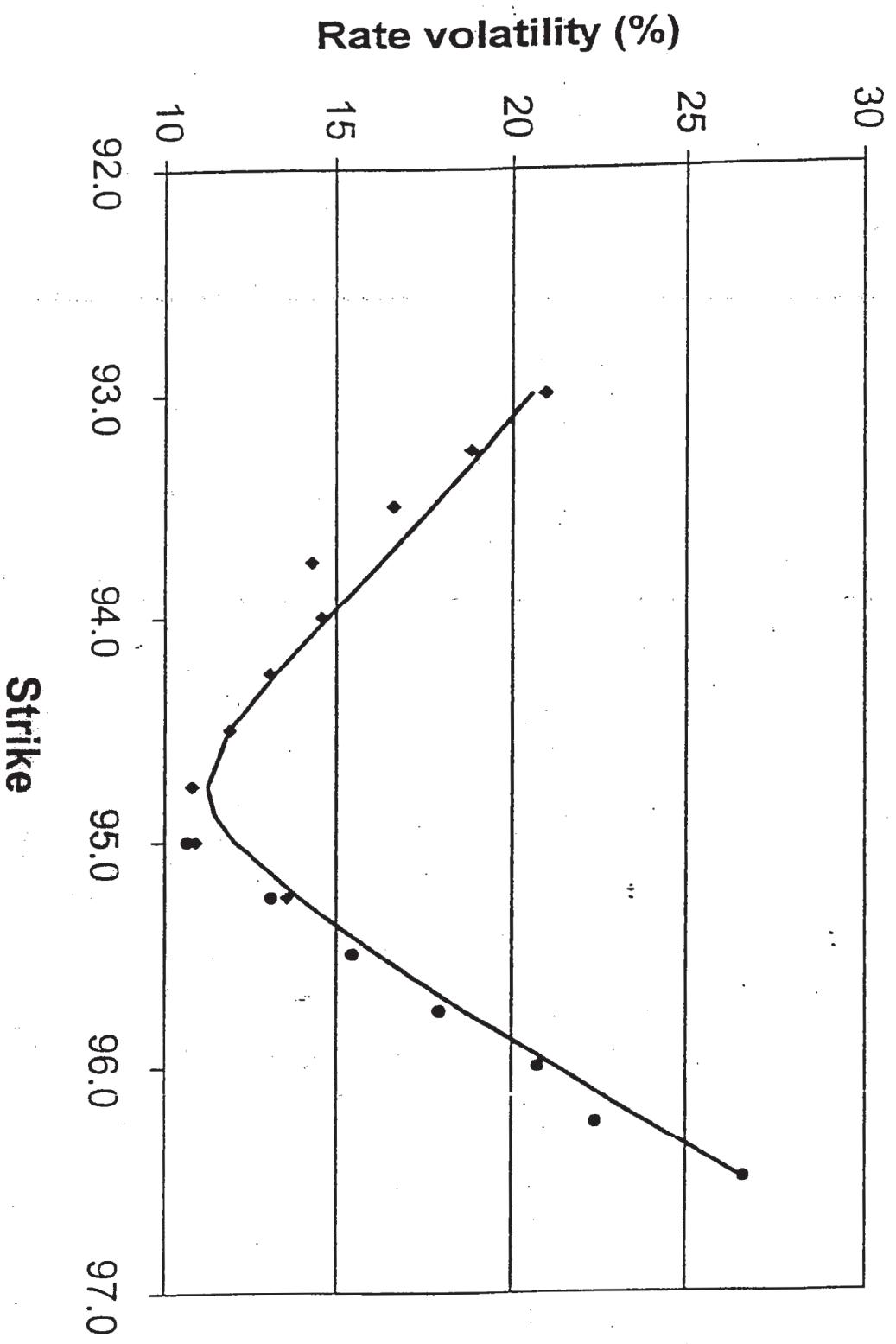
– closed form; extremely fast

M99 Eurodollar option

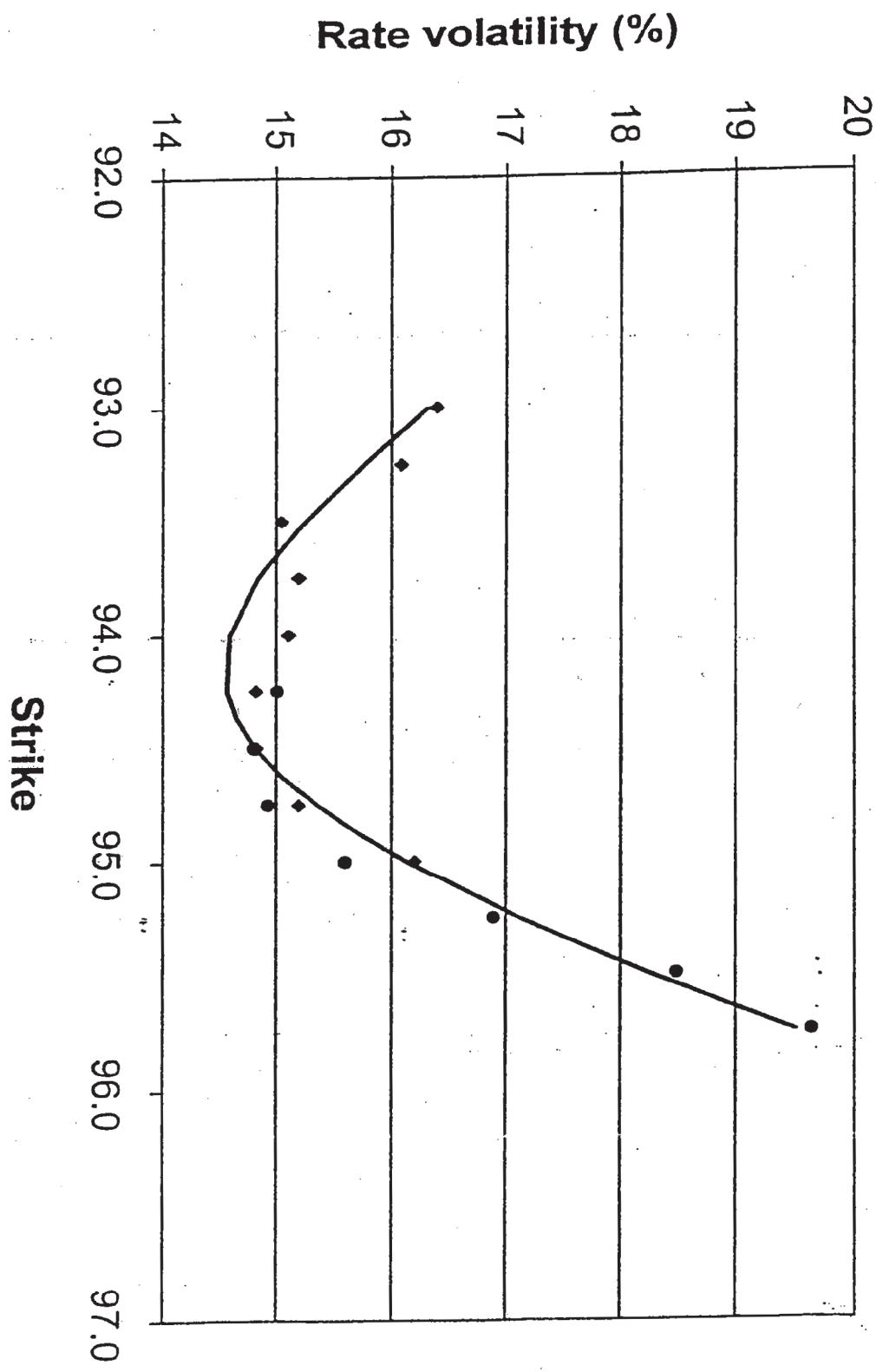


N U M E R I X

U99 Eurodollar option

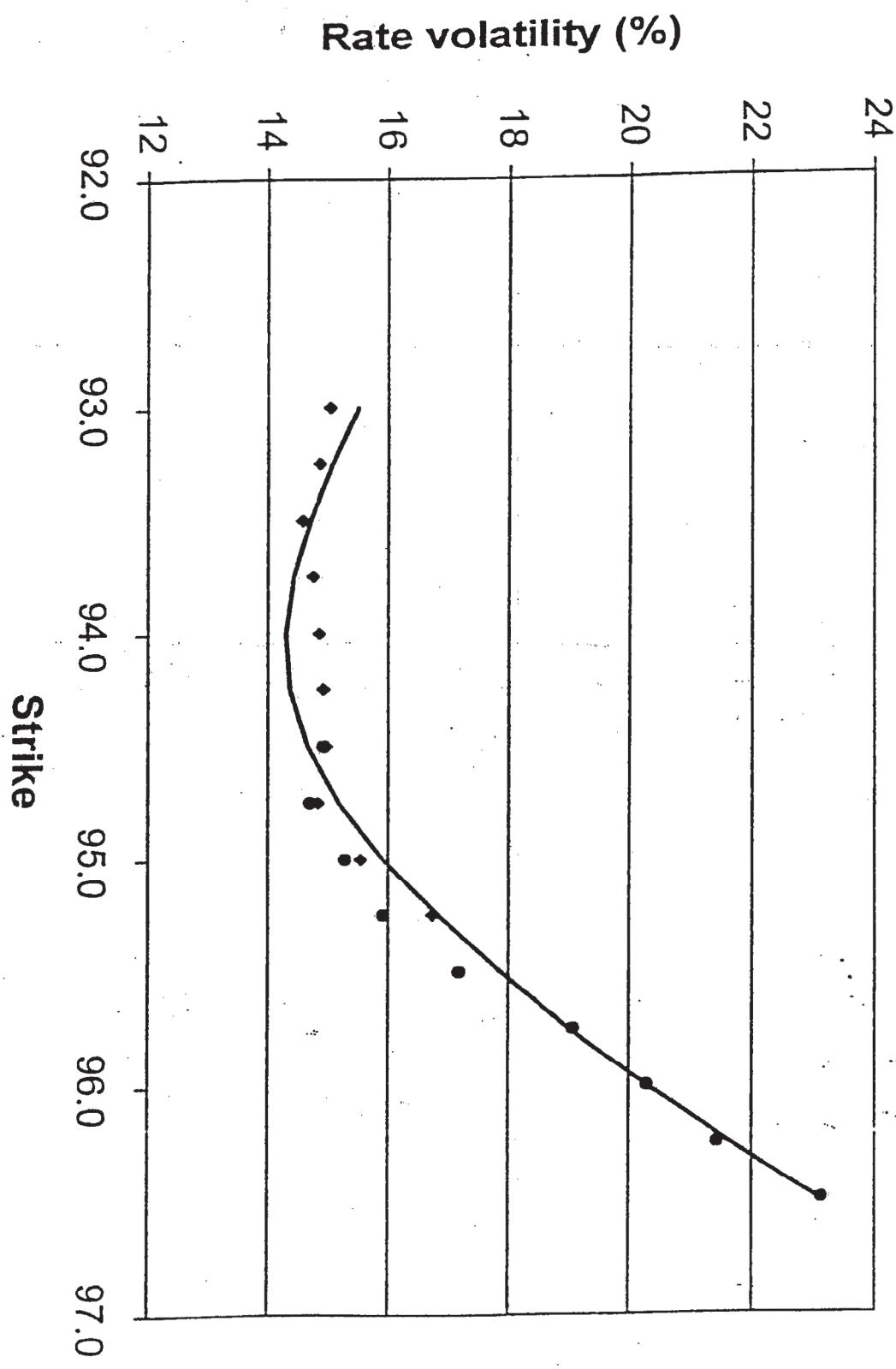


Z99 Eurodollar option



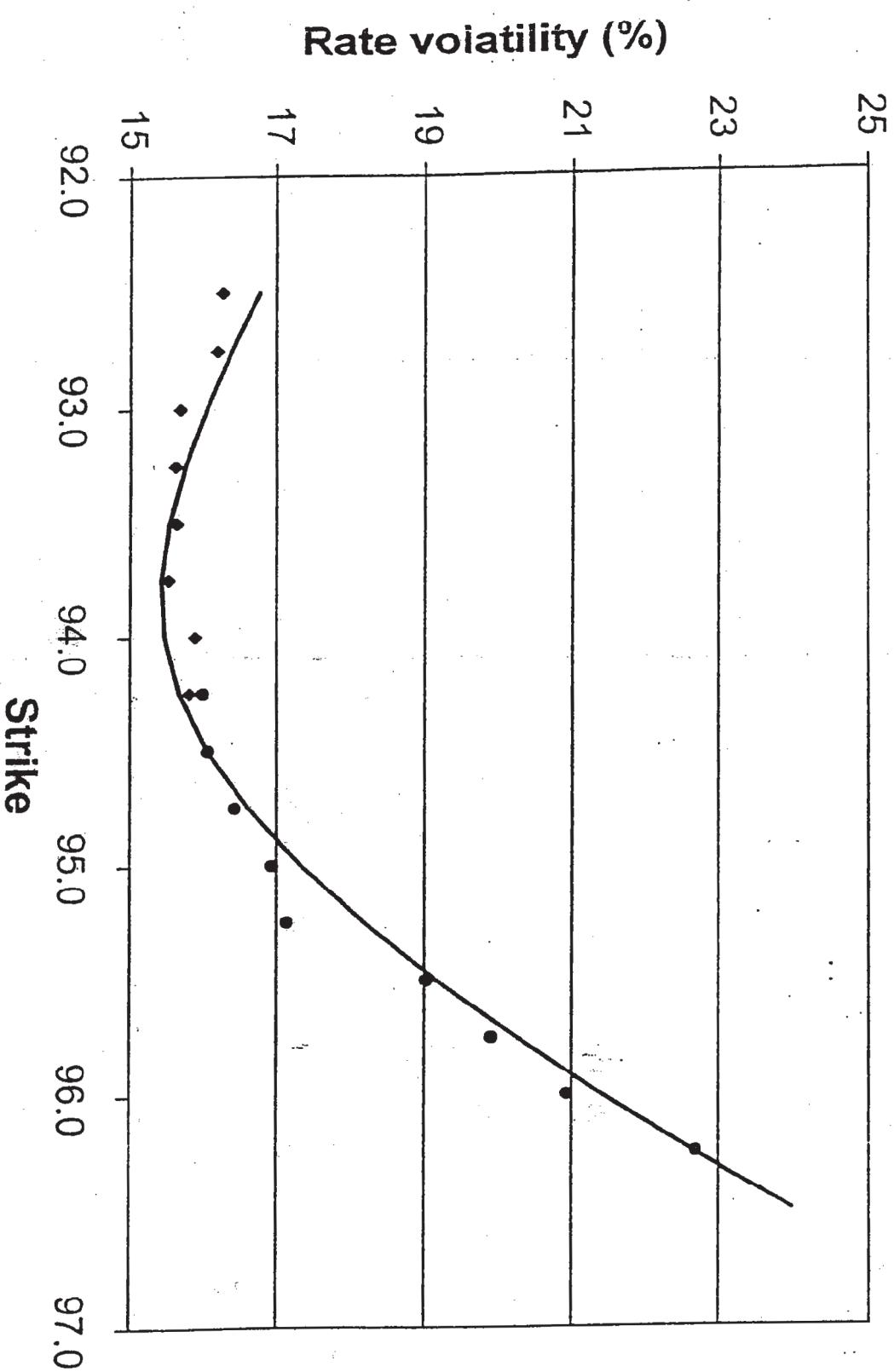
N U M E R I C A

H00 Eurodollar option

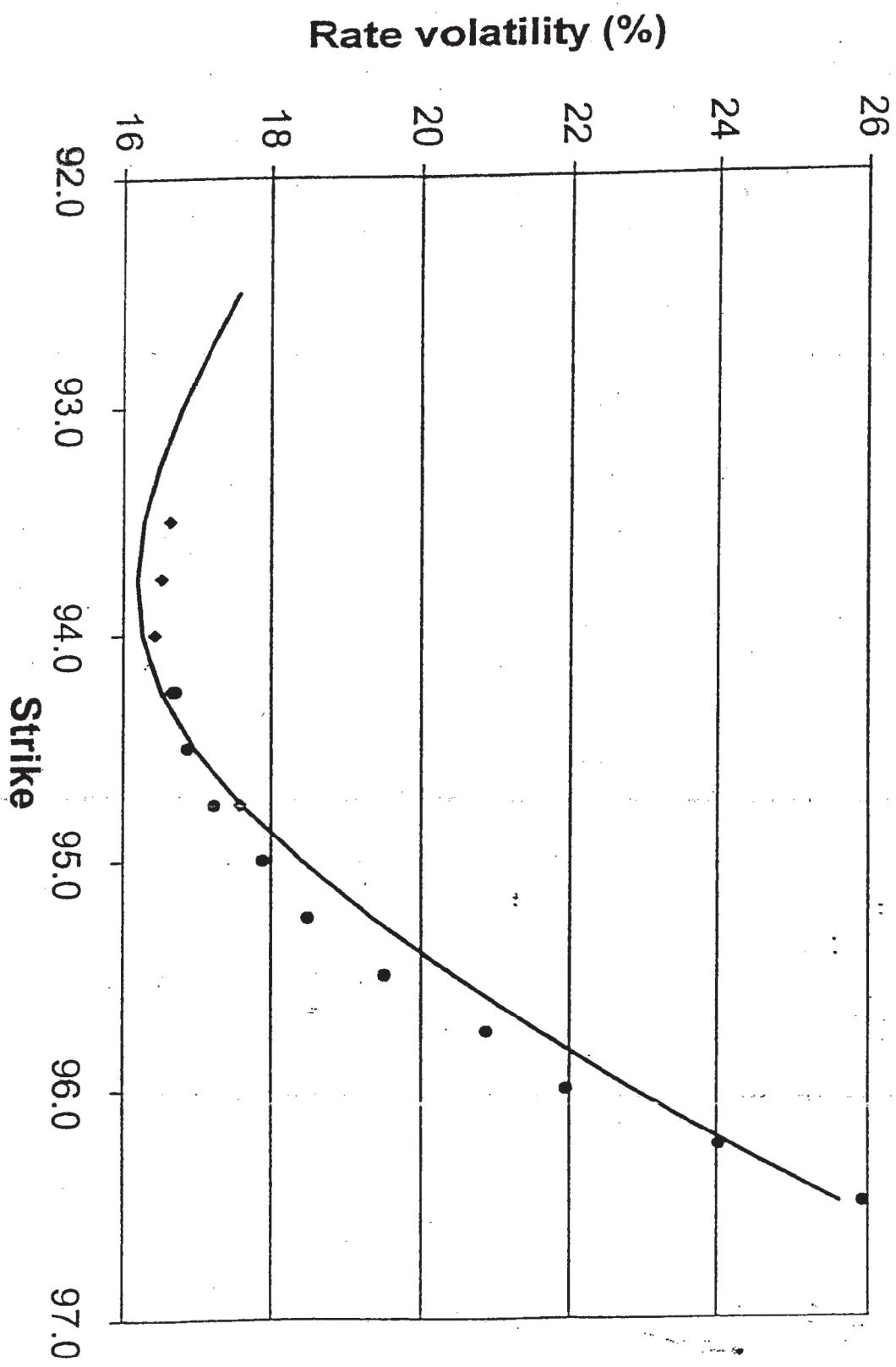


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M00 Eurodollar option

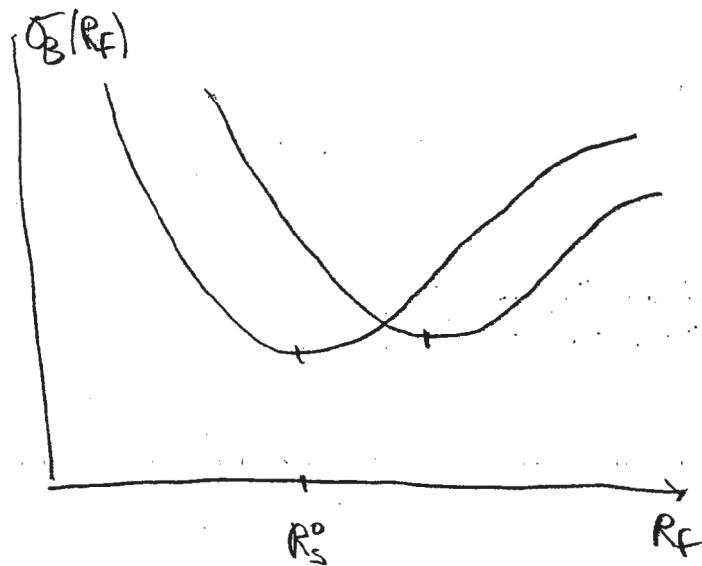


U00 Eurodollar option

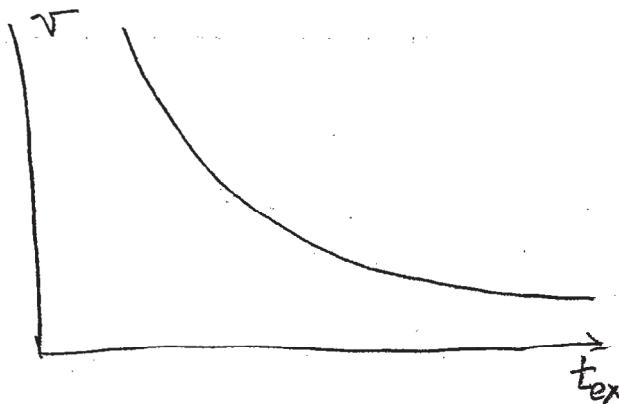


Does it work?

- SABR model
 - matches fixed income market
Liber options, swaptions, caplets/floorlets, treasury options
 - gives stable hedges
 - backbone & skew can be set independently



- Vol of vol vs. t_{ex}
suggests jump model (or truncated Levy model!)



Fitting parameters

- Smile/backbone:

$$\sigma \approx \sigma_{ATM} \left\{ 1 - (1 - \beta - \frac{\rho\nu}{\sigma_{ATM}}) \frac{R_f - R_0}{2R_0} + (2 - 3\rho^2) \frac{\nu^2 (R_f - R_0)^2}{\sigma_{ATM}^2 R_0^2} + \dots \right.$$

with

$$\sigma_{ATM} = \frac{\alpha}{R_0^{(1-\beta)}}$$

- α or σ_{ATM} controls the ATM vol;
- ρ and β both control the skew, but β controls the “backbone”
- ν controls the smile
- typically β is set from $\log \sigma_{ATM}$ vs $\log R_0$ studies ($\frac{1}{3}$, $\frac{1}{2}$, or 1)
- then ρ and ν are fit to the smile
- typically α or σ_{ATM} are updated daily or every few hours,
- ρ and ν are re-fitted every month or as needed
- β doesn't change

Managing smile risk

- Swaption prices are now from 4 matrices

σ_{ATM}	3m	1y	2y	3y	...	10y
1m	5.25%	12.25%	13.50%	14.13%	...	14.25%
3m	7.55%	13.00%	14.13%	14.38%	...	14.50%
6m	11.44%	14.25%	14.18%	15.00%	...	14.75%
1y	16.20%	16.75%	16.38%	16.13%	...	15.50%
:	:	:	:	:	⋮	⋮
10y	14.00%	13.50%	13.00%	12.50%	...	11.00%

- similar matrices for β, ρ, ν
- parameter values are found by interpolating on matrix entries

• Risks

- books have swaptions at all different strikes
- exotic options give rise to risk at various strikes

Bermudan struck at 7.132% is calibrated to swaptions struck at 7.132%. Leads to vega risk at 7.132%

- bumping each parameter in turn, and re-valuing the book yields
 - a matrix of σ_{ATM} risks (vega),
 - a matrix of ρ risks (vanna)
 - a matrix of ν risks (volga)

Levy flights

$$dF = A(t, F)dt + B(t, F)dZ$$

- What is dZ ?
 - stationary, Markovian, independent increment
- Transition density:

$$g(T-t, y-x)dy = \text{prob}\{y < Z(T) \leq y+dy \mid Z(t) = x\}$$

$$g(t, x) = \int_{-\infty}^{\infty} g(\tau, y)g(t-\tau, x-y)dy$$

- Fourier transform:

$$G(t, k) \equiv \int_{-\infty}^{\infty} e^{-ikx} g(t, x)dx$$

$$G(t, k) = G(\tau, k)G(t-\tau, k) \quad \text{for each } k \text{ and } 0 \leq \tau \leq t$$

- For each k , solving the functional equation yields

$$G(t, k) = e^{-t\phi(k)} \quad \text{for some } \phi(k)$$

- *No intrinsic time scale for dZ* \Rightarrow similarity solution!

$$g(t, x) = \frac{1}{t^{1/\alpha}} h(x/t^{1/\alpha})$$

$$G(t, k) = \begin{cases} e^{-c_+ k^\alpha t} \\ e^{-c_- |k|^\alpha t} \end{cases} \quad \text{for some } c_- = c_+^*$$

Truncated Levy distributions

$$g(t, x) = \frac{1}{t^{1/\alpha}} h(x/t^{1/\alpha}) \quad \text{for } 1 \leq \alpha < 2$$

$$G(t, k) = \begin{cases} e^{-c_+ k^\alpha t} & \text{for } k > 0 \\ e^{-c_- |k|^\alpha t} & \text{for } k < 0 \end{cases} \quad \text{for some } c_- = c_+^*$$

$$g(t, x) \sim C_\pm \frac{t}{|x|^{1+\alpha}} \quad \text{no variance!}$$

– experimentally, $\alpha \sim 1.4$

- $g(t, x)$ and option values $V(t, x)$ follow same equation

$$\frac{\partial V(t, x)}{\partial t} = \int K(y) \left\{ V(t, x - y) - V(x) + y \frac{\partial V(t, x)}{\partial y} \right\} dy$$

$$K(y) = \frac{C_\pm}{|y|^{1+\alpha}} \quad \text{for } y \geq 0$$

– $K(y)$ = rate of jumps of size y

– no convergence at $t = t_{\text{exp}}$

- truncate: $K(y) = \frac{C_\pm e^{-\varepsilon|y|}}{|y|^{1+\alpha}}$ for $y \geq 0$

– core spreads out at at^γ , wings spread out slower, at $At^{1/2}$

– for $at^\gamma \ll (t/\varepsilon)^{1/2}$, strong smile/skew (Levy-like)

– for $at^\gamma \gg (t/\varepsilon)^{1/2}$, diffusion like (Brownian motion)

cross-over ~ 3 months

– three parameters: $c_+ = c_-^* = ce^{-i\eta}$, ε

intensity c governs atm vol, asymmetry η governs skew

ε governs both smile and crossover

5.3 Models Beyond Geometric Brownian

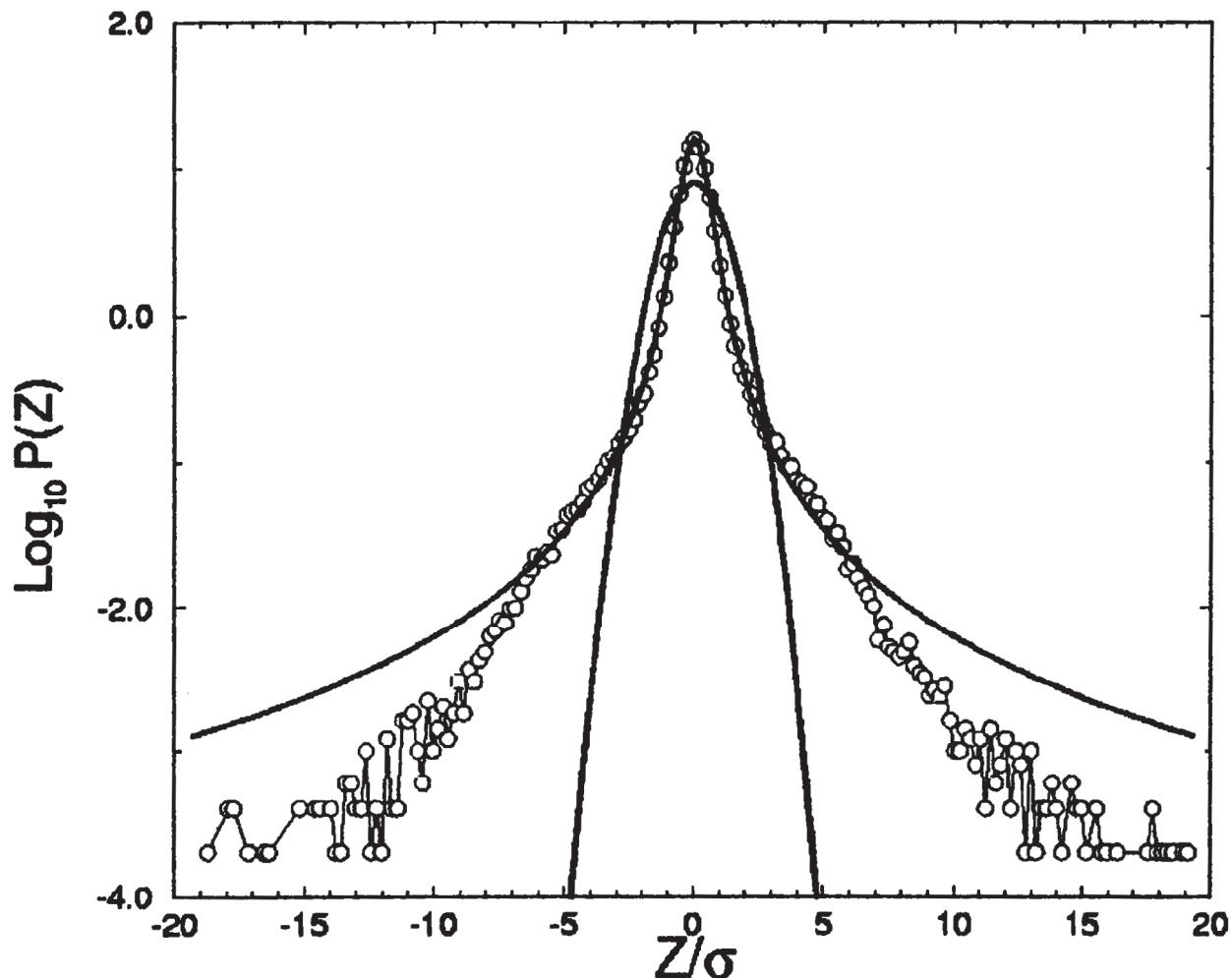


Fig. 5.6. Probability distribution $p_{\Delta t}(\ell)$ of the price variations ℓ [= for the S&P500, determined from all records between 1984 and million records). $p_{\Delta t}(\ell)$ [denoted by $P(Z)$] is plotted versus ℓ/σ where σ ($= 0.0508$) is the standard deviation calculated from the Gaussian distribution (*thick solid line*) corresponding to this value to the best fit with a Lévy distribution (*thin solid line*). The Lévy $\alpha = 1.4$ and $c = 3.75 \times 10^{-3}$, see (5.83)] gives a much better rep data for $\ell/\sigma \leq 6$. If $\ell/\sigma \geq 6$, the distribution of the index deca exponentially. Reproduced with permission from [105]

which decays from 1 (for $t = 0$) to 0 (for $t \rightarrow \infty$). This be zero for all $t > 0$ if the hypothesis were true. In practice