

# **Arbitrage free SABR**

Patrick S. Hagan, Mathematical Institute, Oxford University

Deep Kumar, AVM, L.P., Boca Raton, FL

Andrew L. Lesniewski, Dept. of Mathematics, Baruch College

Diana E. Woodward, Gorilla Science, London, UK

[pathagan1954@yahoo.com](mailto:pathagan1954@yahoo.com)

# SABR model

- SABR models developed to manage skew/smile risk:

$$d\tilde{F} = \tilde{\alpha}C(\tilde{F})dW_1,$$

$$d\tilde{\alpha} = v\tilde{\alpha}dW_2,$$

with

$$dW_1dW_2 = \rho dt$$

- Asymptotic analysis yields approximate formulas for the implied normal volatility

– several variations

– of all the  $O(\varepsilon^2)$ -accurate formulas, our favorite is

$$\sigma_N(K) = \frac{\alpha(f-K)}{\int_K^f \frac{df'}{C(f')}} \cdot \left( \frac{\zeta}{x(\zeta)} \right) \cdot \left\{ 1 + \left[ g\alpha^2 + \frac{1}{4}\rho v\alpha \frac{C(f)-C(K)}{f-K} + \frac{2-3v^2}{24} \right] \tau_{ex} + \dots \right\}$$

with

$$\zeta = \frac{v}{\alpha} \int_K^f \frac{df'}{C(f')}, \quad x(\zeta) = \log \left( \frac{\sqrt{1-2\rho\zeta+\zeta^2} - \rho + \zeta}{1-\rho} \right)$$

$$g = \log \left( \frac{1}{f-K} \int_K^f \frac{\sqrt{C(f)C(K)}}{C(f')} df' \right) \bigg/ \left( \int_K^f \frac{df'}{C(f')} \right)^2$$

– not the simplest, but seems to be the most robust

# CEV Backbone

- Most common case is  $C(F) = F^\beta$

– implied normal vol:

$$\sigma_N(K) = \frac{\alpha(1-\beta)(f-K)}{f^{1-\beta} - K^{1-\beta}} \cdot \left( \frac{\zeta}{x(\zeta)} \right) \cdot \left\{ 1 + \left[ g\alpha^2 + \frac{1}{4}\rho\nu\alpha \frac{f^\beta - K^\beta}{f-K} + \frac{2-3\nu^2}{24} \right] \tau_{ex} + \dots \right\}$$

with

$$\zeta = \frac{\nu}{\alpha} \frac{f^{1-\beta} - K^{1-\beta}}{1-\beta}, \quad x(\zeta) = \log \left( \frac{\sqrt{1-2\rho\zeta + \zeta^2} - \rho + \zeta}{1-\rho} \right)$$

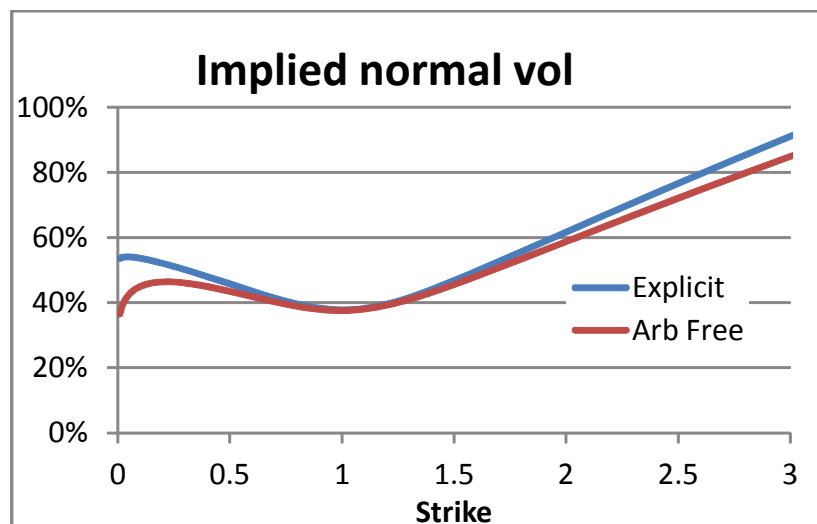
$$g = \frac{(1-\beta)^2}{(f^{1-\beta} - K^{1-\beta})^2} \log \left( (fK)^{\beta/2} \frac{f^{1-\beta} - K^{1-\beta}}{(1-\beta)(f-K)} df' \right)$$

# Arbitrage in the SABR model

- Explicit implied vols  $\sigma_N(K)$  are usually treated as *exact*
  - don't view  $\sigma_N(K)$  as an *approximate* solution to the SABR model
  - view  $\sigma_N(K)$  as the *exact solution* to some *other model* which is approximated by the SABR model
  - For “other model” to be arbitrage free, need:
    - put-call parity (*automatic from using implied vols*)
    - option prices must imply positive probability densities:

$$\frac{\partial^2}{\partial K^2} V = \frac{\partial^2}{\partial K^2} \left( \int_K^\infty (F - K) p(\tau_{ex}, F) dF \right) = p(\tau_{ex}, K) \geq 0 \quad \text{for all } K$$

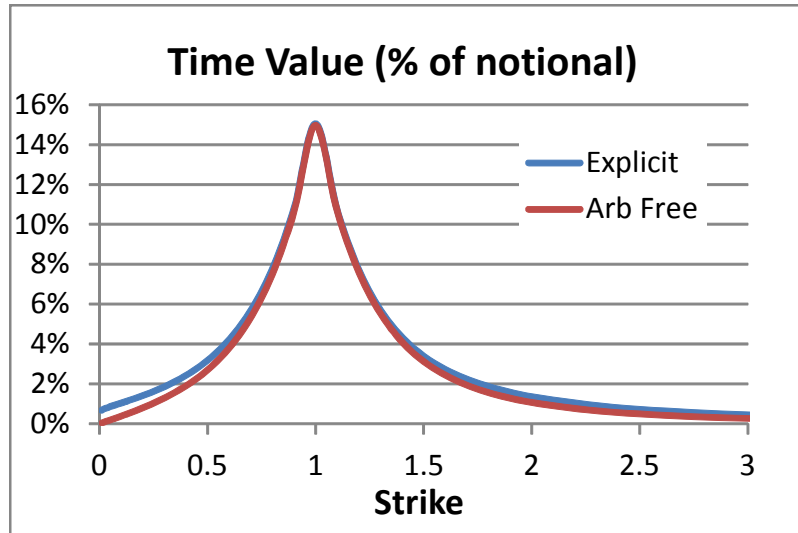
- Can be violated for low strikes, even for nice smiles:



$$\alpha = 35\%, \beta = 25\%, \rho = -10\%, \nu = 100\%$$

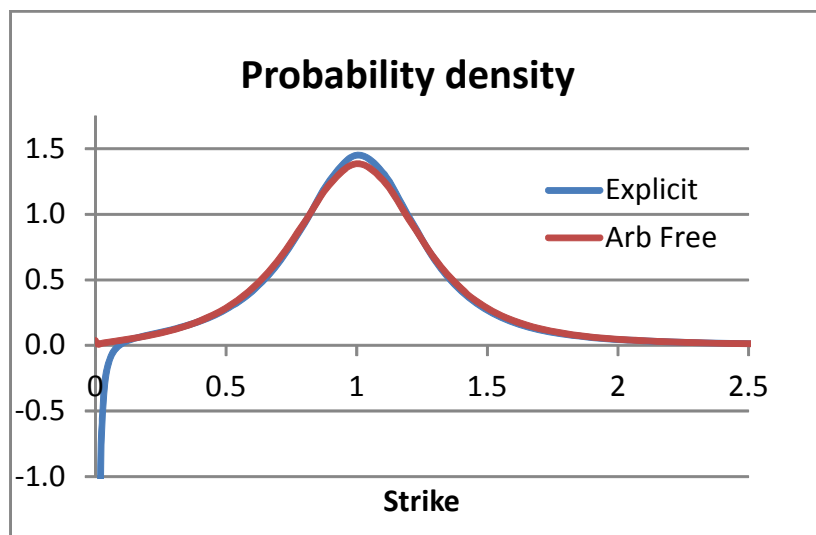
# Implied probability density

- Both  $\sigma_N(K)$  lead to nearly identical prices



$$\alpha = 35\%, \beta = 25\%, \rho = -10\%, \nu = 100\%$$

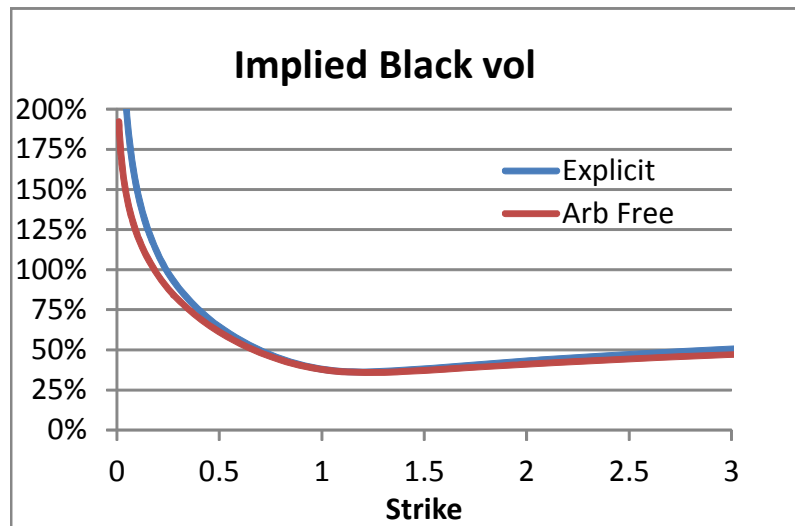
- Yet one leads to negative probability densities, and is not arbitrage free



$$\alpha = 35\%, \beta = 25\%, \rho = -10\%, \nu = 100\%$$

# Black vols

- Using log normal vols doesn't help discern which smiles are arbitrage free



$$\alpha = 35\%, \beta = 25\%, \rho = -10\%, \nu = 100\%$$

# Arbitrage free approach

- SABR model:

$$d\tilde{F} = \varepsilon \tilde{\alpha} C(\tilde{F}) dW_1,$$

$$d\tilde{\alpha} = \varepsilon \nu \tilde{\alpha} dW_2,$$

$$dW_1 dW_2 = \rho dt$$

- Probability density & moments

$$p(T, F, \alpha) dF d\alpha = \text{Prob}\{F < \tilde{F}(T) < F + dF, \alpha < \tilde{\alpha}(T) < \alpha + d\alpha\}$$

$$Q^{(k)}(T, F) dF = \int_0^\infty \alpha^k p(T, F, \alpha) d\alpha$$

- Fökker-Planck equation:

$$p_T = \frac{1}{2} \varepsilon^2 [\alpha^2 C^2(F) p]_{FF} + \varepsilon^2 \rho \nu [\alpha^2 C(F) p]_{F\alpha} + \frac{1}{2} \varepsilon^2 [\alpha^2 Q p]_{\alpha\alpha}$$

– integrate over all  $\alpha$ ,

$$\int_0^\infty [\alpha^2 C(F) p]_{F\alpha} d\alpha = [\alpha^2 C(F) p]_F \Big|_0^\infty = 0,$$

$$\int_0^\infty [\alpha^2 p]_{\alpha\alpha} d\alpha = [\alpha^2 p]_\alpha \Big|_0^\infty = 0$$

- Yields conservation law:

$$Q_T^{(0)} = \frac{1}{2} \varepsilon^2 [C^2(F) Q^{(2)}]_{FF}$$

# Effective forward equation

- Conservation law:

$$Q_T^{(0)} = \frac{1}{2}\varepsilon^2[C^2(F)Q^{(2)}]_{FF}$$

- Use asymptotic methods to analyze backwards equation for  $Q^{(0)}$  and  $Q^{(2)}$ . Obtain:

$$Q^{(2)} = (\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2z^2)e^{\varepsilon^2\rho\nu\alpha\Gamma T}Q^{(0)}\{1 + O(\varepsilon^3)\}$$

where

$$z(F) = \int_f^F \frac{df'}{C(f')}, \quad \Gamma = \frac{C(F) - C(f)}{F - f}$$

- Marginal density

$$Q^{(0)}(T, F)dF = \text{Prob}\{F < \tilde{F}(T) < F + dF\}$$

– satisfies *effective forward equation*

$$Q_T^{(0)} = \frac{1}{2}\left[(\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2z^2)e^{\varepsilon^2\rho\nu\alpha\Gamma T}C^2(F)Q^{(0)}\right]_{FF}$$

- Reduction accurate through  $O(\varepsilon^2)$ ; same as original SABR analysis

- *No corresponding 1-d local volatility model*



# Boundary conditions

- Numerically solve the effective forward equation:

$$Q_T^{(0)} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2 z^2) e^{\varepsilon^2\rho\nu\alpha\Gamma T} C^2(F) Q^{(0)} \right]_{FF}$$

over the domain  $0 < F < F_{\max}$ .

- initial condition:

$$Q^{(0)}(0, F) = \delta(F - f) \quad \text{at } T = 0.$$

- *Absorbing* boundary conditions are *required* for  $\tilde{F}(T)$  to be a Martingale:

$$\begin{aligned} Q^{(0)} &= 0 & \text{at } F = 0 \\ Q^{(0)} &= 0 & \text{at } F = F_{\max} \end{aligned}$$

- Conservation requires:

$$Q(T, F) = \begin{cases} Q^L(T) \delta(F) & \text{at } F = 0 \\ Q^{(0)}(T, F) & \text{for } 0 < F < F_{\max} \\ Q^R(T) \delta(F - F_{\max}) & \text{at } F = F_{\max} \end{cases}$$

with

$$\frac{dQ^L}{dT} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2 z^2) e^{\varepsilon^2\rho\nu\alpha\Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=0^+}$$

$$\frac{dQ^R}{dT} = -\frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2 z^2) e^{\varepsilon^2\rho\nu\alpha\Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=F_{\max}^-}$$

# Option prices

- Numerically solve the PDE

$$Q_T^{(0)} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_{FF},$$

over  $0 < F < F_{\max}$ , with

$$Q^{(0)} = 0 \quad \text{at } F = 0, \quad Q^{(0)} = 0 \quad \text{at } F = F_{\max}$$

and

$$Q^{(0)}(0, F) = \delta(F - f) \quad \text{at } T = 0$$

–  $\delta$ -functions at  $F = 0$  and  $F = F_{\max}$ :

$$\frac{dQ^L}{dT} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=0^+}$$

$$\frac{dQ^R}{dT} = -\frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=F_{\max}^-}$$

- Option prices:

$$V_{call}(\tau_{ex}, K) = \int_K^{F_{\max}} (F - K) Q^{(0)}(\tau_{ex}, F) dF + (F_{\max} - K) Q^R(\tau_{ex})$$

$$V_{put}(\tau_{ex}, K) = \int_0^K (K - F) Q^{(0)}(\tau_{ex}, F) dF + K Q^L(\tau_{ex})$$

– reduced problem has 1 space dimension

numerical solution is essentially instantaneous!

– solving the PDE for  $0 < T < \tau_{ex}$  yields option prices for all strikes  $K$  at  $\tau_{ex}$

# Numerical method

$$Q_T^{(0)} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_{FF}$$

$$Q^{(0)} = 0 \quad \text{at } F = 0, \quad Q^{(0)} = 0 \quad \text{at } F = F_{\max}$$

$$Q^{(0)}(0, F) = \delta(F - f) \quad \text{at } T = 0$$

$$\frac{dQ^L}{dT} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=0^+}$$

$$\frac{dQ^R}{dT} = -\frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho v\alpha z + \varepsilon^2 v^2 z^2) e^{\varepsilon^2 \rho v \alpha \Gamma T} C^2(F) Q^{(0)} \right]_F \Big|^{F=F_{\max}^-}$$

- Use moment preserving Crank-Nicholson scheme

– guarantees probability is conserved exactly, and that  $\tilde{F}(T)$  is exactly a Martingale:

$$Q^L(T) + \int_0^{F_{\max}} Q^{(0)}(T, F) dF + Q^R(T) = 1,$$

$$\int_0^\infty F Q^{(0)}(T, F) dF + F_{\max} Q^R(T) = f.$$

- Maximum principle guarantees that

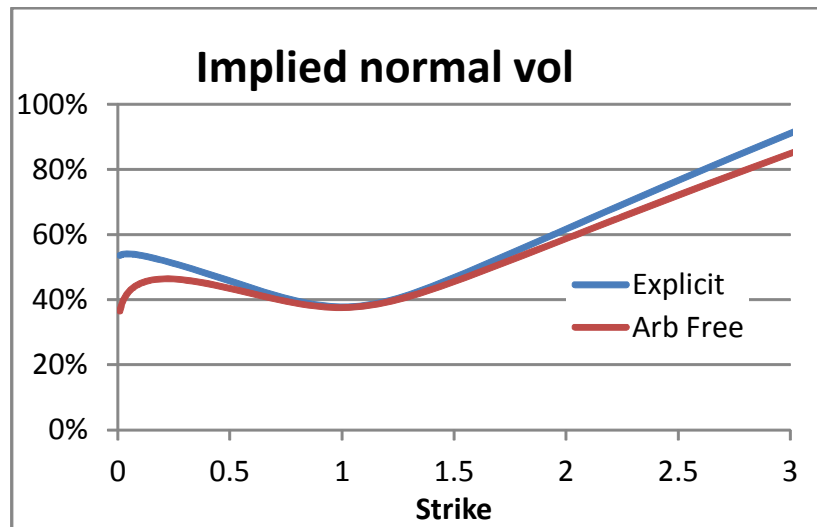
$$Q^{(0)}(T, F) \geq 0 \quad \text{for all } 0 < F < F_{\max}, \text{ all } T > 0,$$

$$Q^L(T) \geq 0, \quad Q^R(T) \geq 0 \quad \text{for all } T > 0$$

- Numerical solution is an exactly arbitrage free model!

# Boundary layer

- Arbitrage free approach yields nearly the same values as the explicit SABR formulas  $\sigma_N(K)$ , except for low strikes and forwards



$$\alpha = 35\%, \beta = 25\%, \rho = -10\%, \nu = 100\%$$

- Effective forward equation:

$$Q_T^{(0)} = \frac{1}{2} \left[ (\alpha^2 + 2\varepsilon\rho\nu\alpha z + \varepsilon^2\nu^2 z^2) e^{\varepsilon^2\rho\nu\alpha\Gamma T} C^2(F) Q^{(0)} \right]_{FF}$$

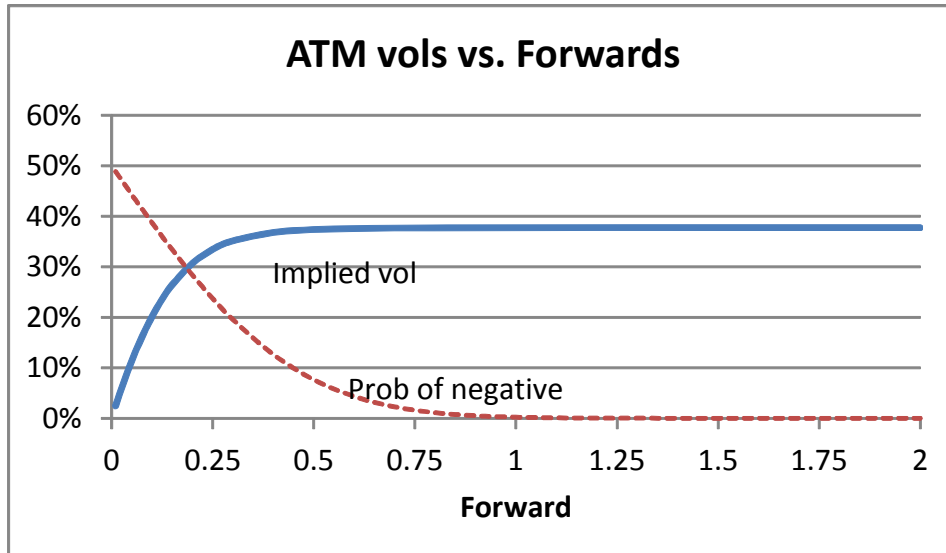
– using asymptotic methods to solve the effective forward equation leads to the *same explicit formulas* for  $\sigma_N(K)$  as in the original analysis, unless the forward or strike is near zero

- Explicit formulas for  $\sigma_N(K)$  do not hold in a boundary layer around zero

– boundary layer occurs where a significant fraction of the paths get absorbed at 0 before expiry

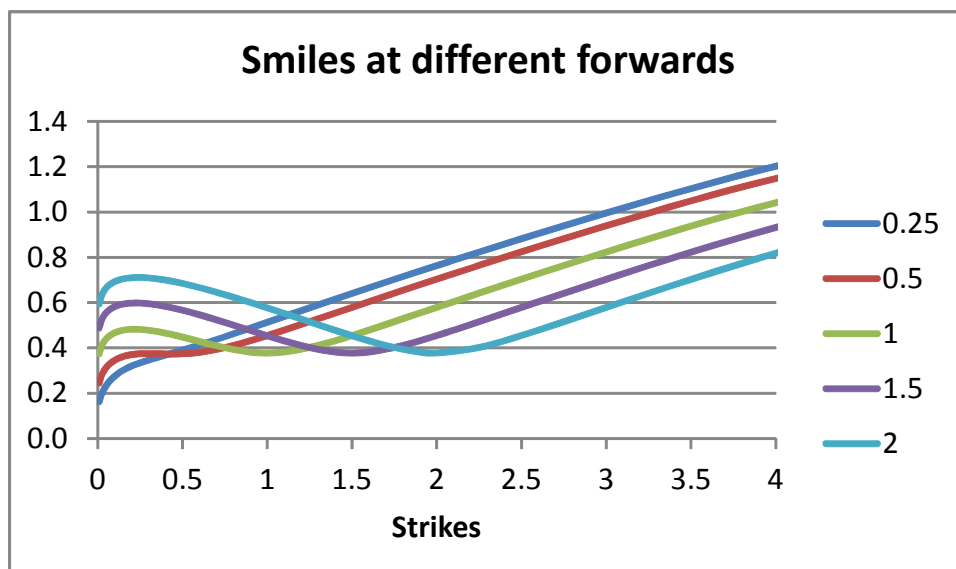
# Boundary layer effects

- At the money vols decrease linearly for small rates



$$\alpha = 35\%, \beta = 0\%, \rho = 0\%, \nu = 100\%$$

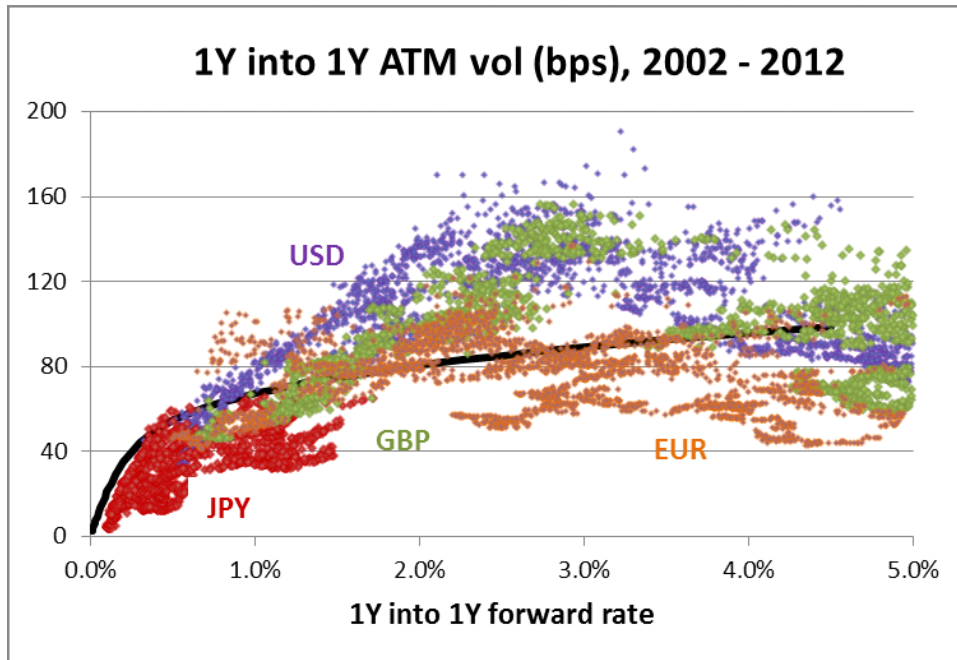
- Knee is often incorrectly ascribed to market switching from normal to log normal behavior in ultra-low rate environments
  - this leads to mispricing *high* strike options in low rate environments



$$\alpha = 35\%, \beta = 0\%, \rho = 0\%, \nu = 100\%$$

# Historical market data

- Arbitrage free SABR closely matches market data

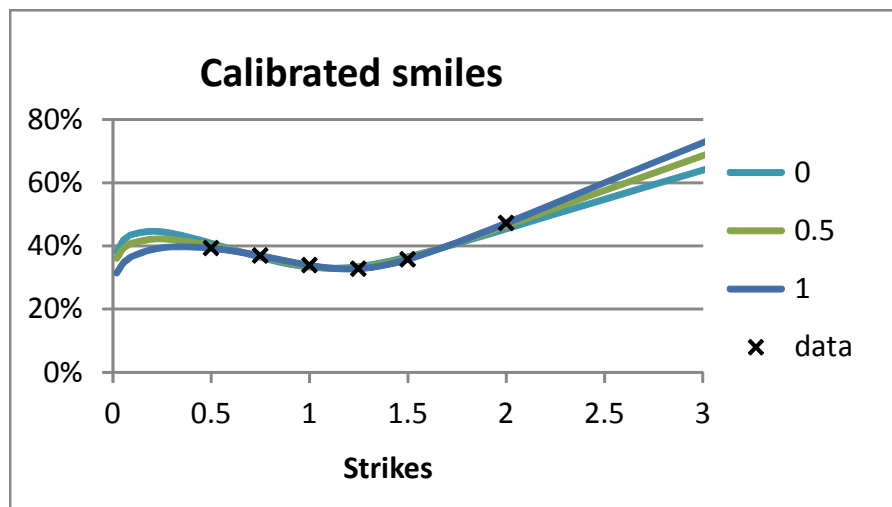


Historic swaption vols for 2002 through 2012

# Calibrating the SABR model

- $\alpha$  controls the at-the-money vol,  $\nu$  controls the smile, but both  $\rho$  and  $\beta$  control the skew

- SABR model calibrated to same market data with  $\beta$  chosen to be 0,  $\frac{1}{2}$ , and 1



SABR model calibrated with  $\beta$  of 0,  $\frac{1}{2}$ , and 1.

– calibrated parameters:

$\alpha$	31.8%	32.9%	35.1%
$\beta$	0	0.5	1
$\rho$	-18.3%	-45.5%	-64.4%
$\nu$	0.777	0.867	0.985

– although tails are somewhat different, all three sets of parameters fit the actual market data well within market noise

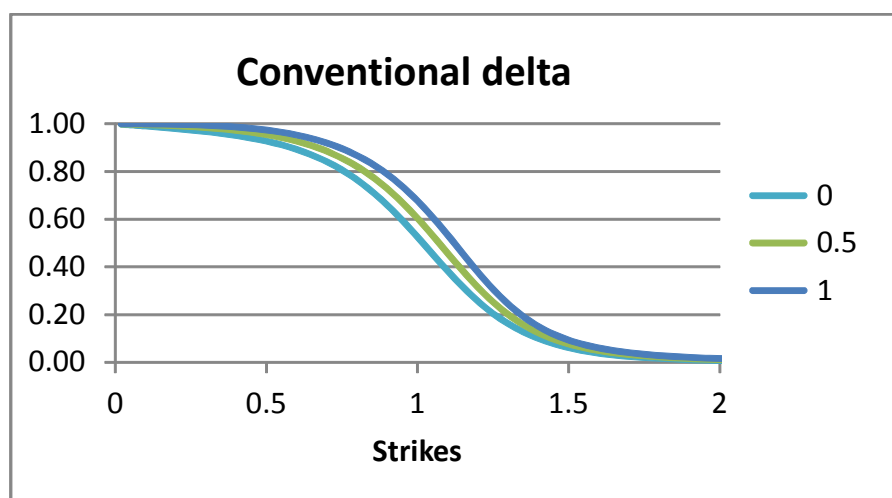
–  $\rho$  can largely compensate for  $\beta$

# Conventional hedging

- Conventional delta,  $\partial V/\partial F$ , based on the scenario

$$\tilde{F} \rightarrow \tilde{F} + \Delta F, \quad \tilde{\alpha} \rightarrow \tilde{\alpha}$$

- conventional delta for the same three sets of SABR parameters



$\partial V/\partial F$  against the strike  $K$  for  $\beta$  of 0,  $\frac{1}{2}$ , and 1.

- Even though all three sets of parameters closely fit the market smile, they lead to different conventional hedges, even near the money
  - choosing the incorrect beta can lead to good fits of the smile, but relatively poor delta hedges



# Alternative delta hedges

$$d\tilde{F} = \tilde{\alpha}C(\tilde{F})dW_1,$$

$$d\tilde{\alpha} = v\tilde{\alpha}dW_2,$$

$$dW_1dW_2 = \rho dt.$$

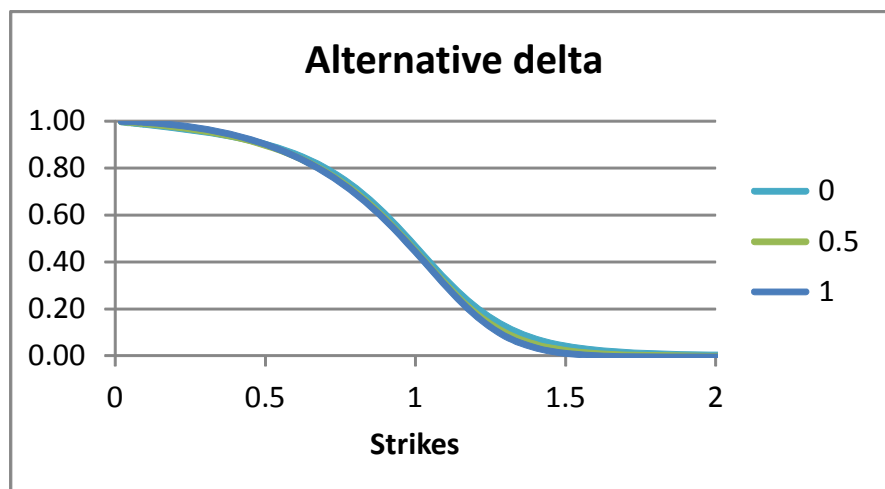
- When  $\tilde{F}$  changes,  $\tilde{\alpha}$  should also change, at least on average

$$d\tilde{\alpha} = v\tilde{\alpha} \left\{ \rho dW_1 + \sqrt{1 - \rho^2} dW_{\perp} \right\} = \left\{ \rho v \frac{d\tilde{F}}{C(\tilde{F})} + \sqrt{1 - \rho^2} v\tilde{\alpha} dW_{\perp} \right\}$$

- Alternative delta based on scenario:

$$\tilde{F} \rightarrow \tilde{F} + \Delta F, \quad \tilde{\alpha} \rightarrow \tilde{\alpha} + \rho v \frac{\Delta F}{C(\tilde{F})}$$

- alternative delta for the same three sets of parameters



Alternative delta,  $\partial V / \partial f + [\rho v / C(f)] \partial V / \partial \alpha$

- Alternative delta is nearly independent of  $\beta$ . It depends mainly on the actual market skew/smile, and not on how the smile is parameterized

- alternative deltas are believed to provide much better hedges