

Outline

- Dividend modeling (very briefly, details in SSRN paper)
- Implied volatility curve design
- No-Arbitrage constraints in price- and vol-space
 - Including some non-trivial, exact results on vol parameter constraints
- Arbitrage-free vol surface fitting in practice

For background see (available at SSRN):

Pricing Vanilla Options with Cash Dividends

Necessary and Sufficient No-Arbitrage Conditions for the SSVI/S3 Volatility Curve

J. Gatheral, A. Jacquier: *Arbitrage-free SVI volatility surfaces*

J. Gatheral: *The Volatility Surface*, Wiley 2006

Implied Vols and Surfaces

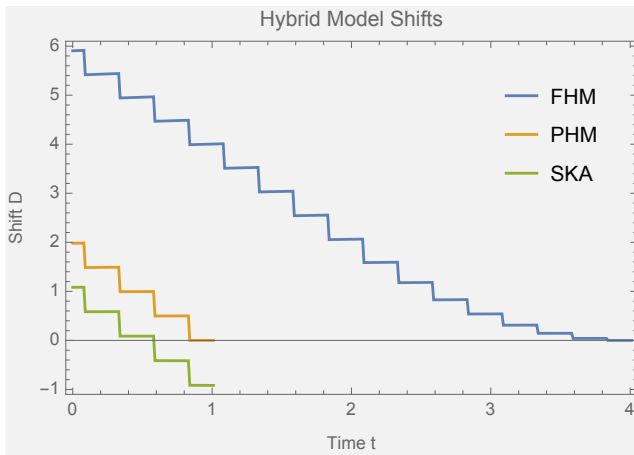
- Implied volatility surfaces (and borrow cost curves) are the standard approach to summarizing the vanilla options market in an intuitive and compact manner.
- They provide the **fundamental building blocks** for trading and risk-managing vanillas (listed and OTC), as well as flow derivatives and exotics.
- Options market makers use a **Black-Scholes framework with some added bells and whistles** for valuation and risk management. (BS as language...)
- For more exotic applications: (i) An arbitrage-free vol surface is equivalent to the existence of a local volatility surface. (ii) Having an arbitrage-free, sensible surface for all (T, K) makes calibrating fancy “SLVJ” models much easier.
- Many of the most liquid options allow American exercise: “**De-Americanization**” is necessary and in principle introduces model-dependence, since American options are (very light) “exotics”.
- So, from now on all implied vols will be the “European-equivalent” ones.
- But what do we mean by **implied vols in the presence of cash dividends... ?**

Dividend Modeling

- Forty years after Black-Scholes there is no consensus on how to model cash dividends (even for vanillas)!
- Cash dividends mean that the observed stock price can not follow geometric Brownian motion (GBM).
- In a vanilla context the question is how to combine the stochastic part of underlier evolution (e.g. *who* follows GBM?) with...
- **Three types of dividends:**
 - A dividend yield – used to model borrow cost
 - Cash dividends – how most dividends are actually paid
 - Discrete proportional dividends
- Most firms use a **blending scheme** to transition from cash dividends on short end to proportional dividends in long term.
- Proportional divs are also useful in times of extreme uncertainty (market-wise or name-specific). E.g. during 2008 crisis.

Dividend Models

- Main two classes of dividend models are:
 - *Spot model*: The dividends come out of the observed stock price. (Need to modify cash dividends at low stock price.)
 - *Hybrid models*: The dividends come out of a “cash buffer”, related to the PV of future dividends: $S_t = \tilde{S}_t + D_t$
- Spot model might seem naively more reasonable, but in practice leads to a lot of complications, since not GBM.
- Hybrid models are much simpler to handle for both vanillas and exotics, since *pure stock* \tilde{S}_t still follows GBM. Can also easily handle credit risk, extension to (light) exotics, local vols, etc.
- We will assume a hybrid model from now on.
- For a detailed review, synthesis, and new results see SSRN paper.



The shifts D_t for various hybrid models with $r=3\%$, $q=1\%$, and a quarterly cash dividend of 0.5 first paid at $t_1=0.085$, using blending scheme (2, 4).

For PHM, SKA: using $T=1.01$.

FHM = HM2, PHM = HM1, SKA = HM3

Volatility Curve Parametrization Wish List

Parametric vol curves are better than non-parametric ones!

- Parameters should have simple, intuitive meaning, esp. first three.
- Parameters should be as “independent” as possible, and stable from day to day (parsimonious).
- Little term-structure, if possible.
- No-arbitrage constraints should be “easy” to incorporate.
- Parametric vols should be easy/fast to compute.
- No hacks! (in wings, etc)
- Vol curves arising from standard “SLVJ”-type model should be fittable within a few bps (at worst).

Our parametrization approach

- Work one term at a time, impose smoothness across terms.
- Factor out overall vol level (ATF) as: $\sigma_0 := \sigma(T, K = F)$.
- Define the “shape” curve $f(z) = f(z|\mathbf{p})$ as a function of **normalized strike**

$$\text{NS} = z := \frac{y}{\hat{\sigma}_0} = \frac{\log(K/F)}{\sigma_0 \sqrt{T}}$$

such that

$$\sigma(z)^2 = \sigma_0^2 f(z|\mathbf{p})$$

- There are no standard definitions – we define the dimensionless “skew” and “smile/convexity” as slope and curvature of the shape curve:

$$f(z) =: 1 + s_2 z + \frac{1}{2} c_2 z^2 + \dots$$

Our parametrization approach (cont'd)

- s_2 and c_2 tend to have mild term-structure (except maybe as $T \rightarrow 0$). They are even comparable across names. Have been range-bound for decades.
- Sometimes it is useful to work with s_1, c_1 defined via

$$\sigma(z) =: \sigma_0 \left(1 + s_1 z + \frac{1}{2} c_1 z^2 + \dots \right)$$

- Trivially: $s_2 = 2s_1, c_2 = 2(c_1 + s_1^2)$.
- Note that

$$\sigma(z) = \sigma_0 + \frac{s_1}{\sqrt{T}} \log(K/F) + \dots,$$

so that an alternative definition of skew

$$\tilde{s}_1 := K \frac{\partial \sigma}{\partial K} \Big|_{K=F} = \frac{s_1}{\sqrt{T}}$$

- No simple relationships between alternative definitions of curvature/convexity/smile.

No-Arbitrage: Basic issues

- In price space arbitrage conditions are well-known.
- It's easy to detect when there is arbitrage.
- But it is not at all clear, a priori, how to remove it.
- There are an infinity of ways of doing so – almost all of them are “bad”, especially when working in price space.
- We find it easier to remove arbitrage in vol-space.
- Due to de-Americanization issue, one has to be a bit careful when moving from price- to vol-space for American options.

No-Arbitrage Constraints in Vol-Space

- **No butterfly arbitrage:** Implied density ρ should be positive:

$$\hat{C}(T, K) = \int_0^\infty dS_T (S_T - K)_+ \rho_T(S_0 \rightarrow S_T)$$

$$\Rightarrow \partial_K^2 \hat{C}(T, K) = \rho_T(S_0 \rightarrow S)|_{S=K}$$

- **No calendar arbitrage:** Total BS variance $w(y) := T\sigma(y)^2$ has to be increasing in T at any fixed y .
- Necessary (but generally not sufficient) constraint on the asymptotic wing behavior of implied vols (TRK 2001, R. Lee, 2004):

$$w(y) \leq 2|y| \quad \text{as } |y| \rightarrow \infty$$

Simple consequences: Implied density 1

- Local vols and implied densities can be calculated most neatly in terms of the total variance $w(y) = T\sigma(z)^2$. Eg the implied density:

$$\rho(y) = \frac{g(y)}{\hat{\sigma}(y)} n(d_-(y)) ,$$

where $n(x) = N'(x)$ is the normal density, and

$$g(y) = \left(1 - \frac{y w'(y)}{2w(y)}\right)^2 - \frac{1}{4} \left(\frac{1}{w(y)} + \frac{1}{4}\right) w'(y)^2 + \frac{1}{2} w''(y)$$

- Absence of butterfly arbitrage: $g(y) \geq 0$ for all y .
- In Black-Scholes case: $g(y) = 1$ for all y .

ATF No-Arbitrage Constraints

- If $w(z) = \hat{\sigma}_0^2 (1 + s_2 z + \frac{1}{2} c_2 z^2 + \dots)$, then

$$g(z=0) = 1 + \frac{1}{2} c_2 - \frac{1}{4} s_2^2 (1 + \frac{1}{4} \hat{\sigma}_0^2)$$

- $g(0) \geq 0$ implies upper bound on slope

$$s_2^2 \leq \frac{4 + 2c_2}{1 + \frac{1}{4} \hat{\sigma}_0^2}$$

or lower bound on curvature ($c_1 = \frac{1}{2} c_2 - \frac{1}{4} s_2^2$)

$$c_1 \geq -1 + \frac{1}{16} s_2^2 \hat{\sigma}_0^2 \approx -1$$

- Very relevant around FOMC and earnings where not just $c_1 < 0$ but even $c_2 < 0$ can happen!

Specific Curves: Parabolas

- What are simplest possible implied vol curves? Need at least 3 parameters for ATF behavior.
- Vendors often use

$$\sigma(y)^n = \sigma_0^n \left(1 + s y + \frac{1}{2} c y^2\right) \quad (\text{or in terms of } z)$$

- Obviously has arbitrage in wings for $n = 1, 2$.
- Slight hope for $n = 4$, but would imply symmetric wings, which is intuitively and empirically wrong.
- Positivity has to be enforced too.
- Must do better...

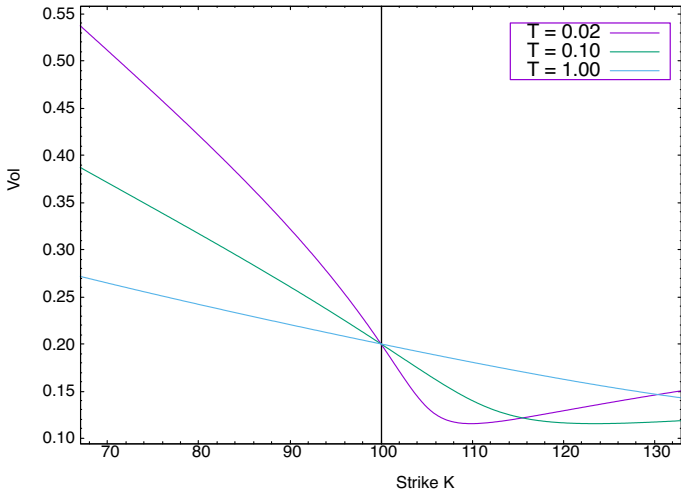
Specific Curves: S3/SSVI

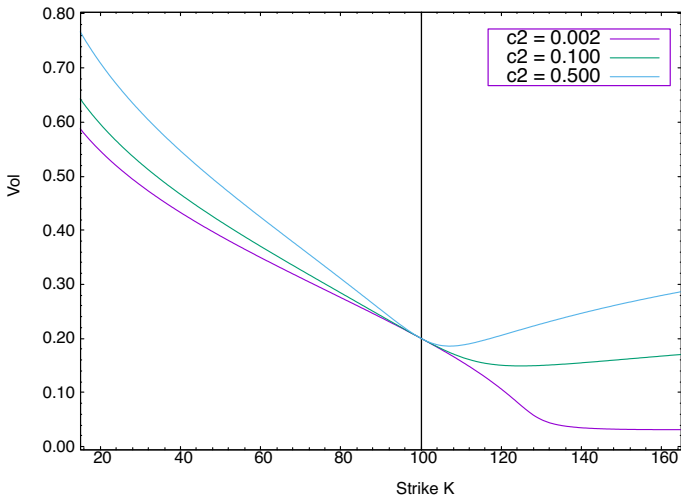
- Simplest sensible curve with 3 parameters ($c_2 \geq 0$):

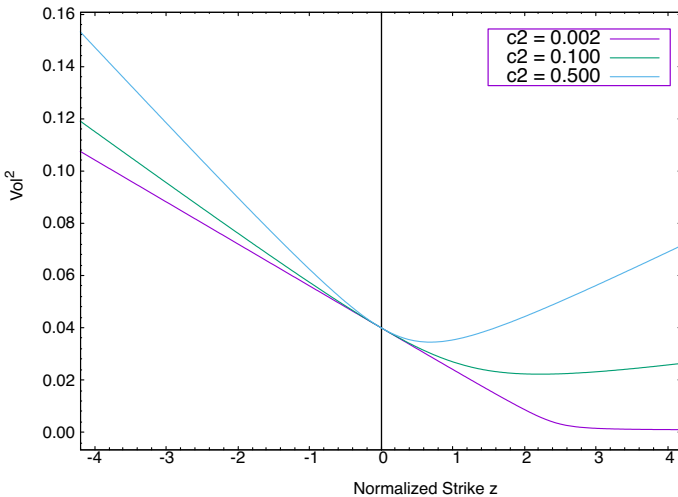
$$\sigma^2(z) = \sigma_0^2 \left(\frac{1}{2}(1 + s_2 z) + \sqrt{\frac{1}{4}(1 + s_2 z)^2 + \frac{1}{2}c_2 z^2} \right)$$

- Was independently discovered by TRK (2003, “S3”) and Gatheral/Jacquier (2013, “SSVI” = Simple SVI).
- Allows surprisingly varied skew shapes, including “takeover-for-cash” curves as $c_2 \rightarrow 0$.
- Allows fitting of vast majority of US equity names.
- Very easy to avoid (butterfly) arbitrage.
- In fact, in terms of the dimensionless variables $\hat{\sigma}_0, s_2, c_2$ can completely answer the butterfly-arbitrage question...

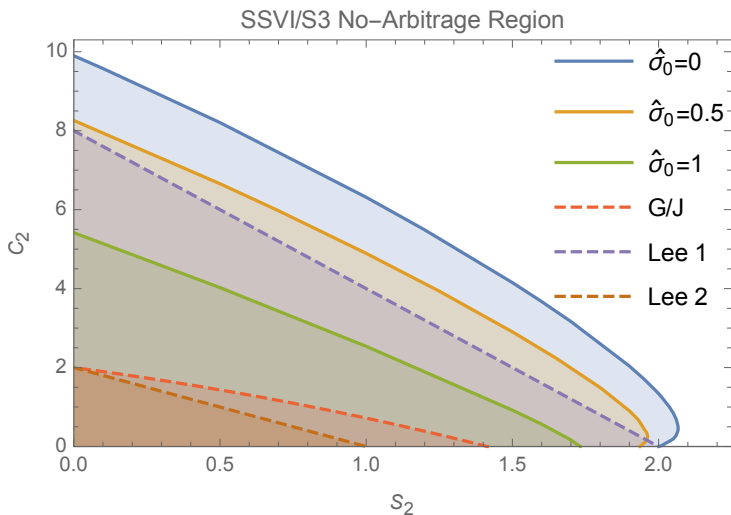
S3 curve for different terms, same parameters



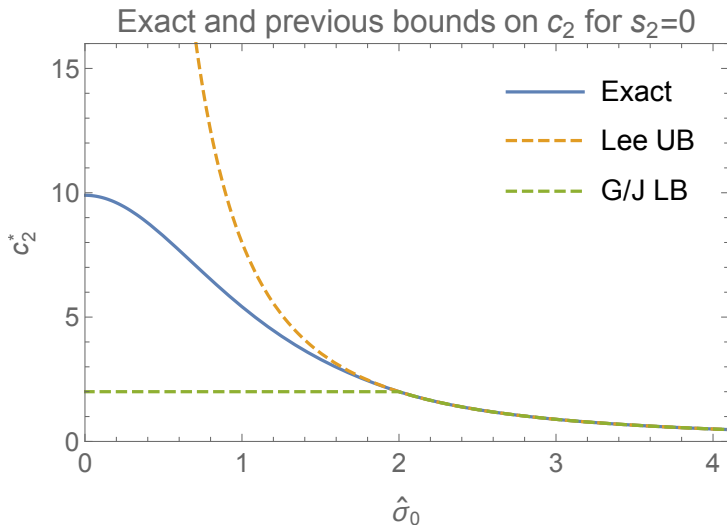
S3 curve for different curvatures c_2 , $T = 0.25$ 

S3 curve for different curvatures c_2 

Necessary and Sufficient No-Arb Conditions for S3/SSVI



Necessary and Sufficient No-Arb Conditions for S3/SSVI



Beyond the Simplest Curves: 5 Parameters (SVI, etc)

- Besides 3 parameters for ATF would be nice to have **independent parameters** C_{\pm} for wings:

$$\sigma(z)^2 \rightarrow \sigma_0^2 C_{\pm} |z| \quad \text{as } z \rightarrow \pm\infty \quad (\hat{\sigma}_0 C_{\pm} \leq 2)$$

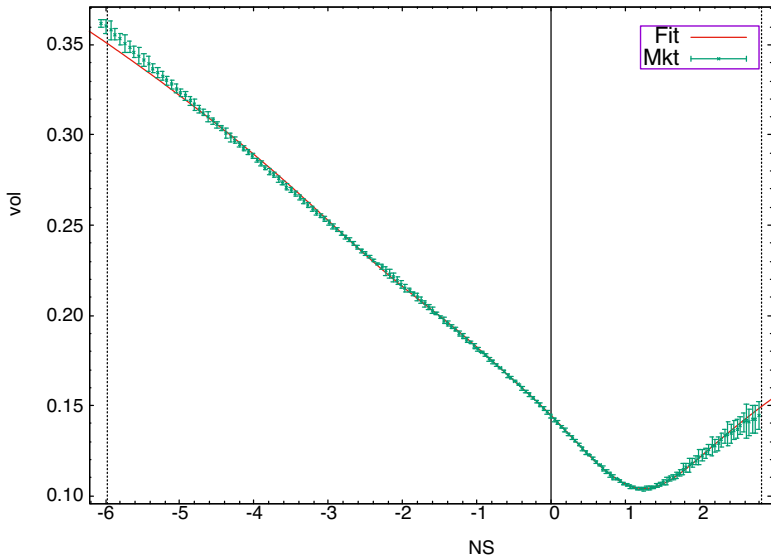
- For S3/SSVI: $C_{\pm} = \sqrt{\frac{1}{4}s_2^2 + \frac{1}{2}c_2} \pm \frac{1}{2}s_2$
- For Gatheral's **SVI** and others (JW/L5, TRK) the C_{\pm} are independent parameters (constrained by $-C_- \leq s_2 \leq C_+$).
- Just some algebra to re-express their “raw” parametrization in terms of natural parameters $\sigma_0, s_2, c_2, C_-, C_+$.
- Can fit some names better than with S3/SSVI.... but not much better in many cases!?
- Certainly can not fit W-shaped curves around events (**still** $c_2 \geq 0$).

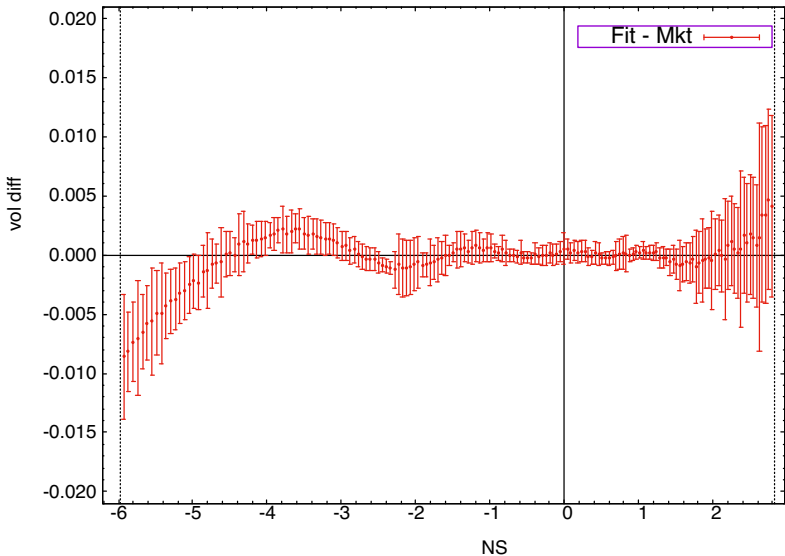
What curves to use for most liquid names?

- For very liquid names (ES, SPX, SPY, other ETFs, AAPL, KOSPI, etc) none of the standard curves (SVI, L5 or amendments) work well, even in the absence of events.
- There is a fundamental problem with the shapes allowed by these curves: Curvature has unique maximum around ATF, but that's not what the market wants!
- Need more flexible shapes that can handle more generic curvature structures, incl. negative curvature around ATF: C5, C6, C7, C8, C10,

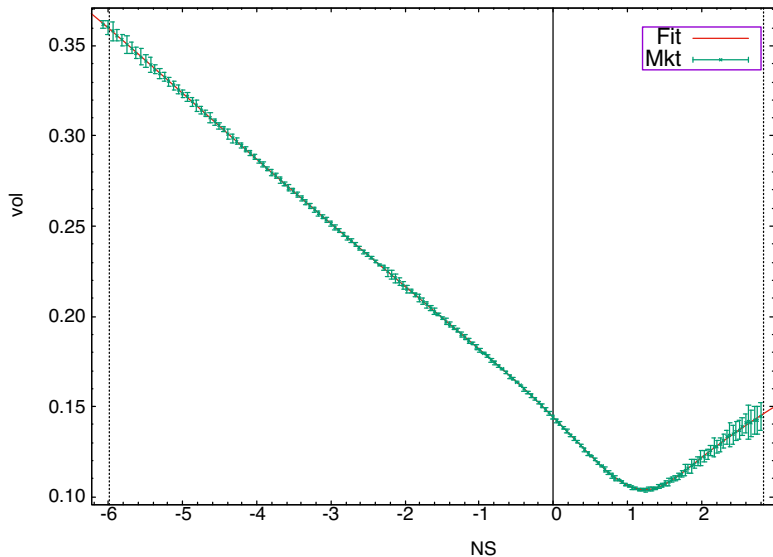
Volatility fitting framework

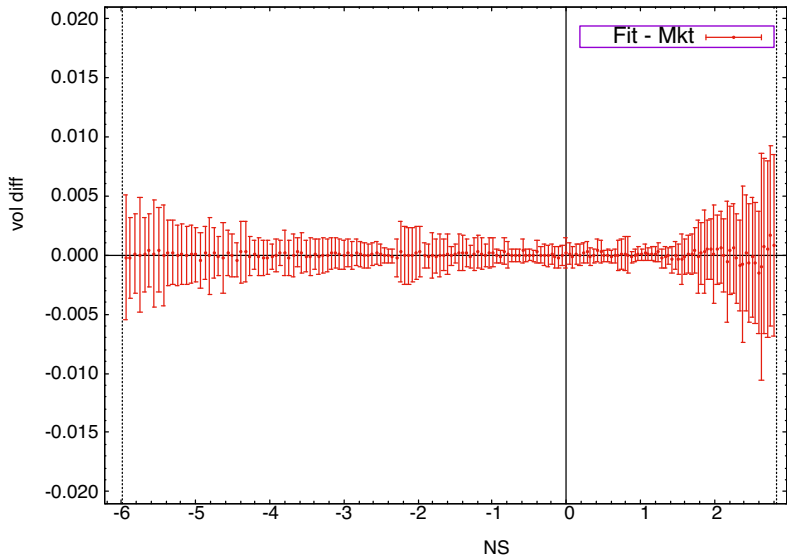
- Input to fitter are implied vols with error bars (after proper div modeling, borrow implication, etc).
- All our vol curves have sensible dimensionless parameters (first three are universal), which allows the use of curve-independent heuristics from decades of vol fitting experience across many names, geographies and asset classes.
- Fit one term at a time, for speed, transfer information between terms, for smoothness and stability.
- Minimize chi-square + soft penalties, for robustness and to allow the fitting of terms with less (effective) data than parameters.
- Good microprices help, but even then various heuristics are needed to deal with data issues in real-time.
- Keeping track of quality-of-fit metrics and error bars for final outputs is crucial for real-time trading applications.

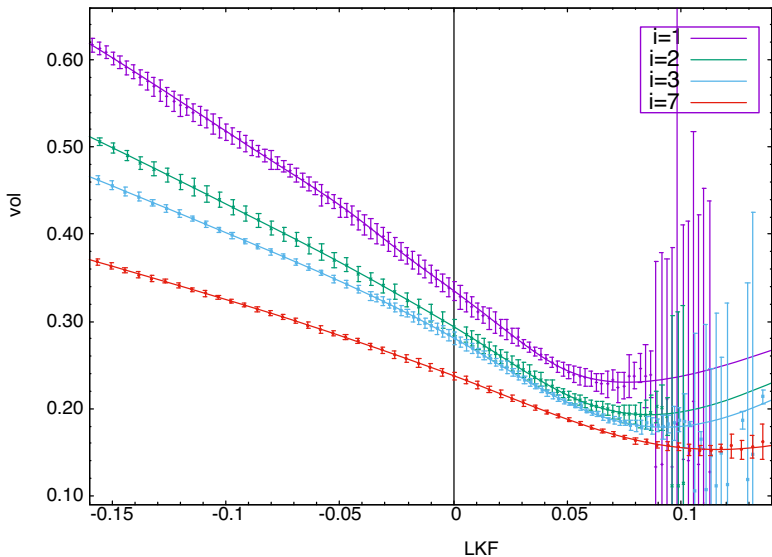
ESM6 20160504-103000 C8: $T=0.1257$, $i=0$, $\chi=0.599$, $avE5=3.2$ 

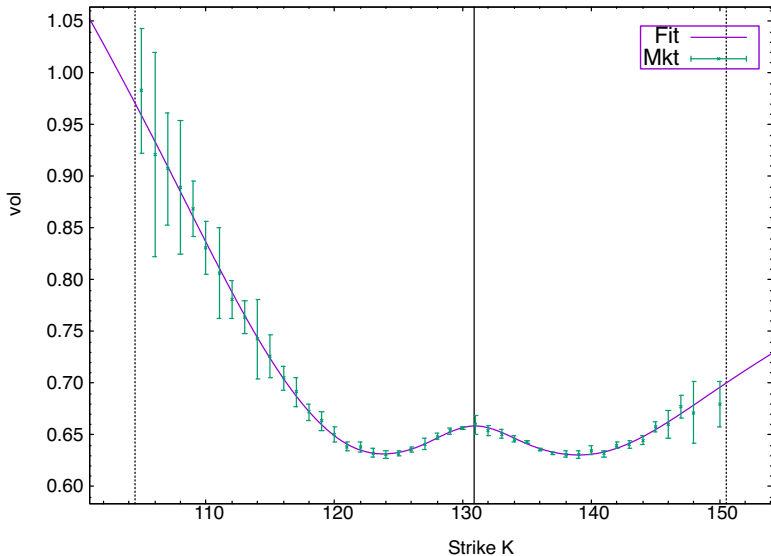
ESM6 20160504-103000 C8: $T=0.1257$, $i=0$, $\chi=0.599$, $avE5=3.2$ 

ESM6 20160504-103000 C12m: $T=0.1257$, $i=0$, $\chi=0.021$, $avE5=1.4$

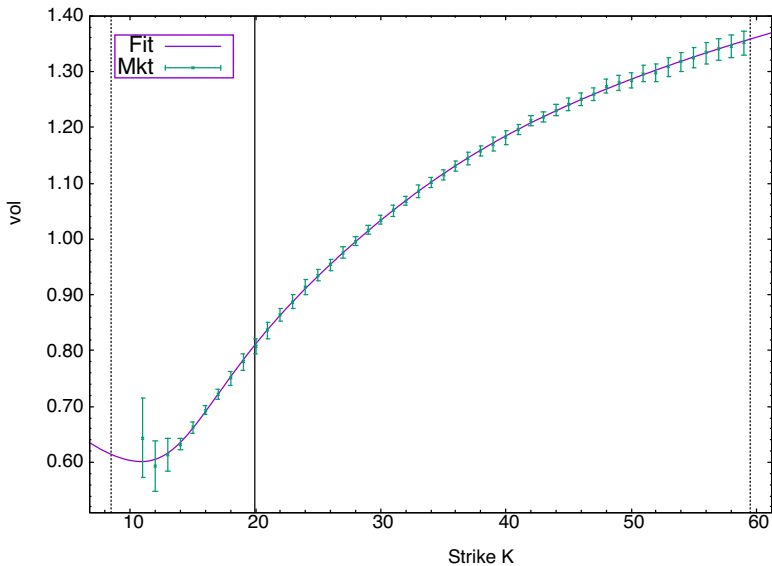


ESM6 20160504-103000 C12m: $T=0.1257$, $i=0$, $\chi=0.021$, $avE5=1.4$ 

SPY 20150825-100000 C10m, $\chi Av=0.048$, $e5Av=3.4$ 

AAPL 20150721-154500 C8: $T=0.0084$, $i=0$, $\chi=0.247$, $avE5=13.2$ 

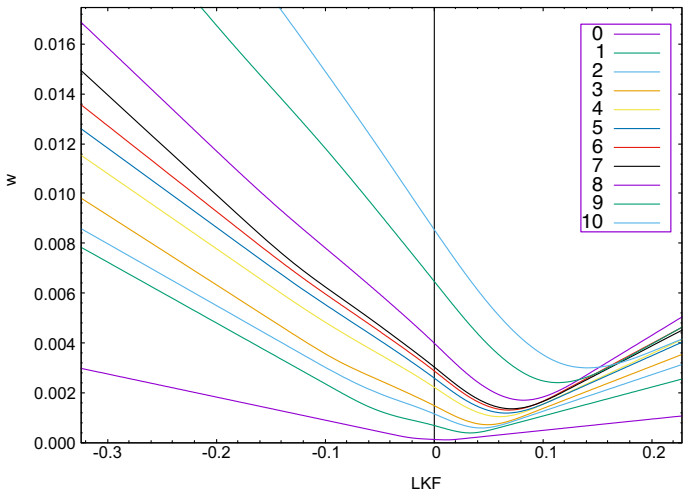
VXX 20151216-135400 C7: $T=0.1784$, $i=7$, $\chi=0.041$, $avE5=20.4$



Arbitrage Elimination in Practice

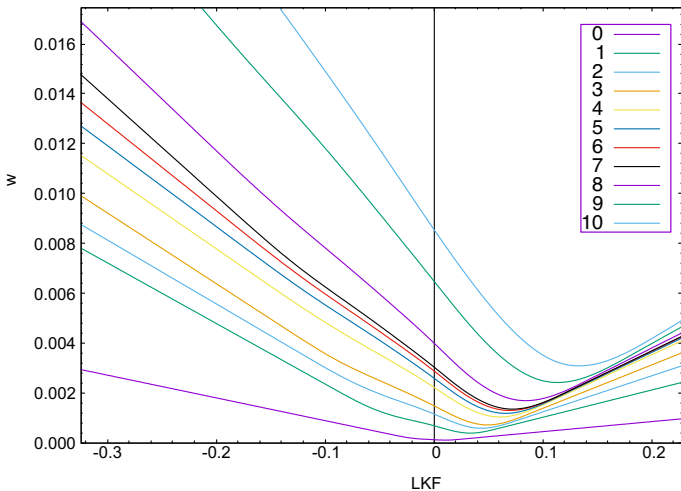
- Our volatility curve framework makes it (almost) impossible to have butterfly arbitrage in the fitted curves.
- Complete calendar arbitrage elimination is accomplished by algorithmic means (“no-arb mode”) considering:
 - With a proper shape parametrization, the term-structure of the (shape) parameters should be “pretty smooth” (except maybe due to events).
 - Enforce total variance constraint taking error bars into account to spread shape information across terms.
- Many potential arbitrages in a vol surface fit of a snapshot can already be eliminated via smart temporal filtering.
- Final result: **arbitrage-free vol surface “closest” to the vol surface that fits market prices best, in parametric form.**
- Some examples, with “regular” and “no-arb” fit mode.

Total Vars SPY 20150820-154500 C8, chiAv=0.096, e5Av=1.5



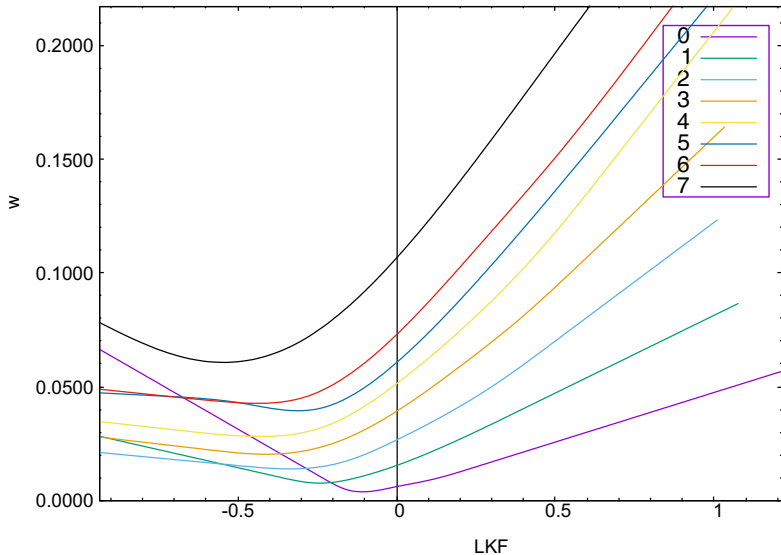
Regular fit mode

Total Vars SPY 20150820-154500 C8, chiAv=0.104, e5Av=1.6

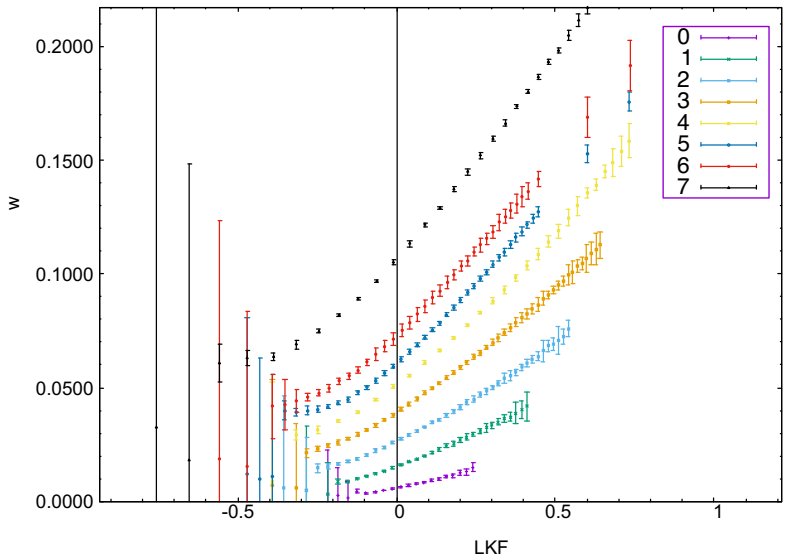


No-arb fit mode

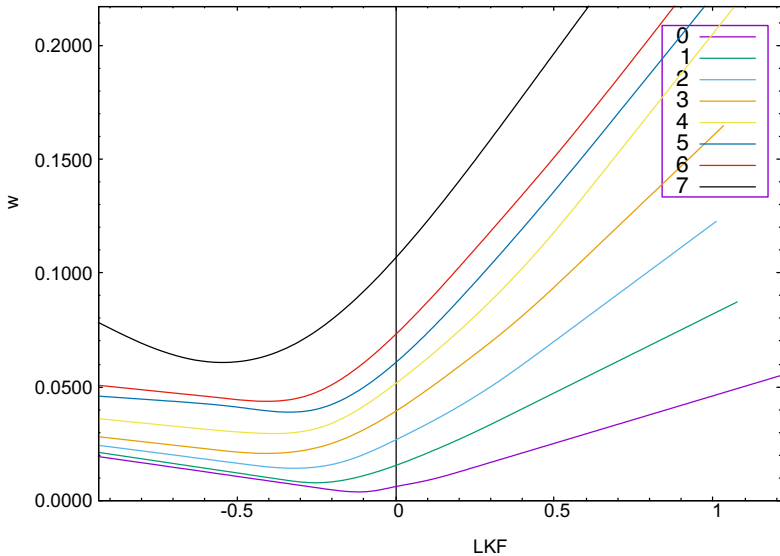
Total Vars VXX 20151216-155500 C7, $\chi_{Av}=0.070$, $e5Av=14.0$



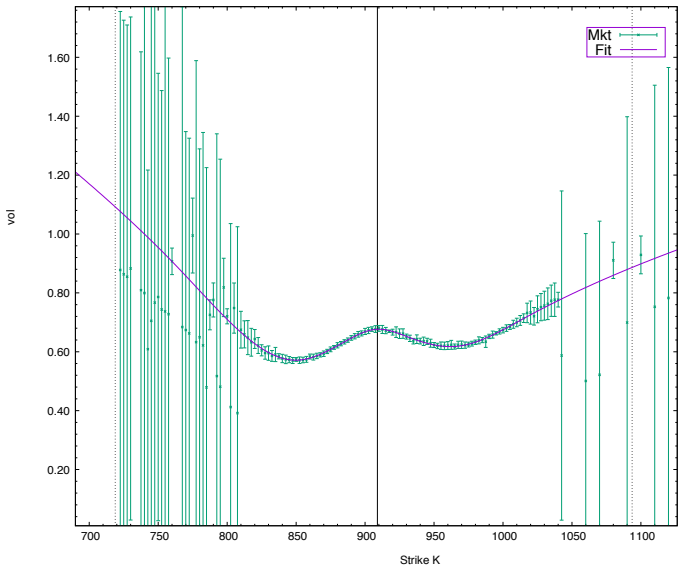
Total Vars VXX 20151216-155500 C7, chiAv=0.070, e5Av=14.0



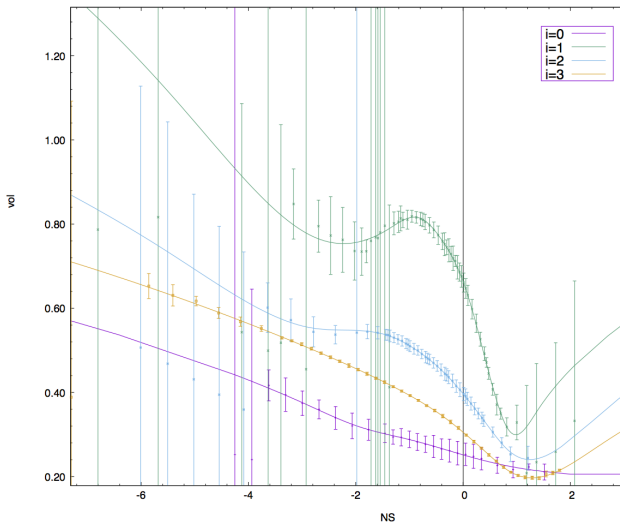
Total Vars VXX 20151216-155500 C7, $\chi_{Av}=0.077$, $e5Av=13.6$



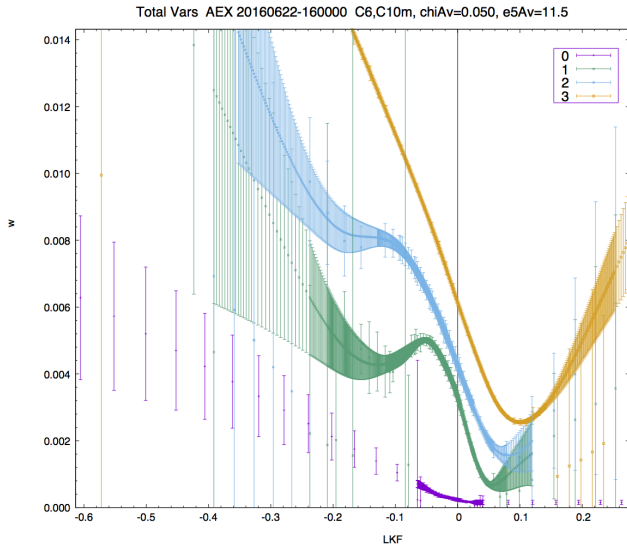
AMZN 20170426-154500 C8: T=0.0056, i=0, chi=0.057, avE5=14.8



AEX 20160622-160000 C6,C10m, $\chi Av=0.050$, $e5Av=11.5$



AEX vols on the day of the Brexit vote!

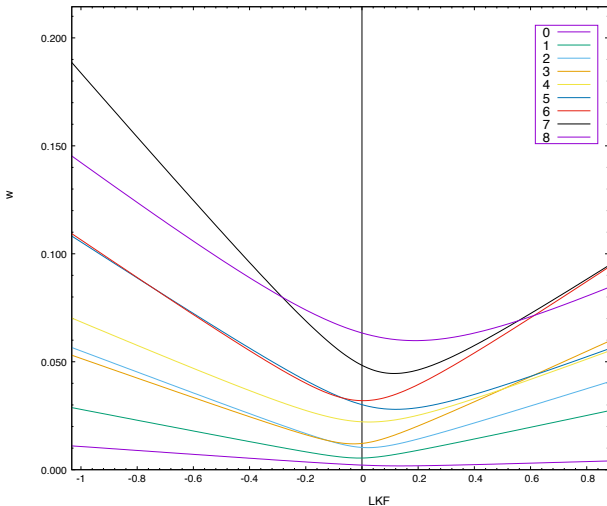


AEX total variances with input and output error bars

Fitting Options on Illiquid Names

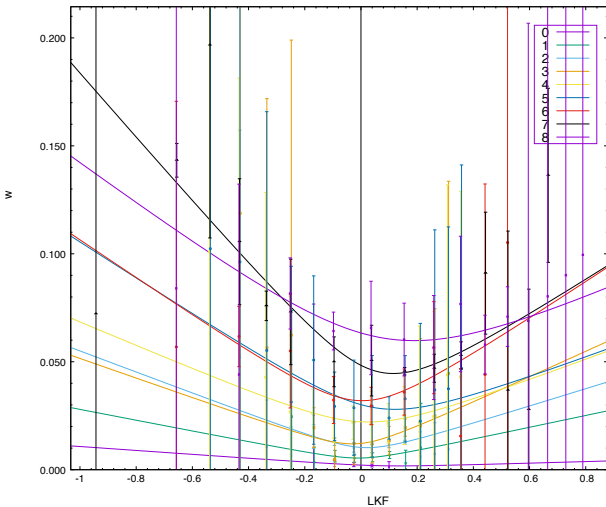
- The vast majority (95%+) of the 4000+ underliers with options in the US are relatively illiquid.
- Mids and even “naive” theos (from some curve fit...) will often have arbitrage (at least when spreads are ignored).
- As far as vol curve fitting is concerned, they can fortunately usually be fitted with the simple 3-parameter curve “S3” (aka “SSVI”).
- No-arbitrage constraints and careful consideration of error bars are extremely helpful in **spreading information in the limited set of more liquid options across all options** and to a complete surface.
- Let’s consider the example of Sprint, ticker S, with closing data. We will show fits in “regular” and “no-arb” mode again.

Total Vars S 20171002-160000 S3, chiAv=0.123, e5Av=550.7



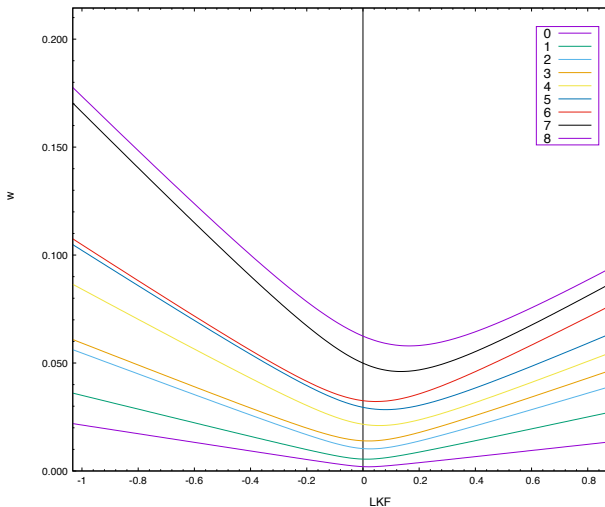
Total variances in regular fit mode

Total Vars S 20171002-160000 S3, chiAv=0.123, e5Av=550.7

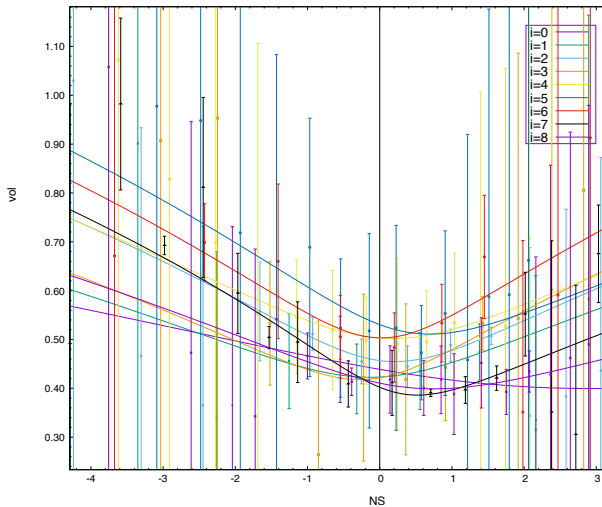


Total variances with error bars in regular fit mode

Total Vars S 20171002-160000 S3, chiAv=0.147, e5Av=582.1

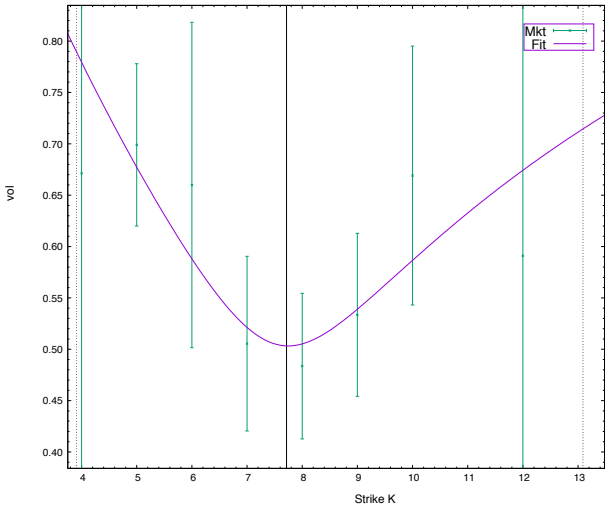


Total variances in no-arb fit mode

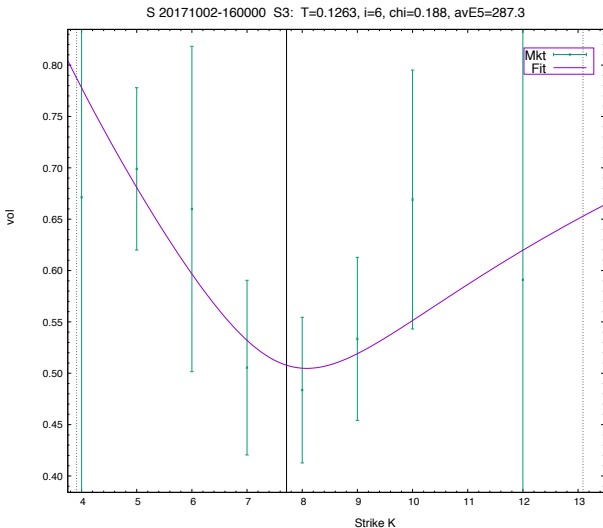
S 20171002-160000 S3, $\chi^2_{\text{Av}}=0.123$, $e5A_{\text{v}}=550.7$ 

Vols in regular fit mode

S 20171002-160000 S3: T=0.1263, i=6, chi=0.129, avE5=272.8



S3 fit of S (Sprint) options, $T = 0.1263y$, $i = 6$, regular fit mode



S3 fit of S (Sprint) options, $T = 0.1263y$, $i = 6$, no-arb fit mode

Other important topics we do not have time for

- Speed: Price and fit the whole US options universe on one box.
- Spot vol dynamics: How do vol surfaces move when the underlier moves? (Simple rules like “sticky-strike” or “sticky-delta” vols are not realistic.) This is relevant for:
 - Realistic parametric scenario generation; overnight moves
 - Smart delta
 - Smart temporal filtering
- How do vol surfaces behave under shifts of “as-of-time” (events...)?
- How does new vol level and skew information in one term spread across the surface?
- How “local” are our parametrizations? Can our parametric volatility surfaces be simply and realistically deformed?

