

Application of Multi-Factor Models

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June 24, 2013

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Introduction

- The Issues
- Market Mechanics
- Spot Price Dynamics
- Application of a Two-Factor Model
- Extrapolation Taken Too Far?

The Issues

Why Do We Need Multi-Factor Models?

- Variety of Options Structures
 - Most franchise energy desks end up dealing in a broad spectrum of structures.
 - A unified vol model is usually better than separate one-factor models for each option type.
- Correlation Structures
 - By construction a single factor model has unit correlation of returns between contracts.
 - The presence of a nontrivial second factor (PCA) should clearly effect valuation of, for example, spread options.

The Issues

Variety of Options Structures

- Energy markets frequently have options with different frequency exercise for the same delivery period.
 - Do you treat each of these separately with its own one-factor model or do you couple them?
- Separate model approach: Independent models for each option type.
 - Benefits:
 - Simple and unambiguous to deploy.
 - Arguably adequate for prop or flow trading of standard options.
 - Liabilities:
 - If you are trading many options products (asians, quarterly swaptions,...) the number of vol surfaces can be very large.
 - All of the vegas of the different options types are unrelated; there is no unifying model and hedging a less liquid vega exposure with a more liquid is outside of the modeling paradigm.

The Issues

Variety of Options Structures

- Multi-Factor Approach: Single underlying modeling framework.
 - Benefits:
 - All options are related. Greeks can be systematically netted and managed.
 - The number of parameters changing daily (or real-time) is of manageable size.
 - Uses liquid market data to sensibly infer prices of less liquid instruments.
 - Liabilities:
 - Challenging to calibrate. Often the slower moving parameters require nonlinear optimizations.
 - Traders with a less analytical bent tend not to like it very much.
 - Overkill for vanilla flow books.

The Issues

Extrapolation

- The primary purpose of a "holistic" risk-neutral model is to extrapolate from liquid instruments to illiquid instruments.
- We have already seen this in several contexts.

- Vol Backwardation

- Our simple one-factor approach to this problem:

$$dF(t, T) = \sigma(T - t)F(t, T)dB_t$$

was used to extrapolate from the prices of vanilla options to valuation of options with non-standard expiry.

- The risk was in the functional form used for $\sigma(T - t)$.
- Only the magnitude of this function was calibrated to market data; the rate of backwardation was estimated statistically.
- Given only say monthly options there is little reason to proceed to more complex models.
- If, however, other options are liquidly traded (e.g. swaptions or daily options), a one-factor model does not have the flexibility to process such information and make useful inference on backwardation.

The Issues

Extrapolation

- Examples (cont)
 - Vol Skew (Single Commodity Vanilla)
 - A set of strikes is used to mark a vol surfaces.
 - Pricing a vanilla option at a different strikes involves interpolating (easy) or extrapolating (hard) the vol grid.
 - Extrapolation can be accomplished either by parametric fits of liquid surfaces or by using a stochastic process that generates "realistic" skews.
 - Vol Skew (Spread Options)
 - The use of vol-lookup by moneyness of opposing leg was purely heuristic approach.
 - There was no single underlying process that justified this approach.

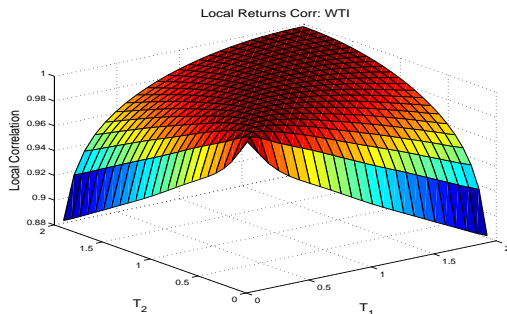
The Issues

Extrapolation

- Examples (cont)

- CSOs

- Most activity is confined to the first few contract months.
- From these we could calibration term correlations. But what do we do about inference to other non-traded pairs of contracts?
- There is a non-trivial behavior of local correlation; see the plot below for daily WTI returns correlation (2007-2010).



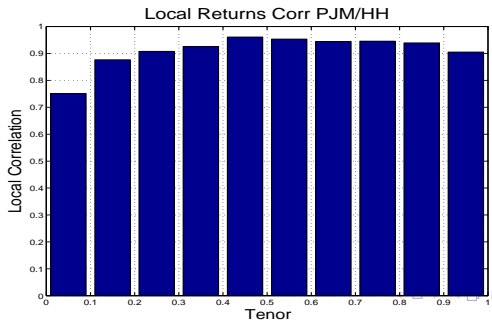
The Issues

Extrapolation

- Examples (cont)

- Tolls

- A similar question arises in cross-commodity spread options.
- If there are broker quotes for tolls it is almost always within a tenor of 2 years.
- There is a non-trivial behavior of local correlation between commodities.
- The following shows the correlation of returns between July PJM and NYMEX NG as a function of tenor to expiry (2007-2010).



The Issues

Working Problem

- Price an Oct2010 forward starting daily straddle on PJM Western Hub 5x16 power as of pricing date 03Feb2010 .
- Details:
 - The quantity will be standard lot size of 50 MW.
 - Monthly Index: ICE PJM Western Hub 5x16 settlement on '-1b'.
 - Daily Index: PJM RT 5x16 daily average price as published by the PJM ISO.
- Comment:
 - This is not a common structure in power.
 - Forward starters are more common in NG markets than fixed strike daily options.

Market Mechanics

Linear Instruments

- Fixed price swaps settle on the monthly average realized spot price for a prescribed bucket.
- Eastern US:
 - Peak bucket is 5x16 (M-F 7AM to 11PM).
 - Offpeak (the "offpeak wrap") is the complement.
 - Often circumstances require trading (or at least viewing) the offpeak wrap as 2x16 and 7x8. These are highly illiquid swaps.
 - Standard contract size: 50MW.
- Western US:
 - Peak bucket is 6x16 (M-Sat 7AM to 11PM).
 - Offpeak (the "offpeak wrap") is the complement.
 - Standard contract size: 25MW.
- Texas (ERCOT):
 - Peak bucket is 5x16 (M-F 6AM to 10PM).
 - Offpeak (the "offpeak wrap") is the complement.
 - Standard contract size: 50MW.

Market Mechanics

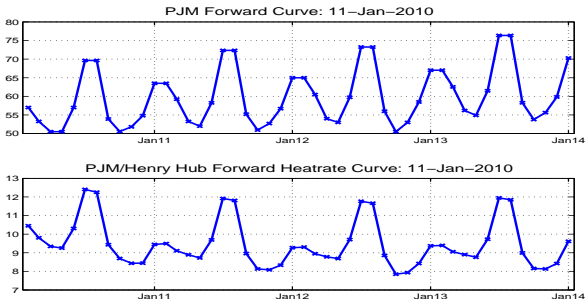
Linear Instruments: Heatrates

- The term "heatrate" refers to a ratio of power price to a natural gas price.
 - The ratio can pertain to market prices, either spot or forward ("market heatrates").
 - It can also refer (as we'll see shortly) to engineering specs of a generator.
- Power swaps often trade as heat-rates:
 - A power buy/sell is associated with a sell/buy of a specified volume of natural gas, both over the same delivery period.
 - The trade will be quoted in heatrate units: $\frac{F_P(t,T)}{F_G(t,T)}$, where P and G denote power and gas respectively.

Market Mechanics

Linear Instruments: Heatrates

- The following figure shows PJM forward prices and PJM/Henry Hub forward heatrates.
 - The seasonality in both arises from (well-founded) expectations that loads will be higher in summer and to a lesser extent winter months.
 - This forces expected clearing prices higher up the stack into more expensive units.



Market Mechanics

Options

- Annual Swaptions
 - Same structure as natural gas in mechanics/valuation.
- Monthly options:
 - Same structure as natural gas.
 - Standard expiry: '-2b'
- Daily Fixed Strike
 - Financially settling and usually manually exercised day-ahead ('-1b')
 - Example: Cash settled value of an auto-exercised call:

$$\sum_{d \in m} \max [F(d, d) - K, 0]$$

- Example: Cash settled value of standard exercise:

$$\sum_{d \in m} 1_{\{E_d\}} [F(d, d) - K]$$

where E_d denotes exercise events.

- Dominant vol exposure: daily (spot).

Market Mechanics

Options

- Monthly Options:

- Exercises at time (T_e) before the beginning of the delivery month into either:
 - A physical forward with delivery during the contract month at a price that is the strike K ;
 - Cash settlement based on the value $F_m(T_e) - K$.
- For example, the value of a call (in either case) is:
 $d(0, T)\tilde{E}(\max[F_m(T_e) - K, 0])$.

- Daily Options:

- A set of daily options usually exercising one business day before delivery.
 - These are usually financial settling on the spread between spot price and the strike: $F(t, t) - K$ where t indexes the delivery day.
- For example, the value of a call is: $d(0, T)\tilde{E}(\sum_{t \in m} \max[F(t, t) - K, 0])$.
- The notation $t \in m$ denotes the active delivery days in the option during month m . For peak (5x16) power options the sum would span the (roughly 20-22) business days in the particular month.

Market Mechanics

Options

- Forward Starters:

- As with daily options these constitute a sequence of daily options.
- The difference is that the strike is set based on the prevailing price for monthly delivery just before the month begins; hence the term "forward starter" as the strike is floating up until the delivery month.
 - We will assume that this is the same as the standard monthly option expiry date $F_m(T_e)$.
 - Settlement is on the spread between spot price and the strike: $F(t, t) - F_m(T_e)$ where t once again indexes the delivery day.
- The value of a call is: $d(0, T)\tilde{E}(\sum_{t \in m} \max[F(t, t) - F_m(T_e), 0])$.

Spot Price Dynamics

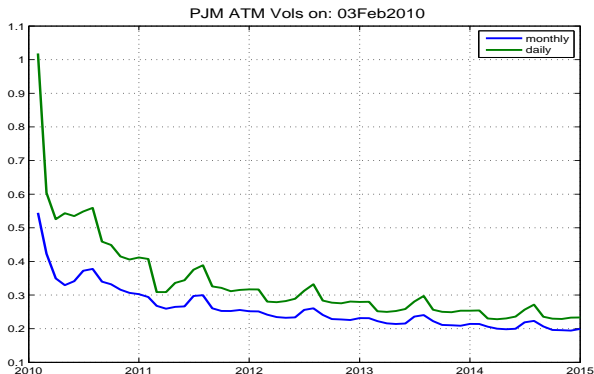
Vernacular and Convention

- The terms "monthly vol" ($\bar{\sigma}_M$) and "daily vol" ($\bar{\sigma}_D$) will always refer to implied vols pertaining to the fixed strike options (the first two categories above).
- Market dichotomy:
 - Power: Monthly and daily fixed strike options are the vanilla options.
 - Natural Gas: Monthly and the forward starter are the vanilla options.

Spot Price Dynamics

Volatility Structure in Power

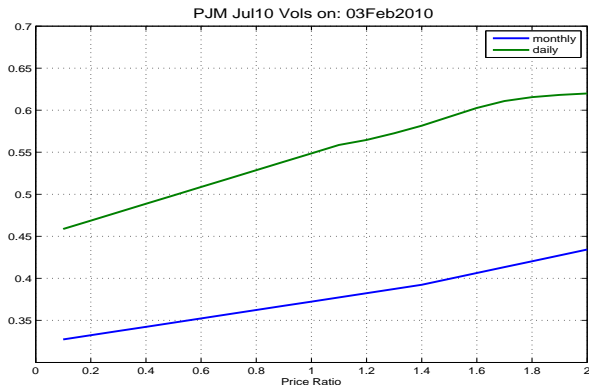
- Power and natural gas exhibit seasonality in volatility similar to seasonality in price.
 - The following figure shows the ATM monthly and daily implied vols for PJM Western Hub on 03Feb2010 .



Spot Price Dynamics

Volatility Structure in Power

- The following figure shows the skew for monthly and daily options in PJM for Jul10 .



Spot Price Dynamics

Spot Volatility

- A simple caricature of spot (daily) prices processes is for the daily returns for the daily spot prices to be i.i.d. normal:

$$F(t, t) = F_m(T_m)e^{\zeta Z_t - \frac{1}{2}\zeta^2}$$

where

- F_m denotes the contract month containing day t
- Z_t is standard normal.
- ζ is the spot volatility.

Spot Price Dynamics

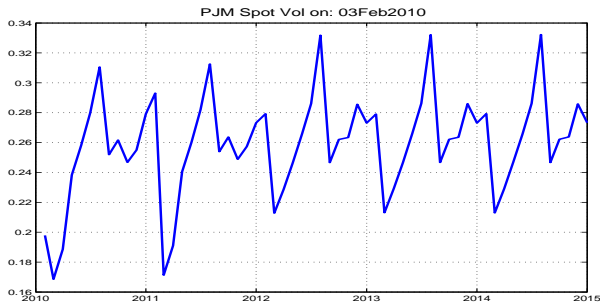
Spot Volatility

- Spot volatility is implied by the monthly and daily vols

$$\zeta^2 = \bar{\sigma}_D^2 T_d - \bar{\sigma}_M^2 T_m$$

where $T_d \approx T_m + \frac{1}{24}$.

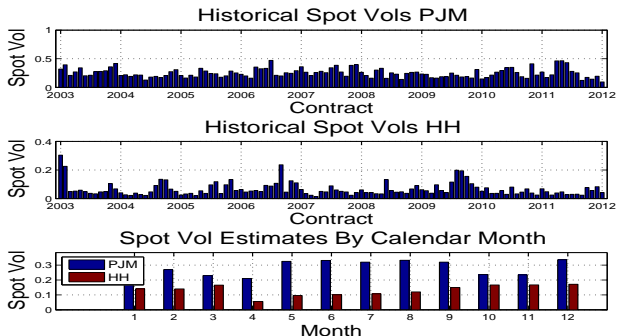
- The following figure shows the ATM spot vol by contract month for PJM



Spot Price Dynamics

Spot Volatility

- For comparison the following figure shows historical spot volatility for PJM and TETM3.
 - The reference price is the BOM contract fixing so that spot returns are defined as $\log \left[\frac{P_d}{F_m(T_e)} \right]$.



Application of a Two-Factor Model

Two-Factor Gaussian Exponential

- Recall this form:

$$\frac{dF(t, T)}{F(t, T)} = \sum_{j=1}^2 \left[\sigma_j(T) e^{-\beta_j(T-t)} dB_s^{(j)} \right]$$

- We have written this in "Big-T" form for now.
- We will often assume that the BMs are independent for simplicity.

- Intuition:

- If $\sigma_2 \equiv 0$, this is a one-factor model identical to that described in the first section:

$$\sigma(T-t) = \alpha e^{-\beta(T-t)}$$

- The second factor will typically have $\beta_2 \gg \beta_1$ and is intended to represent shorter time-scale forward returns.

Application of a Two-Factor Model

Movation

- The motivation behind this model is that:
 - The first factor will be the "long" slowly decaying perturbation to forwards.
 - The second "short" factor will have a larger β and will accommodate the relatively high values of spot volatility.
 - The previous simple model with i.i.d. spot perturbations is formally this model with $\beta_1 = 0$ and $\beta_2 = \infty$

Application of a Two-Factor Model

Calibration: General Comments

- Modeling framework:
 - Number of factors
 - Choice of β 'sdepends upon the market and what you trade.
- Moving to multi-factor models rarely eliminates the need for estimation and/or judgement regarding parameter selection.
 - There are rarely enough liquid tradeables to reliably calibrate all parameters.
 - Your goal is a model with enough flexibility to calibrate to common tradeables and get reasonably close to infrequently observed prices of less standard structures.

Application of a Two-Factor Model

Calibration: A Market Dichotomy

- Long Time-Scale Markets

- Most structures involve time-scales at or beyond monthly delivery.
- Example: Crude oil in which swaps and options refer to current nearby forward contracts.
- The primary modeling goal is valuation/hedging of cross-contract correlation structures.
- Parameters: Mean-reversion rates can be estimated by MLE using historical forwards/futures or set with by fitting Asian options and swaptions.
- Typically $\beta_1 \in [.1, .5]$ and $\beta_2 \in [1, 10]$

Application of a Two-Factor Model

Calibration: A Market Dichotomy

- Long Time-Scale Markets

- Allowing for the BMs to be correlated the two factor model implies a correlation surface:

$$\rho(T, S) = \frac{\sigma_1^2 e^{-\beta_1(T+S)} + \sigma_1 \sigma_2 \rho [e^{-(\beta_1 T + \beta_2 S)} + e^{-(\beta_2 T + \beta_1 S)}] + \sigma_2^2 e^{-\beta_2(T+S)}}{\sigma(T)\sigma(S)}$$

- Since the returns of the T contract can be written as:

$$\sigma_1 e^{-\beta_1 T} \left[dB_t^{(1)} + \frac{\sigma_2}{\sigma_1} e^{(\beta_1 - \beta_2)T} dB_t^{(2)} \right]$$

the free parameters in $\rho(T, S)$ are:

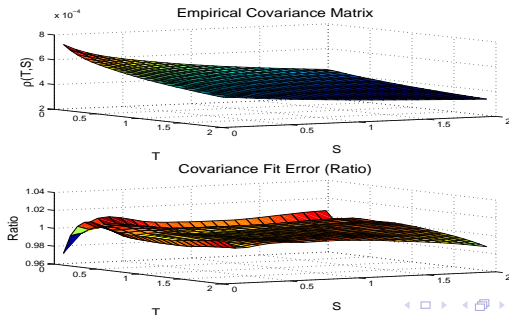
- The difference in the decay rates $\beta_2 - \beta_1$.
- The volatility ratio $\lambda = \frac{\sigma_2}{\sigma_1}$
- The correlation ρ

Application of a Two-Factor Model

Calibration: A Market Dichotomy

- Long Time-Scale Markets

- Minimizing the Frobenius norm of the difference of the empirical and model covariance matrices using WTI returns from Jan2007 to Dec2010 yields:
 - The optimal decay rates are: $\bar{\beta} = [0.106, 1.528]$
 - The estimated correlation between the factors is $\rho = 0.119$



Application of a Two-Factor Model

Calibration: A Market Dichotomy

- Multiple Time-Scale Markets:

- Common tradeables involve daily time-scales usually referencing spot prices.
- Example: Power and, often by association through tolls, natural gas.
- The primary modeling goal is valuation/hedging of cross-commodity correlation structures usually with spot (daily) price exposure.
- Parameters: Mean-reversion rates are harder to estimate due to relevance of both forward and spot prices.
- Typically $\beta_1 \in [.1, .5]$ and $\beta_2 \ggg 10$.

Application of a Two-Factor Model

Working Problem: Calibration

- We have two factors with two volatilities per contract month, namely monthly and daily volatility.
- Calibration involves setting σ_1 and σ_2 to be consistent with the observed monthly and daily option prices.
- Applying a few reasonable approximations this amounts to solving:

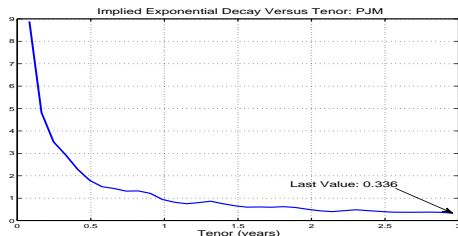
$$\begin{pmatrix} D_1 & D_2 \\ M_1 & M_2 \end{pmatrix} \begin{pmatrix} \sigma_1^2 \\ \sigma_2^2 \end{pmatrix} = \begin{pmatrix} \bar{\sigma}_D^2 \\ \bar{\sigma}_M^2 \end{pmatrix}$$

- Procedure:
 - Obtain $\bar{\sigma}_M$ and $\bar{\sigma}_D$ from market data.
 - Compute D_1, D_2, M_1, M_2 .
 - Solve the linear system above.
 - Use the resulting σ_1 and σ_2 for structure valuation.

Application of a Two-Factor Model

Mean-Reversion Rates

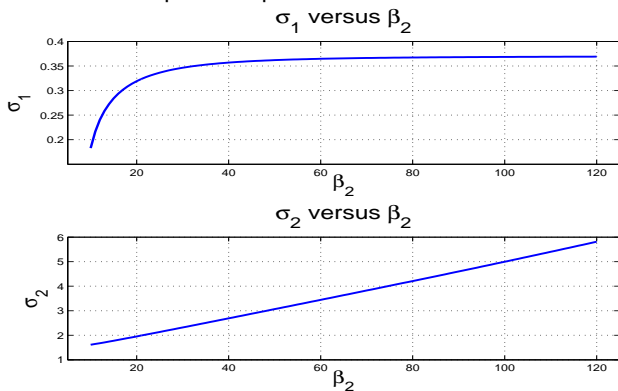
- Selection of β 's requires estimation.
 - Liquid asians or non-standard expiry options would potentially mitigate the need for an empirical basis for β selection.
 - In the absence of such we can either examine term structure of volatility or consider "fundamental" reasons for selection.
 - The following plot shows the implied β versus tenor for PJM WH from historical returns using data from 01Jan2007.
 - The asymptotic implied β is approximately .3.



Application of a Two-Factor Model

Mean-Reversion Rates

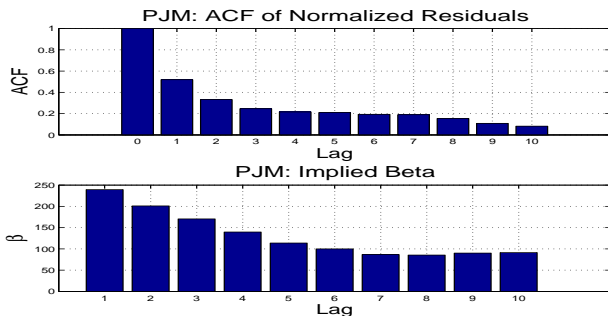
- The following figure shows σ_1 and σ_2 for Oct2010 for varying β_2 with $\beta_1 = .3$.
- Note the increase in both σ 's as the higher β_2 requires more factor vol to hit the required implied vols.



Application of a Two-Factor Model

ACF of Spot Returns

- Spot returns statistics give us some guidance to β_2 :
- What are spot returns for a non-storable commodity?
 - A useful definition for spot returns: $\log \left[\frac{P_d}{F_m(T_e)} \right]$
 - The plot below shows both the ACF as well as the implied β by lag.
 - The results are of comparable magnitude to the weather analysis.



Application of a Two-Factor Model

Working Problem

- Recall that our goal is to price a Jul08 forward starting daily option on PJM Western Hub 5x16 power.
 - We need to evaluate:

$$\tilde{E} \left[\sum_d (F(T_d, T_d) - F_m(T_f))^+ \right]$$

where T_f refers to the date of fixing ('-1b') F_m denotes the monthly contract price.

- Focusing on one particular day d and conditioning on \mathcal{F}_{T_f} we first evaluate:

$$\tilde{E} [(F(T_d, T_d) - F_m(T_f))^+ | \mathcal{F}_{T_f}]$$

Application of a Two-Factor Model

Working Problem

- Conditional Value:

- Since our processes are log-normal, this is rote application of Black for an ATM option—for example, the call value is:

$$C(T_f, F) = e^{-r(T_d - T_f)} (F_m(T_f)\Phi(d_1) - F_m(T_f)\Phi(d_2))$$

where $d_{1,2} = \pm \frac{1}{2}\sigma_{T_f, T_d}\sqrt{T_d - T_f}$.

- It remains to calculate σ_{T_f, T_d} in our multi-factor setting.
 - Note that in the "low-budget" model life is simple: $\sigma_{T_f, T_d} = \zeta$.
- The evolution of F from T_f to T_d is:

$$F(T_f, T_d) = F_m(T_f)e^{\sum_{j=1}^2 \left[\int_{T_f}^{T_d} \sigma_j(m) e^{-\beta_j(T_d - s)} dB_s^{(j)} - \text{"ITO-BLOB"} \right]}$$

- A little integration yields:

$$\sigma_{T_f, T_d}^2 (T_d - T_f) = \sum_j \sigma_j^2(m) \frac{1 - e^{-2\beta_j(T_d - T_f)}}{2\beta_j}$$

Application of a Two-Factor Model

Working Problem

- Unconditioned Value:

- We now have all components for the conditional value:

$$\begin{aligned} & \tilde{E} [\max(F(T_d, T_d) - F_m(T_f), 0) | \mathcal{F}_{T_f}] \\ = & e^{-r(T_d - T_f)} F_m(T_f) (\Phi(d_1) - \Phi(d_2)) \end{aligned}$$

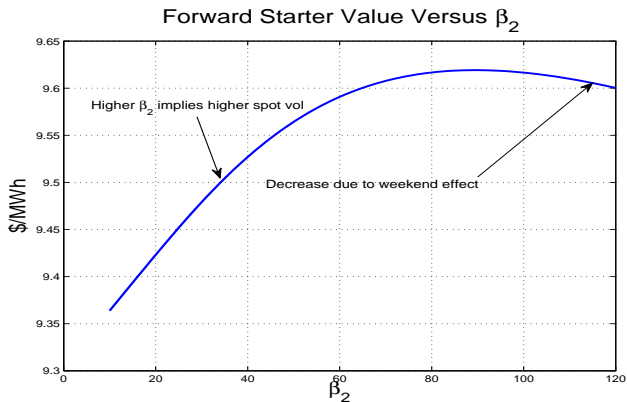
- The only random part for the unconditioned expectation is $F_m(T_f)$ and we know that

$$\tilde{E} [F_m(T_f)] = F_m(0)$$

- The hard (mechanical) part is putting this all together on precisely the right calendar days.
- The following figure shows the pricing results for various β_2 given market data as of 03Feb2010.
- The decrease in value at high values of β_2 is an artifact of weekends:
 - At high mean-reversion rates averaging vols of "daily optionlets" to hit broker vols becomes problematic.

Application of a Two-Factor Model

Working Problem



Application of a Two-Factor Model

- The modeling framework above (arguably) successfully accommodates two important features of the energy markets commonly encountered:
 - Backwardation of volatility.
 - Separation of time-scales—slow mean-reversion of general price level (at all tenors); fast mean-reversion of spot prices.
- There remain open issues:
 - Returns are log-normally distributed and skew is handled via "vol-lookup-by-price ratio."
 - The factors were uncorrelated.
 - This is trivial to fix (the formulas are simply optically less pleasant) and harder to calibrate (especially in the spot price setting).
- One other useful feature of this "HJM"-like approach:
The model is a-priori consistent with the current forward curve.

Extrapolation Taken Too Far?

A Variation of Peaking Options

- Peaking options
 - Generic expression for daily options in which the holder can select a subset of the number of delivery days for exercise.
 - These can be:
 - Auto-exercise (look-back) or manual
 - Fixed strike or forward-starting
- Working Problem:
 - The current date is 01Apr2011 .
 - Value a Henry Hub Gas Daily look-back "peaking" option paying the best J daily forward-starting call payoffs out of the 31 days in Jan12 .

Extrapolation Taken Too Far?

A Variation of Peaking Options

- Working Problem:

- Let \mathcal{I}_J denotes the subset of the 31 days in Jan12 with the largest J values of

$$\max[p_d - F_m(T_e), 0]$$

- T_e denotes the FOM price
 - t_d denotes the spot delivery time for day d
 - p_d the Gas Daily spot index for day d
- The structure will pay:

$$\frac{1}{J} \sum_{d \in \mathcal{I}_J} \max[p_d - F_m(T_e), 0]$$

- We are normalizing the payoff by $\frac{1}{J}$ to afford comparison between structures with different J

Extrapolation Taken Too Far?

A Variation of Peaking Options

- Working Problem:

- Market Data

- NYMEX natural gas for Jan12 delivery is trading at \$5.131/MMBtu.
- The ATM implied vols are: $\bar{\sigma}_M = 0.298$ and $\bar{\sigma}_D = 0.324$

- Calibration

- We will use the Gaussian two-factor model with $\bar{\beta} = [0.30, 40]$
- Calibration to the implied vols yields: $\sigma_{1,2} = [0.34, 1.17]$.

- Simulation

- Using Big-T calibration $F(t, T)$ is driven entirely by two OU processes of the form:

$$dY_j = -\beta_j Y_j dt + dB_t^{(j)}$$

- Given $F_m(0)$ and a realization of the OU processes:

$$[\bar{Y}_{T_F}, \bar{Y}_{t_1}, \dots, Y_{t_{31}}]$$

we compute $F_m(T_e)$ by integrating $F(T_e, T)$ over $T \in [T_1, T_2]$ where T_1 and T_2 are the start and end of the delivery month.

- Recalling that $F_m(T_e)$ is a bid-week price we have taken T_e to be four business days before the start of the delivery month.

Extrapolation Taken Too Far?

A Variation of Peaking Options

- Working Problem:

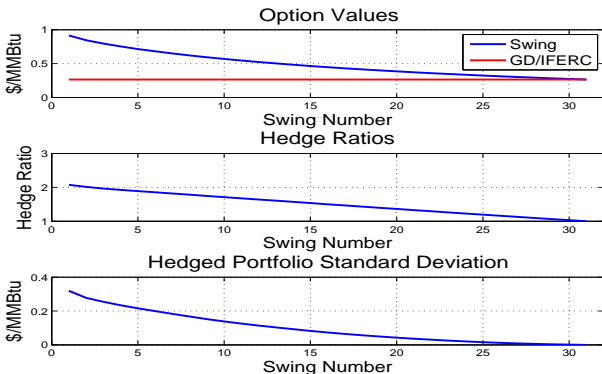
- Simulation

- The daily spot prices p_d for each delivery day d is given by $F(t_d^*, t_d)$ where t_d^* is one-business day before each delivery time t_d since the fixing for the GD index occurs for trading on this day.
- The results are shown below using 100,000 realizations (MC error $\sim .3\%$ of estimated values).
- The top figure shows the value in \$ / MMBtu as a function of swing number.
 - Note that as we have normalized by J the value decreases with J as lower realized outcomes are included in the average. structures.
- The second and third plots are related to just how effectively one can expect to hedge these structures by constructing min-variance hedges using the standard forward starter $J = 31$.

Extrapolation Taken Too Far?

A Variation of Peaking Options

- Working Problem:



Extrapolation Taken Too Far?

A Variation of Peaking Options

- The moral of the story: It is very difficult to hedge the low J structures.
 - The standard forward-starter is ineffective.
 - Δ -hedging within the month is a joke:
 - You have next day prices and (perhaps) balance of of month swaps
 - There is no balance-of-month options markets
 - All that you have really done is used the risk-neutral measure as a reference measure for hedge construction in an incomplete market setting.

Key Points

- If possible model the forward curve explicitly.
 - Spot price models can be compelling intuitively but are burdened with the need to invert inhomogenous drifts for consistency with forward curves.
- It is in general challenging to calibrate all parameters from traded instruments.
 - Long time-scale markets are more amenable to likelihood based estimation.
 - Mixed time-scale markets often rely on more fundamental analysis.
- Multi-factor models provide (arguably) plausible methods for extrapolation of correlations and forward vols.
- Beware of incompleteness in extrapolation
 - Ask the question: How would I actually hedge this payoff given the instruments available?

On Deck

- Lessons from Tolling Deals
- Variable Quantity Swaps and Econometric Models