

Modeling Divergence Swap Rates

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CBOE (2000) calculates the Volatility Index...

$$VIX_t^2 = \frac{2}{T} \sum_j \frac{\Delta K_j}{K_j^2} Q_t(K_j) \approx \frac{2}{T} \int_0^\infty \frac{Q_t(K)}{K^2} dK = -2\mathbb{E}_t^Q \left[\ln \frac{F_T}{F_t} + \sum_{\tau=1}^T \delta_{\tau-1} (F_\tau - F_{\tau-1}) \right]$$

$$D_\rho(y, x) = \frac{y^\rho - x^\rho}{\rho(\rho - 1)} - \frac{x^{\rho-1}(y - x)}{\rho - 1} \quad D_0 = \lim_{\rho \rightarrow 0} D_\rho$$

$$\forall \rho \in \mathbb{R} \quad D_\rho(y, x) = O\left(\ln \frac{y}{x}\right)^2 \text{ for } \ln \frac{y}{x} \rightarrow 0$$

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- Family of power divergence functions D_p defines family of (first-order) quadratic variation swaps. Compare with Bondarenko (2014); Martin (2012); Lee (2010).
- Weights $\frac{\partial^2}{\partial K^2} D_p(K, x) = \frac{1}{K^p}$.

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- Weights $\frac{\partial^2}{\partial K^2} D_p(K, x) = \frac{1}{K^p}$.
- HF limits:

$$\frac{1}{F_t^p} \sum_{\tau=1}^T D_p(F_\tau, F_{\tau-1}) \rightarrow \frac{1}{2} \int_t^T \left(\frac{F_s}{F_t}\right)^p \sigma_s^2 ds + \sum_{t \leq s \leq T} \frac{D_p(F_s, F_{s-})}{F_t}$$

- $1/F^p$ scaling essential \rightarrow notation: $\bar{D}_p = \frac{D_p(F_s, F_t)}{F_t^p}$

Higher-order swaps

- AJD models not bad at fitting TS of Variance Swaps → how many SV factors? Gruber et al. (2015); Andersen et al. (2015).
- Variation in ρ in divergence swaps → another dimension for fitting.

Higher-order swaps

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- Variation in p in divergence swaps \rightarrow another dimension for fitting.
- Take a difference...

$$\frac{1}{2\varepsilon} \left(\frac{D_{p_0+\varepsilon}(F_T, F_t)}{F_t^{p_0+\varepsilon}} - \frac{D_{p_0-\varepsilon}(F_T, F_t)}{F_t^{p_0-\varepsilon}} \right) = O \left(\ln \frac{F_T}{F_t} \right)^3$$

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- ... or differentiate.

$$\frac{\partial}{\partial p} \frac{D_p(F_T, F_t)}{F_t^p} = O\left(\ln \frac{F_T}{F_t}\right)^3$$

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- Rinse and repeat.

$$\frac{\partial^2}{\partial p^2} \frac{D_p(F_T, F_t)}{F_t^p} = O\left(\ln \frac{F_T}{F_t}\right)^4$$

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- Define

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- **Exactly** tradable with static option portfolio and forward trading under any dynamics.

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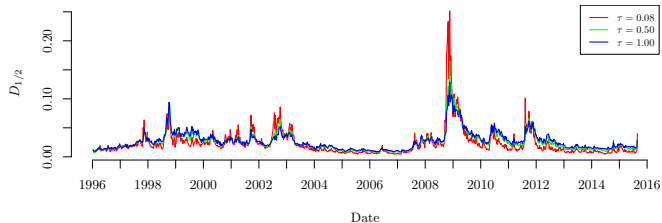
- Fully replicable realized measures allow to define higher-order risk premia.
- D_p , S_p and Q_p highly correlated \rightarrow standardise:

$$\bar{S}_p := \frac{S_p}{D_p^{3/2}} \quad \bar{Q}_p := \frac{Q_p}{D_p^2}$$

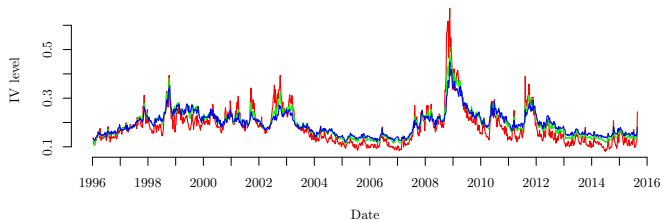
Empirical motivation – S&P 500 options

Price of divergence portfolio (annualised)

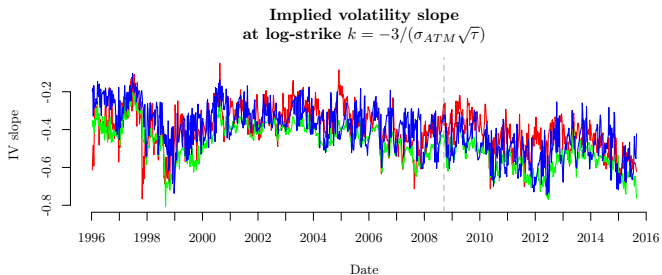
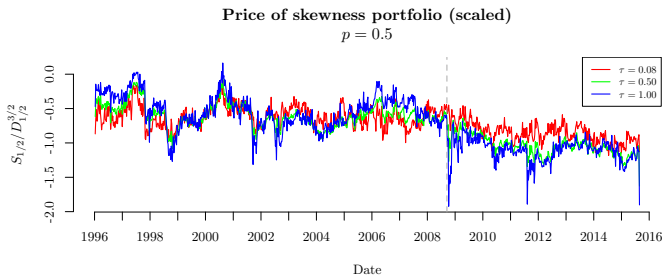
$$p = 0.5$$



Implied volatility (ATM)



Empirical motivation – S&P 500 options



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- Prices of variation swaps at range of p and τ contain the same information as the IV surface;
- Forming option portfolios might alleviate measurement error, particularly for deep OTM contracts;
- Variation swaps are exactly tradable. IV is not. Who knows what IV is anyway?

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Divergence in Affine Jump Diffusions

- Affine Jump Diffusion models are a popular (and somewhat successful) modeling tool.
- State variables: $X_t = [\ln S_t/S_{t-s} \quad V_{1t} \quad \dots \quad V_{Mt}]$, $V_t := X_t[-1]$

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- Latent volatility factor dynamics:

$$\Delta V_{t+1} = \mu_{t+1}^V + \Sigma_{t+1}^V \cdot W_{t+1}, \quad \mu_{t+1}^V, \Sigma_{t+1}^V : \text{non-linear functions of } V_t$$

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- Unscented Kalman Filter – counterfactual assumption of $W_t \sim N(0, \mathbf{I}_{M \times M})$;
- Specification of model under $\mathbb{Q} \rightarrow$ Risk Premia \rightarrow specification under \mathbb{P} ;
- Reasonable performance in simulated settings.

Preliminary results

- First estimations with $p = 1/2$ (symmetric swaps), maturities 1M, 6M;
- In three factor specification possible to identify:
 - ▶ a variance-level factor,
 - ▶ a common factor for \bar{S} and \bar{Q} ,
 - ▶ a term-structure factor driving the slopes of the \bar{S} and \bar{Q} TS.
- Existing challenges:
 - ▶ Improvements in fit of \bar{S} and \bar{Q} at the price of bad \bar{D} fit;
 - ▶ → not enough flexibility in the model's pricing properties.
- Key for improvement:
 - ▶ Careful modeling of jump distributions (Exp-Laplace co-jumps, but see Bollerslev and Todorov (2014));
 - ▶ Inclusion of purely diffusive vol factor that does not drive jumps;
- Further work:
 - ▶ Implementing more flexible jump specifications;
 - ▶ Estimation with a greater number of swap contracts (e.g. $p = 0$, more maturities);
 - ▶ Inference about risk premia.

affineModelR

Numeric backend for handling \mathbb{P} and \mathbb{Q} Characteristic Functions of (almost) arbitrarily specified Affine Jump-Diffusion models. Semi-closed form solutions for up to 3rd derivative wrt the stock price argument. Solutions for $\mathbb{E}_S^{\mathbb{P}} \left[\left(\frac{\Delta S_{t+1}}{S_t} \right)^k (\Delta V_j)^m \right]$ for $m + k \leq 2$, $m, k > 0$. Use with package `transformOptionPricer` for vanilla options.

divergenceModelR

Model-based pricing of divergence and higher-order swaps. Unscented Kalman Filters for estimation of AJD models. Builds against `affineModelR` and `ukfRcpp`.

ukfRcpp

Rcpp implementation of an Unscented Kalman Filter class. Users have to write C++ functions for handling the state dynamics and observation equation, then write an Rcpp function to create an `ukfClass` object and filter to return states or likelihood value.

Rlibcmaes by András Sali

R bindings for the `libcmaes` optimisation library.

Section 1

References

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