# Liquidity considerations in estimating implied volatility 

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Presentation at the $7^{\text {th }}$ conference of
Asia-Pacific Association of Derivatives

26 August, 2011

## Do we need a new implied volatility estimation methodology?

- The first method: ATM options, equally weighted. (CBOE VXO)
- New method: ATM+OTM options, weights are free of a specific option pricing model. (CBOE VIX)
- Why search for a new method?


## Liquidity matters

- Financial markets deliver good prices when liquidity is robust.
- Recently, there have been instances of market liquidity freezing up (eg. $6^{\text {th }}$ May Flash Crash; Sep 2008, Global Financial crisis).
- Market prices are particularly crucial then; but they have to be adjusted for vanishing liquidity.
- Even more constant, cross-sectional variation in liquid for futures and options is high.
- This is a global phenomenon, not one restricted to emerging economies






NIFTY Put options for September 2008



NIFTY Call options for September 2008


## An approach adjusting for cross-sectional liquidity

- Use all options that gives a current market price.
- Near-month and next-month maturities.
- Weight is a simple inverse of percentage spread.
- The liquidity adjusted VIX, SVIX is estimated as :

$$
\begin{aligned}
\sigma_{t j} & =\frac{\sum_{i} w_{i t, j} \sigma_{i t}}{\sum_{i} w_{i t, j}} \\
w_{i t, j} & =\frac{1}{s_{i t, j}}
\end{aligned}
$$

- Where, $s_{i t, j}$ is the spread of the $j^{\text {th }}$ option at time $t$, and $i$ is the maturity of the option, varying between near and next-month.
- This weight incorporates cross-sectional variation in liquidity, automatically adjusts the lower weights for illiquid options.


## Performance evaluation

- Candidates competiting with sVIx:
(1) VXO ,
(2) Vega-weighted VIX (VVIX),
(3) Elasticity-of-volatility-weighted VIX (EVIX)
- Benchmark: Realised volatility (RV) using intra-day returns at one-minute intervals, scaled up to a daily volatility measure.



## Performance evaluations

- Evaluations based on:
(1) Forecasting regressions (Christensen and Prabhala, 1998)
(2) MCS methodology (Hansen et al, 2003)
- Forecasting regressions:
- LHS: log of the volatility candidate
- RHS: RV
- MCS: $\log$ of the volatility candidates against each other.


## Forecasting regression results

| Volatility Indexes | $\mathrm{a}_{0}$ | $\mathrm{a}_{1}$ | Adj. $\mathrm{R}^{2}$ | $\chi^{2}$ | DW |
| :--- | ---: | ---: | ---: | ---: | ---: |
| LVXO | -0.83 | 1.17 | 0.62 | 731.1 | 1.38 |
|  | $(0.00)$ | $(0.00)$ |  | $(0.00)$ |  |
| LVVIX | -0.50 | 1.01 | 0.57 | 249.1 | 1.23 |
|  | $(0.00)$ | $(0.00)$ |  | $(0.00)$ |  |
| LEVIX | -0.69 | 1.05 | 0.43 | 269.0 | 0.99 |
|  | $(0.00)$ | $(0.00)$ |  | $(0.00)$ |  |
| LSVIX | -0.33 | 0.95 | 0.59 | 153.5 | 1.39 |
|  | $(0.00)$ | $(0.00)$ |  | $(0.00)$ |  |


| VIX | MSE | $\mathrm{p}_{T_{r}}$ | MCS $\left(\mathrm{p}_{T_{r}}\right)$ | $\mathrm{p}_{T_{S Q}}$ | MCS $\left(\mathrm{p}_{T_{S Q}}\right)$ |
| :--- | ---: | ---: | ---: | ---: | ---: |
| LVXO | 0.392 | 0.019 | 0.019 | 0.000 | 0.000 |
| LEVIX | 0.304 | 0.011 | 0.019 | 0.000 | 0.000 |
| LVVIX | 0.201 | 0.006 | 0.019 | 0.006 | 0.006 |
| LSVIX | $\mathbf{0 . 1 1 2}$ | - | $\mathbf{1 . 0 0 0}$ | - | $\mathbf{1 . 0 0 0}$ |

## Conclusion

- The liquidity adjusted VIX, SVIX, shows the
(1) Smallest bias vis-a-vis the RV,
(2) The second best $R^{2}$ value in the forecasting regression, and
(3) The best performance in the MCS tests.
- The vega-weighted VVIx has the second best MCS performance, but has the lowest $R^{2}$ in the forecasting regression.
- The vxo has the largest bias and the worst MCS performance, but shows the best $R^{2}$ fit.
- Thus, the SVIX can be taken as an improvement, with
- relatively good performance, and
- the advantage of being easier to implement compared to other existing methods that restrict the set of options used to calculate the VIX value while accounting for illiquidity.

