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A multi-horizon comparison of density forecasts for  
the S&P 500 using index returns and option prices

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## Overview of this presentation

- Context
- Contributions
- The density forecasts
- Econometric methods
- Data
- The empirical results
- Concluding remarks

## Context

- Option markets provide market expectations about the future prices of the underlying asset.
- Option traders can make use of:
  - Historical prices for the underlying asset,
  - Other, forward-looking, information.
- Hence forecasts obtained from option prices may outperform alternatives that are obtained from historical prices.

- There is a considerable literature about volatility forecasting, that compares option-based and historical forecasts.

Surveys in: Poon & Granger, 2003, Taylor, 2005.

- In contrast, there is almost no literature that compares density forecasts.

Two recent examples are:

Anagnou-Basioudis, Bedendo, Hodges & Tompkins, 2005.

Liu, Shackleton, Taylor & Xu, 2005.

- Option-based forecasts of index volatility are often found to be more accurate than historical forecasts,
  - When daily returns define historical forecasts,
  - And even when intraday returns are used:
    - Blair, Poon & Taylor, 2001,
    - Martens & Zein, 2004,
    - Jiang & Tian, 2005.
  
- We may conjecture that similar conclusions apply to density forecasts.

## Contributions

- Estimate risk-neutral densities for all horizons from option prices.

- S & P 500 index, each day from 1990 to 2004.

- Price dynamics of Heston, 1993.

- Applied previously by Bakshi, Cao & Chen, 1997.

- Contrasts with a multitude of single-horizon methods:

Jackwerth & Rubinstein, 1996, Melick & Thomas, 1997,

Ait-Sahalia & Lo, 1998, Bliss & Panigirtzoglou, 2002, etc

- Evaluate ex ante risk-transformations from risk-neutral to real-world densities.
  - Three transformations, 2 parametric and 1 non-parametric.
  - Seven forecast horizons, from 1 day to 12 weeks.
  - Contrasts with ex post transformations, and horizons that coincide with option expiry dates, in:

Bakshi, Kapadia & Madan, 2003,

Bliss & Panigirtzoglou, 2004,

Anagnou-Basioudis et al, 2005, Liu et al, 2005.

- Compare ARCH and option-based density forecasts.
  - Using either daily or intraday returns to produce historical forecasts.
  - Seven forecast horizons.
  - Previous comparisons only use daily returns and they are restricted to horizons that coincide with option expiry dates.

## The density forecasts

- Historical – Obtained from current and past index levels.
- Risk-neutral,  $Q$  – Provided by current option prices.
- Real-world,  $P$  – Given by a transformation of a  $Q$ -density.
- Mixtures – A weighted combination of historical and option-based densities.

## Historical densities

One period of time = forecast horizon.

$t$  counts time periods.

Prices  $F_t$  . Returns  $r_t = \log(F_t / F_{t-1})$ .

Four ARCH specifications for the historical density  $f(r_{t+1} | I_t)$ ,

- constant conditional mean  $\mu$ ,
- two specifications for conditional variances  $h_t$ ,
- conditional density either normal or a t-density.

- Using one-period returns:

$$r_t = \mu + \varepsilon_t$$

$$\frac{h_t}{N_t} = \omega + \frac{(\alpha_1 + \alpha_2 d_{t-1})\varepsilon_{t-1}^2 + \beta_{GJR} h_{t-1}}{N_{t-1}},$$

$$d_{t-1} = 1 \text{ if } \varepsilon_{t-1} < 0, \text{ otherwise } d_{t-1} = 0.$$

The term  $N_t$  represents the number of trading days during period  $t$ .

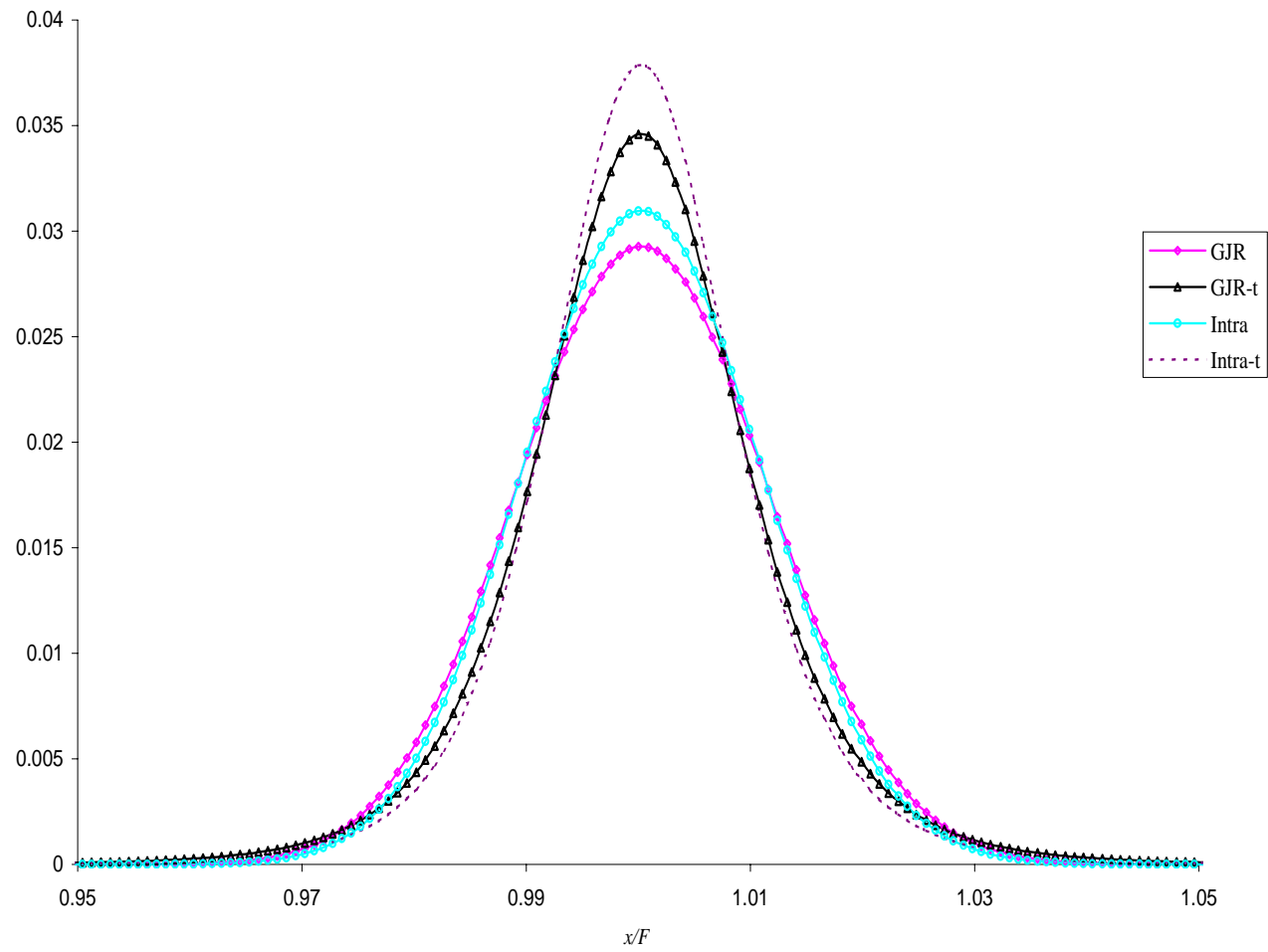
- Using intra-period returns:

$Intra_t$  is the total of some set of squared intra-period returns for period  $t$ .

$$\frac{h_t}{N_t} = \omega + \frac{\gamma Intra_{t-1} + \beta_{Intra} h_{t-1}}{N_{t-1}}.$$

**Figure 7**

**One-day ahead density forecasts obtained from ARCH models on July 30<sup>th</sup>, 2001**



## Risk-neutral densities

- Time  $t$  is now measured in years.
- The clock starts at time zero.
- Futures prices  $F(t)$  are assumed to have  $Q$ -dynamics:

$$dF/F = \sqrt{V} dW_1,$$

$$dV = \kappa(\theta - V)dt + \xi\sqrt{V}dW_2,$$

$$\rho dt = dW_1 dW_2.$$

- Special case:  $\kappa = \xi = 0$ , so  $V$  is constant. **Not in the paper yet.**

- The characteristic function is analytic and defined by:

$$\tilde{g}(\psi) = E^{\mathcal{Q}} [\exp(i\psi F(T)) | F(0), V(0)].$$

- Numerical integration provides the risk-neutral density:

$$g_{\mathcal{Q},T}(x) = \frac{1}{\pi x} \int_0^{\infty} \operatorname{Re}[\exp(-i\psi \log(x)) \tilde{g}(\psi)] d\psi.$$

- ..... and also the probabilities  $P_1(K)$  and  $P_2(K)$  appearing in the formula for European, call option prices:

$$c(K) = e^{-rT} [F_0 P_1(K) - K P_2(K)],$$

where  $K$  is the strike price.

## Real-world densities (drift-correction method)

Not included in the paper yet.

- Include a drift term  $\lambda_1 F V dt$  in the diffusion for the price  $F$ .
- And a term  $\lambda_2 V dt$  in the diffusion for the variance  $V$ .
- The real-world,  $P$ -dynamics are then:

$$\begin{aligned}dF/F &= \lambda_1 V dt + \sqrt{V} dW_1, \\dV &= [\lambda_2 V + \kappa(\theta - V)]dt + \xi \sqrt{V} dW_2, \\ \rho dt &= dW_1 dW_2.\end{aligned}$$

- To implement, must estimate  $\lambda_1$  and  $\lambda_2$  from historical data.

## Real-world densities (calibration methods)

- At time 0, suppose we consider a forecast horizon  $T$ .
- We have these risk-neutral functions for the futures price  $F_T$ :
  - The density  $g_{Q,T}(x)$  and
  - The cumulative distribution function (c.d.f.)  $G_{Q,T}(x)$ .
- Define a random variable by  $U_T = G_{Q,T}(F_T)$ . Suppose it has:
  - Real-world density  $c_T(u)$  and
  - Real-world c.d.f.  $C_T(u)$ .

- Then the real-world functions for the futures price  $F_T$  are its:
  - Density  $g_{P,T}(x) = g_{Q,T}(x)c_T(u)$ , with  $u = G_{Q,T}(x)$ ,
  - c.d.f.  $G_{P,T}(x) = C_T(G_{Q,T}(x))$ .
- These formulae can be implemented if we are willing to assume that the calibration function  $C_T(u)$  is time-invariant and hence can be estimated.
- Some calibration functions can be linked to a utility function that is employed in a representative agent model (Liu et al, 2005).

- The c.d.f. of the Beta distribution defines our preferred, parametric calibration function:

$$C_T(u) = \int_0^u v^{j-1} (1-v)^{k-1} dv / B(j, k).$$

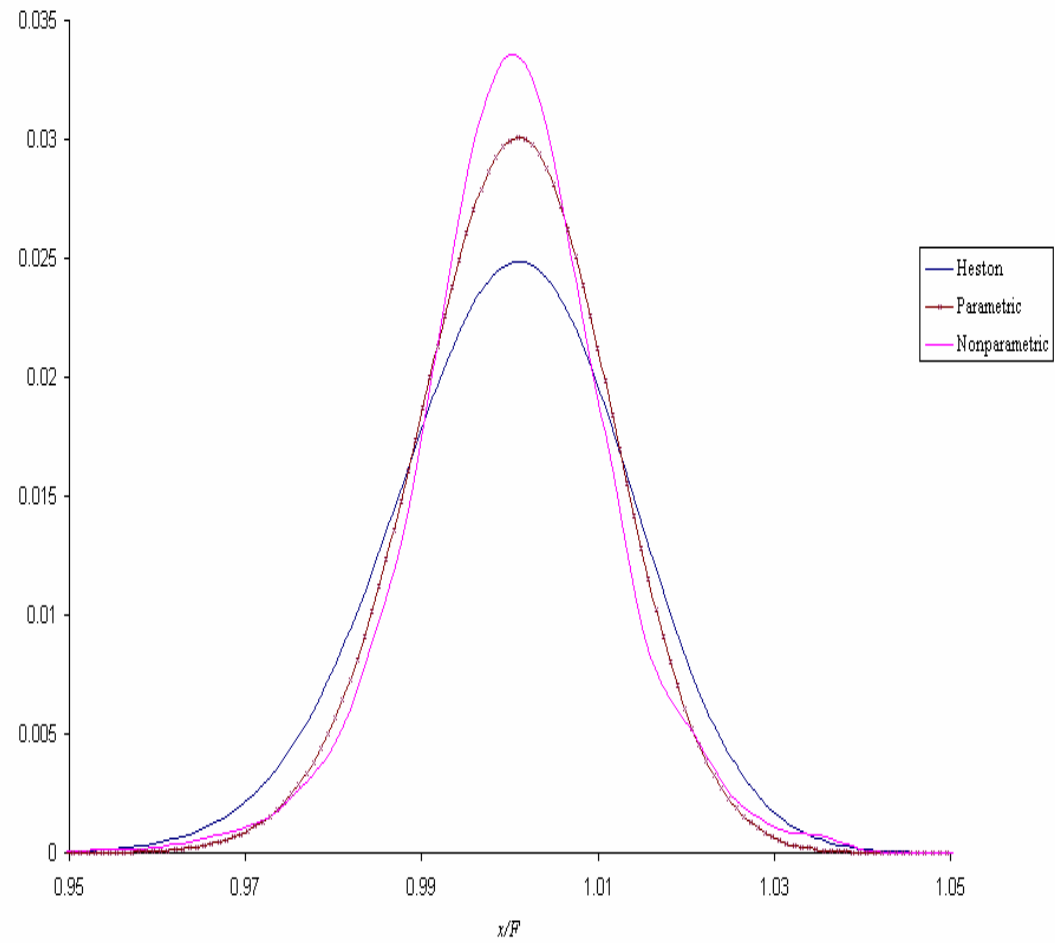
- The calibration parameters,  $j$  and  $k$ , depend on the horizon  $T$ .
  - $B(j, k) = \Gamma(j)\Gamma(k) / \Gamma(j + k)$ .
- The real-world density can be evaluated rapidly using numerical methods. It is:

$$g_{P,T}(x) = \frac{G_{Q,T}(x)^{j-1} (1 - G_{Q,T}(x))^{k-1}}{B(j, k)} g_{Q,T}(x).$$

- A non-parametric calibration function can be estimated by a standard, kernel estimator.
- Figure 8 illustrates the two calibration transformations from a  $Q$ -density to a  $P$ -density.
  - The  $Q$ -densities have variances that are higher than those of credible  $P$ -densities.
  - Consequently, the transformations reduce the variance.

**Figure 8**

**One-day ahead density forecasts obtained from options on July 30<sup>th</sup>, 2001**



## Mixture densities

The mixture density is:

$$g_{mix,T}(x) = \alpha g_{P,T}(x) + (1 - \alpha) g_{ARCH,T}(x), \quad 0 \leq \alpha \leq 1.$$

- If the “best” value of  $\alpha$  is between 0 and 1, then incremental information is provided by both the ARCH and the option-based densities.

## Econometric methods

### Parameter estimation

- Always ex ante. Parameters are re-estimated for each forecast.
- Each day, (1) the volatility for the lognormal  $Q$ -densities is simply the at-the-money, nearest-to-expiry implied volatility.
- (2) least squares provides the five parameters of the Heston  $Q$ -dynamics:
  - by matching option market prices with theoretical prices,
  - using all option expiry dates and all strikes.

- Each day, the maximum likelihood method provides estimates of:
  - ARCH parameters,
  - The parametric risk-transformation parameters,
  - The mixture parameter  $\alpha$ ,
  - Separately for each forecast horizon.
  
- The ML method is applied to all available futures prices; one price is observed at the same time-interval as the forecast horizon.
  
- Thus the density forecasts do not overlap.

## Criteria for evaluating density forecasts

- For a selected, constant forecast horizon....
- Method  $m$  provides a series of density forecasts  $g_{m,t}(x)$ ,
- Made at times  $V \leq t \leq W$ ,
- Evaluated at times from  $V + 1$  to  $W + 1$  inclusive,
- The out-of-sample, log-likelihood is used to rank methods:

$$L_m = \sum_{t=V}^W \log(g_{m,t}(F_{t+1})).$$

Diagnostic tests are applied to observed cumulative probabilities  $u$ .

- For a general method  $m$  these are defined by

$$u_{t+1} = \int_0^{F_{t+1}} g_{m,t}(x) dx, \quad V \leq t \leq W.$$

- Are observations from  $U[0,1]$ , for correctly specified densities...
- ... and are then i.i.d. for non-overlapping densities.
- We evaluate:
  - The Kolmogorov-Smirnov test,
  - The LR3 test of Berkowitz (2001).

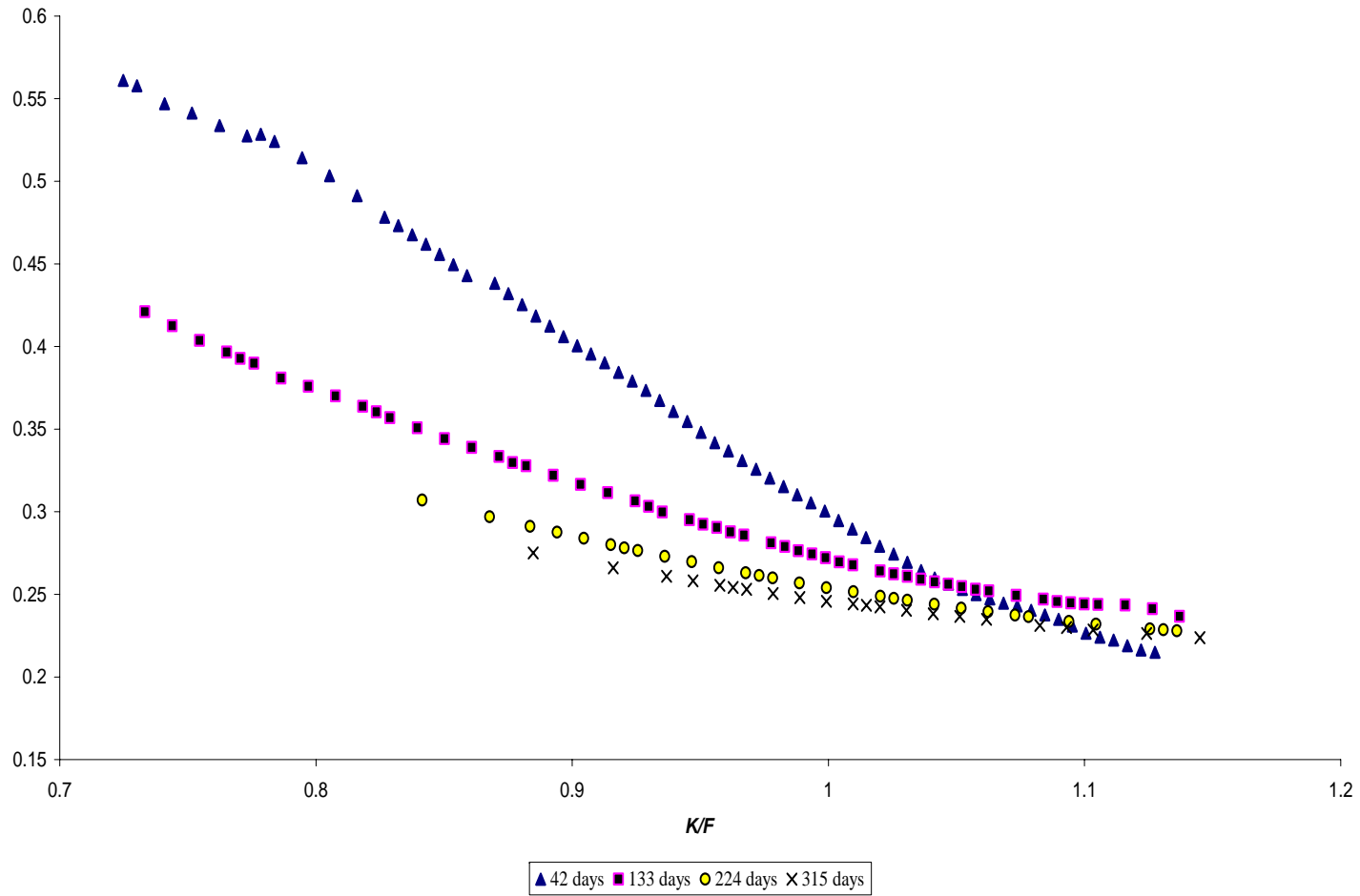
## Data

- Forecasts are made for the prices of S&P 500 index futures:
  - Mar, Jun, Sep, Dec contracts.
- Futures prices are used from 1982 to 2004:
  - Daily settlement prices,
  - Intraday prices, recorded every 5 minutes.

- Prices of American options on futures are used from 1990 to 2004:
  - Daily settlement prices.
  - All out-of-the-money call and put options.
  - European prices calculated by subtracting an estimate of the early exercise premium; the premium is often negligible.
- The daily panels of option prices contain, on average:
  - Prices for 3.1 distinct expiry dates,
  - 37 prices for each expiry date.
- Figure 1 shows typical implied volatilities.

**Figure 1**

**Implied volatilities for out-of-the-money options on November 7, 1997**



## The empirical results

Density forecasts are evaluated:

- From 1991 to 2004.
- For 7 horizons:
  - One trading day,
  - 1, 2, 4, 6, 8 and 12 weeks.
- For non-overlapping forecasts.
- With all parameters estimated ex ante.

## Historical densities

- There are no surprises.
- The usual likelihood comparisons show:
  - Intraday returns are more informative than daily returns.
  - A fat-tailed conditional density is required.
  - ... for all horizons.
- Consequently, we concentrate on the “Intra-t” specification.

## Risk-neutral densities

The median values of the Heston parameters are:

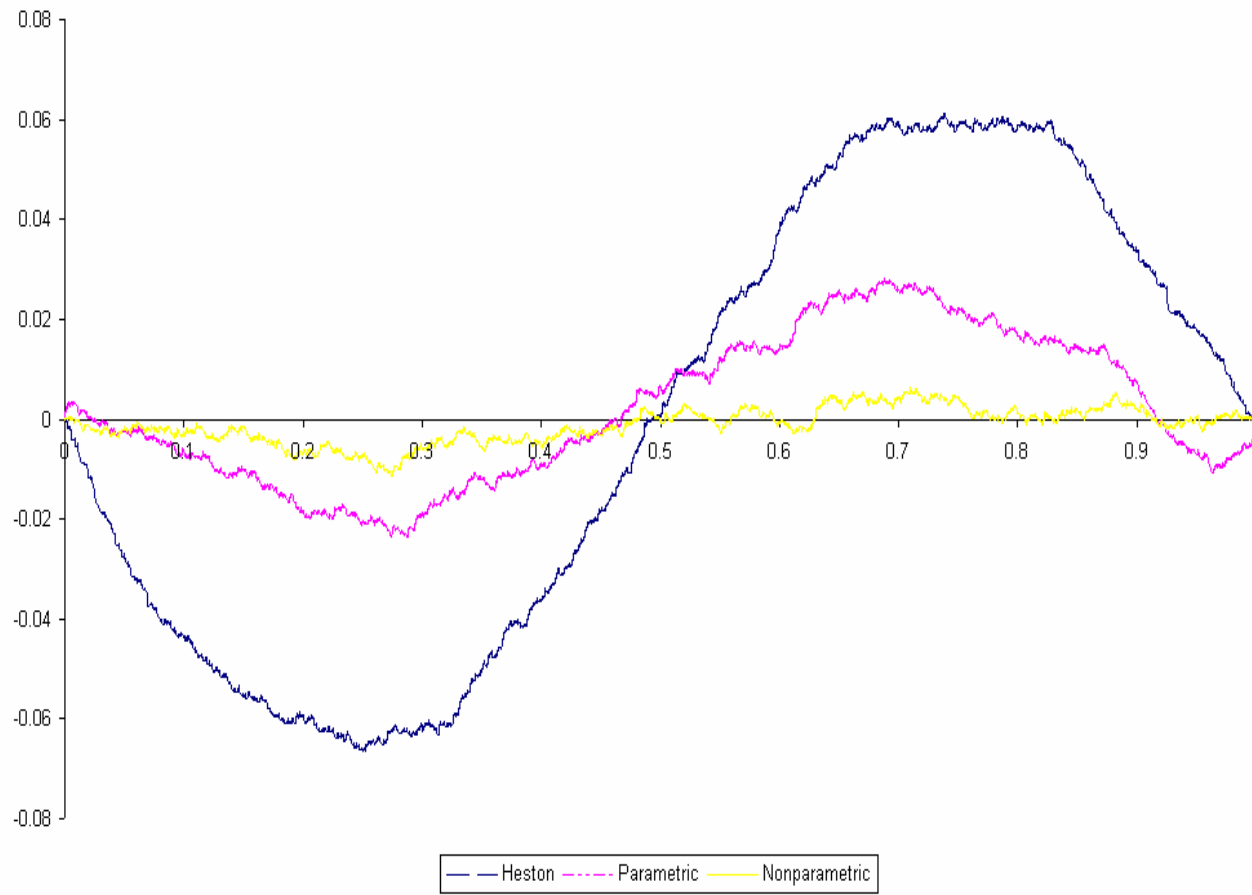
- $\theta = 0.0452$ ; the annualized volatility then reverts towards 21.3%.
- $\kappa = 4.15$ ; the “half-life” of the variance process is then 2 months.
- $\xi = 0.79$ ; this “vol. of vol.” term controls the kurtosis of returns and it is high compared with estimates obtained from returns.
- $\rho = -0.66$ ; which is a consequence of the skewed implied volatility curves.

The empirical  $Q$ -densities are not satisfactory  $P$ -densities.

- The  $Q$ -variance is higher than the real-world variance.
  - So we see too few observations in the tails of the  $Q$ -densities.
  - The “risk transformations” are therefore necessary.
- 
- Define a time series by  $u_{t+1} = G_{Q,t,1}(F_{t+1})$ .
  - Their sample c.d.f. is denoted by  $\tilde{C}(u)$ .
  - The difference between sample and uniform probabilities,  $\tilde{C}(u) - u$ , is shown on Figure 4 for the Heston  $Q$ -densities.

**Figure 4**

**The function  $\tilde{C}(u) - u$  for one-day forecasts obtained from Heston's model and risk-transformations**



## Likelihood comparisons

- Table 4 shows out-of-sample, log-likelihoods, above the GJR level.
- We rank methods using these numbers, and focus on:
  - The historical, Intra-t densities.
  - The lognormal and Heston  $Q$ -densities.
  - The parametric calibration transformation:  $P1$ -densities.
  - The non-parametric calibration transformation:  $P2$ -densities.
  - The drift-correction transformation:  $P3$ -densities.
- Our conclusions depend upon the forecast horizon.

	<u>Intra-t</u>	<u>Lognormal</u>			<u>Heston</u>			
<u>Horizon</u>		<i>Q</i>	<i>P1</i>	<i>P2</i>	<i>Q</i>	<i>P1</i>	<i>P2</i>	<i>P3</i>
1 day	123	27	73	101	-2	104	127	94
1 week	34	17	33	36	18	41	35	41
2 w.	19	14	28	26	15	27	22	26
4 w.	12	12	13	16	16	20	20	23
6 w.	16	16	17	17	20	16	19	20
8 w.	5	5	9	9	7	8	7	7
12 w.	8	6	7	7	10	8	9	8

## One-day horizon: methods ranked by log-likelihood

Heston, $P2$	127	
Intra-t	123	.... Historical data scores highly
Heston, $P1$	104	
Lognormal, $P2$	101	
Heston, $P3$	94	
Lognormal, $P1$	73	
Lognormal, $Q$	27	.... $P$ -densities are much better
Heston, $Q$	-2	than $Q$ -densities

- $P2$  is the best transformation.

### Horizons of 1, 2 and 4 weeks

- The 5  $P$ -density methods outrank the Intra-t densities (14 cases out of 15).
- The 5  $P$ -density methods have similar log-likelihoods.
- Each  $P$ -density method outranks the corresponding  $Q$ -method.

### Horizons of 6, 8 and 12 weeks

- All 8 methods provide similar log-likelihoods.
- Investigate overlapping densities to make more incisive comparisons.

## Mixture densities

Consider:  $\alpha$  \* option-based density +  $(1 - \alpha)$  \* historical density.

Selected results, for the *PI* transform of the Heston densities:

- One-day horizon:
  - Average  $\alpha$  is 0.42.
  - Reject  $\alpha = 0$  and  $\alpha = 1$  at very low significance levels.
- One-week horizon: Average 57%. Reject special cases at the 5% level.
- Longer horizons: the benefits from mixing are less obvious.

## Diagnostic test failures

The  $P2$ -densities fail the least **KS** and **LR3** tests. At the 5% level:

	<u>Intra-t</u>	<u>Lognormal</u>			<u>Heston</u>			
<u>Horizon</u>		$Q$	$P1$	$P2$	$Q$	$P1$	$P2$	$P3$
1 day		<b>K L</b>	<b>K</b>		<b>K L</b>	<b>K</b>		<b>K</b>
1 week	<b>K L</b>	<b>K L</b>	<b>K L</b>	<b>L</b>	<b>K L</b>		<b>L</b>	
2 weeks		<b>K L</b>			<b>K L</b>			
4 weeks	<b>K</b>	<b>K L</b>	<b>K</b>					

## Concluding remarks

- Informative, real-world, densities can be calculated from option prices for a general forecast horizon.
- The Heston  $P2$ -densities outperform historical densities:
  - when out-of-sample, likelihoods are compared,
  - when diagnostic test results are compared,
  - .... even when intraday returns are used.
- Mixture densities provide the best results for short horizons.