Realizing the future with R, C, and Java: A multi-language environment and GUI for high-frequency based volatility modeling *R/Finance, May 17/18, 2013*

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The Components

- heavyModel is a computing environment and GUI for the analysis, estimation, and forecasting of volatility using robust indicators of daily volatility combined with HEAVY models. The environment GUI meshes the strengths of R, C, and Java to create an easy-to-use and fast interface.
 - Realized Measures computed in R (*highfrequency* package, K. Boudt et al.)
 - General High-frEquency BAsed VolatilitY (HEAVY) model (Shephard and Sheppard, 2009)
 - Bayesian and Quasi-likelihood estimation of Heavy models (library written in gnu c using gsl)
 - Java wrapper and GUI using Java Swing by Oracle for purely graphical and 'non-coding' interface for modeling

The Setup

- Denote the daily (log) returns as r_1, r_2, \ldots, r_T , where T total days in sample.
 - Supplement information to the daily returns by a so-called *realized measure* of intraday volatility based on higher frequency data, such as second, minute or hourly data. We denote them as *RM*₁, *RM*₂,..., *RM*_T for the total number of days in the sample. Daily realized measures are an estimation of average of variance autocorrelations during a single day.
 - Easiest is the realized variance computed as $RM_t = \sum_j (X_{t+t_{j,t}} - X_{t+t_{j-1,t}})^2$ where $t_{j,t}$ are the normalized times of trades on day t. Other methods for providing realized measures includes using Kernel based methods for smoother/cleaner results.

The HEAVY Model

• With the realized measures computed for T days, the HEAVY model is given by:

$$Var(r_t | \mathcal{F}_{t-1}^{HF}) = h_t = \omega_1 + \alpha R M_{t-1} + \beta h_{t-1} + \lambda r_t^2$$

$$E(RM_t | \mathcal{F}_{t-1}^{HF}) = \mu_t = \omega_2 + \alpha_R R M_{t-1} + \beta_R \mu_{t-1}$$
(1)

where stability constraints are

- $\alpha, \omega_1 \ge 0, \beta \in [0, 1]$ and $\omega_2, \alpha_R \ge 0$ with $\lambda + \beta \in [0, 1]$ and $\beta_R + \alpha_R \in [0, 1]$.
- \mathcal{F}_{t-1}^{HF} denotes the high-frequency information from the previous day t-1.
- The first equation models the close-to-close conditional variance and is akin to a GARCH type model, whereas the second equation models the conditional expectation of the open-to-close variation.

Forecasting Volatility

 Being important for asset allocation or risk assessment, is carried out through the use of an s-step ahead iteration

$$\begin{pmatrix} h_{t+s|t-1} \\ \mu_{t+s|t-1} \end{pmatrix} = (I+B+\dots+B^s)w+B^{s+1}\begin{pmatrix} h_{t-1} \\ \mu_{t-1} \end{pmatrix},$$

where

$$B = \left(\begin{array}{cc} \beta & \lambda \\ 0 & \beta_R + \alpha_R \end{array}\right)$$

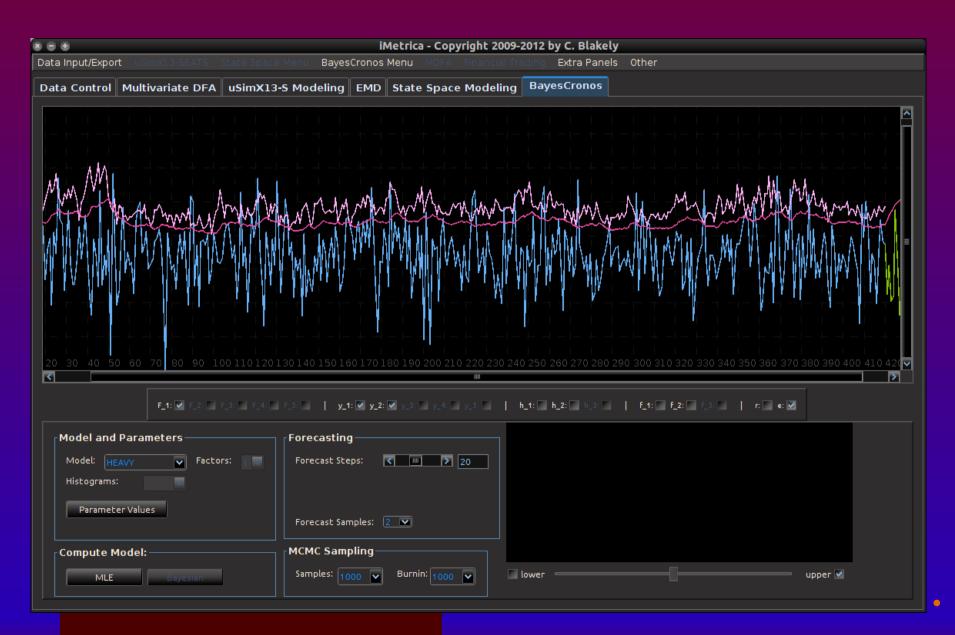
• Forecasting distributions $F(r_{t+s}|\mathcal{F}_{t-1}^{HF})$ or $F(r_{t+1} + r_{t+2} + \cdots + r_{t+s}|\mathcal{F}_{t-1}^{HF})$ done using a model-based bootstrapping method.

The Java Interface
Choose Instrument GOOG.O Compute Realized Volatility Start Date: ccyy 2011 mm 01 dd 01 End Date: ccyy 2012 mm 06 dd 19
Kernel: 'Parzen' Time Scale: minutes Period: S Lags: 3 Prepare Data for MDFA Module

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The Java Interface



The Deliverables

- The (open source) heavyModel package with the Java and C source can be downloaded at sourceforge.net/projects/highfrequency/
 - Visit my 'Hybrid signal extraction' blog at imetricablog.com for more examples and details on this package and much more
 - Personal website c-blakely.com