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## **Advances in Non-Linear Time Series Modelling**

### **Reading Seminars**

May 6, 13, 20, 2025 - 10:00-12:00, Meeting room 7

#### **1<sup>st</sup> Session – Linear and Non-Linear Structural Time Series (Prof. Andrew Harvey)**

In this session we will cover the basics of univariate linear and non-linear structural time series. The linear, and linearised, state space representation and the Kalman Filter. A revision of GARCH and EGARCH models as well as an introduction to score-driven models

Tentative Reading List

Creal, D., S. J. Koopman, and A. Lucas (2013). Generalized autoregressive score models with applications. *Journal of Applied Econometrics* 28, 777-795.

Harvey, A.C. (1993) *Time Series Models*. 2nd Edition, The MIT Press. Ch. 4, 5 and 8

Harvey A. C. (1990). *Forecasting, Structural Time Series Models and the Kalman Filter*. Cambridge University Press; 1990. Ch.3

Harvey, A. C. (2013). *Dynamic models for volatility and heavy tails: with applications to financial and economic time series*. Econometric Society monograph. Ch. 1,2 and 3

Harvey, A (2022) Score-driven time series models. *Annual Review of Statistics and Its Application*, 9, 321-42. doi: 10.1146/annurev-statistics-040120-021023

R. S. Tsay (2005), *Analysis of Financial Time Series*, 2nd Edition, John Wiley & Sons, New York. Ch. 3

#### **2<sup>nd</sup> Session – Score-Driven Models for Volatility and Time Varying Tails (Dario Palumbo)**

In this session we will be focussing on score driven models for modelling time varying volatility in non-linear heavy-tailed time series and their ability to capture properties of financial data. In the lecture we will cover the beta-t-EGARCH model for time varying volatility, also in the presence of time varying location. We will introduce the location/scale GB2 score-driven model for modelling realized volatility of financial assets and their distribution. At last, we will present a score-driven approach for the modelling of time-varying tails and the occurrence of extreme events. In doing so we will compare the performance of score-driven filters in handling heavy-tailed time series and empirically compare their performance against classical methods for the modelling of volatility and extreme events.

Essential Reading List

Harvey, A. C. (2013). *Dynamic models for volatility and heavy tails: with applications to financial and economic time series*. Econometric Society monograph. Ch. 3-5.

Creal, D., Koopman, S.J. and Lucas, A. (2013), Generalised Autoregressive Score Models with Applications. *Journal of Applied Econometrics*., vol. 28, pg. 777-795.

Harvey, A. C. and A. Luati (2014). Filtering with Heavy Tails. *Journal of the American Statistical Association*, vol. 109, no. 507, pg. 1112–22

Harvey, A. C. and D. Palumbo (2023). Score-driven models for realized volatility. *Journal of Econometrics* 237, vol. 2 (Part B), 105448.

Lucas A. and X. Zhang (2016) Score-driven exponentially weighted moving averages and Value-at-Risk forecasting, *International Journal of Forecasting*, vol. 32, no. 2, pp. 293-302.

Palumbo, D. (2021). Testing and Modelling Time Series with Time Varying Tails. *Cambridge Working Paper Series*. <https://doi.org/10.17863/CAM.65414>

#### Extended Reading List

Harvey, A., and Lange, R.-J. (2017) Volatility Modelling with a Generalized  $t$  Distribution. *Journal of Time Series Analysis*, vol. 38, pg. 175–190.

Corsi, F. (2009). A simple approximate long-memory model of realized volatility. *Journal of Financial Econometrics*, vol. 7, pg. 174-196.

#### Follow-up Reading List

Blasques, F., Gorgi, P., Koopman, S. J. and Wintenberger, O. (2018): Feasible invertibility conditions and maximum likelihood estimation for observation-driven models, *Electronic Journal of Statistics*, vol. 12, pg. 1019-1052.

Blasques, F., van Brummelen, J., Koopman, S. J. and Lucas, A. (2022): Maximum Likelihood Estimation for Score-Driven Models, *Journal of Econometrics*, vol. 227, n. 2, pg. 325-346.

Opschoor, A., Janus, P., Lucas, A., & Van Dijk, D. (2017). New HEAVY Models for Fat-Tailed Realized Covariances and Returns. *Journal of Business & Economic Statistics*, vol. 36, n. 4, pg. 643–657.

### **3<sup>rd</sup> Session - Multivariate Score-Driven Models and the Modelling of the Term Structure of Interest Rates (Dario Palumbo)**

This session explores multivariate score-driven models for heavy-tailed multivariate time series. We will introduce the general framework of multivariate score-driven models, emphasizing their ability to capture time-varying volatilities, correlations, and dynamic dependencies across multiple financial variables. The discussion will highlight the advantages of score-driven filters in high-dimensional settings and their relationship with classical multivariate GARCH and dynamic conditional correlation (DCC) models. A key focus will be on the application of score-driven methods to interest rate modelling, particularly in forecasting the term structure of government bond yields. We will examine a score-driven filter for extracting and modelling the term structure of interest rates directly from bond prices. This will include a score-driven extension of term structure factor models, such as the dynamic Nelson-Siegel approach and its variants. Empirical applications will illustrate how these models enhance the characterization of risk dynamics, capturing shifts in macro-financial conditions, time-varying volatility, and the effects of the zero lower bound with greater accuracy.

#### Essential Reading List

Harvey, A. C. (2013). *Dynamic models for volatility and heavy tails: with applications to financial and economic time series*. Econometric Society monograph. Ch. 7

Creal, D., S. J. Koopman, and A. Lucas (2011). A dynamic multivariate heavy-tailed model for time-varying volatilities and correlations. *Journal of Business & Economic Statistics*. Vol. 29, n. 4, pg. 552-563.

Lucas, A., B. Schwaab and X. Zhang (2014) Conditional Euro Area Sovereign Default Risk. *Journal of Business & Economic Statistics*, Vol. 32, No. 2 (April 2014), pp. 271-284

Diebold, F. X. and C. Li (2006). Forecasting the term structure of government bond yields. *Journal of Econometrics*, vol. 130, n. 2, pg. 337-364.

Koopman, S., A. Lucas, and M. Zamojski (2017). Dynamic term structure models with score-driven time-varying parameters: estimation and forecasting. *NBP working paper 258*.

#### Extended Reading List

Bollerslev, T. (1990). Modelling the Coherence in Short-Run Nominal Exchange Rates: A Multivariate Generalized Arch Model. *The Review of Economics and Statistics*, vol. 72, n. 3, pg. 498-505.

Engle, R. (2002) Dynamic Conditional Correlation: A Simple Class of Multivariate Generalized Autoregressive Conditional Heteroskedasticity Models *Journal of Business & Economic Statistics*, vol. 20, n. 3, pg. 339-350

Bauwens, L., S. Laurent, and J.V.K. Rombouts, 2006), Multivariate GARCH models: a survey. *Journal of Applied Econometrics*, vol. 21, pg. 79-109.

Diebold, F. X. and G. D. Rudebusch (2013). *Yield curve modeling and forecasting: The dynamic nelson-siegel approach*. Princeton University Press.

D’Innocenzo, E., A. Luati, and M. Mazzocchi (2023). A robust score-driven filter for multivariate time series. *Econometric Reviews*, vol. 42, n. 5, pg. 441-470.

Gasperoni, F., A. Luati, L. Paci, and E. D’Innocenzo(2021). Score-Driven Modeling of Spatio-Temporal Data. *Journal of the American Statistical Association*, vo. 118, n. 542, pg. 1066–1077.

#### Follow-up Reading List

Krippner, L. (2015). *Zero lower bound term structure modeling: A practitioner’s guide*.

Monfort, A. and F. Pegoraro, (2007) *Switching VARMA Term Structure Models*, *Journal of Financial Econometrics*, vol. 5, n. 1, pg. 105–153.

Monfort, A., F. Pegoraro, J. P. Renne, G. Roussellet, (2017) Staying at zero with affine processes: An application to term structure modelling, *Journal of Econometrics*, vol. 201, n. 2, pg. 348-366.