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for Electricity Price Formation**

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Abstract

The wide range of models needed to support the various short-term operations for electricity generation demonstrates the importance of accurate specifications for the uncertainty in market prices. This is becoming increasingly challenging, since electricity hourly price densities exhibit a variety of shapes, with their characteristic features changing substantially within the day and over time, and the influx of renewable power, wind and solar in particular, has amplified these effects. A general-purpose, analytically tractable representation of the stochastic price formation process would have considerable value for operations control and trading, but existing empirical approaches or the application of standard density functions are unsatisfactory. We develop a general four parameter stochastic model for hourly prices, in which the four moments of the density function are dynamically estimated as latent state variables and furthermore modelled as functions of several plausible exogenous drivers. This provides a transparent and credible model that is sufficiently flexible to capture the shape-shifting effects, particularly with respect to the wind and solar output variations causing dynamic switches in the upside and downside risks. Extensive testing on German wholesale price data, benchmarked against quantile regression and other models in out-of-sample backtesting, validated the approach and its analytical appeal.

JEL codes: C01, C21, C22, C32, C53, Q41, Q47

Keywords: Electricity Prices, Density Estimation, Skewness, Quantiles, Risk

1. Introduction

Price formation in wholesale electricity spot markets is known to be a complex function of many fundamental drivers, interactions, time-varying specifications and stochastic shocks. Various factors characterise the idiosyncratic dynamics, and the reasons why the stochastic models for price formation may be challenging to formulate have invited many explanations, see Lucia and Schwartz (2002), Knittel and Roberts (2005), Chen and Bunn (2010), Panagiotelis and Smith (2008), Benth et al. (2013), Aïd et al. (2013) and Weron (2014) among others.

In particular, power markets are local and resource-dependent. In some markets, the production of electricity may be a commodity spread between gas, oil or coal; in others it

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may be a result of infrastructure investments in nuclear facilities or large reservoirs, whilst elsewhere, and increasingly, it relates to the use of renewable resources such as wind, solar, hydro, biomass, geothermal or tidal currents. Furthermore electricity is produced to meet demand instantaneously; it is not easily storable, and in responding to inelastic consumers, the prices are prone to exhibit substantial volatility. And, with liberalised power markets being far from perfectly competitive, often composed of a small oligopoly of generators, at times of scarcity market power effects can result in price spikes substantially above fundamental levels. In contrast and increasingly, with some producers of electricity having inflexible production facilities, eg district heating facilities or nuclear plants, their aversion to shut down/start-up costs may incentivise them to make negative offers to the market during transient periods of oversupply (particularly from wind), resulting in “downspikes”.

Thus, to the extent that the underlying commodity properties dominate the price formation and these may be nonstationary, power prices will accordingly follow them as random walks; but if the natural renewables dominate, mean reversion will emerge from the dynamics of weather or transient supply outages. Unsurprisingly, therefore, specification tests on daily power series for mean reversion, unit roots, fractional cointegration or trend stationarity have varied in their indications over time, between locations and according to whether spikes have been trimmed out of the data (eg Haldrup and Nielsen (2006); Escribano et al. (2011); De Vany and Walls (1999); Koopman et al. (2007); Bunn and Gianfreda (2010); Nan et al. (2014)). Furthermore, with many markets being in structural transition, as governments seek to incentivise the replacement of fossil fuels with renewables, price formation will veer between different processes as wind, solar and hydro availabilities fluctuate. Finally, the granularity of power markets is fine and, because of the lack of storage, arbitrage between price formation at different times of the day or year is very restricted; thus we see quite different price distributions at offpeak hours during the night from those in the morning, midday or evening peaks. In the context of all of this, therefore, it is easily understandable why attempts to model the power price processes have led to different models for different times of the day and seasons, with regime switching and time varying specifications, nonlinear formulations as well as skewed and fat-tailed distributions all having been applied.

The complexity and evolutionary nature of these various influences on power price formation is well illustrated by the changing shapes of the German price densities since 2007. The German power market provides the main reference for European prices and, having also been at the forefront with its high penetration of renewable energy, is the most attentively observed in the region. In Figure 1 we display three selected daily time series for hours 3, 12 and 19 in 2007, and contrast these with the same series only four years later in 2011. The 2007 series exhibit the conventional patterns of a fossil fuel dominated power system, mostly coal and some gas at that time, with periods of volatility clustering and positive spikes. In contrast, 2011 shows the situation after some substantial penetration by wind and solar facilities. The price distributions are remarkably different. The predominantly positive skewness has transformed to negative skewness. In the supplementary Appendix 6.1, we display the series for all 24 hours and these fully demonstrate the diversity and rapid evolution in time series properties. The penetration of solar facilities in particular, by residential and commercial end-users, is continuing to diminish the midday need for conventional generation and eroding what used to be a daily peak, see Moody’s (2012). Thus, there is complexity in evolution, which materialises annually, as well as the time of day distinctiveness becom-

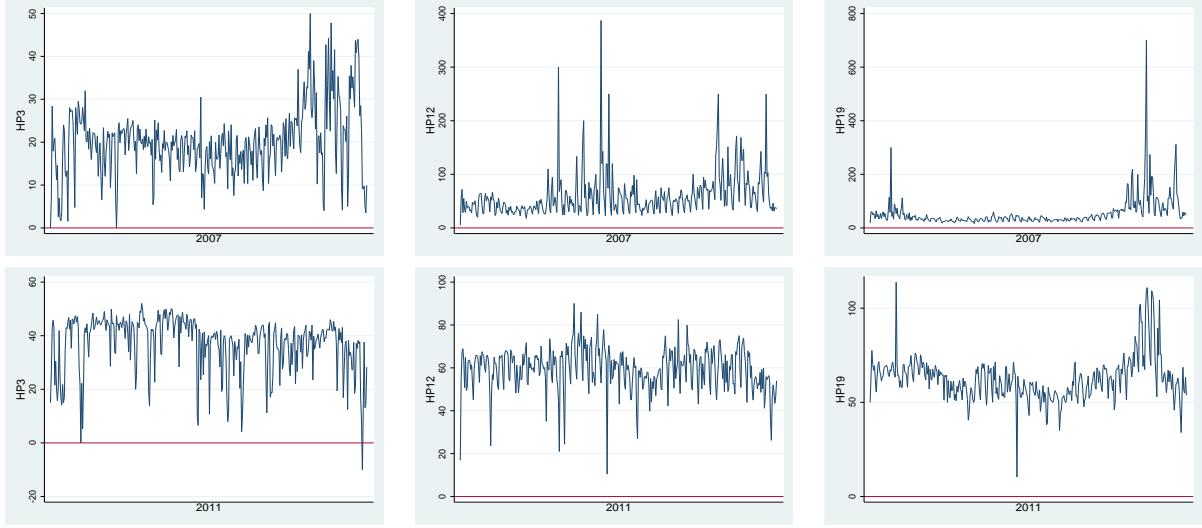


Figure 1: Daily time series for electricity prices (in €/MWh) for hours 3, 12 and 19 in 2007 (first row) and in 2011 (second row). Data source: EPEX (www.epexspot.com)

ing increasingly sensitive to changes in the weather. It is therefore a research question of considerable practical relevance to evaluate if the various hourly price densities can be determined, not apparently as empirical idiosyncrasies, but as variations of a general parametric stochastic process, the parameters of which drive the appropriate changes in shape through their dependence upon the evolution of fundamental exogenous factors. That is the aim of this study.

To pursue this, we have searched for a general three or four parameter density specification with special stochastic and analytical features. From a large selection of distribution functions, we identified a few that are sufficiently flexible to provide adequate fit to the wide range of shapes displayed in the German hourly prices over 2006–2016. Furthermore, using generalized linear multivariate estimation for the four moments, sufficient to define the distributions, we were able to relate these moments to several key fundamental dynamic drivers in a plausible way, as well as to an autoregressive representation of the latent estimates to capture behavioural persistence. With respect to the latter feature, the formulation thereby represents to some extent the widely used conditional heteroscedasticity approach for volatility estimation and generalises it to the higher moments as well; the persistence of stochastic skewness being the particular new feature of power prices that we wish to capture. Overall, the latent four moment dynamic modelling within a Skewed t density representation was considered most appropriate on the balance of its empirical performance and closed form analytical properties. Day ahead predictive densities were then recursively estimated out-of-sample and benchmarked successfully against analogous quantile regression and other methods.

In the next section we consider the practical relevance of more accurate hourly price formation modelling for short-term electricity operations and then review some of the related background research on electricity price modelling, forecasting and density estimation. In Section 3, we discuss the case of Germany and present key results from the distributional analysis of power prices and fundamental drivers. The multifactor modelling approach is

described in Section 4, with the empirical support and the dependence of higher moments on fundamental drivers in Section 4.1, and the relative performance in forecasting density quantiles compared with other benchmarking techniques in Section 4.2. Finally, we summarise the research contributions in Section 5.

2. Operational Contexts of Power Price Density Modelling

Modelling the day-ahead electricity price formation process at hourly resolution has attracted extensive research (see Weron (2014), for a review) motivated by the practical considerations of a wide range of operational decisions for which these models provide support. Typically, in most wholesale power markets worldwide, the main-marker, day-ahead prices emerge as a vector for all 24 hours of a particular day from auctions held around midday of the previous day. Whilst there are usually some demand and supply responses in the subsequent intra-day trading and real-time balancing, substantial operational commitments in practice need to be planned in advance of the day ahead auction, but with outcomes contingent upon those prices. Forecasts or simulations based upon the price formation models of the day-ahead auction prices are therefore essential to such short-term advance planning.

For example, unit commitment decisions for production facilities are often done days in advance, especially if single or two shift daily schedules are being considered (Hobbs et al., 2001; Tseng and Barz, 2002). From a risk management perspective, Stoft (2002) argues that day ahead hourly price forecasts are crucial for generators making offers to the day ahead auction in a way that recovers start-up costs over their expected dispatched periods. With the relative attractiveness of the intra-day and balancing markets, many generators face the decision problem of how much capacity to offer to the day ahead auction and how much to retain for the intra day and balancing opportunities (Soares et al., 2017; Ding et al., 2017). This will depend upon the relative price risks. The economic operation of gas-fired plants require positive spark spreads and the forecasts of power prices will therefore influence not only the offer strategy to the electricity auctions, but also planning the linked activities in the day ahead gas market, pipeline commitments and/or calls upon any swing option contracts for variable gas off-takes (Eydeland and Wolyniec, 2003; Harris, 2006, Jaillet et al., 2004). Similarly, when trading across interconnectors, transmission capacity may have to be acquired in advance of the day ahead auctions, the value of which will be a real option on the anticipated locational spreads across the day-ahead auction prices (Deng et al., 2001; Carmona and Durrleman, 2003; Bunn and Martoccia, 2010). All of these operational decisions are made in advance of the day-ahead auction prices, often with an element of optionality, and as a consequence, their valuations depend upon the probability densities of the hourly prices. The specifications of the stochastic hourly price-formation in the models that have been presented to support these decisions are often, however, quite simple mean-reverting (Ornstein-Uhlenbeck) or seasonal autoregressive processes, suitable for long-term analysis, but without any conditional dependence upon the exogenous factors such as weather and demand forecasts, as well as fuel prices, that are highly informative in the short term.

We expand on three specific illustrative contexts that are currently being actively researched and where short term conditional models of price formation would appear to be crucial for model adequacy:

- *Optimal Battery Storage and Electric Vehicle Charging Operations:* The operation of an electricity pumped storage facility on a daily cycle, based upon day ahead prices, is a well-established textbook example of optimised operational planning (Sioshansi and Conejo, 2017). Recently the linking of batteries to wind facilities has engaged various researchers in the application of stochastic optimisation techniques (Kim and Powell, 2011, Li et al., 2011, Jiang et al., 2013, Ding et al., 2016, Abdullah et al., 2017). The two stage models of Ding et al. (2016) and Abdullah et al. (2017) make use of the 24 day-ahead forecasts, but these forecasts are derived simply as historical densities. Many papers related to operating electric vehicle charging facilities have similar properties. Typical formulations have been as optimal stopping rules for charging and discharging as a function of mean reverting Gaussian spot prices (Jiang and Powell, 2016). In practice, decision making is likely to be more episodic, as with storage, based upon daily expectations for prices (Sioshansi, 2011). The day-ahead storage optimisation is particularly challenging because valuation depends upon a spread for charging and discharging and the agent has to decide upon both bid and offer hourly quantities for the auction. Accurate price risk modelling is clearly important to operate profitably across these short-term spreads.
- *Risk Management by Renewable Energy Producers:* In the absence of a link to storage, wind or solar producers can face considerable short term revenue risk from both prices and output volumes. This can be managed by means of a financial product that has payoffs to cover low volumes and low prices. Such an option contract is an example of an energy “quanto” option¹, extensively discussed in Caporin et al. (2012), Benth et al. (2015) and Brik and Roncoroni (2016). These are offered as bespoke weather derivative products by the insurance industry (eg Munich Re² and Endurance Re³). To understand the payoffs from such options, their prices and the optimal design of strikes, a joint stochastic model for the power price and production (Wind or PV) is required. Typically solved by Monte-Carlo methods, not only are stochastic models for wind and prices needed, but also their correlation. In Benth and Ibrahimb (2017) a simple AR(3) model with constant Gaussian noise is assumed for the hourly prices. Other derivative products, specific for particular hours, designed to help with the short term uncertainties for renewable energy producers include the “cap/floor futures” which have been introduced in Germany⁴ and Australia⁵. These effectively manage the risk, on an hourly basis, of high and low prices, above and below specified thresholds. Pricing these derivatives evidently requires accurate estimation, particularly regarding the tails, of the density functions involved.

¹A long term version of this is sometimes called a *Proxy Revenue Swap*, eg for Capital Power’s Bloom Wind Farm (see <https://www.environmental-finance.com/content/sections/weather-risk-hub/weather-is-the-new-fuel-risk.html>).

²<https://www.munichre.com/weatherandcommodity/en/group/index.html>

³https://platts.com/IM.Platts.Content/ProductsServices/ConferenceAndEvents/emea/EU-Power/presentations/Ralph_Renner.pdf

⁴<https://www.eex.com/en/products/energiewende-products/german-intraday-cap-futures>

⁵https://www.asxenergy.com.au/products/overview_of_the_australian_el

- *Demand-side Engagement:* Operational interest in day-ahead price extremes is also reflected on the demand-side and can be expected to increase with more consumer empowerment and distributed resources. Exposure to price extremes and their operational remedies can often be quite specific and detailed. For example, in Britain, drawing upon the theory of peak load pricing, the TSO recovers transmission charges from the demand side based upon their consumption in the highest three, non-consecutive trading periods in the winter (the so-called “triads”, National Grid, 2015). There is a large incentive to reduce demand in these periods but they are only known ex post, and commercial forecasting services have emerged to provide forecasts⁶. Generally, small distribution-connected turbines are started and run to reduce the net demand of retailers during these periods. But, it has been estimated (Frontier, 2017) that the search cost for these extremes involves targeting over 100 trading periods to ensure coverage of the maximum 3. This is one specific example but, evidently, improved modelling of the extreme price risks has benefits beyond this “triad chasing” to the extent that suppliers can influence demand-side engagement in a more timely and economic manner.

Regarding the need for more accurate price formation models in the above, and other, operational contexts, although there is a large amount of published work on modelling the fundamental drivers of power prices and a growing body of work on the disruptive effects of renewable generation, there is relatively little on the specification of electricity density forecasts. Most of the research on price formation has been in terms of relating their expected values to exogenous factors, such as fuel prices, demand, reserve margin as well as lagged effects, and the model formulations have often justified nonlinear, regime-switching and time-varying specifications (eg Huisman (2008); Karakatsani and Bunn (2008); Chen and Bunn (2010), amongst many). In parallel, stochastic models, often motivated by the need to support derivative pricing, have become increasingly elaborated to take account of non-normality, jumps and mixed processes (eg Benth et al. (2007); Panagiotelis and Smith (2008); Frestad et al. (2010)).

Regarding the particular changes in price formation induced by wind, Gelabert et al. (2011), looking at Spain, demonstrated the negative effects of renewables on price levels. In Texas, similar evidence of negative effects is presented by Baldick (2012), and also by Woo et al. (2011), the latter observing that an increase in wind generation reduces electricity prices but increases the variance and this happens to varying extents throughout the day. The price-wind-demand interrelationship is discussed in the Australian context by Cutler et al. (2011), in which they observe a general lack of correlation between wind and demand, emphasise that demand is the more important driver, but also note periods of low (high) market prices associated with high (low) wind generation at all hours of the day. The additional effects of regional imports and exports, induced by wind variations, have also been investigated by Mulder and Scholtens (2013) in The Netherlands, and by Mauritzen (2013) who identified Nordic hydropower as a natural complement to Danish intermittent wind generation.

From a forecasting perspective, Cruz et al. (2011) compare the predictive accuracy of

⁶Npower Triad Warning Service <https://www.npower.com/business-solutions/buying-energy/demand-management/triadwarningservice/> and Flexitricity Triad Management <https://www.flexitricity.com/en-gb/solutions/triad/>

several univariate, multivariate, linear and nonlinear models for Spanish day-ahead prices, including hourly load and wind generation forecasts as explanatory variables, with results justifying the multivariate specifications. Similarly, Kristiansen (2012) developed a forecasting model for hourly day-ahead prices in Nord Pool based on an autoregressive model with exogenous variables, with the extended specification adding value to the previous work of Weron and Misiorek (2008).

With respect to the evolutionary nature of the model specifications, Paraschiv et al. (2014) considered the effects of both wind and solar generation on the day-ahead price formation in Germany, showing that there has been a continuous electricity price adaption process to market fundamentals, and that price drivers differ across hours with solar and wind generally reducing wholesale electricity prices. They argue that wind effects determine downspikes and even negative prices, whereas solar output balances the high demand during peak hours. Ketterer (2014) also studied the effect of wind in Germany looking at volatility dynamics as well as price levels. She showed that wind power reduces electricity price level but increases its volatility and, through rolling regressions over 3 years, found that the wind effect on mean prices was becoming less negative over time.

Finally, closer to the objectives of this paper, Serinaldi (2011) considered the short-term forecasting of Californian and Italian electricity price densities using the Johnson's S_U distribution, with time varying means and variances, but constant skewness and kurtosis, whilst Panagiotelis and Smith (2008) applied the skew-t distribution in a daily vector autoregressive formulation. Otherwise, rather more researchers have approached the distributional specification through interval forecasts using semiparametric quantile regressions (eg Jónsson et al., 2014; Bunn et al., 2016), also demonstrating the time varying effects of wind, solar and other exogenous variables on particular quantile estimates. The formulation we describe below is an extension of these themes, with a time-varying specification of all four-moments from a parametric density representation of prices, with exogenous drivers and autoregressive latent variable persistence, benchmarked against forecasts from quantile regression and other models, including the simpler specification of Serinaldi (2011).

3. Evolution and Fundamentals of German Prices

In a relatively short period of time, since the turn of this century, the German electricity market has been characterized by a series of radical structural changes: liberalisation, emission trading, a nuclear power phase-out and, most recently, the growth in renewable generation. One consequence of the rapid penetration of wind in particular has been the need to allow negative prices to emerge in the German and coupled spot markets⁷. Negative prices may occur when demand falls and/or wind production is high and they signal an urgent need for generators to reduce output or for consumers to increase demand. However, producers of inflexible plants may prefer to pay for continuing to produce, as this may cost less, or be more practical, than stopping and restarting their plants over a short period of time. Less extreme than negative prices, is negative skewness, and for similar reasons, this is expected to be induced by low demand and/or high wind. Solar could have a similar effect,

⁷Negative pricing has been introduced on the German/Austrian day-ahead market in 2008.

although being a midday producer, it does not generally coincide with low demand, and so its effect may be more manifest in reducing the otherwise positive skewness in those periods.

Germany is therefore an appropriate case study to develop and test a stochastic price formulation model in which there is explicit dependence of the shape of the densities on wind, solar and other short-term fundamental drivers. To this aim, we model the individual hourly electricity prices produced by the coupled German/Austrian day-ahead auction market, from January 2006 to December 2016. These 24 hour price vectors are recorded on a 7-day basis, thereby providing 24 hourly time series each containing 4018 observations⁸. In addition and on the same daily horizon, we have considered actual load, forecasted wind and solar PV generation, coal, gas, CO_2 prices all quoted or converted in €/MWh. This data set has been carefully compiled from different sources, namely the EPEXSpot⁹ (for hourly day-ahead prices), the four German TSOs websites (for actual and forecasted wind and solar generation aggregated on hourly level), ENTSO-e (for hourly actual loads), and Datastream¹⁰ (for coal ICE API2 CIF ARA, TTF natural gas, and CO_2 daily prices).

Figure 2 shows the evolution of average hourly curves for load, solar and wind production, computed yearly over 2006-2016 together with the ‘net load’ faced by conventional thermal generators after the feed-in of wind and solar production. The intra-daily load profiles show a slow decline in levels, whereas those for wind and solar show remarkable increases in output over this period. It should be mentioned that the full time series exhibit the usual annual time-varying patterns following calendar seasons, with higher solar PV generation during summer.

Descriptive statistics of price levels are reported in Table 2 in the supplementary Appendix 6.2 for all individual hours, showing the evolution of empirical sample moments across years.

⁸Regarding the clock changes in spring and autumn, we have excluded from the analysis hour 25, observable at the autumn clock change, and for the spring clock change when the price for hour 3 was missing, we averaged the prices of the previous and following days. Indeed, when the clock is advanced to summer time the data reporting system automatically deletes the values/slots for hour 3 (i.e. from 2.00 am to 3.00 am). Furthermore, in our hour-by-hour approach, the leap year is not awkward as it is simply an extra day influencing the weekly seasonality that we already included in our modelling.

⁹The day-ahead auction DE/AT (Phelix) hourly prices are the reference prices for delivery of electricity on the following day in 24 hour intervals on the German/Austrian TSO zones. The physical deliveries are made within any of the 4 German TSOs zones as well as in the Austrian Power grid. However, EEX is planning to split it into two different zones, and, given our focus on the German market, we limit our analysis on the German TSOs hence considering only actual and forecasted variables (as wind and solar PV) registered in this market by Tennet, 50Herz, Transnet and Amprion. Thus, load and renewable data for Austria are not included.

¹⁰We have also interpolated Datastream quotations over missing weekends and holidays. The tickers of used series are, respectively: LMCYSPT for the settlement prices of coal Intercontinental Exchange API2 cost, insurance and freight Amsterdam, Rotterdam and Antwerp converted in €/MWh using the USEURSP rates from US\$ to Euro (WMR&DS); TRNLTTD for the 1st Future Day settlement prices for the natural gas TTF NL quoted in €/MWh; and finally, EEXEUAS for the EEX-EU CO_2 Emissions E/EUA in €. The four German TSOs used to retrieve wind and solar data are: Tennet (www.tennettso.de), Amprion (www.amprion.net), Transnet BW (www.transnetbw.com) and 50Hertz (www.50hertz.com). Lagged actual load has been used as a proxy for the public unavailability of forecasted load, hence the load observed yesterday is considered the best forecast for today. Through the modelling, we have been attentive to the data that would be available to the market at the time when participants make their bids and offers to the day ahead auction.

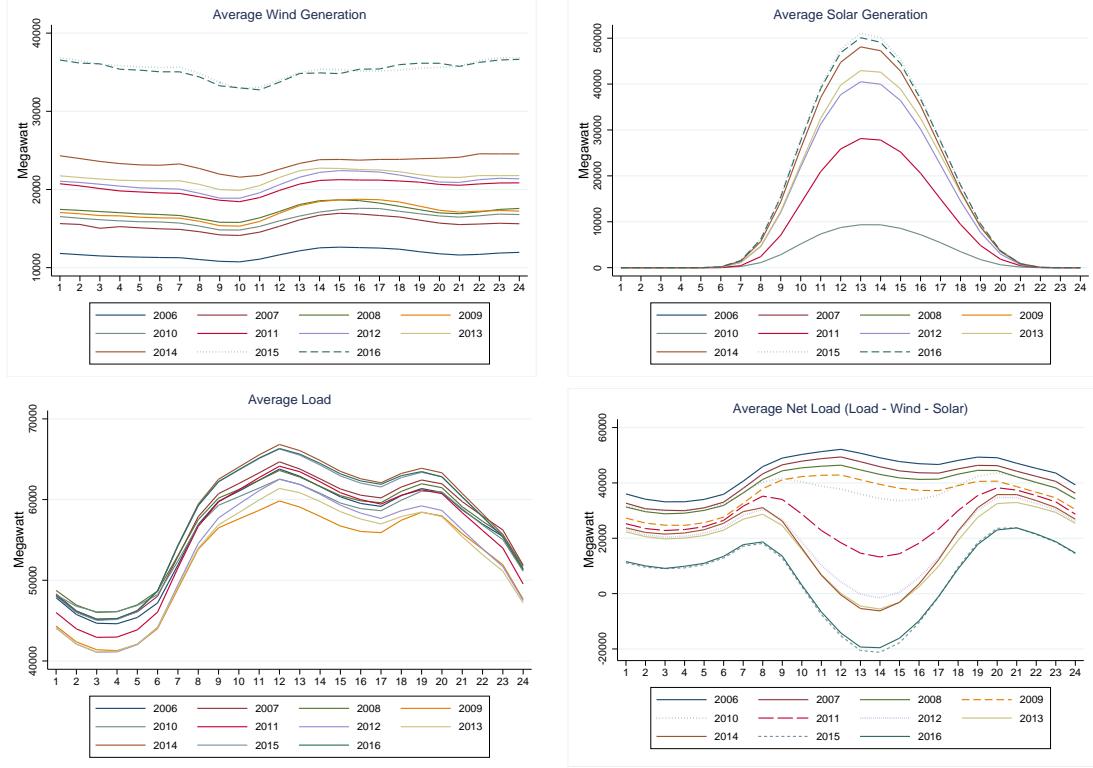


Figure 2: Average Intra-Daily Profiles for Load, Net Load, Wind and Solar Generation. Data sources: Tennet (www.tennettsso.de), Ampriion (www.amprion.net), Transnet BW (www.transnetbw.com) and 50Hertz (www.50hertz.com) for actual wind and solar generation; and ENTSO-E (www.entsoe.eu) for actual load.

Thus, it can be observed that negative skewness, on average, characterized prices densities at times of steep increase and decline in load, ie mid morning and late evening at hours 6, 7, 8, 23 and 24, at the beginning of our sample, but after 2010, negative skewness started to occur across the daytime hours as well, when solar generation created a new cycling requirement on the thermal generators. These observations motivated our modelling choice of selecting non-censored distributions with both positive and negative mean and skewness. Thus, to fit the deseasonalised densities of hourly prices, adjusted for holidays, we have considered 3 different classes of distributions with these features and also the capabilities to work within the multivariate formulations we require for estimation in the next section. The first one is a class of 4-parameter distributions, the *Johnson's SU* (in its original parametrization as in Johnson, 1949 and in its alternative parametrization as in Johnson, 1954; see respectively *JSUo* and *JSU*), the *sinh-arcsinh* (as in Jones and Pewsey, 2009; see SHASHo and SHASHo2), the *skew exponential power* (as in Fernandez et al., 1995; see SEP1 and SEP2), the *skew-t* (as in Azzalini, 1986, in Azzalini and Capitanio, 2003 and in Jones and Faddy, 2003; respectively ST1, ST2 and ST5). The second one is a 3-parameter family represented by the *skew-normal* distributions, specifically the skew normal ‘type 1’ which is a special case of the skew exponential power with $\tau = 2$ (SN1). Thirdly, we selected as a baseline, the 2-parameter *normal* distribution (NO) as this is often used for simplicity in operational models (see previous section).

To identify the best fitting density functions, yearly values of three measures of the

goodness-of-fit (the Kolmogorov-Smirnov *KS*, the Cramer-von Mises *CVM*, and the Anderson-Darling *AD*) are reported in Tables 3-10 in Appendix 6.3. According to these measures of fit and accounting for the absolute maximum, the squared and the weighted squared differences (to give more attention on the tails), as well as the AIC criterion to discourage over-parameterisation, we observed the general superiority of the skew student-t distributions (specifically ST1 and ST2). Also recorded in these tables are details of the computational estimation times (for a PC with Intel Core Duo i7-3520M CPU 2.9 GHz and 8GB RAM) which reveal some of the computational difficulties, particularly with the two JSUs. It is also interesting to observe that SN1 and NO emerged to perform well during midday hours which is when, as we observed earlier, the historic positive skewness had been tempered to become more symmetrical through the impact of solar. When we further repeated the analysis on just the deseasonalized prices over the shorter sample 2010-2016 which we use in the multivariate modelling in the next section (shorter because forecast solar data was only available from 2010), we observed that ST2 and JSU were the two best fitting distributions (the former on hours 1-8 & 24, whereas the latter over the hours 9-23; see Table 11 in Appendix 6.3).

On balance, however, we considered the Skewed-t to be most appropriate on the basis of its general fitting and analytical properties. We note that the skewed-t had previously been used for hourly Australian prices in Panagiotelis and Smith (2008) whilst Serinaldi (2011) used the Johnson's S_U distribution for Californian and Italian electricity price densities. The flexibility of the skew-t distribution to the range of shapes is shown in Figures 3-5, where two forms of the *skew-t* distribution were compared with the JSU, SN1 and NO, for the same motivating sample of hours and years that we displayed in Figure 1. Regarding the skew-t variants, comparing their performances, we decided to focus upon the second skew-t, in which the pdf of the skew-t type 2 distribution, denoted $ST2(\mu, \sigma, \nu, \tau)$, is defined by

$$f_Y(y|\mu, \sigma, \nu, \tau) = \frac{2}{\sigma} f_{Z_1}(z) F_{Z_2}(\omega) \quad (1)$$

for $-\infty < y < +\infty$, where $-\infty < \mu < +\infty$, $\sigma > 0$, $-\infty < \nu < +\infty$ and $\tau > 0$, and where $z = (y - \mu)/\sigma$, $\omega = \nu \lambda^{1/2} z$, $\lambda = (\tau + 1)/(\tau + z^2)$ and f_{Z_1} is the pdf of $Z_1 \sim TF(0, 1, \tau)$ (a t-distribution with $\tau > 0$ degrees of freedom treated as continuous parameter) and F_{Z_2} is the cdf of $Z_2 \sim TF(0, 1, \tau + 1)$. The mean and the variance of Y are given by $E(Y) = \mu + \sigma E(Z)$ and $Var(Y) = \sigma^2 Var(Z)$, where $E(Z) = \nu \tau^{1/2} \Gamma(\frac{\tau-1}{2}) / [\pi^{1/2} (1 + \nu^2)^{1/2} \Gamma(\frac{\tau}{2})]$ for $\tau > 1$ and $E(Z^2) = \tau / (\tau - 2)$ for $\tau > 2$. This distribution is the univariate case of the multivariate skew-t distribution introduced by Azzalini and Capitanio (2003).

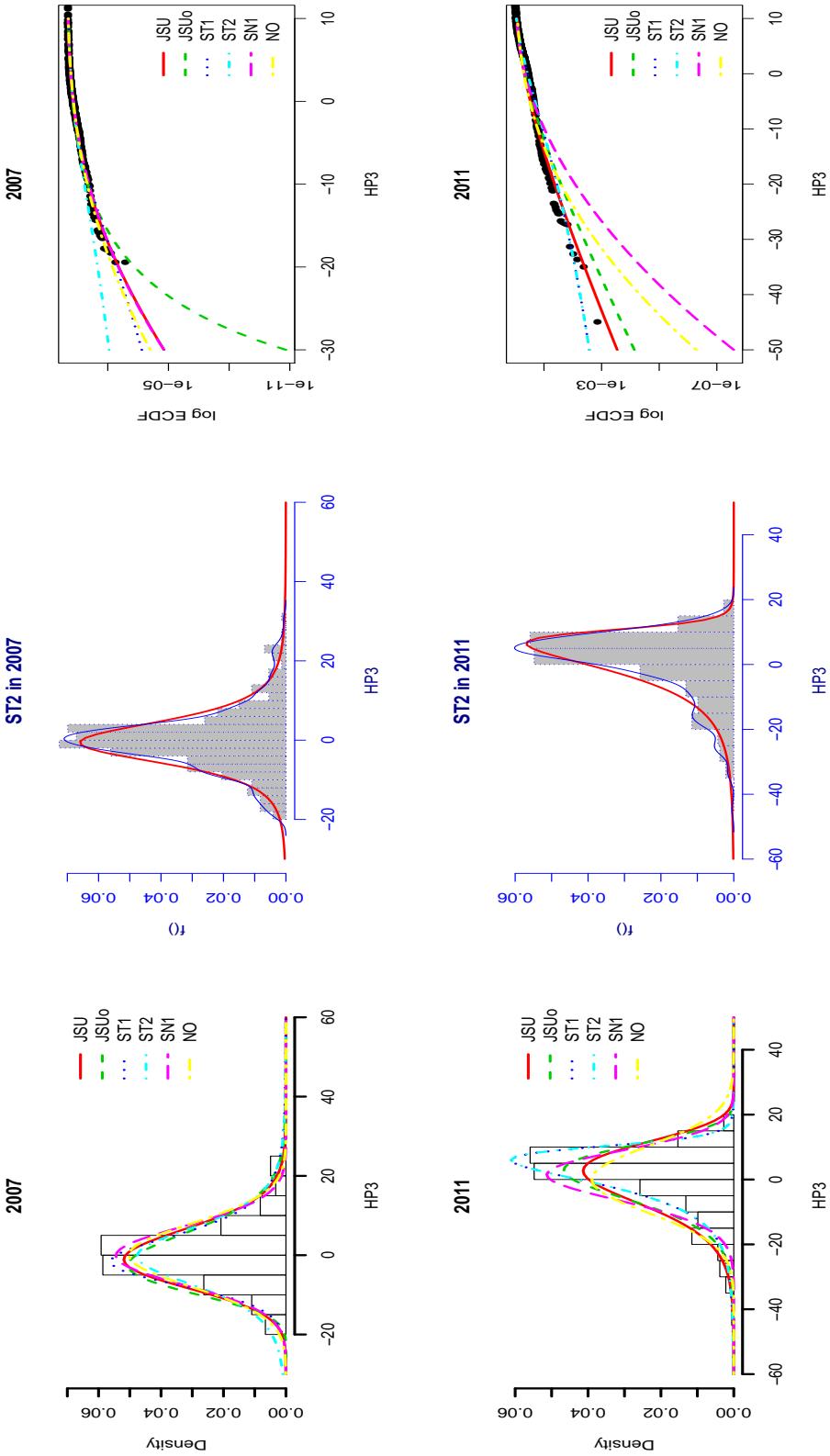


Figure 3: Comparisons of Density Fits (on the first column), the skewed-t fit vs a kernel density (in the middle) and the ECDF with Density in logarithmic scale for hour 3 in 2007 and 2011 (in rows). Descriptive statistics of deseasonalized prices: $\mu = 0.000$, $\sigma = 7.939$, $\nu = 0.546$ and $\tau = 4.487$ in 2007; and $\mu = 0.000$, $\sigma = 10.234$, $\nu = -1.320$ and $\tau = 4.610$ in 2011.

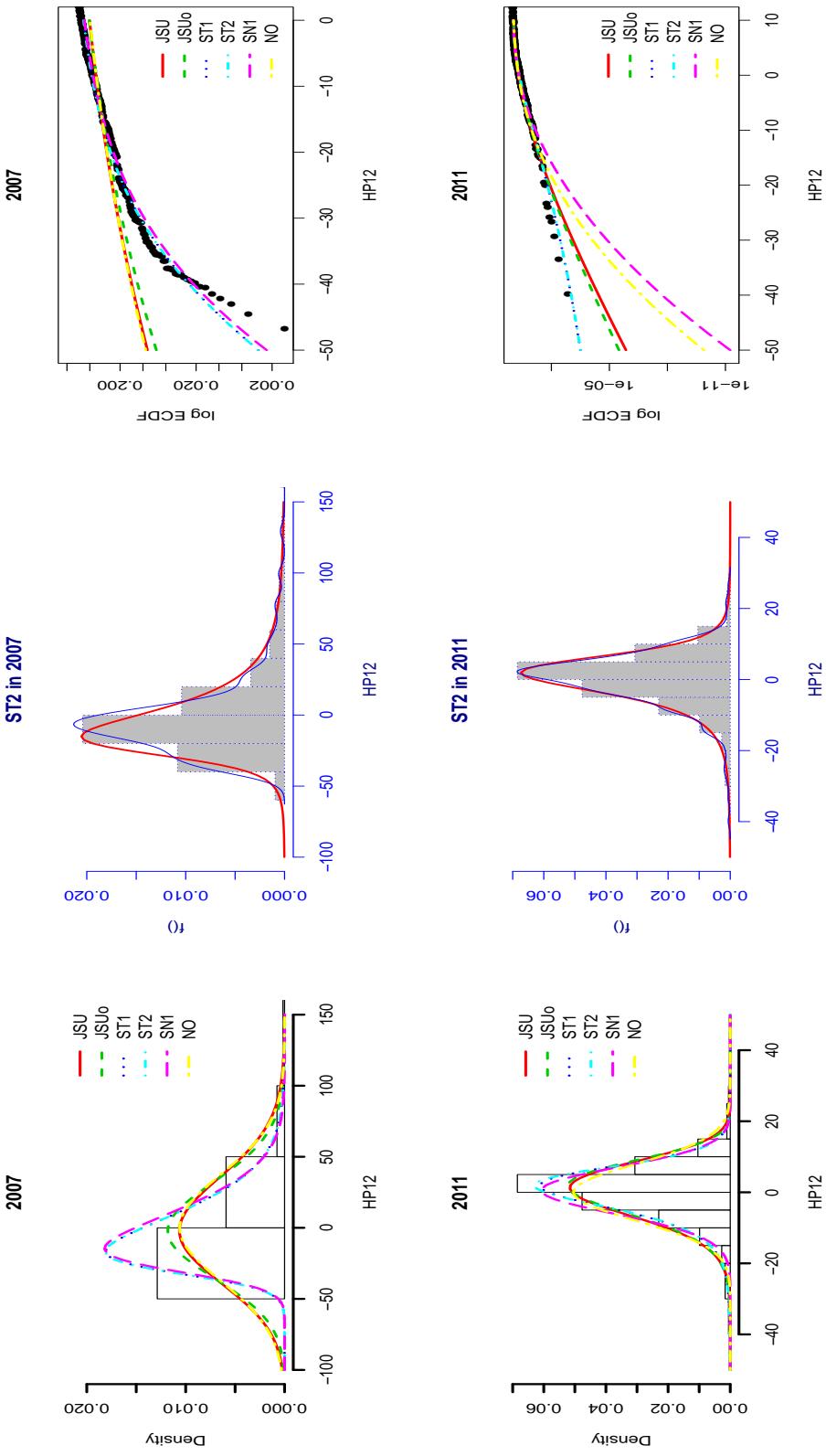


Figure 4: Comparisons of Density Fits (on the first column), the skewed-t fit vs a kernel density (in the middle) and the ECDF with Density in logarithmic scale for hour 12 in 2007 and 2011 (in rows). Descriptive statistics of deseasonalized prices: $\mu = 0.000$, $\sigma = 37.618$, $\nu = 3.642$ and $\tau_7 = 23.575$ in 2007; and $\mu = 0.000$, $\sigma = 7.908$, $\nu = -0.948$ and $\tau = 6.362$ in 2011.

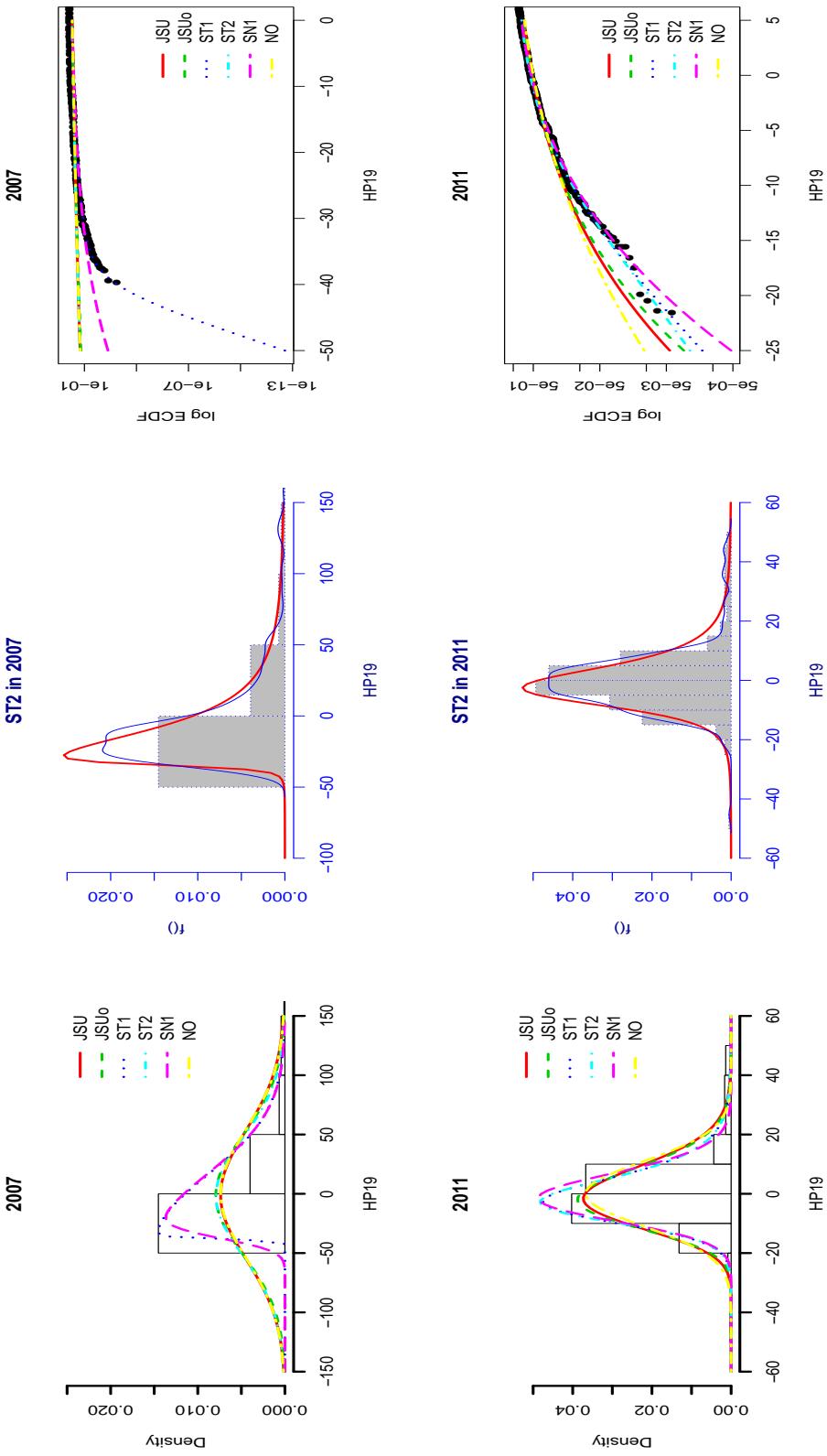


Figure 5: Comparisons of Density Fits (on the first column), the skewed-t fit vs a kernel density (in the middle) and the ECDF with Density in logarithmic scale for hour 19 in 2007 and 2011 (in rows). Descriptive statistics of deseasonalized prices: $\mu = 0.000$, $\sigma = 54.069$, $\nu = 6.172$ and $\tau = 60.714$ in 2007; and $\mu = 0.000$, $\sigma = 10.870$, $\nu = 1.357$ and $\tau = 7.921$ in 2011.

4. Linear Multifactor Dynamic Estimation of the Density Moments

With the attractive flexibility of the four parameter skew-t densities for fitting the various hourly prices, the consequent research question is whether the specification of the first four moments of the skew-t can be well represented in terms of the dynamics of fundamental short term drivers of price formation. To this end, we considered whether these exogenous variables could be formulated and estimated within a generalized model for the first four moments (representing the level μ , the volatility σ , skewness, ν , and kurtosis, τ), resulting from an extension of the Generalized Linear Models in Nelder and Wedderburn (1972); Generalized Additive Models in Hastie and Tibshirani (1986) and Hastie and Tibshirani (1990); and the Generalized Additive Models for Location, Scale and Shape, `gamlss`, as in Rigby and Stasinopoulos (2005) and Stasinopoulos and Rigby (2007). In so doing, we seek to substantially extend the scope of applications undertaken with similar methodology by Serinaldi (2011), Matsumoto et al. (2012) and Scandroglio et al. (2013). Within this framework we formulate the hourly electricity price as a response variable whose distribution function varies according to multiple exogenous factors. From the previous section we choose to represent the response variable (the deseasonalised hourly electricity price) as a the skew-t density with the mean, μ , standard deviation, σ , skewness, ν , and kurtosis, τ modelled as multifactor linear functions as follows.

Let Y be the response variable, it is assumed that independent observations y_i for $i = 1, \dots, n$ have distribution function $F_Y(y_i; \theta^i)$

with $\theta^i = (\theta_1^i, \dots, \theta_p^i)$ a vector of p distribution parameters accounting for position, scale and shape. Generally, p is less than or equal to four, since the 4-parameter families provide enough flexibility.

Given an n length vector of the response variable $\mathbf{y}^T = (y_1, \dots, y_n)$, let $g_k(\cdot)$ for $k = 1, \dots, p$ be monotonic link functions relating the distribution parameters to explanatory variables and random stochastic effects to account for extra not explained variability through an additive model given by

$$g_k(\theta_k) = \eta_k = \mathbf{X}_k \beta_k + \sum_{j=1}^{J_k} \mathbf{Z}_{jk} \gamma_{jk} \quad (2)$$

where θ_k and η_k are vectors of length n , e.g. $\theta_k^T = \{\theta_k^1, \dots, \theta_k^n\}$, $\beta_k^T = \{\beta_{1k}, \dots, \beta_{J_k k}\}$ is a parameter vector of length J_k , \mathbf{X}_k is a known design matrix of order $n \times J_k$, \mathbf{Z}_{jk} is a fixed known $n \times q_{jk}$ design matrix and γ_{jk} is a q_{jk} -dimensional random variable.

The linear predictors η_k , for $k = 1, \dots, p$ comprise a parametric component $\mathbf{X}_k \beta_k$ and additive components $\mathbf{Z}_{jk} \gamma_{jk}$; the first term represents a linear function of explanatory variables and the second one represents random effects. For sake of parsimony, we used only linear components in link functions. We assumed identity link functions, $g_1(\cdot)$ and $g_3(\cdot)$, for the expected hourly prices and their skewness; whereas logarithmic link functions were used for $g_2(\cdot)$ and $g_4(\cdot)$ to ensure positivity for standard deviation and kurtosis, obtained by taking the exponential of the filtered log series.

Based upon the conventional considerations of market fundamentals and with regard to the information that would generally be available to market participants by the time they make offers and bids into the day ahead auction, we have expressed the expected hourly

price in reduced form as a function of its value observed on the day before (y_{t-1}), as well as on electricity load observed on the day before ($load_{t-1}$, in thousands of MW), forecasts for wind and solar PV generation ($fwind_t$, and $fsolar_t$, also in thousands of MW) available to the market prior to the auction, lagged prices of coal, gas and CO_2 allowances (respectively $coal_{t-1}$, gas_{t-1} , co_{2t-1}), together with calendar holidays and weekends in a dummy variable, hol_t , which takes value 1 for weekends and German holidays¹¹.

Formally, in its basic formulation, this dynamic multi-factor skew t “MFST” model has $g_1(\theta_1) = \eta_1 = E(y_t) = \mu_t$, with a time-varying latent mean

$$\mu_t = \alpha_1 + \gamma_1 y_{t-1} + \beta_{11} hol_t + \beta_{12} load_{t-1} + \beta_{13} fwind_t + \beta_{14} fsolar_t \quad (3)$$

$$+ \beta_{15} coal_{t-1} + \beta_{16} gas_{t-1} + \beta_{17} co_{2t-1}. \quad (4)$$

We assumed that the dispersion, estimated dynamically as a time varying latent volatility (standard deviation) state variable, is a function of fundamental drivers through $g_2(\theta_2) = \eta_2 = \log(\sqrt{Var(y_t)}) = \log(\sigma_t)$, with

$$\log(\sigma_t) = \alpha_2 + \beta_{21} hol_t + \beta_{22} load_{t-1} + \beta_{23} fwind_t + \beta_{24} fsolar_t + \beta_{25} coal_{t-1} + \beta_{26} gas_{t-1} + \beta_{27} co_{2t-1}. \quad (5)$$

Distinct from previous formulations (eg Serinaldi, 2011), we extend the scope of dynamic multifactors to the third and fourth moments through the time varying latent parameters, $g_3(\theta_3) = \eta_3 = \nu_t$ and $g_4(\theta_4) = \eta_4 = \log(\tau_t)$ as follows

$$\nu_t = \alpha_3 + \gamma_3 hol_t + \beta_{31} load_{t-1} + \beta_{32} fwind_t + \beta_{33} fsolar_t + \beta_{34} coal_{t-1} + \beta_{35} co_{2t-1} + \beta_{36} gas_{t-1} \quad (6)$$

$$\log(\tau_t) = \alpha_4 + \gamma_4 hol_t + \beta_{41} load_{t-1} + \beta_{42} fwind_t + \beta_{43} fsolar_t + \beta_{44} coal_{t-1} + \beta_{45} co_{2t-1} + \beta_{46} gas_{t-1}. \quad (7)$$

In order to investigate possible persistence in the latent moments, we also considered a new expanded formulation to include the autoregressive dynamics of the latent moments into the above mean, variance, skewness and kurtosis equations, thus defining an autoregressive multifactor skew-t model, “AR-MFST”. In the “AR-MFST” models, the 1-lag autoregressive variables μ_{t-1} , σ_{t-1} , ν_{t-1} and τ_{t-1} are included in each of the corresponding equations 4-7 above. Furthermore, extending this concept to include possible implicit effects between these latent moments, we also considered a vector autoregression version in which the first lags of all four of the latent moments are included in each latent moment equation, 4-7, to give the “VAR-MFST” formulation¹².

¹¹To avoid over-parametrization, we have included weekly seasonality together with holidays in one single dummy variable, equal to one over weekends and holidays such as: New Year, Good Friday, Easter Monday, Labour Day, Ascension Day, Whit Monday, German unit day, Christmas Eve, Christmas Day, Boxing day and New Years Eve. Given our focus on the German TSOs market, we have not considered Austrian public holidays nor regional holidays, such as Pentecost, Corpus Christi Assumption of Mary, Reformation day, All Saints day and Repentance Day.

¹²In the former case, μ_{t-1} is replacing lagged prices. And autoregressive terms have been extracted after the estimation of the basic MFST model presented in eqs. 4-7.

All parameters are estimated by maximizing the likelihood function¹³, through an adaptation of Cole and Green (1992) algorithm, which uses the first, second and cross derivatives of the likelihood function with respect to the distribution parameters, $\theta(\mu, \sigma, \nu, \tau)$. As the computation of cross derivatives proved to be intractable, we adopted a generalization of the algorithm developed in Rigby and Stasinopoulos (1996a), and Rigby and Stasinopoulos (1996b) for fitting the mean and dispersion additive models which does not require the cross derivatives. Initial values for autoregressive sample parameters have been obtained by fitting the model to the entire sample, then lagged series have been included (and later in the forecasting performance, updated through the rolling iterations).

Optimisation in ML estimations such as this are often sensitive to starting values, leading to local optima, or even singularities with small samples, but with the length of time series in our study, neither of these were problems¹⁴. Overfitting, on the other hand, was a serious consideration and for that reason we restricted our factor specifications to plausible drivers on the basis of what is known about electricity price formation and we undertook extensive out-of-sample testing in a forecasting context with some challenging benchmark testing.

Whilst it should be recalled that the main objective of this modelling is to elucidate a general-purpose price formation model, flexible enough to accommodate the range of shapes that hourly power prices may take, in a way that relates these shape changes parametrically to fundamental exogenous drivers, the estimation process also derives the time-varying latent estimates of the moments. For several financial engineering applications in particular, such latent estimates of the instantaneous volatility and skewness states are attractive alternatives to historic estimates based upon lagged finite rolling windows. For example, Figure 6 shows the time series of the latent moments for hour 12, whereas similar series are shown for hour 19 in Figure 10 in Appendix 6.1.

¹³The estimation has been carried out using the statistical software R and some of the "gamlss" library models (<http://www.gamlss.org/>).

¹⁴There are two algorithms to maximize the likelihood function: the first is the CG algorithm, a generalization of the Cole and Green (2002) algorithm which uses first, (expected or approximated) second and cross derivatives of the likelihood function with respect to the distribution parameters; the second one is the RS algorithm, a generalization of the algorithm used by Rigby and Stasinopoulos (1996a) and Rigby and Stasinopoulos (1996b) for fitting the mean and the dispersion of additive models and it does not use cross derivatives. Singularities in the likelihood function similar can potentially occur, especially when the sample size is small as in our yearly analysis. The RS algorithm has an outer cycle which maximizes the penalized likelihood with respect to β_k and γ_{jk} for $j = 1, \dots, J_k$ in the model successively for each θ_k in turn, for $k = 1, \dots, p$; and at each calculation in the algorithm, the current updated values of all quantities are used. Roughly, the RS algorithm starts initializing the fitted values θ_k and random effects γ_{jk} in the first outer and inner cycle iterations by regressing partial residuals against a ‘weighted’ design matrix to obtain updated parameter estimates. Then, the linear predictors η_k are evaluated and updated, and the cycle is repeated until the change in the penalized likelihood is sufficiently small. At this point, convergence is obtained and results produced. On the contrary, in the CG algorithm all weight matrices are evaluated and updated after fitting of all θ_k . Further details on the functioning of algorithms and the maximization of the likelihood can be found in Rigby and Stasinopoulos (2005) and in Stasinopoulos et al. (2008). In all our estimation, we used the RS algorithm and control parameters set to: 0.5 as for the the convergence criterion for the algorithm (*c.crit* and *cc*), and the tolerance level for the backfitting algorithm (*bf.tol*); 1000 for *n.cyc*, for *cyc*, as the number of cycles of the algorithm, and for the number of cycles of the backfitting algorithm (*bf.cyc*). Finally, *Inf* was used for the global deviance tolerance level (*gd.tol*), to allow the algorithm to converge even if the global deviance changes dramatically during the iterations.

Residual analysis for enlarged models is presented in Appendix B.5, Figures 11 and 12, which contain the ACF and PACF of residuals from the main models under consideration; in addition, ACF and PACF for level, squared and cubic residuals are shown in Figures 13 and 14. The PACFs indicated some residual correlations which was expected. We did not include any nonlinear specification in the models, particular with respect to load. Supply functions are known to be nonlinear at low and high levels and whilst we chose a simple linear response as a robust assumption, it will presumably be underspecified. There are also interaction effects. In some of the subsequent modelling we have therefore included AR(7) to capture serial effects of up to a week. For additional insight, Table 17 in Appendix 6.4.3 presents the descriptive statistics of normalized (quantile) residuals for a sample of hours. The *normalized quantile* residuals are given by $\hat{r}_i = \Phi^{-1}(u_i)$ where Φ^{-1} is the inverse cumulative distribution function of a standard normal variate, $u_i = F(y_i|\hat{\theta}^i)$ and $F(y|\theta)$ is the cumulative distribution function with $\theta = (\mu, \sigma, \nu, \tau)$. If the “MFST” models are correctly specified, then the normalized quantile residuals should behave as standard normal ones, see Dunn and Smyth (1996). In our results, these residuals generally exhibit zero mean, unit variance, zero asymmetry and kurtosis equal to three (except for the simpler SN1 and NO distributions) thus indicating model adequacy (the results for other hours are similar and available on request).

4.1. Dependence of Higher Moments on Exogenous Fundamental Drivers

In order to avoid over-elaboration and to justify the multifactor, time-varying representation of the moments, we developed a series of progressively more complex specifications. According to the multifactor formulations of eqs. 4-7, four models were successively estimated:

- M1: time-varying mean, but all other moments constant;
- M2: time-varying mean and standard deviation, with constant skewness and kurtosis;
- M3: time-varying mean, variance and skewness, constant kurtosis;
- M4: all four moments time varying.

In all cases, we also specified the models with and without each of the renewable energy drivers (wind and solar) and the results confirmed their incremental impacts (these have not been reported for lack of space, but they are available on request).

Expectations for these factor effects follow from previous work, but as such only inform price levels and variance (eg from Karakatsani and Bunn (2010), Ketterer (2014), Paraschiv et al. (2014), Cló et al. (2015) and Bunn et al, 2016, among others) as this research is the first to consider the higher skewness and kurtosis drivers. Thus, previous research would imply:

1. positive autoregression at lag one in price levels and the higher moments, reflecting some *adaptive behavior and persistence*;
2. positive effect of load (i.e. demand) on the mean power price because of the increasing fundamental marginal cost supply function, and on volatility, reflecting the conventional “inverse leverage” observation that at times of high demand and prices, volatility also increases (in contrast to the usual leverage in equity markets with higher volatility at lower prices);

	Hour 3				Hour 12				Hour 19			
	μ	$\log(\sigma)$	ν	$\log(\tau)$	μ	$\log(\sigma)$	ν	$\log(\tau)$	μ	$\log(\sigma)$	ν	$\log(\tau)$
hol_t	-	+	-	+	-	+	-	-	-	\pm	-	+
y_{t-1}	+				+				+			
μ_{t-1}	+	+	-	+	-	-		-	\pm	-	-	-
$\log(\sigma_{t-1})$	+	+			-	+		-	+	+	-	-
ν_{t-1}	-	+			+	+	-		-	-	-	+
$\log(\tau_{t-1})$				-		+			+	+		+
$load_{t-1}$	+	-	+	-	+	-	-	+	+	\pm	\pm	-
$fwind_t$	-	+	-	+	-	+	-	-	-	-	-	+
$fsolar_t$					-	-	-		-	-	-	+
gas_t_{-1}	+	+	-	+	+	+	+	-	+	+	+	-
$coal_t_{-1}$	+	\pm	-	-	+	\pm	-	\pm	+	-	-	
$co2t_{-1}$	+	+	-	-	+	+			+	+	+	-

Table 1: High level summary of significant signs for the full model M4 and its AR and VAR variations.

3. fuel prices and CO_2 generally increasing the mean power price because of their input cost;
4. wind (and solar) reducing electricity prices (especially for peak hours as consequence of balancing excess demand), but increasing the volatility.

A high level summary of the significant signs of the full model specifications for the MFST, AR-MFST and VARM-MFST models is shown in Table 1, whereas the full results for estimated coefficients and t-values are in Tables 12-14 in Appendix 6.4.1 for a sample of hours. The entries in Table 1 are for the significant (5%) coefficients and their signs. Where plus and minus signs are overlayed, the indications from all three MFST variations differ. The R-squares (the generalised R-squared in Nagelkerke (1991)), the Global Deviance (defined as a function of the log-likelihood, formally $GD = -2\log\mathcal{L}$), and information criteria about the progressive modelling are reported in Tables 15 and 16 in Appendix 6.4.2. These results present a generally coherent interpretation, mostly consistent with expectations but with some new insights. The most important observations are the shape-shifts induced by wind and solar. Both wind and solar production do indeed reduce the skewness of hourly electricity prices and this is more evident at hour 12 when solar is at its maximum level. In addition, both increase the kurtosis of these prices at peak hour 19. The consistency of the results across all of the modelling progressions, with the various inclusions of wind and solar, adds considerably to the robustness of the fundamental specification. The key factor influences are:

- 1' Lagged price is positive for all four MFST models on price level, indicating adaptive behaviour consistent with expectations. However, for the latent moment autoregressions the results are mixed. Holidays, as expected, reduce business activities and therefore price levels. Latent volatility was positively persistent on itself, as expected and had a positive effect on price levels in periods 3 and 19, but negative in period 12. For the higher moments, some mixed results might suggest over-fitting with the AR MFST and particularly VAR MFST.
- 2' Load (i.e. demand) has a positive coefficient on the price levels, consistent with conventional expectations and previous research, but negative on price variance for hours 3 and 12. Hour 3 is when more negative price spikes are observed compared to hours 12 and 19 and low load levels over night will therefore manifest higher volatility.

- 3'** Among fuel prices, we find that *natural gas* increases mean and volatility. During the day it increases skewness, whereas it reduces the kurtosis. At night it reduces skewness and increases kurtosis. These effects seem plausible as peaking generation during the day may be associated with positive spikes, whereas its use at night might be counteracting negative price tendencies. *coal* increases mean but it reduces skewness. We also confirm that CO_2 affects price levels and volatility.
- 4'** Wind lowers prices and lowers skewness. Hence the appearance of negative skewness, as noted earlier, can be confirmed as a wind-induced factor. Solar has effects during the day, with negative effects on price levels, volatility and skewness. The wind and solar effects are consistent across all modelling specifications.

It is questionable whether the gains of expanding the dynamic specifications in the model from 3 to 4 parameters, and in extending the specifications to AR and VAR latent moments is worthwhile. In one respect, whilst the interpretations of the kurtosis factor coefficients are generally significant, intuitions regarding their sign are equivocal, and the gains in fit from $M3$ to $M4$ are small. The dynamic three parameter improvement over a two parameter model is substantial, and the skewness story is persuasive; but less so for the kurtosis in going to four parameters. Likewise, the AR and VAR inclusions of the latent estimates produced equivocal estimations and this suggests specification or overfitting problems, despite their conceptual appeal. Thus, in the next section we report results on out-of-sample forecast testing performed on rolling windows of 365 days¹⁵. The motivation is not primarily to suggest a forecasting method, but to test the robustness of the specifications.

4.2. Forecasting Performance

For robustness and as a check against over-fitting, we test the performance of the MFST, AR-MFST and VAR-MFST models through one period ahead, out-of-sample forecasting, using rolling window estimations and also considering all one-, two-, three- and four-moment specifications. We compare against two benchmarking techniques, one being the well-established semi parametric quantile regression method to derive empirical interval limits and the other being a conventional approach for estimating the conditional means and variances of electricity prices using ARMA-GARCH type models.

Benchmarking against quantile regressions is a challenging test insofar as these estimates of particular points on the density function (eg 5%, 95% intervals) make no distributional assumptions and are purely empirical. The MFST approach will only be as accurate as quantile regression at specific quantiles if its parametric specification is appropriate. More precisely, we compare M1- AR1- and VARM1-MFST versus Quantile Regressions, including an AR(7) structure together with drivers, holidays and seasonality. When the first two moments are considered, an ARX-EGARCHX(1,1), specified with a student-t and a skewed-student-t for the innovations, will estimate volatility persistence and its formulation clearly recalls the $\log(\sigma_t)$ in the M2-MFST versions, which will only capture this to the extent that the drivers of heteroscedasticity are in the exogenous multifactor dynamics. But, in the

¹⁵Not reported here were results based on rolling windows of 730 days and an expanding window of 365 days. These confirmed those found in the rolling approach presented here.

ARMA-GARCH framework, the innovations cannot change density shape as effectively as in the MFST, and then the AR- and VAR-MFST should capture some of the persistence effect through the inclusion of lagged latent moments. In addition, we compare the (AR)M2-MFST (that is specified with two moment equations) with the model formulated in Serinaldi (2011) based on the Johnsons' S_U with daily and hourly dummies, quadratic and cubic loads, and price functionals.

Quantile methods, following Koenker and Bassett (1978), are extensively applied as regression models for expressing specific percentiles of a response variable as a function of exogenous factors. Their main attractive features are firstly their semiparametric formulation of interval estimates of the predictive distribution; and secondly, the fact that they distinguish the impact of explanatory variables to different intervals of the distribution. Thus, in the conventional way, we let $q \in [0, 1]$ be a quantile of interest, Y_t be the dependent variable (that is the hourly electricity price levels) and X_t a d -dimensional vector of explanatory variables (e.g. load, forecasted wind and solar generation; gas, coal, and CO_2) with a constant included. The linear conditional quantile function is given by $Q_q(Y_t|X_t) = X_t\beta_q$ and we have used the fundamental drivers as in eq.(4) to forecast and compare selected quantiles:

$$Q_q(y_t) = \alpha^q + \sum_{i=1}^7 \gamma_i^q y_{t-i} + \beta_1^q hol_t + \beta_2^q load_{t-1} + \beta_3^q fwind_t + \beta_4^q fsolar_t + \beta_5^q coal_{t-1} + \beta_6^q gas_{t-1} + \beta_7^q co_{2t-1} \quad (8)$$

As observed by Chernozhukov and Umantsev (2001), data scarcity may be a problem in estimating the extreme tails of the distribution.

Benchmarking with the ARX-EGARCHX(1,1) model we test both student-t and skew student-t error distributions. Again, we have specified an AR(7) structure with the same factors as eq.8, and in this case with the conditional variance following an EGARCH(1,1) process also driven by the load, wind, solar and lagged fuel drivers. Formally, the conditional mean being as follows

$$y_t = c + \sum_{i=1}^7 \phi_i y_{t-i} + \lambda_1 hol_t + \lambda_2 load_{t-1} + \lambda_3 fwind_t + \lambda_4 fsolar_t + \lambda_5 coal_{t-1} + \lambda_6 gas_{t-1} + \lambda_7 co_{2t-1} + \theta \varepsilon_t \quad (9)$$

and the conditional variance as follows

$$\begin{aligned} \log(\sigma_t^2) &= \omega + \alpha \log(\sigma_{t-1}^2) + \beta g(\varepsilon_{t-1}) + \varphi_1 hol_t + \varphi_2 load_{t-1} + \varphi_3 fwind_{t-1} + \varphi_4 fsolar_{t-1} \\ &\quad + \varphi_5 coal_{t-1} + \varphi_6 gas_{t-1} + \varphi_7 co_{2t-1} \end{aligned} \quad (10)$$

with $g(\varepsilon_t) = \theta \varepsilon_t + \varrho[|\varepsilon_t| - E(|\varepsilon_t|)]$ with the conditional density set to be a Student-t, $\varepsilon_t | I_{t-1} \sim t(0, \sigma_t^2)$, or a skew Student-t, $\varepsilon_t | I_{t-1} \sim st(0, \sigma_t^2)$.

We expect that the four parameter, dynamic, multifactor skew-t models ("MFST" and "AR-MFST") to perform better than the ARMAX-GARCHX method, to the extent that the latter is only able to capture mean and variance dynamics. Even if the GARCH element would add the extra responses to conditional volatility and the persistence of shocks, the AR-MFST model is expected to exhibit competitive performance due to the extra flexibility in the density shape and the lagged latent moment persistence.

Having a full dataset of 2557 observations from the beginning of 2010 (because of the solar data not being available before), we perform an in-sample estimation using a rolling

window of 365 days. We then forecast the next observation as an out-of-sample test, and recursively advance the process through the time-series¹⁶.

We assess the calibration of the forecast densities by simply using conventional tests, such as Kupiec (1995) unconditional (UC), Christoffersen (1998) conditional coverage (CC) and the Engle and Manganelli (2004) dynamic quantile (DC) tests on yearly basis. In the coverage tests, the null hypothesis is that the calibration frequencies at each quantile are correct. Thus, in the results, a p-value greater than 0.01 indicates that we cannot reject the null hypothesis under a significance test at 1% (hence we are looking for high p values for well calibrated coverage). Tables 18-20, 21-23 and 24-26 in Appendix 6.5 summarise the results for hour 3, 12 and 19, for lower, middle and upper quantiles respectively. Results for a sample of other hours are available in Tables 27, 28, and 29. On this basis, we observe that this forecasting process on a rolling window of 365 observations provides poor results for *quantile regression*, most likely for the sample size reasons indicated in Chernozhukov and Umantsev (2001).

For hour 3, most practical interest would be in the lower quantiles, for the occurrence of downspikes, and here the MFST and the AR-MFST outperform the quantile regression and the GARCH methodology. For hour 12, practical interest would be in both the high and low quantiles, and here again the MFST and AR-MSFT performances in terms of coverage are better than the two benchmarks especially at lower quantiles (even with the shortest rolling window). For hour 19, most practical interest would be in the high quantiles for the risk of high prices, and we see again that the MFST models outperform the benchmarks. In addition, with the lower quantiles, the flexibility and forecasting ability of MFST models continues to be beneficial with the benchmarks being outperformed. Furthermore, we observe that the ARM3-MFST outperforms the ARM4-MFST in the out-of-sample forecasting testing, and this provides empirical evidence, as suspected, that there is little to be gained through modelling with the fourth moment being dynamic, and potential overfitting can be avoided. Likewise, the VAR representation proved to be less robust than the AR for out-of-sample, and together with the mixed interpretation of the coefficients in Table 1, appears to reflect overspecification.

Finally, to summarise the main comparisons, we show in Figure 7 the theoretical number of hits (exceedances at particular quantiles) versus the empirical frequencies across the quantiles, for three key models. For illustration we plot the graphs for hour 12 in 2011 and 2015 (similar results for other hours are available on request, showing the same indications). This clearly displays the superiority of the proposed ARM3-MFST, being closest to the theoretical curve, compared to the JSU density model used by Serinaldi (2011) and the EGARCH with skew student t which would perhaps be the most appropriate of the conditional volatility models typically used for Value-at-Risk analysis.

¹⁶Thus, we start by estimating the models using the first 365 observations, and from this forecast the 1%, 2%, 5%, 25%, 50%, 75%, 95%, 98% and 99% quantiles of the 366th observation. Thereafter, we estimate the models using observation 2 to 366 to forecast quantiles of observation 367, and so on. The recursive (V)AR-MFST models have been initialized by using the lagged filtered moments obtained from the MSFT estimation on the whole sample; then, the autoregressive terms have been recursively updated by rolling estimations and used into the forecasting procedures.

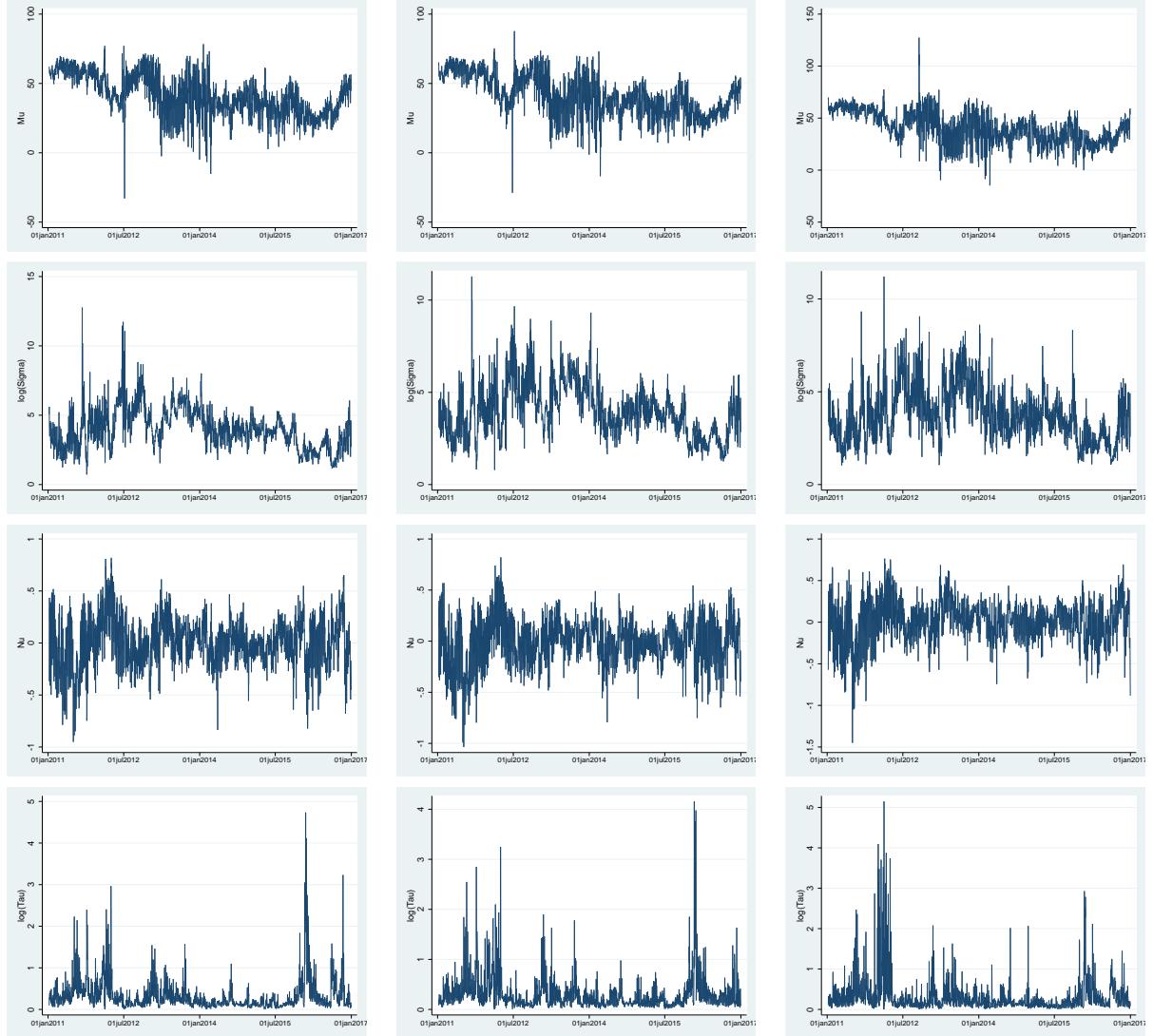


Figure 6: Filtered time series of mean ($\hat{\mu}_t$) on the first rows, volatility ($\log(\hat{\sigma}_t)$) on the second rows, skewness ($\hat{\nu}_t$) on the third rows and kurtosis ($\log(\hat{\tau}_t)$) on the last rows, from a skew-t representation of prices at hour 12 based on the MFST (first column), AR-MFST (second) and VAR-MFST models (third column) estimated on a rolling window of 365 days and continuously updated from 2010 towards the end of our sample. Static filtered moments can also be extracted once the model is estimated on the full sample.

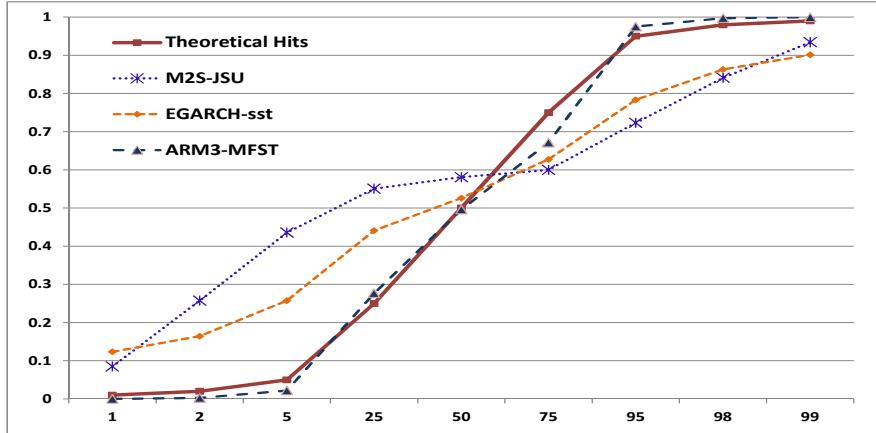
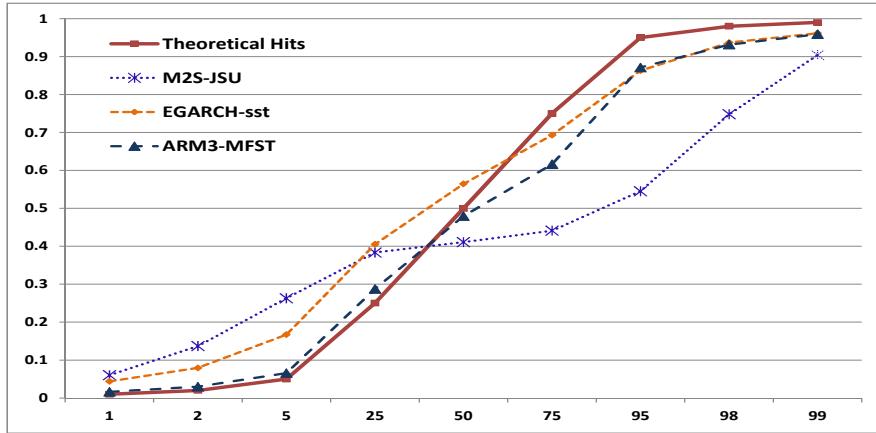


Figure 7: Theoretical versus Empirical Hits of Selected Models across Quantiles for hour 12 in 2011 (first row) and in 2015 (second row). Theoretical Hits are simply 1%, 2%, 5%, 25%, 50%, 75%, 95%, 98%, 99%; M2S-JSU is the model proposed by Serinaldi (2011); EGARCH-sst is the AR(7)X-EGARCH(1,1)-X with a skew student-t distribution as formulated in eqs. 9-10; ARM3-MFST is the ‘MFST’ model with three moment equations and autoregressive terms included (as formulated in Section 4).

5. Final Comments and Conclusion

The development of an accurate, flexible and analytically tractable representation for electricity price processes has considerable practical value for operations control, risk management and financial products, and the increasing penetration of renewable energy, through solar and wind, is not only adding to the complexity of price formation, but also the greater need for short-term hedging products. Closed form solutions are desirable for hedging, but the choice of an appropriate density function is awkward in power markets because of the rapid shape changes that may occur over time due to intermittent wind and solar output as well as the usual demand and fuel price shocks. The increasing interest in battery storage operations and EV charging logistics is also motivating the use of analytics that require accurate short-term models of the hourly prices. Furthermore, whilst generators have always needed to plan unit commitment several days in advance, retailers are now looking to anticipate demand-side engagement at the day ahead stage, and both generators and retailers are carefully considering their offer quantities to the day ahead auctions in the light of reserving capacities for the profitable intra day and balancing markets.

With this context and methodological need in mind, we have examined the applicability of a stochastic price formation model based upon dynamic latent moments estimated within a skewed-t density to accommodate the range of shapes that hourly power prices exhibit, and furthermore related these shape changes dynamically to a set of exogenous drivers. This is conveniently and transparently achieved through linear multifactor representations of the first four moments of this density, estimated as dynamic latent variables. Benchmarked against quantile regression and ARMAX-GARCHX, the method performed well. In particular, the ability to capture the swings between positive and negative skewness by time of day and according to the amount of renewable energy generation is an appealing feature and could provide a useful ingredient into the daily optimisation of trading positions and operational scheduling. Furthermore, since the evolution of market structure is represented through the exogenous variables, the frequent re-specification that a more empirical approach would require is avoided. The MFST approach therefore offers shape flexibility and stability of specification. Furthermore, introducing lagged latent moments into the model, as AR-MFST, is attractive for interpreting persistence in skewness and volatility, as well as adaptive evolution in the mean, and this was robust to out-of-sample testing.

The approach was not outperformed by the nonparametric quantile regression benchmark, which does not offer analytical solutions, but would have been expected to estimate the percentiles more precisely. In fact our proposed approach was clearly better in terms of out-of-sample performance. This is reassuring for the general functional form of the skew-t and its multifactor drivers. Regarding its comparison to an ARX-EGARCHX, also in this case, the (V)AR-MFST models outperformed, showing that extreme quantiles were better calibrated by the multifactor skew-t models compared to the GARCH model. We observe that the inclusion of conditional volatility and lagged higher moments made the specification more dynamic and superior to the GARCH approach.

It is an open question if the dynamic specification of the fourth moment adds value, with the three moment version performing better out of sample, but for situations where this may be useful, eg portfolio optimisation models that explicitly use the first four moments (Giesecke et al. (2014)), the formulation appears sufficiently robust. As regards, the choice

of further exogenous variables, the reserve margin and the flow of imports/exports in the German context would be potentially valuable and may improve accuracy. There is certainly scope for a more comprehensive multifactor modelling of the price formation process, but the main methodological objectives of this research have nevertheless been well supported by the specification, as applied.

Beyond the electricity case which motivated this study, this general approach should have quite wide appeal in capturing the shape-shifting properties in the market price densities not only of a wide range of commodities but also for prices in any market where the key ingredient is a role for significant and time-varying exogenous drivers in determining dynamic shifts in the density shapes.

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6. APPENDICES

6.1. Reference Figures



Figure 8: 24 daily time series in 2007, one for each of the separately determined hourly prices.

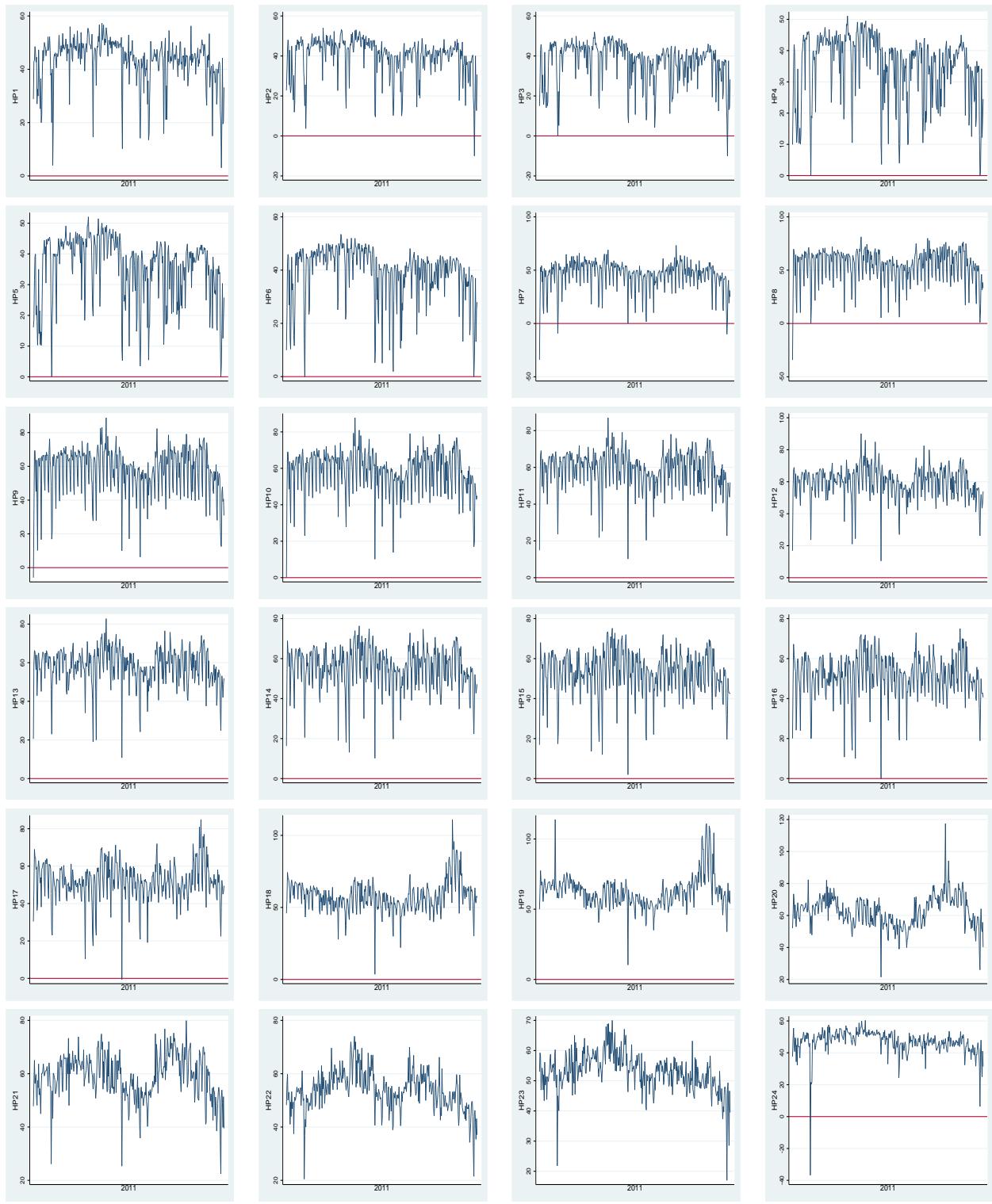


Figure 9: 24 daily time series in 2011, one for each of the separately determined hourly prices.

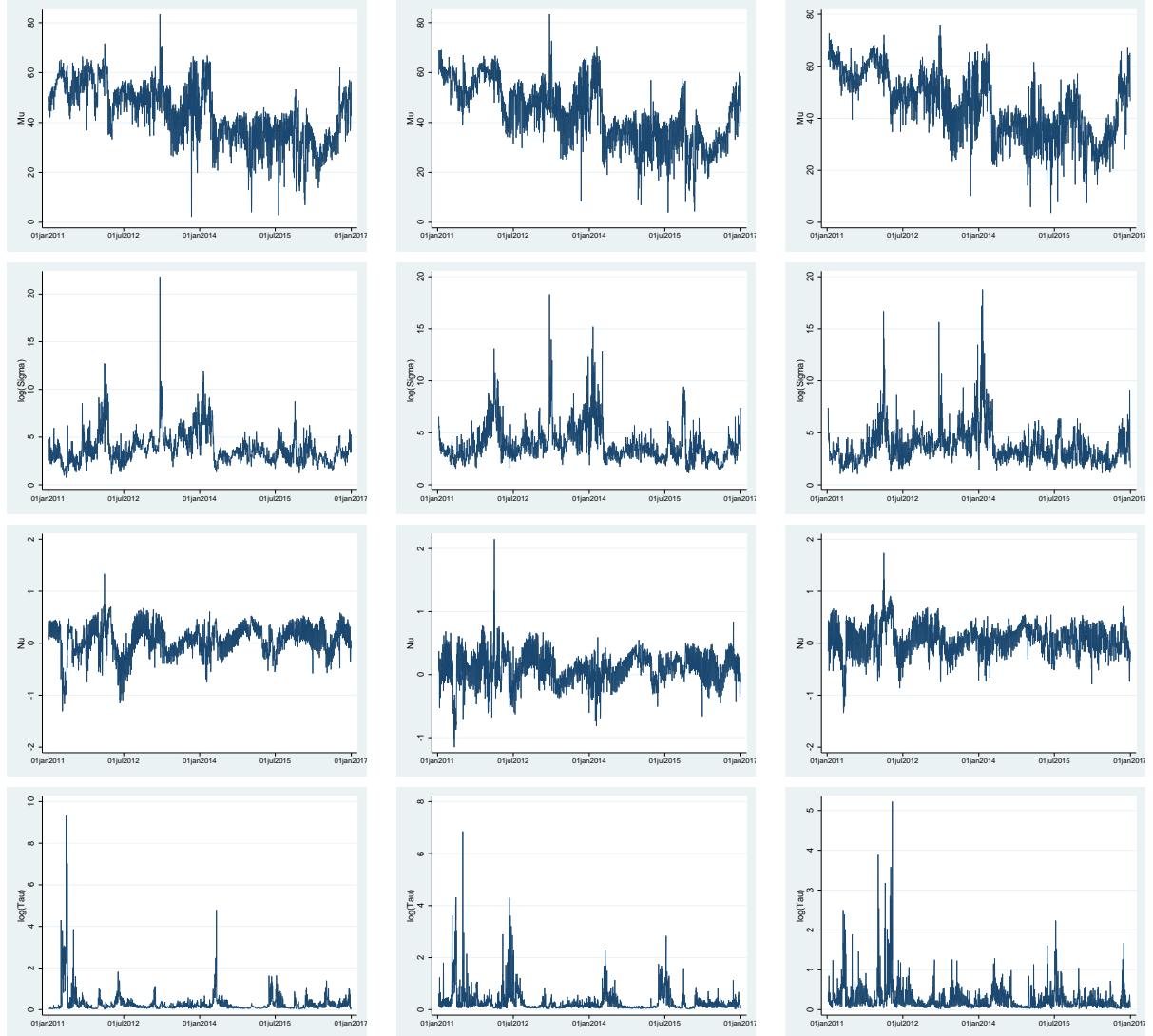


Figure 10: Filtered time series of mean ($\hat{\mu}_t$) on the first rows, volatility ($\log(\hat{\sigma}_t)$) on the second rows, skewness ($\hat{\nu}_t$) on the third rows and kurtosis ($\log(\hat{\tau}_t)$) on the last rows, from a skew-t representation of prices at hour 19 based on the M4-MFST (first column), ARM4-MFST (second) and VARM4-MFST models (third column) estimated on a rolling window of 365 days and continuously updated from 2010 towards the end of our sample. Static filtered moments can also be extracted once the model is estimated on the full sample.

6.2. Yearly Descriptive Statistics across 24 Hours

Hours	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015	2016
	Mean										
1	36.09	26.42	50.23	30.30	37.46	43.28	34.31	28.99	26.10	25.17	23.96
2	31.46	22.59	43.52	25.05	33.77	39.87	30.69	26.25	23.92	23.27	22.30
3	28.07	20.19	38.93	20.85	30.56	36.91	28.38	24.15	22.43	21.90	21.26
4	25.03	18.12	35.41	19.02	27.63	34.85	26.21	23.29	21.14	21.29	20.58
5	25.65	18.13	36.31	19.66	28.01	35.44	26.61	23.74	21.69	21.69	21.00
6	30.90	22.88	43.08	24.45	32.00	39.16	29.74	26.15	23.89	23.81	22.41
7	36.18	25.69	50.70	30.57	38.29	45.49	36.90	34.18	30.55	30.08	27.03
8	52.41	39.21	70.19	41.52	46.19	54.31	46.96	43.60	37.53	36.86	32.95
9	58.50	45.91	75.64	45.00	50.07	57.44	51.38	46.71	39.65	38.85	34.94
10	64.99	49.11	81.77	47.97	52.23	58.32	51.04	45.75	38.53	36.99	34.09
11	69.35	52.44	86.11	50.45	53.28	58.72	49.77	43.84	36.87	34.93	32.28
12	80.78	59.83	93.43	53.56	54.86	59.48	49.50	43.39	36.57	34.21	31.74
13	68.40	50.89	84.96	50.25	52.84	57.92	46.65	40.36	34.19	31.58	29.51
14	64.74	48.06	79.73	46.78	49.94	55.18	44.02	38.10	32.37	30.10	27.96
15	60.94	44.70	75.70	43.25	47.57	52.83	42.05	36.41	31.51	29.42	27.15
16	57.13	41.37	71.04	40.28	45.76	51.29	42.04	36.88	32.08	30.52	28.21
17	54.41	41.38	69.62	40.05	46.00	51.78	43.26	38.50	33.61	32.23	29.74
18	58.27	51.53	76.14	45.53	50.82	57.52	49.87	45.41	38.66	37.78	34.50
19	69.39	53.33	80.78	49.90	54.54	62.55	56.12	50.95	43.37	41.61	36.83
20	58.78	46.39	78.22	48.88	53.52	62.37	56.39	51.29	44.20	42.53	37.19
21	55.07	40.15	74.32	44.98	49.42	58.93	50.98	46.29	39.76	38.70	34.01
22	48.30	34.26	65.82	40.36	45.81	53.63	46.35	41.46	35.61	34.91	30.84
23	46.53	32.90	63.61	40.08	46.23	52.85	45.10	39.14	33.78	33.24	29.36
24	37.57	26.20	52.93	33.78	40.85	46.85	37.97	31.86	28.33	27.35	25.66
	Standard Deviation										
1	10.76	9.23	12.87	11.95	7.28	8.70	14.58	8.20	7.13	9.13	7.74
2	11.20	8.13	14.15	16.98	8.83	9.68	18.63	9.43	7.88	9.33	8.52
3	11.26	7.97	15.24	30.92	10.28	10.34	19.79	11.01	8.24	9.96	9.24
4	11.52	7.70	15.11	17.63	10.60	10.71	20.99	10.78	9.00	9.21	9.14
5	11.02	7.87	14.93	14.86	10.07	10.54	19.04	10.56	8.36	9.18	8.76
6	11.52	9.25	15.17	12.76	10.32	10.11	18.65	9.99	8.29	8.32	8.84
7	16.52	15.03	23.68	20.55	13.40	13.40	22.40	14.09	11.11	11.46	11.11
8	25.05	28.81	30.52	23.94	15.84	15.51	22.01	18.17	14.45	14.20	13.44
9	28.89	34.81	28.59	19.00	14.66	13.83	19.35	18.14	13.81	13.88	13.42
10	40.30	30.07	28.68	16.35	12.28	11.62	16.21	16.30	12.61	12.37	12.20
11	59.40	31.40	29.73	15.43	11.33	10.31	15.19	15.51	11.92	11.84	12.02
12	128.53	41.02	35.17	16.00	11.05	9.99	14.22	15.89	11.77	11.57	11.96
13	49.95	25.73	25.70	13.70	9.97	9.64	13.62	14.80	11.65	11.10	12.82
14	46.54	27.11	25.26	14.25	11.11	10.64	14.04	15.53	13.43	12.33	14.47
15	47.91	25.96	26.43	13.76	11.51	11.09	13.95	17.19	13.74	12.92	15.80
16	42.89	23.66	24.34	12.94	11.74	10.87	13.49	16.69	13.44	12.70	14.44
17	28.21	27.12	22.19	13.41	12.39	10.28	13.57	15.34	12.31	11.82	14.89
18	26.61	64.62	31.94	17.64	15.50	10.73	16.60	17.97	12.61	12.41	15.42
19	131.19	55.41	26.75	19.39	13.40	11.83	20.13	18.06	12.70	12.06	12.08
20	21.96	30.50	23.56	16.72	10.78	9.79	15.85	15.67	11.56	11.39	10.54
21	17.46	18.01	19.63	11.69	7.49	8.42	11.49	12.07	8.30	9.02	8.31
22	13.69	11.72	14.54	7.07	5.39	7.33	9.44	10.28	6.98	8.28	7.11
23	12.53	9.76	11.60	7.10	4.98	6.30	8.30	8.45	6.23	7.99	6.37
24	9.82	7.61	11.28	8.24	5.72	7.70	10.69	7.94	6.86	7.96	7.12
	Skewness										
1	0.56	1.77	-0.88	-5.48	-0.66	-1.84	-7.75	-11.18	-1.86	-1.79	-1.49
2	0.30	0.90	-0.97	-6.09	-1.07	-1.71	-9.04	-1.51	-2.13	-1.94	-1.97
3	0.15	0.50	-2.48	-13.48	-1.07	-1.42	-8.86	-2.53	-2.07	-2.24	-2.66
4	0.14	0.42	-2.33	-4.85	-0.77	-1.10	-8.11	-2.03	-3.09	-1.82	-2.38
5	0.08	0.52	-2.46	-4.13	-0.69	-1.14	-8.17	-1.64	-2.56	-2.21	-2.00
6	-0.29	0.55	-0.89	-4.29	-1.16	-1.67	-8.64	-1.24	-2.87	-1.42	-2.25
7	-0.21	0.93	-0.84	-5.00	-1.27	-1.85	-6.76	-1.11	-1.24	-1.11	-2.32
8	-0.07	3.00	-0.33	-2.66	-0.68	-1.50	-1.71	-0.29	-0.66	-0.68	-1.05
9	1.93	4.88	0.19	-1.22	-0.50	-1.18	1.15	-0.10	-0.35	-0.43	-0.54
10	5.13	2.55	0.64	0.68	-0.23	-1.21	0.61	-0.11	-0.12	-0.24	0.40
11	11.07	2.58	0.81	1.19	-0.04	-1.15	0.61	-0.12	-0.05	-0.15	0.46
12	12.24	3.35	1.55	1.42	0.06	-1.01	0.52	0.30	0.09	-0.03	0.49
13	8.07	1.97	0.83	1.25	0.01	-1.27	0.35	-0.07	-0.31	-0.14	-1.15
14	8.76	2.44	0.47	1.01	-0.35	-1.08	0.19	-0.36	-1.57	-1.32	-1.87
15	10.64	2.16	0.49	0.71	-0.30	-1.03	0.05	-1.81	-1.71	-1.88	-2.82
16	9.64	2.00	0.28	0.57	-0.33	-1.05	-0.05	-1.65	-1.85	-1.34	-0.84
17	3.69	2.91	0.27	0.75	0.10	-0.79	0.43	-0.45	-0.74	-0.30	-0.47
18	2.06	7.55	6.58	1.44	1.48	0.41	1.54	0.75	0.44	0.37	1.38
19	16.47	6.11	2.84	1.73	1.19	1.23	2.90	0.79	0.64	0.41	0.92
20	0.90	3.16	3.02	2.14	0.92	0.20	1.87	0.85	0.67	0.44	0.68
21	0.61	1.67	1.27	1.31	0.77	-0.68	1.01	0.29	-0.03	-0.25	0.00
22	0.68	1.44	0.50	1.12	0.33	-0.39	-0.31	0.14	-0.41	-0.42	-0.91
23	0.85	1.57	0.17	1.32	0.22	-0.85	-0.74	-0.19	-0.89	-0.27	-0.60
24	0.66	1.42	-0.13	-1.95	-0.63	-4.46	-5.38	-1.91	-2.28	-1.08	-1.70
	Kurtosis										
1	3.61	8.22	4.84	71.08	3.32	7.03	89.13	4.67	7.17	8.46	9.36
2	3.44	5.94	5.22	57.11	4.66	6.22	105.79	6.60	9.54	8.03	12.53
3	2.74	4.48	21.84	223.38	4.63	4.79	102.88	15.72	9.82	10.35	18.26
4	2.44	3.90	20.44	41.09	3.47	3.47	86.69	13.98	23.16	7.73	17.27
5	2.49	4.67	21.82	35.88	2.95	3.63	89.33	10.07	18.23	12.46	15.70
6	3.05	5.20	3.78	47.10	3.87	5.65	99.38	6.07	20.83	5.71	18.00
7	2.64	4.59	3.70	50.96	4.19	8.29	66.90	6.20	5.61	5.06	18.50
8	2.51	22.35	2.95	30.85	3.66	6.56	27.15	2.81	3.76	3.42	11.49
9	15.69	47.70	3.43	18.27	3.83	5.10	10.20	3.02	2.82	3.05	8.93
10	46.66	13.01	4.43	6.40	3.96	5.92	7.10	2.95	3.15	3.30	4.83
11	166.68	12.63	4.39	5.34	4.29	5.87	6.61	2.90	3.36	3.32	4.54
12	167.81	20.11	7.27	5.63	3.74	6.36	5.92	5.02	3.28	3.13	4.37
13	88.86	8.92	4.83	5.09	3.82	6.46	4.95	3.40	5.46	3.65	15.65
14	109.26	13.64	3.19	4.93	4.24	5.11	4.36	3.67	13.76	12.44	20.53
15	158.61	9.78	3.14	4.99	4.09	5.17	4.24	14.34	14.04	17.31	30.61
16	136.73	9.38	2.77	5.58	4.59	5.65	4.36	14.38	14.72	11.72	12.65
17	29.62	16.78	2.80	5.77	5.16	6.18	6.00	4.95	7.96	4.06	10.88
18	13.01	78.17	82.38	5.86	7.69	7.21	9.72	4.43	4.11	4.42	6.21
19	293.55	58.69	17.11	7.26	5.90	7.98	18.27	3.82			

6.3. Yearly Goodness-of-fit Statistics and Computational Times

	SEP1	SEP2	SHASHo	SHASHo2	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG
HP 1														
2010	sec	1.95	0.88	0.84	0.45	2.94	0.62	0.97	1.22	0.53	0.76	0.08	0.08	0.10
	KS	0.1043	0.1311	0.1019	0.0999	0.0345	0.0401	0.0261	0.0252	0.2004	0.0795	0.0767	0.1533	0.0765
	CVM	1.3808	2.1892	1.1370	1.0399	0.0812	0.0954	0.0269	0.0261	4.6939	0.5771	0.5852	3.1919	0.6642
	AD	9.6540	12.9543	6.3452	5.2505	0.4870	0.5755	0.3833	0.3927	30.8319	3.3187	3.3320	20.3288	4.0299
	AIC	2560	2571	2518	2503	2462	2466	2472	2474	2773	2488	2486	2625	2483
2011	sec	1.90	0.63	0.79	0.39	> 100	0.39	1.16	0.97	0.43	0.69	0.08	0.06	0.09
	KS	0.5537	0.3347	0.3385	0.1570	0.1279	0.0550	0.0549	0.3067	0.0774	0.1696	0.2198	0.1967	0.3663
	CVM	50.9773	17.7589	13.7136	2.5864	1.6536	0.2247	0.2226	11.7844	0.6778	3.1368	6.4255	4.2294	16.7820
	AD	105.9948	60.6156	13.7888	10.1099	2.3374	2.3246	63.9873	9.2319	17.8803	35.1438	20.3593	77.4224	
	AIC	11508	3423	2697	2548	2536	2461	3046	2580	2616	2742	2541	2994	
2012	sec	16.62	51.18	0.81	0.55	17.63	0.48	0.99	1.01	0.60	1.76	0.08	0.08	0.06
	KS	0.6393	0.5794	0.2199	0.2114	0.1392	0.1394	0.4675	0.0463	0.2209	0.2530	0.2940	0.9816	
	CVM	38.5683	42.4846	5.6398	5.0497	1.6275	1.6311	26.7599	0.1587	5.6783	10.7433	0.9326	119.2940	
	AD	193.3144	30.5878	27.3843	8.6018	8.6160	126.4099	1.8496	30.7987	56.9362	44.4275	> 1900		
	AIC	>>	3678	2969	2742	4495	2465	2995	3068	2836	7588			
2013	sec	er	er	0.80	0.68	94.60	0.65	1.03	1.22	0.58	0.88	0.08	0.07	0.08
	KS	0.2128	0.2762	0.1899	0.2105	0.0814	0.0662	0.0295	0.0304	0.2564	0.1044	0.1127	0.1906	0.1297
	CVM	7.9361	12.6973	3.7454	4.8465	0.6839	0.4256	0.0429	0.0416	7.6231	1.2110	1.4358	4.3985	1.7883
	AD	17.7897	22.2952	3.6564	2.6397	0.4862	0.4810	45.1560	7.6349	8.3450	25.8997	9.3280	36.0136	
	AIC	3175	3223	2597	2605	2512	2489	2916	2567	2566	2700	2531	2792	
2014	sec	23.79	er	0.81	0.54	er	0.43	1.13	1.44	0.61	0.98	0.08	0.06	0.08
	KS	0.6335	0.3497	0.3474	0.1613	0.120	0.0704	0.0707	0.3227	0.0977	0.1739	0.2426	0.2005	0.3985
	CVM	60.6188	19.5052	14.2807	3.0956	2.2901	0.4359	0.4391	11.9734	1.0463	3.0671	6.4328	4.5950	19.2615
	AD	63.8515	15.9596	15.9596	13.2023	3.772	3.2818	64.0149	12.0111	19.9117	34.9124	21.9710	88.5767	
	AIC	>>	8851	2513	2341	2253	2846	2373	2417	2542	2543	2555		
2015	sec	4.35	er	0.93	0.58	0.35	0.94	1.01	0.38	0.61	0.06	0.07	0.07	0.10
	KS	0.2890	0.3161	0.1425	0.1232	0.0683	0.0692	0.3132	0.1060	0.1555	0.2033	0.1730	0.3935	
	CVM	10.4052	11.8170	2.0225	1.4370	0.4010	0.3982	12.9457	1.0474	2.4432	6.0687	3.1722	19.4923	
	AD	89.7501	53.9219	10.5565	7.8600	2.6301	2.5690	69.1543	7.3633	13.1756	33.4590	16.0647	89.6177	
	AIC	6579	2804	2595	2590	2534	2532	3151	2634	2645	2769	2599	3115	
2016	sec	2.61	3.96	0.87	0.46	1.09	0.42	0.83	0.86	0.47	0.95	0.06	0.08	0.06
	KS	0.3542	0.3176	0.1086	0.0927	0.0723	0.0696	0.1533	0.0910	0.1187	0.1866	0.1490	0.4191	
	CVM	20.3441	10.4512	1.0312	0.7630	0.4876	0.4574	12.9752	0.5959	1.1565	5.0358	2.0256	22.4516	
	AD	48.5671	5.5252	4.1592	2.6349	2.4987	69.4496	3.8469	6.4984	28.9854	11.0424	102.4420		
	AIC	5892	2740	2502	2500	2471	2470	3115	2535	2537	2662	2518	3108	
HP 2														
2010	sec	1.46	2.47	0.62	0.44	er	0.57	0.67	0.75	0.49	0.70	0.06	0.08	0.09
	KS	0.1967	0.2493	0.1563	0.1731	0.0640	0.0555	0.0429	0.0457	0.2463	0.0977	0.0985	0.1672	0.0970
	CVM	6.0047	9.7827	2.8068	3.5846	0.2902	0.2028	0.1073	0.1229	6.9717	0.8188	0.8548	3.6102	1.1048
	AD	44.8887	13.5729	17.1491	1.4014	1.0447	0.7084	0.8307	42.3148	4.5120	4.7178	22.6638	6.2409	29.2194
2011	sec	1.63	1.26	0.62	0.45	er	0.39	1.81	2.12	0.53	0.75	0.06	0.08	0.09
	KS	0.5076	0.3641	0.3177	0.1688	0.1430	0.0665	0.0666	0.2965	0.0989	0.1845	0.2187	0.2029	0.3559
	CVM	39.7275	17.3078	11.9079	2.5446	1.7575	0.2871	0.2844	10.6147	0.9205	3.2174	6.0129	3.9994	14.5305
	AD	78.8731	52.6372	12.9447	10.1919	2.6181	2.6078	58.7266	11.8056	17.6847	33.1361	19.0758	66.9070	
	AIC	6232	2740	2620	2609	2541	3081	2660	2691	2821	2621	3034		
2012	sec	9.60	er	27.67	1.19	0.46	0.06	0.79	0.83	0.47	0.70	0.06	0.08	0.09
	KS	0.7787	0.9945	0.2168	0.2117	0.1605	0.1055	0.2616	0.2614	27.4254	0.1587	7.2688	12.5223	11.5658
	CVM	59.5991	120.0109	7.2922	6.3019	34.6691	34.6152	13.0893	13.0913	129.0777	1.8496	39.8791	65.4567	57.0516
	AD	>>	>>	3144	3142	2831	2832	4703	2465	3170	3232	2084	7736	
	AIC	>>	>>	2752	2604	2564	3129	2658	2660	2653	2792	3056		
2013	sec	1.67	0.71	0.81	0.61	> 150	0.47	0.72	0.80	0.64	0.71	0.08	0.07	0.09
	KS	0.4962	0.4013	0.2482	0.1049	0.0964	0.0855	0.0858	0.2939	0.1133	0.1215	0.2067	0.1264	0.3058
	CVM	41.4389	24.0605	8.8667	1.1364	0.9043	0.3702	0.3723	9.9005	1.4381	1.6227	4.2936	2.0785	12.9561
	AD	41.8339	41.8339	5.9565	4.9003	3.1641	3.1871	56.3943	9.1054	8.9442	26.2086	11.0855	61.2545	
	AIC	>>	>>	2752	2594	2564	3129	2658	2660	2653	2792	3056		
2014	sec	er	er	1.23	0.47	er	0.5	0.93	0.91	0.72	0.72	0.09	0.1	0.09
	KS	0.7671	0.8333	0.3566	0.1603	0.1401	0.0653	0.0653	0.3467	0.0847	0.1702	0.2451	0.1987	0.4473
	CVM	78.8824	91.8229	16.6291	2.7307	2.1475	0.2149	0.2141	13.6960	0.6803	3.1409	6.1295	4.2783	25.9378
	AD	76.3807	58.8355	11.7469	7.3552	6.5116	2.6083	2.6115	9.1220	22.7225	4.7225	31.0819	30.7777	30.7777
	AIC	>>	>>	40.8121	36.4886	13.4274	13.4411	128.9497	1.8496	41.0196	66.4242	58.3554	> 2000	
2015	sec	er	er	0.73	0.87	er	0.67	0.56	0.66	1.76	0.75	0.08	0.07	0.08
	KS	0.6561	0.3590	0.0990	0.1028	0.0777	0.0779	0.3785	0.0961	0.1057	0.2157	0.1199	0.5551	
	CVM	41.6352	19.3388	1.2196	0.9982	0.5327	0.5329	17.0629	0.9204	1.3967	4.8512	2.4580	52.8333	
	AD	90.1463	7.4434	5.7117	4.1830	4.1961	86.9382	6.2943	8.4513	29.8888	14.0135	283.4390		
	AIC	>>	>>	3077	2732	2726	3605	2752	2773	2892	2897	3897		
2014	sec	er	er	1.69	1.04	37.89	1.02	0.98	1.06	1.61	1.12	0.08	0.11	0.13
	KS	0.7653	0.8928	0.3113	0.1310	0.1349	0.0885	0.0885	0.3520	0.0800	0.1492	0.2240	0.1045	0.1111
	CVM	83.7038	99.2333	16.1892	2.5493	1.7315	0.2676	0.1973	13.8646	0.7222	2.5672	5.5370	3.5586	29.0196
	AD	75.0351	11.7859	9.9569	2.4595	2.4511	3.1327	3.1329	73.8313	8.9729	14.0008	31.8295	18.1992	141.3160
	AIC	>>	>>	2909	2650	2641	2557	3095	2469	2512	2634	3189		
2015	sec	1.13	1.21	0.78	0.58	19.31	0.78	1.40	1.58	0.78	0.84	0.08	0.08	0.11
	KS	0.8379	0.8812	0.3627	0.1410	0.1342	0.0499	0.0499	0.3594	0.0781	0.1499	0.2356	0.1729	0.4806
	CVM	94.4345	97.0608	18.8198	2.6									

	SEPI	SEP2	SHASHo	SHASHo2	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG		
HP 4																
2010	sec	9.74	er	0.93	0.88	6.77	0.61	0.87	1.17	0.58	0.76	0.08	0.06	0.07	0.1	
	KS	0.1207	0.1509	0.0856	0.1146	0.0333	0.0344	0.0394	0.0474	0.2193	0.0866	0.0866	0.1670	0.0719	0.1761	
	CVM	1.9020	2.9639	0.7369	1.3779	0.0637	0.0632	0.1015	0.1553	4.9719	0.6317	0.6320	3.0000	0.6382	3.7895	
	AD	12.5125	17.0817	3.9722	6.5940	0.4469	0.4321	0.8183	1.1257	32.5812	3.6354	3.6362	19.8885	4.0828	18.6208	
	AIC	2867	2881	2756	2761	2721	2721	2735	2738	3070	2756	2754	2893	2741	2923	
2011	sec	1.59	0.87	0.72	0.59	74.45	0.5	1.05	1.15	1	0.59	0.09	0.08	0.08	0.08	
	KS	0.2045	0.2274	0.1734	0.2071	0.1113	0.1004	0.0598	0.0601	0.2433	0.1477	0.1441	0.2093	0.1479	0.2390	
	CVM	4.9371	6.5266	2.9125	4.4781	1.0009	0.8106	0.1849	0.1881	5.6471	2.1763	2.1438	4.2326	2.1761	5.9168	
	AD	26.4488	33.4352	12.7205	19.7600	5.0823	4.8065	1.6221	1.6406	35.5928	12.4633	12.0514	24.9330	11.1334	29.3874	
	AIC	2789	2810	2600	2716	2631	2602	2666	2688	3300	2763	2761	2900	2717	2924	
2012	sec	er	er	0.61	> 80	0.72	1.47	0.96	0.57	0.33	0.09	0.08	0.08	0.08	0.08	
	KS	0.8000	0.6756	0.2046	0.1923	0.1256	0.1256	0.4689	0.0463	0.2355	0.3075	0.2954	0.9743	0.2954	0.9743	
	CVM	64.3944	58.5051	7.0742	6.2841	2.0964	2.0985	26.9361	0.1587	7.1161	12.6411	11.5732	118.1021			
	AD			283.5335	40.4860	36.5685	12.0376	12.0502	127.1166	1.8496	40.7426	66.2035	57.5411	> 700		
	AIC			>>	3997	3228	2914	2914	4697	2465	3257	3322	3073	7376		
2013	sec	er	0.62	0.89	0.55	> 200	0.57	0.57	0.62	1.54	0.72	0.09	0.08	0.06	0.08	
	KS	0.8870	0.4113	0.3202	0.0855	0.0904	0.0769	0.0736	0.3603	0.1063	0.0930	0.1814	0.0901	0.0901	0.5797	
	CVM	> 100	19.7004	15.7141	0.5893	0.5324	0.5045	0.5271	14.9981	0.7890	0.7671	3.6982	1.2739	57.6342		
	AD			74.2578	3.5322	2.9158	3.3162	3.5938	78.3246	4.3443	4.5371	24.2414	7.9567	> 300		
	AIC			>>	3057	2712	2687	2688	3543	2747	2750	2877	2704	4008		
2014	sec	er	er	0.9	1.03	er	0.7	0.49	0.56	0.7	0.92	0.08	0.06	0.07	er	
	KS	0.7923	0.4263	0.1378	0.1356	0.0888	0.0888	0.4096	0.0909	0.1435	0.2413	0.1653	0.1653	0.1653	0.1653	
	CVM	62.7834	24.2725	1.9257	1.6512	0.4388	0.4390	19.3243	0.7114	2.0834	5.5590	3.4822	88.0624			
	AD			111.9962	11.1687	9.3618	4.3842	4.3847	95.8934	6.8059	12.0344	32.6887	18.3608	62.5921		
	AIC			>>	2978	2539	2431	2431	3523	2526	2575	2687	2488	4490		
2015	sec	0.98	0.67	0.95	0.52	27.81	0.81	er	0.75	0.58	0.59	0.06	0.08	0.08	0.11	
	KS	0.6582	0.6307	0.3021	0.1321	0.1213	0.0652	0.0653	0.3262	0.0789	0.1442	0.2242	0.1560	0.1560	0.3750	
	CVM	69.0908	50.3467	13.6487	1.8487	1.3801	0.2247	0.2242	12.3242	0.8497	2.2599	5.3449	3.1991	21.2121		
	AD			62.7400	10.1942	7.9779	2.5013	2.5003	66.6485	9.2079	13.0111	30.9201	16.3746	99.7264		
	AIC			>>	2784	2589	2507	2507	3167	2621	2648	2775	2577	3177		
2016	sec	1.64	0.65	1.49	0.59	2.66	0.71	0.96	1.13	0.6	1.02	0.08	0.07	0.08	0.11	
	KS	0.8017	0.4045	0.1507	0.1397	0.1001	0.1010	0.3808	0.1238	0.1575	0.1871	0.1778	0.1778	0.1778	0.1741	
	CVM	65.6259	21.0197	1.9191	1.5186	0.8714	0.8563	18.2593	0.9983	2.0247	6.0170	3.0664	76.6072			
	AD			94.6257	9.6661	7.5677	4.5508	4.4793	91.8512	5.5836	10.3508	33.5938	15.7217	> 400		
	AIC			>>	3022	2619	2565	2565	3479	2642	2653	2770	2606	4219		
HP 5																
2010	sec	er	er	1.43	er	> 100	0.56	0.73	0.7	2.95	0.67	0.11	0.07	0.07	0.08	
	KS	0.0944	0.1146	0.0961	0.0764	0.0370	0.0279	0.0379	0.0419	0.1989	0.0755	0.0755	0.0752	0.0752	0.1475	
	CVM	1.1590	1.1612	0.7643	0.6466	0.0528	0.0444	0.0522	0.1488	4.1636	0.5539	0.5539	2.8866	0.5153	2.6604	
	AD	7.4178	9.4164	4.8496	3.4398	0.5388	0.4049	0.8594	1.1115	28.0520	3.2294	3.2311	19.3156	3.5634	14.1088	
	AIC	2736	2744	2723	2704	2682	2695	2983	2983	2704	2842	2698	2822			
2011	sec	1.78	0.72	0.96	0.46	> 200	0.53	0.68	0.72	1.05	0.69	0.06	0.07	0.07	0.08	
	KS	0.2008	0.2320	0.1616	0.1933	0.1012	0.0883	0.0558	0.0562	0.2402	0.1408	0.1392	0.1392	0.1372	0.2287	
	CVM	5.0548	6.9943	2.7795	4.2929	0.8396	0.6466	0.1286	0.1301	5.8866	4.9142	4.9142	4.1271	4.2027	6.0205	
	AD	27.6920	36.2844	12.1794	18.8648	4.1508	3.7962	1.1789	1.2106	36.7400	11.1488	10.8851	24.8258	10.2650	29.6992	
	AIC	2801	2825	2691	2709	2667	2666	3021	3021	2742	2740	2878	2698	2822		
2012	sec	31.2	74.84	1.01	1.19	er	0.85	0.67	0.72	1.25	1.86	0.1	0.07	0.08	er	
	KS	0.7432	0.6797	0.6797	0.1868	0.1732	0.1022	0.1021	0.4682	0.0463	0.1876	0.2926	0.2814	0.2814	0.9778	
	CVM	53.4748	63.6934	5.9456	5.1884	1.2801	1.2782	26.9331	0.1587	5.9829	11.6338	10.1940	118.2857			
	AD			322.0774	34.9762	31.0550	7.7130	7.7049	127.1051	1.8496	35.2018	62.0799	51.6763	> 300		
	AIC			>>	4347	3153	2860	2860	4684	3182	3247	3002	3002	7378		
2013	sec	1.47	0.83	1.46	0.74	er	0.76	0.66	0.76	0.93	0.79	0.06	0.08	0.08	0.1	
	KS	0.7298	0.8525	0.2747	0.0953	0.0945	0.0735	0.0684	0.3259	0.1140	0.1080	0.1776	0.0973	0.4415		
	CVM	84.7781	105.2470	11.1016	0.5250	0.4685	0.4290	0.4550	12.2342	0.7995	0.7761	3.4898	1.0956	32.5742		
	AD			53.7360	2.9419	2.5454	2.9308	6.6475	4.4272	4.3997	22.9379	6.7880	> 160			
	AIC			>>	2950	2691	2675	2677	3374	2727	2858	2692	2692	3562		
2014	sec	er	er	1.04	0.97	er	0.8	0.64	0.55	0.78	0.92	0.06	0.08	0.08	er	
	KS	0.7773	0.3817	0.1322	0.1313	0.0828	0.0829	0.3906	0.0843	0.1392	0.2266	0.1524	0.1524	0.1524	0.7040	
	CVM	60.2882	21.3293	1.7193	1.5044	0.4637	0.4633	17.6099	0.8749	1.9229	5.0631	2.9084	77.4100			
	AD			99.0188	9.5598	8.1941	4.8234	4.8253	88.8364	7.9863	10.6761	30.0715	15.4657	> 500		
	AIC			>>	2874	2475	2513	2513	3379	2611	2639	2630	2630	4167		
2015	sec	er	> 170	1	0.39	er	0.58	0.95	0.96	0.61	0.82	0.07	0.07	0.07	0.09	
	KS	0.5871	0.3617	0.1196	0.1155	0.0677	0.0678	0.3703	0.0724	0.1282	0.2189	0.1495	0.1495	0.1495	0.6061	
	CVM	34.6601	19.2105	1.7322	1.4146	0.2740	0.2740	16.0455	0.7163	1.9999	5.2638	3.0257	58.8175			
	AD			88.9700	9.7665	8.0192	3.4736	3.4765	82.6491	7.4749	11.3890	30.8338	15.7673	> 300		
	AIC			>>	2937	2590	2513	2513	3379	2611	2639	2630	2630	3896		
2016	sec	22.86	9.37	1.66	0.99	er	0.71	0.59	0.69	1.09	1.1	0.06	0.06	0.06	0.02	
	KS	0.6960	0.3935	0.1383	0.1289	0.1005	0.1016	0.3740	0.1233	0.1460	0.1894	0.1606	0.1606	0.1606	0.7136	
	CVM	48.3794	19.0627	1.7331	1.4365	1.0039	0.9737	17.6625	1.1272	1.8013	5.6729	2.7295	77.0855			
	AD			86.2933	8.5103	7.0486	4.9628	4.9628	89.3606	5.9599	8.9804	31.6439	14.1862	> 400		
	AIC			>>	2991	2589	2549	2549	3425	2614	2618	2738	2596	4212		
HP 6																
2010	sec	er	0.67	0.99	er	er	0.42	er	0.0367	0.0318	0.1956	0.0862	0.0860	0.1565	0.0832	0.1721
	KS	0.1077	0.1376	0.1486	0.1020	0.0502	0.0586	0.0367	0.0357	0.2600	0.0902	0.1320	0.2002	0.1475	0.2731	
	CVM	1.2604	2.1135	2.1470	1.3615	0.1923	0.3134	0.0979	0.0709	4.9943	0.6266	0.6233	3.4470	0.8025	3.2385	
	AD	10.5366	13.9707	11.5335	7.2015	1.1566	1.7720	0.7793	0.6343	32.1228	3.6850	3.6750	21.1784	4.7227	17.0167</	

Table 4: Goodness-of-fit statistics and Computational Time for hours 4-5-6. KS = Kolmogorov-Smirnov, CVM = Cramer-von Mises, and AD = Anderson-Darling statistics; AIC = Bayesian Information Criterion. “>” and “>>” mean respectively ‘greater’ and ‘grater than thousands’. “er” means that an error occurred in the process of fitting the density, then it has been reported in the computational times (in elapsed seconds) of estimated models. The general model formulation (without indication of coefficients and adapted accordingly to the number of moments in used distribution) is: $\mu_t = E(y_t) = y_{t-1} + hol_t + fwind_t + fsolart + load_{t-1} + coal_{t-1} + gast_{t-1} + co2_{t-1}$, $\log(\sigma_t) = \nu_t = \log(\tau_t) = hol_t + fwind_t + fsolart + load_{t-1} + coal_{t-1} + gast_{t-1} + co2_{t-1}$. SN1 used the first three equations, whereas NO, LO, GU and RG used only the first two.

	SEPI	SEP2	SHASHo	SHASHo2	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG
HP 7														
2010	sec	er	er	0.56	er	0.68	0.62	0.71	0.63	0.63	0.06	0.07	0.07	0.09
	KS	0.0752	0.0912	0.1924	0.1632	0.0691	0.1282	0.0744	0.0698	0.2032	0.0698	0.1574	0.1164	0.1570
	CVM	0.5737	0.7299	3.8903	2.840	0.4199	1.6399	0.5261	0.4391	6.1206	0.4922	0.4903	3.7541	1.5716
	AD	12.5869		18.8995	14.1395	2.3583	9.4350	3.3794	2.8187	37.6391	2.8151	2.8151	22.2702	8.4786
	AIC	2978		2980	2727	2607	2684	2615	2612	2930	2614	2612	2748	2665
2011	sec	0.72	2.11	0.54	2.4	0.66	0.92	1.15	0.53	0.91	0.08	0.08	0.08	0.09
	KS	0.7288	0.8001	0.3168	0.0851	0.0678	0.0334	0.0329	0.3197	0.0705	0.0958	0.1723	0.1331	0.4383
	CVM	83.1678	95.2119	11.5326	6.6412	0.3403	0.0781	0.0753	12.4506	0.3308	0.8548	4.5012	1.6179	28.0037
	AD			53.1121	3.8073	2.2633	0.7338	0.7281	67.2618	2.8777	5.3406	26.8478	9.0252	130.9830
	AIC	>>	22737		2929	2685	2681	2654	2655	3321	2722	2725	2851	3410
2012	sec	25.06	18.09	0.62	0.44	11.23	0.39	0.91	1.14	0.72	1.36	0.06	0.08	0.08
	KS	0.7681	0.6574	0.2375	0.2238	0.1487	0.1488	0.4688	0.0463	0.2384	0.2937	0.3282	0.9777	
	CVM	60.3252	49.8804	6.2658	5.4580	1.7007	1.7022	26.8757	0.1587	6.3057	12.0031	10.3090	117.7743	
	AD			> 220	36.3452	32.2419	10.6596	126.8747	1.8496	36.5996	63.6392	51.3834	> 300	
	AIC	>>	3879		3185	3182	2930	4694	2466	3214	3278	3038	27216	
2013	sec	er	er	0.77	1.68	er	1.03	1.05	0.81	0.99	1.36	0.09	0.08	0.11
	KS	0.5836	0.4780	0.2744	0.0558	0.0966	0.0507	0.0484	0.3098	0.0427	0.0530	0.1658	0.1139	0.3727
	CVM	66.5509	40.5945	7.0983	0.2320	0.7238	0.2587	0.2422	10.9645	0.1192	0.1234	3.0961	0.7421	22.5975
	AD			36.3694	1.6609	6.2181	1.7268	1.6661	> 60	1.0314	1.2406	21.1882	5.6781	> 108
	AIC	>>	2919		2687	2785	2680	2682	3334	2707	2706	2837	2713	3368
2014	sec	27.06	2.75	2.01	1.23	> 100	0.69	0.67	0.74	0.59	0.7	0.06	0.06	0.08
	KS	0.5365	0.1318	0.2989	0.0854	0.1359	0.0559	0.0577	0.3135	0.0560	0.0870	0.1840	0.1441	0.3506
	CVM	56.1613	1.9358	8.7181	0.4091	1.7066	0.2598	0.2681	11.1843	0.1977	0.4819	4.1285	1.2977	17.1937
	AD			19.9211	41.3323	3.2375	11.9159	1.7344	1.7744	61.5034	2.4080	4.2174	25.5815	8.2111
	AIC	>>	3920		2645	2560	2471	2419	3038	2476	2476	2452	2963	
2015	sec	er	er	15.89	2.02	0.56	1.58	1.14	0.66	0.64	0.83	0.06	0.08	0.08
	KS	0.2752	0.2932	0.1793	0.1993	0.0562	0.0556	0.0587	0.0703	0.2762	0.0709	0.1724	0.0950	0.2422
	CVM	12.1973	18.2420	3.4826	4.6419	0.1908	0.1649	0.3144	0.4347	7.6377	0.4407	0.4549	3.2894	0.8635
	AD			> 160	17.7708	22.9289	1.2046	1.0443	1.7031	2.3209	45.7958	3.0922	3.2932	21.7750
	AIC	4900		4971	2648	2553	2551	2554	3042	2599	2597	2732	2568	2907
2016	sec	32.62	1.26	2.21	0.64	46.35	0.86	1.03	1.1	0.72	0.83	0.07	0.1	0.12
	KS	0.7732	0.4460	0.1102	0.1368	0.0872	0.0845	0.3965	0.0854	0.1148	0.1982	0.1544	0.8195	
	CVM	61.3611		22.2028	1.4433	2.1732	0.8006	0.7672	19.6343	0.7643	1.4376	5.6762	2.6115	95.2370
	AD			99.5174	8.0098	12.3993	4.5409	4.3802	97.5620	4.6339	8.0723	32.4566	14.8361	> 600
	AIC	>>	3123		2659	2737	2612	2611	3648	2680	2686	2800	2667	4723
HP 8														
2010	sec	2.26	1.28	0.95	0.95	er	1.06	1.06	1.37	0.64	1.02	0.06	0.06	0.08
	KS	0.0664	0.0855	0.1979	0.1606	0.0455	0.0801	0.0423	0.0394	0.1954	0.0588	0.0587	0.1641	0.1276
	CVM	0.4651	0.7973	3.3874	2.0813	0.1821	0.4606	0.1114	0.0801	5.0683	0.3420	0.3419	3.1402	0.2481
	AD	5.7753		17.9253	11.5070	0.1048	3.2946	0.1645	0.7876	32.7814	1.8214	1.8211	20.0594	10.6722
	AIC	2891		2894	2839	2810	2715	2750	2724	3031	2720	2720	2859	2765
2011	sec	3.83	1.22	0.63	2.51	25.12	0.98	0.8	1.17	0.45	0.73	0.08	0.06	0.08
	KS	0.7405	0.3138	0.0976	0.0943	0.0526	0.0502	0.0526	0.3286	0.1081	0.1101	0.1855	0.1184	0.4660
	CVM	86.5030	12.7256	0.8444	0.5743	0.1559	0.1561	0.1561	12.8831	0.7243	1.3153	4.4290	1.7502	33.6274
	AD			59.1597	4.7174	3.2507	1.3487	1.3864	69.1771	4.6302	6.5663	26.6984	9.6135	> 160
	AIC	>>	2956		2695	2691	2659	2659	3353	2734	2737	2737	2737	3539
2012	sec	er	er	0.44	0.37	0.7	0.43	1	1.61	0.39	1.15	0.07	0.1	er
	KS	0.9973	0.4679	0.1479	0.1505	0.1155	0.1104	0.1479	0.0439	0.1493	0.1493	0.2423	0.6423	0.9584
	CVM	> 120	22.8926	3.3166	3.4879	1.9205	1.6004	24.6410	0.1587	3.3449	8.8556	58.0876	> 10	
	AD			> 100	20.2013	21.1637	12.4271	10.6276	> 110	1.8496	20.1758	48.2311	> 120	
	AIC	>>	3751		3095	3128	3013	2996	4406	2465	3104	3193	4199	6479
2013	sec	er	er	0.47	0.4	9.45	0.63	1.25	1.32	0.76	0.78	0.06	0.08	0.07
	KS	0.0778	0.0799	0.1285	0.0918	0.0807	0.0979	0.0678	0.0886	0.2022	0.0684	0.0685	0.1602	0.0996
	CVM	0.6309	0.6153	2.3371	1.0441	0.6979	0.8599	0.5155	0.6014	4.3108	0.5276	0.5276	2.6632	0.9936
	AD			13.5252	6.5539	3.7773	7.3672	8.3294	4.0120	29.4136	3.0589	3.0587	18.9109	5.7618
	AIC	2897		2893	2858	2900	2717	2576	2573	3027	2558	2558	2718	2845
2014	sec	1.62	0.94	2.37	1.51	er	1.36	0.94	1.07	0.78	0.78	0.06	0.07	0.06
	KS	0.1691	0.1949	0.3047	0.1645	0.0513	0.1006	0.0562	0.0523	0.2502	0.0512	0.0512	0.1638	0.1049
	CVM	3.8198	7.6138	9.7525	2.3710	0.1224	1.0482	0.1706	0.1743	7.5051	0.1161	0.1164	3.1463	4.9332
	AD			50.6024	14.8664	1.0810	7.8303	1.3845	1.3462	44.8921	1.1781	1.1801	20.7476	5.4684
	AIC	6215		6257	2900	2717	2576	2573	3027	2718	2718	2718	2718	2845
2015	sec	57.6	er	0.49	0.34	er	0.9	1.37	0.92	0.48	0.61	0.06	0.08	0.08
	KS	0.0893	0.1275	0.0879	0.1049	0.0254	0.0278	0.0353	0.0448	0.2153	0.0585	0.0584	0.1614	0.0784
	CVM	1.1654	2.2407	1.0259	1.1063	0.0340	0.0595	0.0683	0.0984	5.4766	0.2118	0.2113	2.9504	3.0405
	AD	11.6563	16.5843	6.6147	6.7323	0.2536	0.5250	0.5981	0.7187	35.1751	1.3207	1.3196	19.6718	15.7813
	AIC	2962		2976	2703	2702	2641	2646	3004	2656	2654	2702	2660	2823
2016	sec	76.41	83.93	0.81	0.41	> 100	0.54	0.79	0.94	0.45	0.98	0.08	0.06	0.1
	KS			1.0000	0.3740	0.0998	0.1288	0.0719	0.0885	0.3601	0.0911	0.1014	0.2035	0.1623
	CVM			> 100	14.0426	1.2409	1.6410	0.6014	0.9002	17.2711	1.0052	1.1597	5.0421	3.4564
	AD			66.7838	7.0025	9.7371	5.1441	3.2896	87.7998	5.6934	6.6224	29.4270	18.5967	> 100
	AIC	>>	3160		2752	2806	2714	2713	3594	2762	2760	2882	2860	4324
HP 9														
2010	sec	21.81	24.44	1.7	1.02	84.38	0.92	1.12	1.42	0.95	0.84	0.06	0.08	0.08
	KS	0.0444	0.0628	0.2039	0.1784	0.0365	0.0723	0.0390	0.0352	0.2030	0.0433	0.0404	0.1471	0.1283
	CVM	0.2346	0.4484	4.0150	2.8513	0.0606	0.4168	0.0634	0.0519	5.5236	0.1234	0.1237	3.0125	0.6900
	AD			21.77										

	SEPI	SEP2	SHASHo	SHASH62	JSU	JSUo	ST1	ST2	ST5	SNI	NO	LO	GU	RG			
HP 10																	
2010	sec	72.65	1.2	1.45	0.28	> 200	0.83	0.97	1.04	0.84	0.7	0.06	0.07	0.08	0.06		
	KS	0.0503	0.0624	0.1972	0.1573	0.0276	0.0505	0.0444	0.0397	0.270	0.0337	0.0337	0.1514	0.1293	0.0772		
	CVM	0.2593	0.5877	4.3911	2.8226	0.0471	0.4333	0.1050	0.0961	5.9571	0.0850	0.0849	3.1086	2.1320	0.7829		
	AD			23.4270	16.0225	0.4089	2.9392	0.9735	0.8397	37.4577	0.7282	0.7279	20.1967		5.3250		
	AIC	3211	3214	2803	2776	2635	2671	2640	2639	3004	2642	2640	2777	2779	2702		
2011	sec	69.39	er	5.24	1.64	3.27	0.42	1.5	1.64	0.52	0.94	0.1	0.08	0.08	0.09		
	KS	0.4043	0.1124		0.0522	0.0440	0.0201	0.0204	0.2755	0.0716	0.0765	0.1750	0.0968	0.2797			
	CVM	31.5557	1.7880		0.2612	0.1387	0.0172	0.0207	9.4574	0.4059	0.5691	3.8703	1.0218	11.2688			
	AD				1.5750	0.8963	0.1391	0.1688	53.9736	2.5255	3.4554	23.9321	6.3186	52.3735			
	AIC		>>	4662	2565	2548	2549	3080	2601	2600	2731	2585	2978				
2012	sec	19.29	er	0.89	0.4	7.26	0.92	0.68	1.21	0.46	1.06	0.06	0.08	er	0.08		
	KS	0.8253	0.3315	0.1032	0.0879	0.0592	0.0630	0.3324	0.0463	0.1069	0.1921	0.4425	0.1476				
	CVM	102.1823	11.4260	0.9149	0.6272	0.4892	0.2474	3.2396	14.3101	0.1587	1.0182	5.2631	27.0913	1.9506			
	AD			54.1023	5.8183	4.1656	1.7444	3.4890	75.2884	1.8496	6.6542	30.6382		11.6370			
	AIC		>>	3130	2853	2851	2809	3535	2465	2885	3006	3531	2882				
2013	sec	er	0.69	0.75	0.49	er	1.07	0.8	0.84	0.73	0.61	0.06	0.08	0.08	0.08		
	KS	0.0694	0.0695	0.1314	0.0936	0.0681	0.0784	0.0851	0.0785	0.1802	0.0574	0.0575	0.1410	0.0783	0.0812		
	CVM	0.2513	0.2500	2.0387	0.7047	0.3248	0.6720	0.2684	0.2479	3.0853	0.1752	2.3456	0.5864	0.5423			
	AD	2.1025	2.0998	10.0898	3.3673	1.9644	5.2252	3.0222	2.7309	22.7405	1.1668	1.1670	17.0352	4.7692	3.9262		
	AIC	2865	2865	2866	2863	2932	2905	2908	3127	2856	2854	2996	2925	2885			
2014	sec	er	er	er	18.91	0.87	er	1.04	0.73	0.8	1.04	0.63	0.06	0.08	0.07		
	KS	0.0450	0.0448	0.2035	0.1582	0.0343	0.0638	0.0624	0.0559	0.1987	0.0321	0.0321	0.1464	0.0911	0.0767		
	CVM	0.1297	0.1362	4.5719	2.7480	0.1094	0.4892	0.2474	3.2396	3.6495	0.0597	0.0596	2.4997	0.7502	0.4925		
	AD	2.4311	2.4801	22.4636	13.9794	0.7751	3.2381	2.3508	2.2261	25.6064	0.6342	0.6387	17.4354	5.7328	3.7391		
	AIC	2678	2678	2740	2632	2676	2660	2665	2914	2631	2629	2767	2696	2679			
2015	sec	er	er	er	0.47	0.47	er	1.01	0.8	0.84	0.43	0.69	0.06	0.08	0.06		
	KS	0.0485	0.0466	0.1986	0.1642	0.0275	0.0824	0.0577	0.0411	0.2033	0.0337	0.1389	0.1002	0.0884			
	CVM	0.1563	0.1684	4.1766	2.5294	0.0336	0.6233	0.1552	0.1262	4.2876	0.0543	0.0542	2.7466	0.9546	0.7306		
	AD	3.2949	3.3616	20.2099	12.2884	0.3132	4.0804	1.6463	1.4172	28.9195	0.4415	0.4414	18.4196	6.1122	4.7451		
	AIC	2777	2777	2788	2756	2678	2739	2702	2978	2679	2677	2817	2741	2749			
2016	sec	21.41	14.03	5.21	er	93.51	1.26	0.76	0.83	0.54	0.89	0.08	0.08	0.12	0.07		
	KS	0.2056	0.2695	0.1228	0.1585	0.0846	0.0748	0.0562	0.0552	0.2436	0.1141	0.1137	0.1715	0.2371	0.0999		
	CVM	5.6213	10.1727	1.7080	3.4811	0.5761	0.4886	0.2078	0.1857	7.9964	0.1380	0.1322	4.1668	6.5916	1.3198		
	AD			9.6122	17.7386	2.9205	2.5730	1.0314	0.9016	46.9116	5.4246	5.4066	24.6907	31.2461	7.4699		
	AIC	4117	4161	2929	2849	2732	2733	2714	2712	3141	2763	2761	2895	2993	2757		
HP 11																	
2010	sec	47.18	70.18	0.47	0.32	1.06	1.28	0.95	1.17	0.95	0.71	0.06	0.06	0.08	0.07		
	KS	0.0894	0.1093	0.3459	0.1564	0.0556	0.1190	0.0545	0.0499	0.2348	0.0631	0.0632	0.1561	0.1697	0.0970		
	CVM	0.7792	1.8038	13.6504	3.2676	0.1747	1.2615	0.2249	0.2082	6.9006	0.2017	0.2019	3.3967	3.1287	1.0141		
	AD			65.2800	18.7759	1.1430	8.2509	1.4677	1.3490	41.9763	1.4909	1.4918	21.5030		6.2869		
	AIC	3866	3877	2950	2775	2609	2696	2607	2606	3007	2619	2617	2752	2789	2678		
2011	sec	> 100	er	0.58	0.38	8.89	0.48	2.34	2.56	0.36	0.86	0.07	0.06	0.08	0.08		
	KS	0.3214	0.1144	0.1895	0.0555	0.0454	0.0249	0.0219	0.2691	0.0753	0.0790	0.1645	0.0944	0.2563			
	CVM	19.9330	1.8311	4.5118	0.3062	0.2003	0.0304	0.0255	8.7766	0.5351	0.6834	3.9525	1.0958	8.5834			
	AD			11.0602	22.7053	1.7442	1.1878	0.2199	0.2098	50.7722	3.1869	4.0066	24.1639	6.5356	40.4594		
	AIC	7004	7004	2688	2513	2512	2495	2985	2549	2547	2679			2535			
2012	sec	7.03	56.19	7.05	0.52	0.78	0.84	0.79	1.15	0.52	1.05	0.06	0.08	er	0.06		
	KS	0.8211	0.2055	0.0411	0.1237	0.0382	0.0369	0.3149	0.0463	0.0806	0.1974	0.4146	0.1221				
	CVM	106.2392	9.5626	0.4117	1.6358	0.1578	0.1451	12.3665	0.1587	0.4892	4.2141	24.8778	1.2953				
	AD			>>	3085	2837	2947	2811	2811	3478	2465	2867	2994	3505	2858		
	AIC																
2013	sec	54.18	94.15	0.53	0.42	> 120	0.93	1.24	1.89	0.5	0.61	0.06	0.07	0.07	0.06		
	KS	0.0498	0.0498	0.1316	0.0746	0.0658	0.0834	0.0730	0.0924	0.1702	0.0445	0.0445	0.1402	0.0881	0.0732		
	CVM	0.2400	0.2396	2.2357	0.6379	0.4323	0.6550	0.4405	0.3663	2.4872	0.1999	0.1998	2.1463	0.4465	0.3824		
	AD	1.6247	1.6236	11.2423	2.9485	2.5933	4.8432	4.0809	3.5991	19.4797	1.2644	1.2644	16.2537	3.9384	3.3046		
	AIC	2857	2857	2910	2862	2876	2928	2923	3115	2862	2860	3004	2916	2892			
2014	sec	er	er	er	5.07	1.36	er	1.12	0.67	0.87	1.47	0.67	0.07	0.08	0.06		
	KS	0.0605	0.0615	0.2095	0.1651	0.0491	0.0671	0.0603	0.0547	0.1995	0.0424	0.0424	0.1450	0.0903	0.0741		
	CVM	0.1937	0.2011	4.8993	3.0166	0.1615	0.4814	0.3034	0.3060	3.7196	0.0962	0.0961	0.2493	0.7861	0.5029		
	AD	2.5966	2.6573	24.2718	15.3131	1.0415	3.1903	2.5853	2.5353	26.0167	0.8269	0.8264	17.4978	5.8157	3.7742		
	AIC	2688	2688	2758	2718	2635	2680	2663	2924	2635	2633	2771	2704	2684			
2015	sec	er	er	er	0.8	0.36	er	0.73	1.09	1.07	0.78	0.97	0.07	0.08	0.08		
	KS	0.0543	0.0609	0.1755	0.1448	0.0379	0.0727	0.0498	0.0454	0.1960	0.0519	0.0518	0.1447	0.1033	0.0718		
	CVM	0.2200	0.2575	3.9107	2.2096	4.5295	0.5151	0.4365	0.1012	0.0780	9.6160	0.6346	0.7836	4.3122	0.2634	0.4659	
	AD	2.5256	2.6901	18.9831	10.7169	0.1731	0.2723	0.1531	0.2723	54.5331	3.7518	4.5758	17.7572	5.8200	3.6920		
	AIC	2729	2729	2774	2703	2524	2527	2496	2492	3002	2549	2547	2678	2564	2861		
2016	sec	> 100	71.61	1.09	0.97	23.82	1.25	1.09	1.3	0.7	0.9	0.07	0.08	0.12	0.1		
	KS	0.1638	0.2209	0.1329	0.1484	0.0707	0.0695	0.0441	0.0402	0.2424	0.1030	0.1030	0.1753	0.2182	0.0944		
	CVM	3.5903	6.6320	2.3704	2.6339	0.3718	0.3054	0.0879	0.0743	6.9191	0.8783	0.8786	3.8793	5.3784			

	SEPI	SEP2	SHASHO	SHASH62	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG
HP 13														
2010	sec	6.83	88.29	0.84	0.54	40.02	1.82	1.53	0.88	0.5	0.61	0.07	0.07	0.07
	KS	0.0476	0.0529	0.1691	0.1363	0.0406	0.0995	0.0668	0.050	0.2054	0.0450	0.0450	0.1564	0.1173
	CVM	0.1977	0.2704	3.6881	2.4433	0.1258	0.9880	0.2815	0.1176	0.554	0.132	0.1352	3.1596	1.5799
	AD			19.4998	13.4850	0.8423	6.8207	2.1577	1.2938	35.3496	1.0417	1.0416	20.3154	9.6731
	AIC	2907	2908	2696	2671	2562	2638	2575	2572	2905	2568	2566	2703	2623
2011	sec	26.29	0.67	0.42	0.34	24.44	0.78	0.89	0.95	0.42	1.04	0.08	0.08	0.06
	KS	0.3637	0.1109	0.2019	0.0740	0.0681	0.0468	0.0517	0.2681	0.0999	0.0985	0.1687	0.0998	0.2384
	CVM	23.2394	1.7751	5.1405	0.3558	0.2308	0.1344	0.1446	8.8388	0.6169	0.7356	3.8633	1.1413	8.9291
	AD			22.2215	19.1792	2.0987	8.5340	2.4539	1.6086	49.2535	1.8019	1.8005	21.3379	5.7886
	AIC	>>	>>	2934	2939	2793	2903	2791	2799	3294	2811	2809	2944	3155
2012	sec	er	6.15	0.7	> 120	0.94	0.67	0.92	1.61	1.04	0.08	0.08	0.11	0.07
	KS	0.2558	0.2929	0.2085	0.1928	0.0729	0.1064	0.0728	0.0475	0.2684	0.0483	0.0485	0.1526	0.2367
	CVM	11.3868	18.4062	4.1662	3.4001	0.3421	1.1210	0.3927	0.1758	8.3866	0.2141	0.2139	3.1945	7.8670
	AD			38.0211	13.9006	1.2298	6.8236	1.0033	1.2506	42.6400	1.3162	1.3145	19.5911	7.4999
	AIC	>>	>>	2934	2939	2793	2903	2791	2942	2911	3170	2859	2857	2999
2013	sec	er	8.99	1.17	0.49	er	0.46	0.47	1.04	0.64	0.66	0.06	0.06	0.11
	KS	0.0791	0.0797	0.0953	0.0514	0.0692	0.0717	0.0796	0.0739	0.1925	0.0573	0.0573	0.1411	0.0799
	CVM	0.4537	0.4543	1.1431	0.3268	0.4594	0.5052	0.4448	0.4396	3.4207	0.2872	0.2873	2.1734	0.4130
	AD			6.5695	2.3493	2.9543	3.8621	3.7299	24.9515	1.9632	1.9634	16.8207	3.8236	
	AIC	2884	2884	2889	2871	2864	2901	2911	3170	2859	2857	2999	2884	
2014	sec	er	7.1	0.67	1	5.76	1.51	1.03	0.86	0.64	0.81	0.06	0.06	0.1
	KS	0.1140	0.1275	0.2818	0.1351	0.0504	0.0962	0.0466	0.0472	0.2599	0.0486	0.0485	0.1594	0.1022
	CVM	1.1636	2.7071	7.4497	2.1421	0.1529	0.8163	0.1328	0.1965	7.0104	0.1577	0.1571	2.8760	3.3831
	AD			38.0211	13.9006	1.2298	6.8236	1.0033	1.2506	42.6400	1.3162	1.3145	19.5911	20.5136
	AIC	5718	5740	2859	2788	2738	2645	2642	2848	2654	2652	2786	2718	
2015	sec	35.32	3.49	0.59	0.36	7.64	0.68	1.19	0.57	0.76	0.09	0.09	0.07	0.1
	KS	0.0040	0.0684	0.1555	0.1188	0.0655	0.063	0.0530	0.0455	0.1936	0.0614	0.0614	0.1591	0.0983
	CVM	0.3164	0.3631	3.4355	1.7063	0.2423	0.5344	0.2064	0.2536	4.0011	0.2493	0.2493	2.6486	0.9956
	AD			3.5021	3.6988	14.8245	8.5011	1.5576	4.6962	1.9343	27.6365	1.6322	18.3446	6.4368
	AIC	2731	2731	2738	2714	2671	2692	2699	2700	2670	2668	2809	2720	
2016	sec	er	62.73	1.36	0.76	> 130	0.76	1.59	1.59	0.81	1.11	0.06	0.21	0.11
	KS		0.9320	0.3914	0.1009	0.1276	0.0706	0.0925	0.3598	0.0998	0.0999	0.2128	0.1590	0.7479
	CVM		> 100	14.0623	1.2878	1.5703	0.5653	0.9776	16.9905	1.1696	1.1707	4.8637	3.2644	77.0552
	AD			66.4082	7.4275	9.5616	3.3455	5.8965	86.6441	6.8532	6.8593	28.6817	17.8404	441.0596
	AIC		>>	3237	2837	2893	2800	2803	3674	2848	2846	2968	2928	
HP 14														
2010	sec	11.48	er	2.04	0.92	er	0.67	1.17	1.01	0.94	er	0.07	0.07	0.08
	KS	0.0469	0.0468	0.3801	0.1736	0.0322	0.0654	0.0567	0.0637	0.2290	0.0333	0.0333	0.1501	0.0916
	CVM	0.1199	0.1446	15.7608	3.3779	0.0520	0.5785	0.1470	0.3438	6.2837	0.0701	0.0701	3.1008	1.1523
	AD			7.3046	78.1265	18.7133	0.5045	4.2966	1.3726	3.1374	39.0968	0.6773	0.6773	7.6918
	AIC	3454	3454	3165	2733	2585	2633	2593	2604	2970	2592	2590	2726	
2011	sec	4.73	58.67	0.43	0.88	> 100	0.73	0.91	0.96	0.47	0.7	0.06	0.08	0.14
	KS	0.2049	0.2558	0.1652	0.1713	0.0440	0.0362	0.0470	0.0541	0.2537	0.0714	0.0758	0.1762	0.0963
	CVM	7.3313	12.0170	3.4298	3.6409	0.1472	0.0870	0.1050	0.1427	7.3115	0.4936	0.4936	3.4803	6.2168
	AD			17.0496	18.1984	0.9203	0.5755	0.6029	0.8234	44.0862	3.1063	3.4409	22.3179	5.5917
	AIC	3686	3737	2574	2586	2487	2485	2487	2932	2530	2529	2663		
2012	sec	er	17.58	1.25	0.89	er	0.74	1.08	1.48	2.2	0.98	0.06	0.13	0.08
	KS	0.2539	0.2638	0.3886	0.1916	0.0636	0.1062	0.0695	0.0411	0.2735	0.0498	0.0498	0.1591	0.2212
	CVM	9.6495	18.8161	17.2347	2.9999	0.3599	1.0881	0.4524	0.4440	8.0448	0.2224	0.2533	3.0442	7.3351
	AD			> 200	79.107	17.5300	2.9795	8.5423	2.9240	2.0182	47.5420	2.0137	2.0137	20.3042
	AIC	8692	8765	3066	2913	2790	2882	2776	2776	2788	2786	2786	3124	
2013	sec	er	0.37	0.44	> 110	0.37	er	0.78	0.57	0.68	0.08	0.08	0.09	0.07
	KS	0.0693	0.0686	0.1388	0.0813	0.0785	0.0918	0.0800	0.0915	0.1974	0.0658	0.0658	0.1537	0.1011
	CVM	0.6348	0.6338	2.7305	0.8834	0.6922	0.5688	0.5931	0.7064	3.4743	0.4939	0.4939	2.2371	0.5434
	AD			3.9993	3.9959	0.16	5.5782	4.0108	4.5536	4.9931	> 25	3.0343	3.0342	4.5066
	AIC	2886	2886	2968	2908	2908	2907	2941	2907	3180	2915	2915	3004	
2014	sec	er	er	er	0.94	er	1.12	1.39	1.23	0.64	1.01	0.06	0.07	er
	KS	0.9059	0.3955	0.0880	0.1159	0.0701	0.0701	0.0700	0.3652	0.0624	0.0863	0.1957	0.1442	0.6563
	CVM		> 90	15.1828	0.4071	0.8924	0.3037	0.2940	16.1546	0.2182	0.3714	0.4036	2.4065	
	AD			70.6951	3.3690	2.9235	7.2238	2.1097	2.0460	82.8453	1.8748	3.2967	25.8534	
	AIC		>>	3088	2721	2800	2688	3169	2748	2744	2862	2747	3980	
2015	sec	11.87	21.8	0.78	0.43	64.52	0.94	1.27	1.07	0.66	0.99	0.08	0.06	er
	KS	0.8953	0.3917	0.0824	0.1149	0.0821	0.0807	0.0807	0.3651	0.0677	0.0820	0.1883	0.1359	0.7222
	CVM	98.9341	15.2374	0.4288	0.7811	0.4496	0.4336	15.9531	0.3262	0.3646	3.6455	1.1225	73.9154	
	AD			70.9565	2.8799	6.1226	2.5599	2.4933	82.3336	0.2936	2.6719	23.8690	7.7545	
	AIC		>>	3102	2717	2696	2697	3612	2739	2738	2864	2740	4263	
2016	sec	er	er	er	0.97	0.59	1.29	1.53	0.78	1.09	0.1	0.07	er	
	KS	0.6742	0.4614	0.1288	0.1557	0.1070	0.1067	0.3856	0.1101	0.1304	0.2347	0.1694	0.8533	
	CVM	45.0243	45.16203	19.7371	0.4480	0.7007	0.7650	0.7781	17.5609	0.6594	1.5155	5.5892	96.0404	
	AD			89.0716	9.3613	12.2560	6.9967	7.3276	96.7877	7.2012	8.9807	32.1070	> 600	
	AIC		>>	3372	2908	2971	2867	3876	2925	2925	3043	2966	4974	
HP 15														
2010	sec	52.89	31.17	1.61	0.76	68.99	0.73	0.89	1.09	0.83	0.72	0.08	0.08	0.08
	KS	0.0564	0.0567	0.2174	0.1909	0.0933	0.0857	0.0635	0.0477	0.2200	0.0336	0.0387	0.1448	0.1132
	CVM	0.1792	0.2538	5.6815	4.4362	0.0859	0.7697	0.4456	0.092					

	SEP1	SEP2	SHASHo	SHASHo2	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG		
HP 16																
2010	sec	26.08	35.24	1.45	0.71	7.05	0.83	1.03	1.12	1.18	0.83	0.07	0.06	0.07	0.09	
	KS	0.0668	0.0852	0.3390	0.2125	0.0518	0.0092	0.0446	0.0399	0.0298	0.0098	0.0098	0.1451	0.1457	0.1178	
	CVM	0.3359	0.5392	13.2991	5.3241	0.1896	0.7842	0.1660	0.1238	6.6722	0.2548	0.2546	3.5183	2.2249	1.3379	
	AD			64.1514	26.5369	1.1716	4.7793	1.2043	0.9323	40.7208	1.631	1.6346	21.8255	12.4152	7.7766	
	AIC	3503	3505	3020	2805	2639	2687	2636	2632	3009	2647	2645	2780	2763	2734	
2011	sec	> 120	> 260	1.41	0.97	er	0.79	0.9	1.05	1.06	0.88	1.06	0.06	0.09	0.06	
	KS	0.2037	0.2469	0.1146	0.1905	0.0459	0.0407	0.0409	0.0457	0.2526	0.0460	0.0538	0.1520	0.1054	0.2160	
	CVM	8.0548	13.7828	1.0433	3.4378	0.1075	0.0906	0.0806	0.0941	7.8142	0.1885	0.2437	3.4422	0.7457	6.9522	
	AD			7.0857	17.8594	0.8300	0.7077	0.5163	0.5594	46.4573	1.5668	1.9235	21.9901	5.0965	33.6018	
	AIC	5375	5436	2711	2638	2515	2511	2513	2513	2978	2544	2542	2676	2538	2830	
2012	sec	er	er	0.7	1.14	2.87	1.03	0.92	1.06	1.28	0.78	0.08	0.08	0.13	0.08	
	KS	0.2294	0.2749	0.5711	0.1748	0.0401	0.0956	0.0420	0.0435	0.2709	0.0359	0.0359	0.1485	0.2401	0.0871	
	CVM	10.4634	16.9921	55.3655	3.0380	0.1188	1.158	0.1551	0.1478	8.3062	0.0970	0.0973	3.1300	8.5582	0.7069	
	AD				17.8697	0.8556	7.9771	1.1301	1.0998	48.8683	0.9010	0.9028	20.7621	4.9850		
	AIC	>>		3540	2875	2712	2817	2708	2710	3206	2726	2724	2859	3090	2753	
2013	sec	3.08	er	0.6	0.52	er	1.13	0.92	1	0.85	1.47	0.06	0.08	0.08	er	
	KS			0.6655	0.4377	0.0705	0.0923	0.0795	0.0793	0.3938	0.0765	0.0678	0.1815	0.1366	0.8518	
	CVM		> 40	19.8103	0.3563	0.5139	0.5957	0.6008	18.1028	0.5040	0.3474	3.3001	0.8419	> 90		
	AD			> 90	3.0675	5.2609	3.5571	3.6012	91.3107	3.1781	3.0886	23.2975	7.0456	> 700		
	AIC		>>	3363	2896	2869	2868	2869	4021	2917	2919	3042	2901	5100		
HP 17																
2010	sec	13.73	7.52	1.59	0.71	1.84	0.81	1.26	1.23	1.25	1.03	0.09	0.08	0.08	0.12	
	KS	0.1487	0.2127	0.2733	0.1425	0.0800	0.0728	0.0534	0.0500	0.2465	0.0976	0.0975	0.1609	0.2290	0.0921	
	CVM	2.6714	5.9413	10.1330	3.2979	0.4949	0.6013	0.1751	0.1397	7.9830	0.7424	0.7418	4.1570	5.5665	1.2317	
	AD			54.0124	18.3030	2.6921	3.3510	0.9089	0.6989	46.8011	4.1611	4.1599	24.4887	26.5762	7.3253	
	AIC	4534	4561	3133	2875	2722	2736	2703	2699	3122	2741	2739	2846	2771		
2011	sec	5.18	33.96	1.38	0.92	> 200	1.22	0.81	0.96	0.82	0.83	0.06	0.08	0.06	er	
	KS			0.4079	0.3539	0.0778	0.1190	0.0701	0.0788	0.097	0.0611	0.0767	0.1840	0.1336	0.7005	
	CVM		16.7883	11.2401	24.9856	3.4952	0.1413	1.1963	0.1042	8.7976	0.1645	0.1650	3.5664	1.0741	6.9960	
	AD			113.2372	20.8130	1.2370	8.2489	0.9291	0.7728	51.0297	1.4799	1.4828	22.5877	7.0623	33.8320	
	AIC		>>	3100	2724	2808	2701	2702	3612	3014	2539	2537	2598	2844		
2012	sec	> 100	11.1	1.41	0.86	8.63	0.65	1.07	1	0.62	1.55	0.08	0.08	0.11	er	
	KS			0.5527	0.3569	0.1184	0.1473	0.0808	0.0779	0.3470	0.1148	0.1148	0.2211	0.1851	0.6564	
	CVM		51.6103	11.2212	1.9110	2.1750	0.7449	0.6605	15.6888	1.8262	1.8269	5.2714	4.3196	58.3558		
	AD			54.5586	10.5503	12.4839	4.9778	4.6054	81.1302	10.1560	10.1586	30.1652	22.3122	> 290		
	AIC		>>	3264	2914	2969	2868	2849	3630	2921	2919	3042	3012	4113		
HP 18																
2010	sec	30.34	82.29	0.93	0.49	3.02	1.48	1.24	1.25	0.89	0.76	0.07	0.08	0.13	0.08	
	KS			0.6075	0.6785	0.0415	0.1159	0.0401	0.0433	0.0433	0.3677	0.0909	0.1291	0.1943	0.3709	0.1564
	CVM		60.9070	72.7608	11.8176	1.4958	0.9832	0.1354	0.1313	11.9370	0.7604	1.9234	5.2619	18.2202	2.7194	
	AD			53.7287	7.8142	5.4161	1.3861	1.3636	64.8716	6.3395	10.5619	30.1565	13.6732			
	AIC		>>	3111	2923	2916	2869	2869	2966	2977	3104	3446				
2011	sec	58.84	18.08	0.89	0.8	1.41	0.53	1	1.36	0.4	0.98	0.07	0.06	0.08	0.08	
	KS			0.7784	0.2254	0.0936	0.0922	0.0520	0.0483	0.2952	0.1026	0.1026	0.1796	0.2985	0.1374	
	CVM		94.8822	6.6623	0.6910	0.8427	0.1347	0.1106	11.1984	0.8389	0.8386	4.7271	10.7375	2.4279		
	AD			21.0507	12.7069	4.5096	5.3832	0.9653	0.7940	61.6710	5.5153	5.5145	18.9287	5.0202	9.0501	
	AIC	2995	2996	2848	2812	2695	2785	2715	2716	3060	2699	2697	2836	2749	2830	
2016	sec	86.35	er	1.51	1.08	3.37	0.79	1.23	1.46	0.96	1.28	0.09	0.08	0.11	er	
	KS			0.6111	0.9213	0.3128	0.1332	0.1516	0.0811	0.0806	0.3420	0.1337	0.1338	0.2329	0.2151	0.5287
	CVM		69.1100	114.7526	8.3838	2.1091	2.4203	0.5974	0.5852	14.3811	2.1343	2.1360	5.5257	5.5208	35.9622	
	AD			43.2982	11.7663	13.9244	4.5243	4.3743	75.5322	11.9330	11.9378	31.2674	27.5555	165.3423		
	AIC		>>	3258	2949	3002	2886	2867	3572	2956	2954	3080	3085	3785		
HP 18																
2010	sec	30.34	82.29	0.93	0.49	3.02	1.48	1.24	1.25	0.89	0.76	0.07	0.08	0.13	0.08	
	KS			0.6075	0.6785	0.0415	0.1159	0.0401	0.0433	0.0433	0.3677	0.0909	0.1291	0.1943	0.3709	0.1564
	CVM		60.9070	72.7608	11.8176	1.4958	0.9832	0.1354	0.1313	11.9370	0.7604	1.9234	5.2619	18.2202	2.7194	
	AD			53.7287	7.8142	5.4161	1.3861	1.3636	64.8716	6.3395	10.5619	30.1565	13.6732			
	AIC		>>	3111	2923	2916	2869	2869	2966	2977	3104	3446				
2012	sec	58.84	er	0.68	0.56	5.79	1.22	1.16	1.48	0.52	0.88	0.06	0.06	0.11	0.08	
	KS			0.8508	0.2260	0.1234	0.1053	0.0393	0.0390	0.3558	0.1317	0.1313	0.1894	0.5210	0.1654	
	CVM		100.1575	10.7359	1.4989	0.9502	0.1019	0.0999	15.3785	1.7309	1.7232	5.6522	40.3765	2.9302		
	AD				8.6239	5.6756	1.2009	1.1968	79.8263	10.2231	10.2004	32.5393	15.2236			
	AIC		>>	2939	2933	2707	2871	2867	3667	2988	3106	3848				
2013	sec	0.1937	er	0.53	0.29	er	1.25	1	0.95	0.91	0.73	0.08	0.06	0.09	0.08	
	KS			0.2226	0.1765	0.0599	0.0548	0.0710	0.0854	0.2622	0.0777	0.0776	0.1872	0.2184	0.1185	
	CVM		5.6999	9.5769	2.8433	0.2702	0.2374	0.3906	0.5433	6.6811	0.5296	0.5272	3.5252	5.7395	1.0656	
	AD			14.4337	1.9006	1.8049	2.2718	2.9301	41.0454	4.4106	4.4016	22.6280	29.3912	6.6990		
	AIC	3633	3676	3044	2987	2784	2994	2999	3419	3039	3037	3173	3262	3007		
2014	sec	10	42.53	1.37	0.59	1.16	1.32	0.65	0.8	0.81	0.84	0.08	0.06	0.08	0.08	
	KS			0.1129	0.1498	0.1103	0.0413	0.0393	0.0458	0.0519	0.2238	0.0768	0.0767	0.1689	0.1685	0.0723
	CVM		1.7153	3.0067	1.2967	0.1209	0.0877	0.0799	0.1034	5.4600	0.5091	0.5078	3.0654	3.5003		
	AD			8.1955	2.1965	0.1728	0.6681	0.6319	0.7755	34.8235</						

	SEPI	SEP2	SHASHO	SHASH62	JSU	JSUo	ST1	ST2	ST5	SNI	NO	LO	GU	RG
HP 19														
2010	sec	4.12	3.29	1.38	>190	1.22	1.89	1.86	2.06	1.69	0.08	0.11	0.1	0.9
	KS	0.3882	0.4243	0.4364	0.2200	0.0935	0.0913	0.0833	0.0843	0.2798	0.1131	0.1121	0.1800	0.3024
	CVM	24.2719	30.8214	32.9072	6.8064	0.8348	0.7045	0.4322	0.4441	8.8725	1.3449	1.3449	3.8118	12.1343
	AD				4.1951	3.6783	3.2392	3.3112	51.3539	7.2872	7.2873	23.8887		1.5636
	AIC	7947	8044	4825	2975	2838	2835	2820	3336	2890	2888	3023	3288	2849
2011	sec	33.75	2.77	0.69	0.45	16.69	0.82	1.11	1.4	0.66	1.51	0.08	0.08	0.17
	KS				0.2055	0.2599	0.1005	0.0881	0.0472	0.0461	0.3094	0.1104	0.1103	0.1911
	CVM				4.4876	6.9619	0.9103	0.6690	0.1364	0.1315	11.7099	1.1723	1.1712	5.2932
	AD				25.7360	36.4192	6.4129	5.0856	1.3596	1.3219	63.9166	8.4989	8.4947	13.5056
	AIC				3106	2989	2747	2747	2692	2689	2782	2780	2904	1.8172
2012	sec	31.08	er	2.29	4.32	0.37	3.29	0.77	0.8	0.59	0.78	0.08	0.07	er 0.07
	KS				0.8998	0.4726	0.1556	0.1361	0.0445	0.1229	0.4054	0.1582	0.1583	0.2362
	CVM				92.4132	27.8655	3.1640	2.3213	0.1395	2.2142	20.0709	3.3302	3.3313	7.9301
	AD					19.0161	14.5728	2.0296	15.0650	99.3397	20.1120	20.1450	43.6178	77.4136
	AIC				>>	3532	3126	3119	2975	3104	4045	3174	3172	28.5466
2013	sec	6.79	1.27	0.56	0.36	er	1.52	0.82	1.03	0.97	0.86	0.06	0.06	0.08 0.06
	KS	0.1276	0.1580	0.1270	0.1556	0.0422	0.0456	0.0527	0.0689	0.2389	0.0654	0.0653	0.1805	0.1798
	CVM	2.8026	4.5604	1.3930	2.4212	0.1277	0.1254	0.2345	0.3897	5.4239	0.5087	0.5066	3.2350	4.2900
	AD				7.1119	11.8717	1.0095	1.0586	1.5678	2.2566	34.8530	3.9093	3.9017	21.0950
	AIC	3240	3262	3034	3006	3006	3024	3029	3386	3057	3055	3193	3234	3030
2014	sec	1.92	1.1	0.69	0.58	>100	1.13	er	0.86	0.86	0.77	0.06	0.08	0.09 0.06
	KS	0.1198	0.1339	0.0677	0.1031	0.0403	0.0389	0.0547	0.0614	0.1990	0.0933	0.0931	0.1747	0.1661
	CVM	1.5650	2.0688	0.4708	0.9853	0.1389	0.1180	0.1408	0.1758	4.0186	0.8240	0.8213	2.9260	2.9526
	AD	8.8296	10.9298	2.7455	4.6130	0.9065	0.7461	1.1235	3.1355	4.5463	4.5379	19.3902	15.5070	4.3699
	AIC	2767	2775	2731	2729	2716	2734	2737	3016	2752	2750	2888	2871	2736
2015	sec	sec	er	er	0.52	0.63	1.01	1.01	1.03	0.99	0.09	0.09	0.08	0.06
	KS	0.0781	0.0986	0.0809	0.0907	0.0246	0.0451	0.0355	0.0319	0.2060	0.0543	0.0542	0.1394	0.1417
	CVM	0.7294	1.2712	0.7042	0.7352	0.0448	0.1037	0.0667	0.0770	4.6561	0.1996	0.1985	2.8368	2.4048
	AD			4.0903	4.1841	0.3118	0.6785	0.8233	1.2457	1.2419	19.0026	12.9506	3.2598	
	AIC	2890	2898	2789	2789	2753	2756	2770	3083	2771	2769	2908	2907	2773
2016	sec	8.7	0.74	2.03	1.23	2.4	0.83	0.98	1.1	0.74	0.78	0.07	0.06	0.14 0.08
	KS	0.2441	0.3021	0.1895	0.2103	0.0959	0.0791	0.0420	0.0422	0.2570	0.1248	0.1247	0.1986	0.2589
	CVM	9.5332	14.4336	4.1696	5.3512	0.8863	0.6142	0.1297	0.1268	7.8884	1.6219	4.3749	7.8199	1.9202
	AD			19.4422	24.5948	4.3899	3.3401	1.0286	1.0078	46.4725	8.8245	8.8187	25.9236	37.1815
	AIC	3543	3595	2837	2851	2758	2755	2734	3176	2817	2815	2950	3056	2776
HP 20														
2010	sec	69.87	1.73	0.96	0.49	93.09	1.11	0.86	0.88	0.98	0.81	0.06	0.08	0.08 0.07
	KS	0.1828	0.2138	0.1145	0.1461	0.0817	0.0772	0.0856	0.0872	0.2434	0.1199	0.1198	0.1971	0.1977
	CVM	4.0056	5.7762	1.6478	2.8323	0.4595	0.4478	0.4828	0.5052	5.8084	1.2232	1.2215	3.4325	4.8902
	AD	23.3879	8.4136	13.5645	2.4696	2.5189	3.1024	3.2309	36.8075	6.9085	6.9029	22.1631	24.3473	7.2449
	AIC	2895	2917	2731	2739	2686	2683	3067	2742	2740	2879	2927	2710	
2011	sec	3.77	0.47	0.42	2.14	1.01	1.08	1.23	1.31	0.75	0.07	0.06	er 0.06	
	KS	0.1676	0.1931	0.3967	0.1363	0.0404	0.0785	0.0389	0.0355	0.2595	0.0357	0.0355	0.1432	0.1945
	CVM	3.9768	7.6441	17.2557	3.0070	0.0641	0.7484	0.0859	0.1128	7.9465	0.0523	0.0523	2.8786	6.1166
	AD			83.9392	19.3687	0.5892	0.8609	0.9265	47.2847	0.5625	0.5626	19.8404	6.4185	
	AIC			>>	3239	2850	2652	2739	2652	3145	2659	2657	2793	2984
2012	sec	7.25	er	0.75	0.35	1.08	0.72	1.31	1.31	0.56	0.91	0.07	0.07	er 0.06
	KS	0.9631	0.3844	0.0883	0.0695	0.0425	0.0425	0.0677	0.0682	0.3625	0.0677	0.0679	0.2164	0.4246
	CVM		114.9456	19.0074	1.1838	1.7080	0.0944	0.7614	16.4758	0.3841	1.2119	5.3030	57.8714	2.3039
	AD				7.2443	4.7617	1.0183	5.7795	84.2222	8.3364	31.3748	13.5765		
	AIC				>>	3297	2949	2945	2897	2942	2990	3109	4176	19.9390
2013	sec	er	er	er	0.72	0.38	>220	1.31	1.51	1.39	1.47	0.09	0.14	0.12 0.13
	KS	0.1729	0.2133	0.1378	0.1668	0.0375	0.0281	0.0405	0.0484	0.2523	0.0530	0.0531	0.1661	0.2009
	CVM	5.2932	9.1589	1.6840	2.8958	0.0465	0.0403	0.1061	0.1617	6.7756	0.2765	3.1420	5.4750	0.6471
	AD			9.3134	14.9140	0.4440	0.3748	0.7802	1.0797	41.6062	2.1141	20.8635	4.5288	
	AIC	3957	4000	2988	3003	2911	2918	2923	3352	2950	2948	3084	3195	2931
2014	sec	sec	er	er	2.84	1.06	er	1.8	1.22	1.33	0.96	0.09	0.12	0.13 0.11
	KS	0.1149	0.1424	0.1058	0.0986	0.0453	0.0408	0.0377	0.0410	0.2110	0.0835	0.0834	0.1600	0.1712
	CVM	1.4037	2.3466	1.1505	1.0860	0.0925	0.0807	0.0595	0.0728	4.9523	0.5062	0.5046	3.0883	3.0594
	AD			6.9443	6.0291	0.4874	0.5280	0.4689	0.5465	32.1657	2.8543	19.9418	16.2640	3.9483
	AIC	2801	2814	2716	2703	2651	2655	2658	3066	2776	2767	2810	2814	2672
2015	sec	1.25	1.37	0.9	14.56	1.48	1.48	2.13	1.33	2.14	0.77	0.06	0.19	0.09 0.07
	KS	0.1171	0.1693	0.2106	0.1144	0.0407	0.0384	0.0301	0.0368	0.2282	0.0681	0.0682	0.1535	0.2017
	CVM	2.0319	4.0674	7.0514	1.7217	0.0767	0.0885	0.0445	0.0530	6.4055	0.2875	0.2883	3.0446	4.8772
	AD			40.4729	9.8743	0.4408	0.6503	0.4023	0.4432	39.6918	1.5230	19.9540	3.7765	
	AIC	3581	3603	3015	2803	2705	2710	2708	3105	2722	2720	2857	2958	2733
2016	sec	> 240	54.74	0.85	0.53	1.23	1.47	1.06	2.53	1.15	0.84	0.06	0.08	0.08 0.08
	KS	0.2052	0.2681	0.2195	0.1738	0.0850	0.0748	0.0389	0.0370	0.2498	0.1151	0.1152	0.1935	0.2479
	CVM	6.1670	10.5452	6.7896	3.4113	0.5704	0.4381	0.0999	0.0915	7.7567	1.1637	1.1643	4.1186	6.7753
	AD			37.9025	17.5083	3.0575	2.5033	0.7167	0.6637	45.8690	6.3091	6.3109	24.6440	31.8067
	AIC	3791	3834	2948	2679	2680	2709	2657	2655	3086	2715	2713	2848	2944
HP 21														
2010	sec	1.69	1.48	2.56	0.94	32.41	1.13	1.27	1.4	0.98	0.68	0.07	0.08	0.08 0.08
	KS	0.1499	0.1761	0.1017	0.1288	0.0513	0.0453	0.0307	0.0348	0.2219	0.0974	0.0973	0.1713	0.1898
	CVM	2.5694	4.1130	1.3465	1.9030	0.1514	0.1136	0.0393	0.0533	5.4381	0.8428	0.841		

	SEPI	SEP2	SHASHo	SHASHo2	JSU	JSUo	ST1	ST2	ST5	SN1	NO	LO	GU	RG
HP 22														
2010	sec	41.86	19.13	0.86	0.48	> 200	0.64	1.26	1.42	0.63	0.75	0.08	0.11	0.08
	KS	0.0895	0.1039	0.1228	0.1186	0.0487	0.1247	0.0535	0.0471	0.2143	0.0524	0.0585	0.1492	0.1631
	CVM	0.6681	1.3726	2.5111	1.5919	0.1991	1.3547	0.2814	0.1244	0.0187	3.2558	2.7250	0.9072	0.0899
	AD			14.1189	9.4448	1.1608	8.8061	1.9714	1.3859	37.5883	1.3050	1.3064	20.6752	14.6565
	AIC	2717	2726	2238	2313	2229	2329	2237	2243	2589	2239	2237	2374	2398
2011	sec	52.01	15.58	1.08	0.63	11.51	0.67	1.24	1.26	0.56	0.71	0.06	0.06	0.08
	KS	0.0838	0.1081	0.1267	0.1225	0.0561	0.1036	0.0590	0.0454	0.2263	0.0466	0.0464	0.1468	0.0837
	CVM	0.8363	1.9940	2.0909	1.8636	0.1655	1.1233	0.2616	0.0933	6.7247	0.1497	0.1497	3.1424	0.9517
	AD			12.5190	11.2326	0.9689	7.6385	1.9257	0.8151	41.2219	0.9483	0.9500	20.4056	6.3103
	AIC	3729	3744	2573	2571	2462	2558	2470	2465	2866	2470	2468	2604	2511
2012	sec	er	er	1.03	0.67	er	0.49	1.11	1.36	1.06	0.98	0.07	0.08	0.08
	KS	0.0803	0.1309	0.2288	0.2101	0.0483	0.0907	0.0431	0.0389	0.2624	0.0528	0.0527	0.1568	0.1992
	CVM	0.6605	2.2344	6.0640	5.1345	0.2387	1.0141	0.1377	0.1018	9.0976	0.2736	0.2735	3.7929	4.5710
	AD			32.5091	27.9379	1.6748	6.3693	0.9741	0.7495	52.3133	1.9642	1.9635	23.3896	15.5156
	AIC	>>	>>	2819	2808	2588	2648	2571	2565	3064	2597	2595	2727	2840
2013	sec	er	er	0.53	0.34	> 200	0.53	0.8	0.97	1.56	0.65	0.06	0.08	0.06
	KS	0.0464	0.0464	0.2051	0.1801	0.0323	0.0752	0.0599	0.0600	0.1916	0.0321	0.0321	0.1332	0.0830
	CVM	0.1322	0.1322	5.4370	3.4558	0.0378	0.6418	0.3171	0.3171	4.3632	0.0314	0.0314	2.6660	0.7894
	AD			27.8823	17.6619	0.2660	4.2548	3.3569	3.3569	29.4440	0.2359	0.2359	18.2219	5.1283
	AIC	2758	2758	2812	2764	2639	2699	2679	2691	2640	2638	2778	2712	2708
2014	sec	8.08	1.75	0.98	0.53	er	0.45	0.7	0.9	0.8	0.6	0.07	0.08	0.07
	KS	0.0824	0.1144	0.1090	0.0978	0.0269	0.0391	0.0353	0.0341	0.2107	0.0518	0.0513	0.1450	0.0833
	CVM	0.9348	1.9032	1.2057	1.1242	0.0302	0.1070	0.0483	0.0551	5.3619	0.1774	0.1866	2.9997	0.5045
	AD	10.9022		7.4035	6.8405	0.2524	0.8000	0.5166	0.5060	34.3166	1.1735	1.2352	19.5918	3.7535
	AIC	2585	2597	2356	2352	2288	2294	2295	2297	2305	2303	2303	2437	2312
2015	sec	6.03	1.75	0.89	0.53	er	0.66	1.12	1.17	0.48	0.63	0.07	0.06	0.07
	KS	0.0642	0.0902	0.1597	0.1243	0.0322	0.0554	0.0393	0.0533	0.2079	0.0503	0.0504	0.1490	0.0745
	CVM	0.4459	0.9634	2.8241	1.7470	0.0632	0.3581	0.0784	0.0572	5.4298	0.1824	0.1829	3.2113	0.6587
	AD	7.7992	10.0479	15.4738	10.1008	0.4250	2.2794	0.7558	0.5734	34.6471	1.2217	1.2238	20.3835	4.5647
	AIC	2780	2786	2624	2600	2505	2529	2510	2510	2839	2515	2515	2651	2548
2016	sec	9.61	21.6	1.32	0.5	> 200	0.74	0.91	1.51	0.65	0.72	0.08	0.06	0.08
	KS	0.1349	0.2049	0.3879	0.1421	0.0415	0.0517	0.0379	0.0345	0.2431	0.0538	0.0564	0.1618	0.0937
	CVM	3.7492	7.7812	16.7983	2.7201	0.1608	0.2316	0.0490	0.0306	7.8577	0.2600	0.2953	3.6182	0.8334
	AD			78.1542	15.8524	1.2447	1.8497	0.5471	0.3541	46.5602	1.7655	1.9790	22.5540	6.0479
	AIC	5080	5120	2820	2562	2428	2436	2417	2415	2863	2444	2442	2576	2468
HP 23														
2010	sec	10.13	6.24	0.49	0.26	er	0.6	1.28	1.01	0.41	0.62	0.07	0.06	0.08
	KS	0.0717	0.0829	0.1717	0.1378	0.0733	0.1298	0.0942	0.0452	0.2170	0.0787	0.0787	0.1841	0.1535
	CVM	0.4030	0.6529	3.4625	2.4252	0.2716	1.3021	0.3954	0.1348	6.2837	0.2961	0.2961	3.5252	2.2925
	AD			17.5490	12.9825	1.7332	8.3031	2.5878	1.2903	38.7752	1.9746	1.9753	21.6879	12.8638
	AIC	2752	2755	2320	2306	2202	2284	2211	2203	2555	2210	2208	2344	2259
2011	sec	6.52	1.22	0.76	0.53	> 100	0.42	0.96	1.39	0.5	0.97	0.06	0.07	0.08
	KS	0.3436	0.3982	0.0873	0.2151	0.0506	0.1003	0.0426	0.0375	0.2733	0.0436	0.0502	0.1550	0.0968
	CVM	23.6223	32.3389	0.3284	4.8234	0.2646	1.3031	0.1415	0.1061	10.0699	0.1912	0.2676	3.7548	11.6500
	AD			25.8272	15.4738	1.7193	8.4103	1.0400	0.8098	56.7560	1.3135	1.7742	23.3260	7.2841
	AIC	>>	>>	8167	2554	2363	2458	2347	2347	2379	2378	2378	2405	2795
2012	sec	er	er	0.84	0.39	> 100	0.58	1.27	1.33	0.73	0.84	0.06	0.06	0.06
	KS	0.1617	0.1543	0.3743	0.2028	0.0463	0.0709	0.0345	0.02453	0.0557	0.0557	0.0557	0.1632	0.0937
	CVM	1.6273	4.0354	16.5040	8.6680	0.2139	0.4273	0.0579	0.0446	7.7297	0.3755	0.3754	3.6958	0.9510
	AD			78.1569	41.3523	1.4296	2.8547	0.4128	0.3000	45.9747	2.4829	2.4829	22.9550	22.5110
	AIC	4779	4800	297	2756	2554	2560	2518	2518	2957	2551	2549	2683	2743
2013	sec	24.54	er	0.64	0.48	er	0.42	1.45	1.46	0.7	0.98	0.06	0.06	0.06
	KS	0.0748	0.0744	0.1544	0.1228	0.0435	0.0956	0.0654	0.0621	0.2052	0.0430	0.0430	0.1499	0.0917
	CVM	0.2653	0.3626	2.2901	1.5358	0.1422	1.1430	0.3012	0.2404	5.2990	0.1480	0.1481	3.1487	0.9911
	AD	8.3111	8.6993	11.5351	8.3462	1.0371	7.7056	2.5539	2.1312	33.9628	1.1658	20.0905	6.2098	10.3404
	AIC	2788	2790	2610	2602	2536	2553	2453	2453	2863	2542	2540	2678	2661
2014	sec	14.76	0.84	0.58	0.52	er	0.98	1.3	1.66	0.57	0.55	0.06	0.07	0.06
	KS	0.1428	0.2118	0.1651	0.1753	0.0573	0.0516	0.0386	0.0354	0.2408	0.0699	0.0756	0.1569	0.1105
	CVM	3.8279	7.7029	2.6683	2.8894	0.2422	0.2343	0.0951	0.0830	7.2268	0.4040	0.4863	3.7961	1.0074
	AD			14.4674	14.9949	1.5365	1.4652	0.5910	0.5170	43.1742	2.9494	3.3516	23.0000	6.0309
	AIC	3312	3349	2359	2360	2256	2257	2247	2247	2284	2282	2282	2280	2492
2015	sec	80.34	> 160	0.46	0.32	87.59	0.43	0.92	1.25	0.69	0.61	0.07	0.06	0.08
	KS	0.0558	0.0705	0.2381	0.2159	0.0482	0.0470	0.0413	0.0413	0.2086	0.0535	0.0535	0.1489	0.0954
	CVM	0.3201	0.6862	6.3442	5.0422	0.1408	0.7178	0.1597	0.1135	5.7256	0.2330	0.2329	3.4006	1.0204
	AD			29.9111	23.9733	0.8959	4.2146	1.2471	0.9324	35.9844	1.5330	1.5330	21.0481	6.1835
	AIC	2857	2860	2681	2654	2524	2522	2851	2527	2525	2525	2526	2588	2630
2016	sec	1.48	27.47	1.62	1.36	er	1.03	0.78	0.83	0.87	0.7	0.07	0.08	0.11
	KS	0.0874	0.1291	0.1851	0.1487	0.0502	0.0787	0.0353	0.0314	0.2210	0.0643	0.0643	0.1540	0.0912
	CVM	0.8192	2.0286	3.3427	2.3763	0.1948	0.4491	0.0976	0.0638	6.5602	0.3464	0.3466	3.5920	0.9031
	AD			18.1965	13.5403	1.2076	2.8294	0.7626	0.5016	40.1838	2.1165	2.1186	22.0654	5.9282
	AIC	3036	3047	2511	2494	2373	2395	2368	2366	2384	2384	2520	2422	2541
HP 24														
2010	sec	GT	er	0.58	0.87	0.84	0.65	0.76	1.05	0.47	0.84	0.07	0.06	0.08
	KS	0.0816	0.1214	0.3506	0.1192	0.0411	0.0482	0.0326	0.0329	0.2249	0.0572	0.0584	0.1522	0.0791
	CVM	0.9426	2.0527</td											

hours	KS					CVM					AD				
	JSU	JSU _o	ST1	ST2	ST5	JSU	JSU _o	ST1	ST2	ST5	JSU	JSU _o	ST1	ST2	ST5
1	0.0603	0.0822	0.0294	0.0291	0.4210	1.9100	4.3528	0.5453	0.5277	148.5070	12.8945	29.8425	3.9139	3.8630	727.8232
2	0.0794	0.0688	0.0260	0.0259	0.4584	4.2045	2.9519	0.3020	0.2985	176.7370	27.9584	19.9817	3.1511	3.1483	841.7994
3	0.0885	0.0620	0.0255	0.0255	0.4610	5.7829	2.1785	0.2036	0.2023	179.0372	38.1926	14.4810	2.8697	2.8709	851.0370
4	0.0808	0.0735	0.0184	0.0182	0.4619	5.5244	4.2863	0.1258	0.1224	180.2509	37.4213	28.9289	1.7491	1.7386	855.9105
5	0.0733	0.0639	0.0194	0.0190	0.4567	3.8933	2.6782	0.1071	0.1021	176.3345	26.8176	18.2258	1.2072	1.1912	840.1847
6	0.0645	0.0769	0.0170	0.0167	0.4600	2.7978	4.5500	0.1375	0.1369	178.3401	21.0471	33.4914	1.3620	1.3644	848.2432
7	0.0636	0.0775	0.0205	0.0203	0.4549	3.0892	5.1891	0.1415	0.1403	175.8737	24.0319	38.7167	1.5832	1.5664	838.3356
8	0.0328	0.0464	0.0264	0.0227	0.3877	0.5175	2.1116	0.1587	0.1478	128.0669	inf	18.8580	inf	2.0238	644.3292
9	0.0261	0.0832	0.0293	0.0407	0.2952	0.2976	6.5921	0.6850	0.7924	72.8623	3.0777	inf	5.1236	10.6042	409.4537
10	0.0343	0.0648	0.0520	0.0558	0.2121	0.9334	2.7193	1.6108	1.1460	33.5401	6.2472	inf	14.9900	12.0242	223.8301
11	0.0467	0.0621	0.0695	0.0641	0.1910	1.7901	2.7249	1.7070	1.8615	25.7141	11.3990	inf	17.6122	17.9298	183.3643
12	0.0566	0.0632	0.0742	0.0688	0.1897	2.4602	2.9940	2.2254	24.308	24.1867	15.5861	inf	21.1385	21.8025	175.5929
13	0.0513	0.0642	0.0647	0.0650	0.2146	2.4087	2.8771	2.5482	1.9983	32.0021	15.4343	19.2628	20.9921	24.5901	217.4465
14	0.0466	0.1034	0.0537	0.0557	0.2738	1.4318	7.5778	2.1409	2.2748	58.8315	10.4153	65.1354	15.3257	16.2581	346.9655
15	0.0426	0.0726	0.0529	0.0540	0.3414	0.8479	3.6727	1.7213	1.8447	96.3967	8.0750	35.2673	11.0307	11.7799	512.1744
16	0.0336	0.0756	0.0454	0.0480	0.3075	0.7256	5.0136	1.3505	1.5281	78.6884	5.9686	43.0665	8.8552	9.8153	435.6142
17	0.0327	0.0593	0.0478	0.0526	0.2545	0.6182	2.0496	1.4480	1.3419	51.1149	4.4347	inf	10.3259	15.5496	309.8492
18	0.0307	0.0245	0.0334	0.0388	0.2623	0.4800	0.4532	0.8896	1.1539	53.7730	3.7851	3.1608	5.3925	6.7847	321.4886
19	0.0352	0.0313	0.0376	0.0386	0.3264	0.8963	0.5020	0.9327	1.1456	85.7751	8.0516	4.4464	5.8537	7.1876	466.0824
20	0.0170	0.0217	0.0355	0.0425	0.2643	0.1570	0.1950	0.6848	1.0053	55.6637	1.5136	1.5098	4.7497	6.7215	331.0547
21	0.0239	0.0337	0.0362	0.0338	0.2122	0.1983	0.6995	0.3370	0.4910	37.0782	5.7331	4.9873	5.6475	240.7256	
22	0.0209	0.0560	0.0509	0.0466	0.1806	0.2626	3.2594	0.8717	0.7708	27.8992	1.4116	22.7194	10.9785	9.6025	192.3555
23	0.0472	0.0612	0.0667	0.0606	0.1679	1.4743	4.0055	1.6974	1.6342	22.7213	7.6127	29.9739	18.1659	16.5928	165.2484
24	0.0489	0.1069	0.0384	0.0376	0.3323	1.3256	7.6553	1.2046	1.1320	92.5652	8.2081	53.8549	6.7853	6.5296	495.1395

Table 11: Comparing JSU, JSU_o, ST1, ST2 and ST5 over the full sample of deseasonalized electricity prices 2010-2016. KS = Kolmogorov-Smirnov, CVM = Cramer-von Mises, and AD = Anderson-Darling statistics; AIC = Bayesian Information Criterion. “Inf” means ‘infinite number’.

6.4. Progressive Modelling

6.4.1. Estimated Coefficients

		MI	M2	M3	M4	AR-MI	AR-M2	AR-M3	AR-M4	VAR-MI	VAR-M2	VAR-M3	VAR-M4		
Eq.		μ_t	drivers	-5.381	5.763	3.811	-10.582	-11.470	-10.240	-9.721	-3.426	-3.807	-4.202		
		$load_{t-1}$	0.375	0.215	1.225	1.283	0.215	0.160	0.130	0.388	1.234	1.2504	1.556		
		$cond_{t-1}$	2.266	2.9678	2.252	2.846	2.160	2.780	2.551	2.533	2.400	2.075	2.235		
		$gast_{t-1}$	0.182	8.748	0.2173	14.422	0.429	18.238	0.452	19.583	0.205	0.291	13.575	0.451	
		$co2_{t-1}$	0.506	16.333	0.618	20.469	0.654	15.258	0.633	17.809	0.701	0.739	19.259	0.451	
		hol_{t-1}	-1.271	-6.908	-0.561	-2.924	0.228	0.839	0.369	1.449	-0.411	-0.296	0.995	0.451	
		$fwind_{t-1}$	-0.200	-35.186	-0.146	-23.545	-0.147	-18.077	-0.137	-18.531	-0.211	-30.460	-0.148	-1.159	
		$fsoilar_{t-1}$	0.092	11.797	0.093	11.721	0.092	12.103	0.091	11.682	-0.025	-1.036	0.028	-1.159	
		g_{t-1}	μ_{t-1}	0.092	11.797	0.093	11.721	0.092	12.103	0.091	11.682	-0.025	-1.036	0.028	-1.159
		σ_{t-1}	ν_{t-1}	τ_{t-1}	$load_{t-1}$	$cond_{t-1}$	$gast_{t-1}$	$co2_{t-1}$	hol_{t-1}	$fwind_{t-1}$	$fsoilar_{t-1}$	g_{t-1}	μ_{t-1}	ν_{t-1}	
log(σ_t)		1.772	43.763	2.510	12.691	1.873	-16.162	-0.063	-0.044	-0.048	-0.044	-0.044	-0.044	-0.044	
		$const$													
		$load_{t-1}$													
		$cond_{t-1}$													
		$gast_{t-1}$													
		$co2_{t-1}$													
		hol_{t-1}													
		$fwind_{t-1}$													
		$fsoilar_{t-1}$													
		μ_{t-1}													
		σ_{t-1}													
		ν_t													
log(τ_t)		-2.966	-11.181	-2.714	-12.288	5.374	5.064	4.535	4.229	-2.805	-10.837	-2.627	-12.292	5.526	
		$load_{t-1}$													
		$cond_{t-1}$													
		$gast_{t-1}$													
		$co2_{t-1}$													
		hol_{t-1}													
		$fwind_{t-1}$													
		$fsoilar_{t-1}$													
		μ_{t-1}													
		σ_{t-1}													
		ν_{t-1}													
		τ_{t-1}													

Table 12: MFST Progressive Modelling for HP3.

		Hour 12										Hour 13										AR-M1					AR-M2					AR-M3				
		M1					M2					M3					M4					VAR-M4					VAR-M3									
Eq.	drivens	16.316	9.128	14.688	7.456	5.768	9.552	2.057	2.863	15.556	7.841	14.017	6.406	4.345	1.657	8.248	12.536	2.260	5.600	4.941	3.341	18.795	5.600	4.941	3.341	18.506	4.423	3.266	1.222	0.183	17.588	3.266	1.222	0.183		
μ_t	$load_{t-1}$	0.188	0.197	0.354	0.294	0.419	0.256	0.240	0.204	0.215	0.167	0.275	0.322	0.425	0.446	0.257	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222	0.222					
	$coal_{t-1}$	2.306	18.04	2.368	18.780	2.766	16.608	2.691	15.740	2.665	18.779	2.736	19.652	3.175	3.072	1.7310	17.713	3.072	1.7310	17.713	3.072	1.7310	17.713	3.072	1.7310	17.713	3.072	1.7310	17.713	3.072	1.7310					
	$gast_{t-1}$	0.483	13.707	0.460	9.078	0.383	8.681	0.307	5.460	0.505	13.233	0.471	8.762	0.393	0.334	6.455	0.841	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478				
	$co2_{t-1}$	0.570	11.218	0.572	11.719	0.663	10.393	0.734	11.328	0.685	12.384	0.698	12.966	0.696	0.696	12.902	0.807	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478	0.779	11.478						
	hol_t	-13.118	-46.095	-12.243	-10.472	-7.139	-14.989	-8.197	-12.208	-13.328	-45.764	-12.419	-39.097	-7.059	-15.441	-8.000	-11.770	-12.294	-36.813	-14.326	-12.294	-36.813	-14.326	-12.294	-36.813	-14.326	-12.294	-36.813	-14.326	-12.294						
	$fwind_{t-1}$	-0.266	-43.106	-0.249	-39.970	-0.183	-39.970	-0.249	-26.061	-0.253	-38.402	-0.253	-38.402	-0.253	-0.253	-24.641	-0.193	-24.707	-0.178	-24.707	-0.178	-24.707	-0.178	-24.707	-0.178	-24.707	-0.178	-24.707	-0.178	-24.707						
	$fsoar_{t-1}$	-0.242	-35.253	-0.233	-35.253	-0.211	-26.675	-0.205	-25.995	-0.242	-33.667	-0.234	-33.667	-0.234	-0.234	-22.471	-0.213	-26.311	-0.234	-26.311	-0.234	-26.311	-0.234	-26.311	-0.234	-26.311	-0.234	-26.311								
	y_{t-1}	0.115	8.087	0.114	8.277	0.093	8.440	0.098	8.076	-0.018	-0.832	-0.023	-1.072	-0.037	-1.430	-0.031	-1.104	-0.027	-1.027	-0.031	-1.027	-0.031	-1.027	-0.031	-1.027	-0.031	-1.027	-0.031	-1.027							
	μ_{t-1}																																			
	σ_{t-1}																																			
	τ_{t-1}																																			
$\log(\sigma_t)$	$const$	1.750	70.752	1.851	8.671	1.760	8.438	1.559	5.731	1.768	70.792	1.597	6.131	1.544	4.733	1.419	4.694	1.419	4.694	1.419	4.694	1.419	4.694	1.419	4.694	1.419	4.694	1.419	4.694	1.419						
	$load_{t-1}$																																			
	$coal_{t-1}$																																			
	$gast_{t-1}$																																			
	$co2_{t-1}$																																			
	hol_t																																			
	$fwind_{t-1}$																																			
	$fsoar_{t-1}$																																			
	μ_{t-1}																																			
	σ_{t-1}																																			
	ν_{t-1}																																			
	τ_{t-1}																																			
$\log(\tau_t)$	$const$	0.085	0.490	0.089	0.455	3.513	4.021	2.169	1.972	0.090	0.510	0.102	0.512	3.021	3.640	1.895	1.777	0.097	0.473	1.415	1.357	1.415	1.357	1.415	1.357	1.415	1.357	1.415	1.357	1.415						
	$load_{t-1}$																																			
	$coal_{t-1}$																																			
	$gast_{t-1}$																																			
	$co2_{t-1}$																																			
	hol_{t-1}																																			
	$fwind_{t-1}$																																			
	$fsoar_{t-1}$																																			
	μ_{t-1}																																			
	σ_{t-1}																																			
	ν_{t-1}																																			
	τ_{t-1}																																			
$\log(\tau_t)$	$const$	1.877	16.181	2.097	14.946	2.353	14.525	-1.014	-0.759	1.892	16.103	2.137	14.743	2.381	14.168	0.443	0.296	2.135	14.684	2.340	14.502	-4.065	-1.555	-4.065	-1.555	-4.065	-1.555	-4.065	-1.555	-4.065						
	$load_{t-1}$																																			
	$coal_{t-1}$																																			
	$gast_{t-1}$																																			
	$co2_{t-1}$																																			
	hol_{t-1}																																			
	$fwind_{t-1}$																																			
	$fsoar_{t-1}$																																			
	μ_{t-1}																																			
	σ_{t-1}																																			
	ν_{t-1}																																			
	τ_{t-1}																																			

Table 13: MFST Progressive Modelling for HP12.

Eq.	μ_t	Hour 19										Hour 20									
		M1	M2	M3	M4	t_{coal}	t_{coal}	t_{coal}	t_{coal}	t_{coal}	t_{coal}	AR-M1	AR-M2	AR-M3	AR-M4	VAR-M1	VAR-M2	VAR-M3	VAR-M4		
log(σ_t)	ν_t	7.260 0.396 σ_{t-1} ν_{t-1}	-7.773 18.763 0.398 σ_{t-1} ν_{t-1}	2.289 13.111 0.331 σ_{t-1} ν_{t-1}	5.186 10.476 0.508 σ_{t-1} ν_{t-1}	-12.566 -10.374 -10.374 σ_{t-1} ν_{t-1}	7.210 6.793 0.182 σ_{t-1} ν_{t-1}	6.715 8.630 0.187 σ_{t-1} ν_{t-1}	6.715 8.630 0.187 σ_{t-1} ν_{t-1}	6.715 8.630 0.187 σ_{t-1} ν_{t-1}	6.715 8.630 0.187 σ_{t-1} ν_{t-1}	-12.674 -12.674 -12.674 σ_{t-1} ν_{t-1}	5.163 5.163 5.163 σ_{t-1} ν_{t-1}								
log(τ_t)	ν_t	45.050 2.040 σ_{t-1} ν_{t-1}	2.297 -0.039 σ_{t-1} ν_{t-1}	8.754 -1.620 -0.006 σ_{t-1} ν_{t-1}	1.818 -0.014 -0.044 σ_{t-1} ν_{t-1}	6.540 -1.601 -0.014 σ_{t-1} ν_{t-1}	2.225 -0.282 -0.246 σ_{t-1} ν_{t-1}	8.189 -1.194 -0.009 σ_{t-1} ν_{t-1}	2.025 -0.004 -0.033 σ_{t-1} ν_{t-1}	43.058 -1.194 -0.033 σ_{t-1} ν_{t-1}	1.305 -1.194 -0.033 σ_{t-1} ν_{t-1}	6.043 -6.043 -6.043 σ_{t-1} ν_{t-1}	6.453 6.453 6.453 σ_{t-1} ν_{t-1}								
log(σ_t)	ν_t	8.784 1.391 σ_{t-1} ν_{t-1}	1.514 0.314 σ_{t-1} ν_{t-1}	10.005 -0.307 -0.307 σ_{t-1} ν_{t-1}	0.314 -0.043 -0.043 σ_{t-1} ν_{t-1}	0.845 -0.744 -0.744 σ_{t-1} ν_{t-1}	1.240 -3.768 -3.768 σ_{t-1} ν_{t-1}	1.389 -1.157 -1.157 σ_{t-1} ν_{t-1}	8.029 -0.095 -0.095 σ_{t-1} ν_{t-1}	9.285 -0.095 -0.095 σ_{t-1} ν_{t-1}	-5.759 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}	-6.068 -6.068 -6.068 σ_{t-1} ν_{t-1}
log(τ_t)	ν_t	1.339 σ_{t-1} ν_{t-1}	16.857 1.701 σ_{t-1} ν_{t-1}	1.5514 1.723 σ_{t-1} ν_{t-1}	1.723 1.723 σ_{t-1} ν_{t-1}	16.544 1.726 -0.046 σ_{t-1} ν_{t-1}	4.024 4.026 -0.348 σ_{t-1} ν_{t-1}	1.349 1.703 -0.032 σ_{t-1} ν_{t-1}	17.003 1.725 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}	1.726 1.726 -0.032 σ_{t-1} ν_{t-1}

Table 14: MFST Progressive Modelling for HP19.

6.4.2. Informative Statistics of Enlarged Models

	Base	R-sqr AR	VAR	Global Deviance			AIC Base	AIC AR	VAR	SBC Base	SBC AR	SBC VAR
				Base	AR	VAR						
Hour 1												
M1	0.7996	0.7818		15151	15362		15173	15384		15237	15448	
M2	0.8377	0.8217	0.8223	14613	14848	14838	14647	14884	14878	14746	14989	14995
M3	0.8466	0.8326	0.8360	14469	14686	14634	14515	14736	14696	14649	14882	14878
M4	0.8476	0.8337	0.8378	14451	14670	14606	14509	14734	14694	14679	14921	14951
Hour 2												
M1	0.7684	0.7591		15580	15673		15602	15695		15666	15760	
M2	0.8250	0.8174	0.8185	14866	14967	14952	14900	15003	14992	14999	15109	15109
M3	0.8360	0.8283	0.8330	14700	14810	14739	14746	14860	14801	14880	15006	14982
M4	0.8386	0.8320	0.8398	14659	14755	14633	14717	14819	14721	14887	15006	14978
Hour 3												
M1	0.7405	0.7284		15919	16029		15941	16051		16005	16115	
M2	0.8077	0.7988	0.7995	15153	15264	15255	15187	15300	15295	15287	15405	15411
M3	0.8193	0.8115	0.8152	14996	15098	15047	15042	15148	15109	15176	15294	15290
M4	0.8227	0.8153	0.8186	14946	15046	15000	15004	15110	15088	15173	15297	15345
Hour 4												
M1	0.6922	0.6817		16335	16413		16357	16435		16421	16500	
M2	0.7714	0.7625	0.7632	15576	15667	15659	15610	15703	15699	15709	15808	15816
M3	0.7855	0.7769	0.7811	15414	15508	15460	15460	15558	15522	15594	15704	15703
M4	0.7888	0.7801	0.7864	15374	15470	15397	15432	15534	15485	15601	15721	15742
Hour 5												
M1	0.6987	0.6888		16194	16271		16216	16293		16280	16357	
M2	0.7721	0.7646	0.7651	15483	15558	15554	15517	15594	15594	15616	15700	15711
M3	0.7855	0.7797	0.7867	15328	15390	15307	15374	15440	15369	15508	15586	15551
M4	0.7889	0.7734	0.7898	15287	15462	15270	15345	15526	15358	15514	15713	15615
Hour 6												
M1	0.7668	0.7609		15634	15688		15658	15712		15728	15782	
M2	0.8188	0.8165	0.8165	14990	15014	15013	15028	15054	15057	15139	15171	15186
M3	0.8334	0.8312	0.8325	14776	14801	14781	14828	14857	14849	14980	15021	15047
M4	0.8371	0.8341	0.8368	14719	14757	14716	14785	14829	14812	14978	15039	15092
Hour 7												
M1	0.7729	0.7673		16701	16753		16725	16777		16795	16847	
M2	0.8056	0.8082	0.8092	16304	16259	16246	16342	16299	16290	16453	16416	16419
M3	0.8248	0.8318	0.8344	16039	15925	15886	16091	15981	15954	16243	16144	16152
M4	0.8288	0.8339	0.8374	15981	15893	15838	16047	15965	15934	16240	16176	16215
Hour 8												
M1	0.8045	0.8056		17540	17516		17564	17540		17635	17610	
M2	0.8197	0.8224	0.8228	17334	17285	17278	17372	17325	17322	17483	17442	17451
M3	0.8382	0.8460	0.8525	17058	16922	16812	17110	16978	16880	17262	17142	17078
M4	0.8420	0.8487	0.8577	16997	16877	16720	17063	16949	16816	17256	17159	17097
Hour 9												
M1	0.7957	0.7998		17610	17545		17634	17569		17704	17640	
M2	0.8077	0.8092	0.8106	17455	17423	17404	17493	17463	17448	17604	17580	17577
M3	0.8241	0.8333	0.8427	17227	17079	16930	17279	17135	16998	17431	17299	17197
M4	0.8286	0.8404	0.8485	17162	16969	16835	17228	17041	16931	17421	17251	17212
Hour 10												
M1	0.7901	0.7949		17346	17275		17370	17299		17440	17369	
M2	0.8006	0.8013	0.8048	17215	17195	17150	17253	17235	17194	17364	17351	17322
M3	0.8126	0.8139	0.8349	17057	17028	16722	17109	17084	16790	17261	17248	16989
M4	0.8195	0.8222	0.8401	16961	16911	16640	17027	16983	16736	17220	17193	17017
Hour 11												
M1	0.8039	0.8034		17143	17140		17167	17164		17237	17234	
M2	0.8150	0.8122	0.8176	16995	17023	16948	17033	17063	16992	17144	17179	17121
M3	0.8328	0.8261	0.8450	16737	16827	16533	16789	16883	16601	16941	17047	16800
M4	0.8379	0.8322	0.8443	16658	16736	16545	16724	16808	16641	16917	17018	16921
Hour 12												
M1	0.8132	0.8106		17079	17107		17103	17131		17174	17201	
M2	0.8215	0.8192	0.8241	16965	16988	16918	17003	17028	16962	17114	17145	17091
M3	0.8341	0.8367	0.8432	16778	16729	16626	16830	16785	16694	16981	16949	16892
M4	0.8367	0.8380	0.8428	16736	16708	16633	16802	16780	16729	16995	16990	17009

Table 15: Informative Statistics of Enlarged Models. R-sqr is the “Cox-Snell” or Nagelkerke’s R squared, Global Deviance is $GD = -2\mathcal{L}\mathcal{L}$, AIC and SBC are the Akaike and Schwarz Bayesian Information Criteria for estimated models, as formulated in eqs 2-7 and described therein.

	Base	R-sqr AR	VAR	Global Deviance			AIC			SBC		
				Base	AR	VAR	Base	AR	VAR	Base	AR	VAR
Hour 13												
M1	0.8262	0.8222		16825	16876		16849	16900		16919	16970	
M2	0.8345	0.8311	0.8345	16700	16745	16693	16738	16785	16737	16849	16902	16865
M3	0.8510	0.8421	0.8549	16433	16573	16358	16485	16629	16426	16637	16793	16625
M4	0.8526	0.8515	0.8559	16405	16417	16340	16471	16489	16436	16663	16699	16717
Hour 14												
M1	0.8321	0.8282		16845	16896		16869	16920		16939	16990	
M2	0.8381	0.8357	0.8382	16752	16781	16742	16790	16821	16786	16901	16938	16915
M3	0.8590	0.8579	0.8644	16400	16411	16293	16452	16467	16361	16604	16631	16559
M4	0.8639	0.8631	0.8705	16309	16316	16176	16375	16388	16272	16568	16599	16552
Hour 15												
M1	0.8267	0.8227		16870	16920		16894	16944		16964	17014	
M2	0.8328	0.8297	0.8316	16779	16817	16789	16817	16857	16833	16928	16974	16962
M3	0.8541	0.8540	0.8605	16431	16425	16309	16483	16481	16377	16635	16644	16576
M4	0.8608	0.8612	0.8705	16311	16296	16120	16377	16368	16216	16570	16579	16497
Hour 16												
M1	0.8129	0.8111		16826	16843		16850	16867		16920	16937	
M2	0.8200	0.8186	0.8205	16728	16739	16713	16766	16779	16757	16877	16895	16886
M3	0.8365	0.8359	0.8429	16482	16483	16372	16534	16539	16440	16686	16703	16639
M4	0.8463	0.8478	0.8589	16324	16292	16099	16390	16364	16195	16582	16574	16476
Hour 17												
M1	0.7924	0.7919		16972	16971		16996	16995		17066	17066	
M2	0.8017	0.8001	0.8009	16855	16869	16859	16893	16909	16903	17004	17026	17032
M3	0.8176	0.8187	0.8227	16642	16620	16563	16694	16676	16631	16846	16839	16830
M4	0.8214	0.8217	0.8258	16589	16577	16518	16655	16649	16614	16848	16860	16895
Hour 18												
M1	0.7465	0.7442		17881	17898		17905	17922		17975	17992	
M2	0.7683	0.7657	0.7666	17652	17674	17664	17690	17714	17708	17801	17831	17836
M3	0.7858	0.7831	0.8006	17451	17477	17262	17503	17533	17330	17655	17696	17529
M4	0.7990	0.7965	0.8021	17289	17314	17243	17355	17386	17339	17548	17596	17620
Hour 19												
M1	0.7489	0.7417		17829	17895		17853	17919		17923	17989	
M2	0.7714	0.7654	0.7726	17589	17650	17569	17627	17690	17613	17739	17806	17742
M3	0.7915	0.7866	0.7895	17354	17408	17373	17406	17464	17441	17558	17627	17640
M4	0.8040	0.8001	0.8052	17196	17241	17176	17262	17313	17272	17455	17523	17553
Hour 20												
M1	0.7223	0.7156		17585	17640		17609	17664		17679	17734	
M2	0.7457	0.7413	0.7494	17361	17399	17317	17399	17439	17361	17510	17556	17490
M3	0.7695	0.7641	0.7658	17111	17163	17145	17163	17219	17213	17315	17383	17411
M4	0.7784	0.7727	0.7766	17010	17068	17024	17076	17140	17120	17269	17351	17401
Hour 21												
M1	0.7553	0.7494		16424	16478		16448	16502		16518	16572	
M2	0.7712	0.7673	0.7716	16252	16290	16242	16290	16330	16286	16401	16447	16415
M3	0.7805	0.7776	0.7882	16146	16174	16050	16198	16230	16118	16350	16393	16317
M4	0.7878	0.7854	0.7945	16060	16083	15973	16126	16155	16069	16319	16366	16350
Hour 22												
M1	0.7862	0.7732		15464	15608		15486	15630		15550	15695	
M2	0.7996	0.7894	0.7935	15299	15419	15369	15333	15455	15409	15432	15560	15526
M3	0.8138	0.8032	0.8155	15111	15246	15082	15157	15296	15144	15291	15442	15325
M4	0.8177	0.8070	0.8154	15058	15196	15082	15116	15260	15170	15285	15447	15427
Hour 23												
M1	0.8082	0.8013		15032	15118		15054	15140		15119	15204	
M2	0.8182	0.8123	0.8138	14897	14972	14952	14931	15008	14992	15030	15113	15109
M3	0.8335	0.8292	0.8298	14671	14732	14722	14717	14782	14784	14852	14928	14965
M4	0.8330	0.8304	0.8330	14679	14714	14674	14737	14778	14762	14906	14965	15019
Hour 24												
M1	0.8286	0.8191		14656	14788		14678	14810		14743	14874	
M2	0.8438	0.8365	0.8382	14419	14531	14504	14453	14567	14544	14552	14672	14661
M3	0.8556	0.8500	0.8528	14219	14312	14264	14265	14362	14326	14400	14508	14507
M4	0.8548	0.8500	0.8564	14233	14310	14200	14291	14374	14288	14460	14561	14545

Table 16: Informative Statistics of Enlarged Models. R-sqr is the “Cox-Snell” or Nagelkerke’s R squared, Global Deviance is $GD = -2\mathcal{L}\mathcal{L}$, AIC and SBC are the Akaike and Schwarz Bayesian Information Criteria for estimated models, as formulated in eqs 2-7 and described therein.

6.4.3. Residuals Analysis

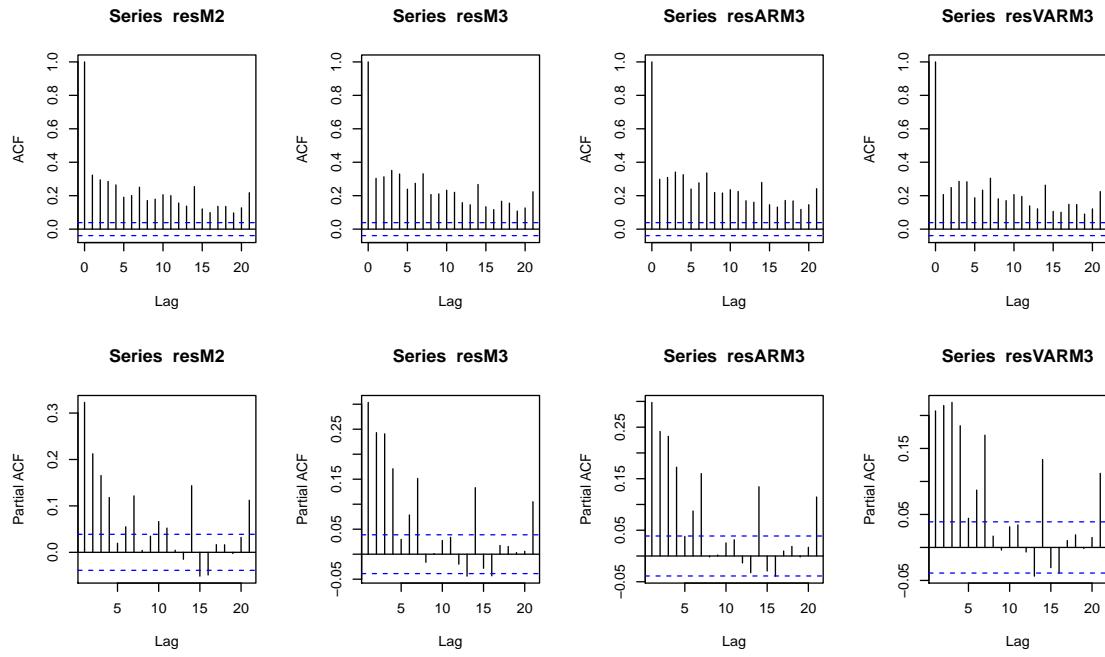


Figure 11: Autocorrelation and Partial autocorrelation Functions for Models M2, M3, ARM3 and VARM3 for hour 3.

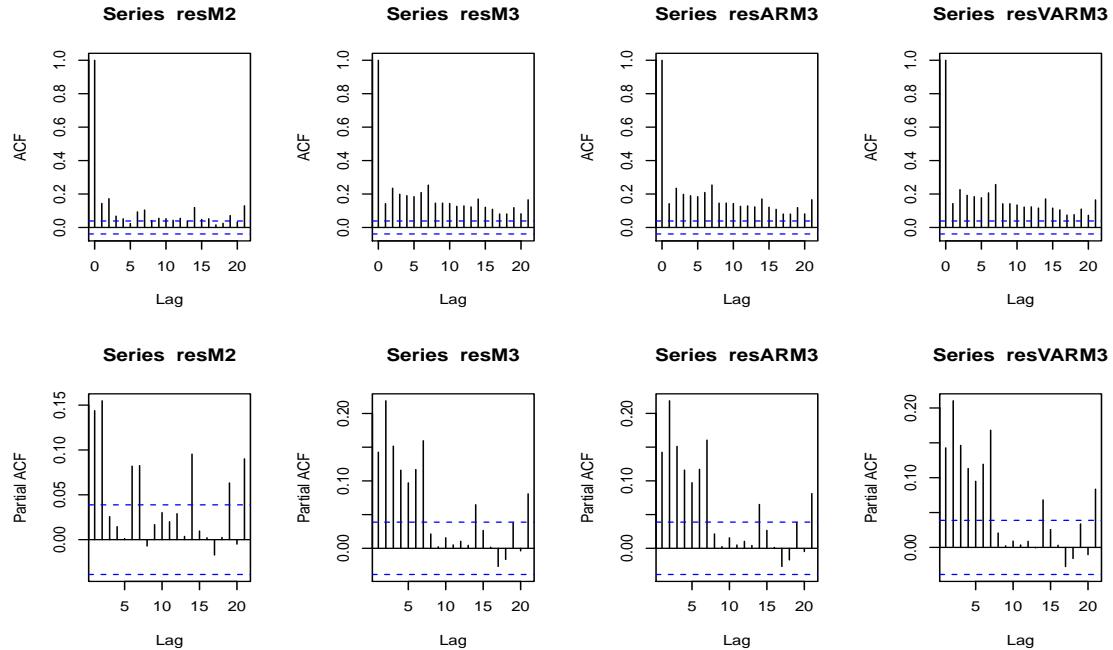


Figure 12: Autocorrelation and Partial autocorrelation Functions for Models M2, M3, ARM3 and VARM3 for hour 12.

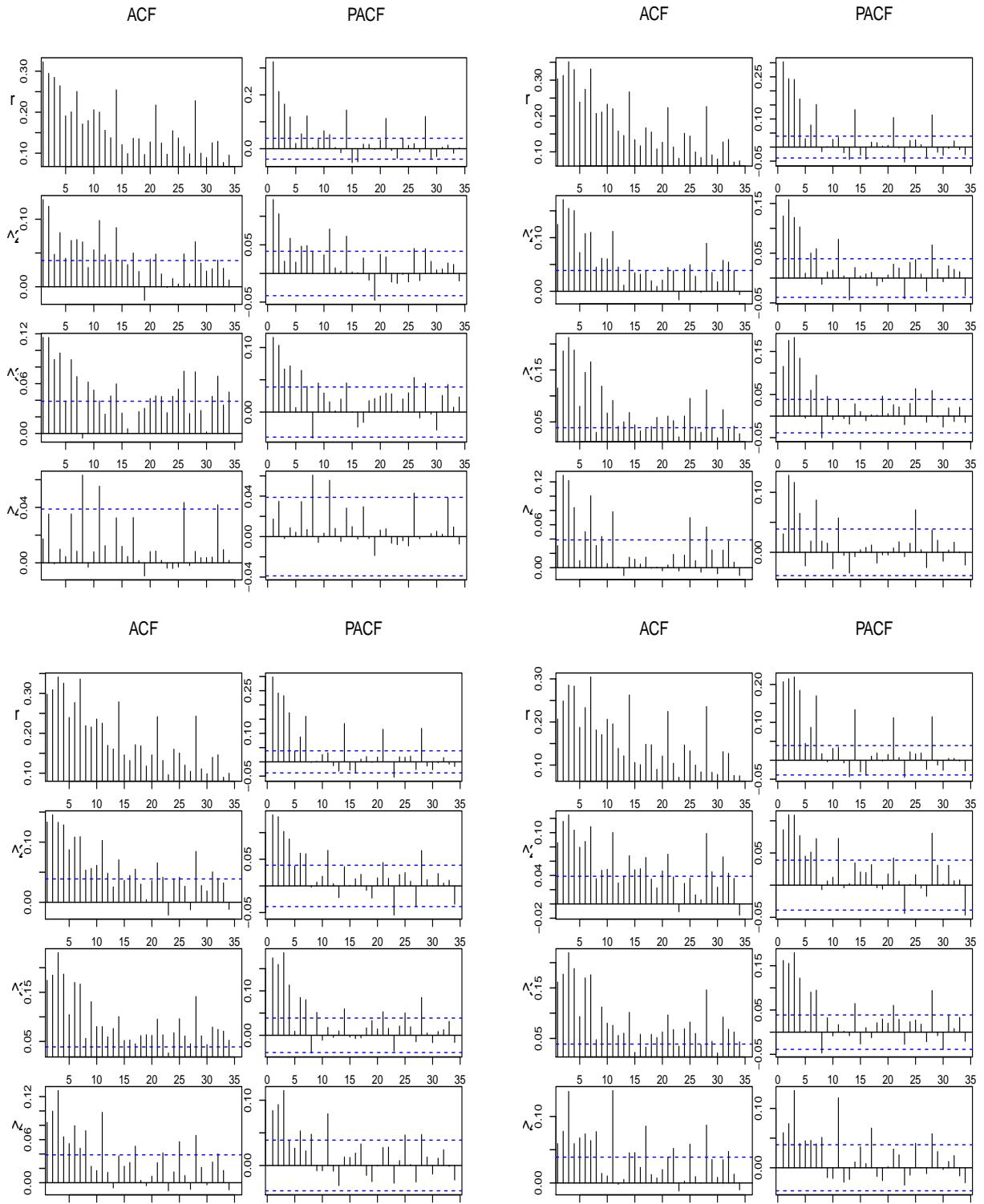


Figure 13: ACF and PACF for Residuals of Models M2 (top-left), M3 (top-right), ARM3 (bottom-left) and VARM3 (bottom-right) for hour 3. Levels of residuals on first rows, r_t^2 on second rows, r_t^3 on third rows and finally r_t^4 on last rows.

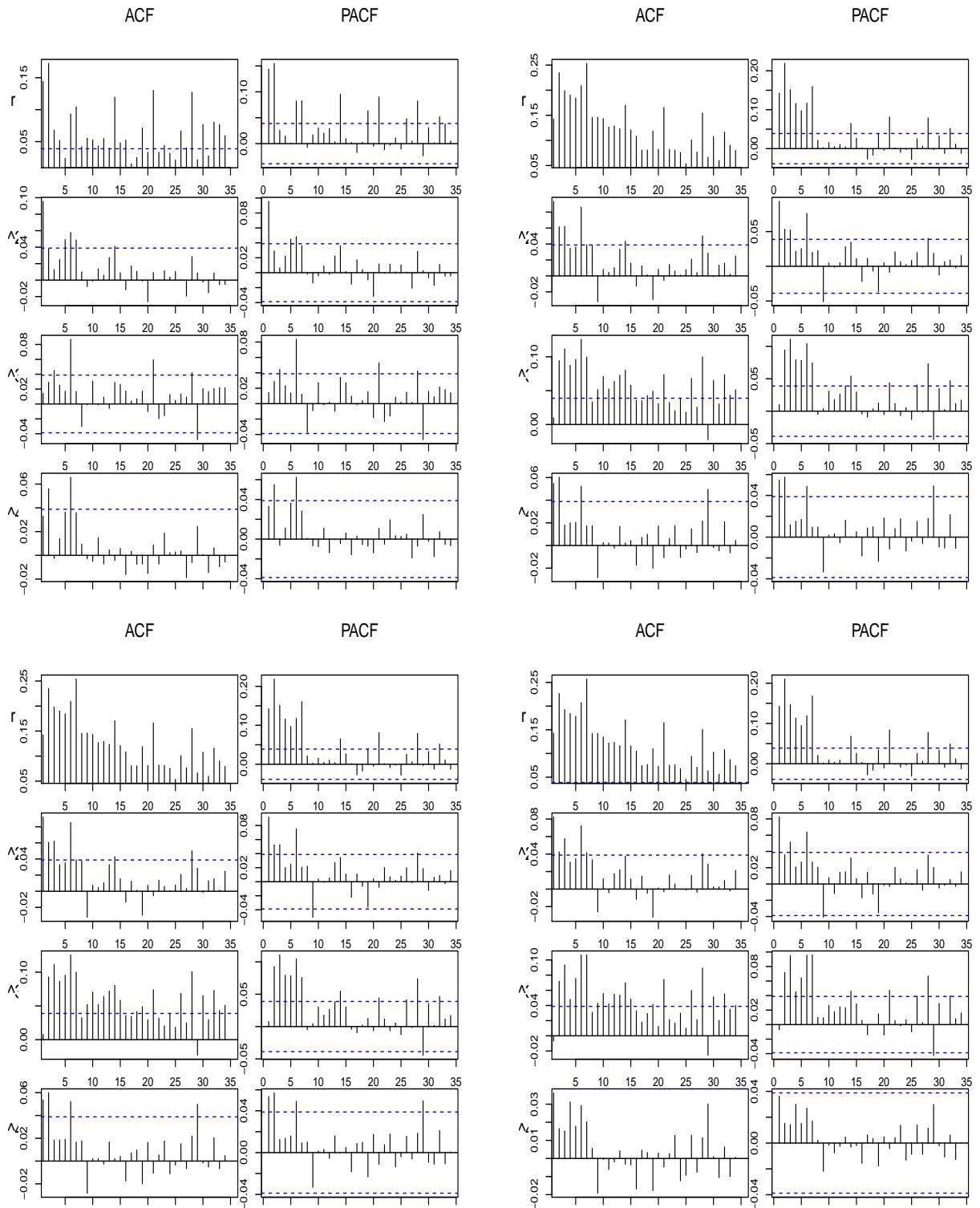


Figure 14: ACF and PACF for Residuals of Models M2 (top-left), M3 (top-right), ARM3 (bottom-left) and VARM3 (bottom-right) for hour 12. Levels of residuals on first rows, r_t^2 on second rows, r_t^3 on third rows and finally r_t^4 on last rows.

	Base				AR				VAR			
	Mean	Var	Skew	Kurt	Mean	Var	Skew	Kurt	Mean	Var	Skew	Kurt
	Hour 3											
JSU	M1	-0.004	1.001	-0.061	3.055	-0.002	0.999	-0.075	3.085			
	M2	-0.029	1.007	0.027	2.955	-0.029	1.007	0.025	2.965	-0.028	1.006	0.024
	M3	-0.029	1.011	0.026	2.941	-0.030	1.010	0.024	2.973	-0.029	1.007	0.025
	M4	-0.036	0.994	-0.017	3.210	-0.033	0.997	0.007	3.084	-0.029	0.989	-0.034
JSUo	M1	-0.002	0.998	-0.068	2.994	-0.002	0.998	-0.084	3.020			
	M2	-0.002	0.996	0.033	3.272	-0.002	0.996	0.032	3.325	-0.002	0.995	0.005
	M3	-0.001	0.997	0.036	2.908	-0.001	0.998	0.050	2.868	-0.001	0.998	0.005
	M4	-0.002	0.993	0.139	3.505	-0.001	0.996	-0.013	3.411	-0.001	0.994	0.053
ST1	M1	-0.006	1.001	-0.055	2.835	-0.008	1.000	-0.068	2.830			
	M2	-0.008	0.988	0.113	3.232	-0.009	0.987	0.117	3.255	-0.010	0.986	0.108
	M3	-0.009	0.974	-0.019	3.226	-0.012	0.971	-0.032	3.279	-0.014	0.958	-0.019
	M4	-0.009	0.978	-0.033	3.143	-0.016	0.972	-0.026	3.224	-0.027	0.974	-0.055
ST2	M1	-0.004	0.995	-0.058	2.876	-0.005	0.993	-0.068	2.886			
	M2	-0.009	0.992	0.091	3.189	-0.009	0.990	0.094	3.210	-0.011	0.988	0.088
	M3	-0.011	0.977	-0.027	3.165	-0.014	0.973	-0.041	3.222	-0.016	0.960	-0.028
	M4	-0.013	0.976	-0.045	3.125	-0.020	0.967	-0.019	3.223	-0.028	0.964	-0.031
SN1	M1	0.002	1.068	-10.506	> 240	0.001	1.067	-10.271	> 230			
	M2	-0.001	0.991	-0.068	4.254	-0.003	0.990	-0.052	4.222	-0.004	0.987	-0.054
	M3	0.007	0.982	-0.258	4.141	0.004	0.977	-0.251	4.035	0.000	0.962	-0.237
	NO	M1	0.000	1.000	-10.527	> 240	0.000	1.000	-10.296	> 230		
NO	M2	-0.044	0.998	-0.855	5.024	-0.044	0.998	-0.810	4.861	-0.041	0.999	-0.815
												4.874
Hour 12												
JSU	M1	0.000	0.999	0.054	3.189	0.001	0.999	0.065	3.199			
	M2	0.001	0.999	0.040	3.144	-0.003	0.999	0.043	3.136	0.004	0.999	0.049
	M3	-0.001	1.002	0.017	3.104	-0.004	1.002	0.016	3.085	0.000	1.004	0.027
	M4	0.006	0.973	0.014	3.287	-0.005	0.973	0.018	3.188	-0.003	1.007	0.003
JSUo	M1	0.000	1.000	0.054	3.093	0.000	1.000	0.067	3.100			
	M2	0.000	1.000	0.037	3.032	0.000	1.000	0.067	3.031	0.000	1.000	0.022
	M3	0.000	0.997	0.005	3.103	0.001	0.997	-0.002	3.148	0.001	0.996	-0.015
	M4	0.000	0.997	0.095	3.309	0.000	0.997	0.057	3.356	0.001	0.995	-0.053
ST1	M1	0.000	1.001	0.008	3.079	0.000	1.001	0.021	3.080			
	M2	0.000	1.002	0.010	3.059	-0.001	1.002	0.028	3.051	0.000	1.002	0.009
	M3	0.012	0.984	0.127	3.220	0.016	0.980	0.377	3.536	0.024	0.980	0.350
	M4	0.012	0.973	0.483	3.889	0.004	0.968	0.503	3.860	0.011	0.970	0.395
ST2	M1	0.001	1.001	0.039	3.085	0.001	1.001	0.056	3.090			
	M2	0.001	1.002	0.035	3.064	0.001	1.002	0.053	3.059	0.001	1.002	0.038
	M3	0.010	0.985	0.146	3.202	0.016	0.980	0.357	3.492	0.023	0.981	0.337
	M4	0.000	0.974	0.355	3.619	0.007	0.973	0.482	3.889	0.012	0.970	0.389
SN1	M1	-0.002	0.968	0.309	5.880			0.250	5.334			
	M2			0.134	4.705			0.100	4.104	0.000	1.003	0.092
	M3			0.114	3.784	-0.008	0.980	0.328	4.824	0.002	0.991	0.077
NO	M1	inf				inf				inf		
	M2	inf				inf				inf		
Hour 19												
JSU	M1	0.000	0.997	0.096	3.203	0.000	0.998	0.110	3.227			
	M2	0.012	1.002	0.027	3.063	0.011	1.002	0.030	3.074	0.010	1.002	0.021
	M3	0.007	1.006	0.016	3.110	0.005	1.005	0.003	3.167	0.004	1.004	0.000
	M4	0.013	1.004	0.040	2.909	-0.001	0.991	0.027	2.871	0.005	0.996	0.040
JSUo	M1	0.001	0.999	0.115	3.136	0.001	0.999	0.130	3.155			
	M2	0.003	0.988	0.097	4.500	0.003	0.989	0.081	4.151	0.002	0.997	-0.087
	M3	0.002	0.992	0.407	3.831	0.002	0.990	0.380	3.780	0.002	0.993	0.260
	M4	-0.001	0.992	0.252	3.621	-0.001	0.991	0.256	3.554	-0.001	0.989	0.246
ST1	M1	0.001	1.003	0.063	2.971	0.001	1.003	0.074	2.976			
	M2	0.005	0.995	-0.086	3.209	0.008	0.991	-0.112	3.282	0.003	0.998	-0.071
	M3	-0.001	0.969	-0.077	3.325	0.003	0.955	-0.136	3.377	-0.001	0.965	-0.100
	M4	0.030	0.947	-0.007	3.447	0.014	0.920	0.307	3.647	0.017	0.960	-0.078
ST2	M1	0.005	0.994	0.086	2.983	0.005	0.993	0.100	2.983			
	M2	0.007	0.995	-0.062	3.152	0.010	0.992	-0.085	3.209	0.005	0.998	-0.050
	M3	0.033	0.957	0.504	3.624	0.026	0.959	0.484	3.629	0.023	0.959	0.497
	M4	0.016	0.924	0.352	3.850	0.012	0.922	0.295	3.630	0.013	0.921	0.269
SN1	M1		1.593	10.244	-0.004	0.940	1.615	10.361				
	M2	0.006	0.987	0.042	4.423	0.010	0.982	-0.004	4.128	0.001	0.996	0.030
	M3	-0.013	0.973	0.329	4.274	-0.011	0.977	0.351	3.976	-0.006	0.989	0.091
NO	M1	inf			inf					inf		
	M2	0.023	1.000	0.643	4.870	0.028	1.000	0.533	4.491	0.013	1.000	0.553
												4.306

Table 17: Descriptive Statistics of (Randomized Quantile) Residuals for Enlarged Models and selected hours.

6.5. Coverage Tests

The formulated models in Tables 18-26, without parameter indications and under the skew Student-t distribution unless diversely specified, are

- with just the first moment:

- with just the first moment:
 - M1: $\mu_t = c + \sum_{i=1}^7 y_{t-i} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$;
 - ARM1: $\mu_t = c + \mu_{t-1} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$ with μ_{t-1} filtered from M1;
 - QReg: $Q_q(y_t) = \alpha^q + \sum_{i=1}^7 \gamma_i^q y_{t-i} + \beta_1^q hol_t + \beta_2^q load_{t-1} + \beta_3^q fwind_t + \beta_4^q fsolar_t + \beta_5^q coal_{t-1} + \beta_6^q gas_{t-1} + \beta_7^q co_{2t-1}$;

- with the first two moments:

- with the first two moments:
 - M2st: μ_t as in M1 and $\log(\sigma_t) = hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$;
 - M2jsu: as M2 but with the Johnson's S_U distribution;
 - M2Ser: $\mu_t = c + y_{t-1} + y_{t-2} + y_{t-7} + miny_{t-1} + load_t + load_t^2 + load_t^3 + \sum_{i=1}^6 d_i$ and $\log(\sigma_t) = c + y_{t-1} + abs(y_{t-1} - y_{t-2}) + \sum_{i=1}^6 d_i$ under the Johnson's S_U ;
 - ARM2: μ_t as in ARM1, and $\log(\sigma_t) = c + \log(\sigma_{t-1}) + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$ with μ_{t-1} and $\log(\sigma_{t-1})$ filtered from M2;
 - VARM2: $\mu_t = \log(\sigma_t) = c + \mu_{t-1} + \log(\sigma_{t-1}) + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$, with μ_{t-1} and $\log(\sigma_{t-1})$ filtered from M2;
 - AR-EGARCH-st and -sst: as in eqs. (9) and (10) under the Student-t and the skew Student-t distributions;

- with the three moments:

- with the three moments:
 - M3: $\mu_t = c + y_{t-1} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$, and $\log(\sigma_t) = \nu_t = c + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$;
 - ARM3: μ_t and $\log(\sigma_t)$ as in ARM1 and ARM2, $\nu_t = c + \nu_{t-1} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$ with μ_{t-1} , $\log(\sigma_{t-1})$ and ν_{t-1} filtered from M3;
 - VARM3: $\mu_t = \log(\sigma_t) = \nu_t = c + \mu_{t-1} + \log(\sigma_{t-1}) + \nu_{t-1} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$, with μ_{t-1} , $\log(\sigma_{t-1})$ and ν_{t-1} filtered from M3;

- with four moments:

- with four moments:
 - M4: μ_t , $\log(\sigma_t)$, and ν_t as in M3, and with $\log(\tau_t) = c + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$;
 - ARM4: μ_t , $\log(\sigma_t)$ and ν_t as in ARM3, $\log(\tau_t) = c + \tau_{t-1} + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$ with μ_{t-1} , $\log(\sigma_{t-1})$, ν_{t-1} and $\log(\tau_{t-1})$ filtered from M4;
 - VARM4: $\mu_t = \log(\sigma_t) = \nu_t = \log(\tau_t) = c + \mu_{t-1} + \log(\sigma_{t-1}) + \nu_{t-1} + \log(\tau_{t-1}) + hol_t + load_{t-1} + fwind_t + fsolar_t + coal_{t-1} + gas_{t-1} + co_{2t-1}$, with μ_{t-1} , $\log(\sigma_{t-1})$, ν_{t-1} and $\log(\tau_{t-1})$ filtered from M4.

Years	Models	Q_1			Q_2			Q_5		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000006	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.042737	0.047887	0.000000	0.021708	0.000301	0.000000	0.063088	0.000001	0.000000
2012	ARM3	0.238900	0.107340	0.000000	0.003907	0.000000	0.000000	0.038200	0.000000	0.000000
	VARM3	0.005059	0.001559	0.000000	0.001519	0.000014	0.000000	0.000213	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000008	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000008	0.000000	0.000000	0.000029	0.000000	0.000000	0.000005	0.000000	0.000000
	M2st	0.001823	0.000003	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.117696	0.000001	0.000000	0.197757	0.000190	0.000000	0.000153	0.000015	0.000000
2015	ARM3	0.117696	0.000001	0.000000	0.338885	0.000109	0.000000	0.000013	0.000000	0.000000
	VARM3	0.117696	0.000001	0.000000	0.338885	0.000109	0.000000	0.000013	0.000001	0.000000
	M4	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000030	0.000000	0.000000
	ARM4	0.000008	0.000000	0.000000	0.000009	0.000000	0.000000	0.000153	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000002	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000008	0.000000	0.000000	0.001938	0.003929	0.000000	0.857958	0.208593	0.128320
	M2st	0.006756	0.025518	0.934814	0.068612	0.086869	0.001038	0.000000	0.000000	0.000000
	M2jsu	0.006756	0.004890	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.098483	0.254687	0.988059	0.539538	0.373329	0.000914	0.000000	0.000000	0.000000
	VARM2	0.098483	0.254687	0.988316	0.539538	0.373329	0.000906	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000178	0.000828	0.138134
	ARM3	0.006756	0.025518	0.934814	0.03122	0.012651	0.786306	0.060904	0.019116	0.039357
	VARM3	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.005822	0.008170	0.344486
	M4	0.501312	0.743997	0.021828	0.539538	0.659422	0.745787	0.382296	0.576701	0.928168
	ARM4	0.117696	0.083839	0.001082	0.107486	0.191873	0.360920	0.518500	0.269329	0.332937
	VARM4	0.117696	0.256373	0.078650	0.025949	0.047842	0.090188	0.272025	0.500320	0.802992
2017	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.006054	0.017444	0.000102	0.001997	0.000159	0.000000	0.050582	0.057404	0.000000
	M2st	0.097586	0.252834	0.991500	0.067627	0.183640	0.967964	0.000000	0.000000	0.000000
	M2jsu	0.006054	0.012375	0.000184	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000008	0.000041	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.860310	0.941947	0.004213	0.011879	0.013910	0.000476	0.000000	0.000000	0.000000
	VARM2	0.860310	0.941947	0.004471	0.055625	0.032678	0.000119	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.006681	0.025263	0.934247	0.175408	0.381984	0.148638	0.000718	0.000001	0.000000
2018	ARM3	0.006681	0.025263	0.934247	0.175408	0.381984	0.148473	0.000718	0.000001	0.000000
	VARM3	0.006681	0.025263	0.934247	0.611055	0.794848	0.565671	0.000345	0.000000	0.000000
	M4	0.018014	0.048560	0.000474	0.342897	0.481181	0.308151	0.414556	0.641707	0.967434
	ARM4	0.018014	0.027413	0.001663	0.055625	0.032678	0.000211	0.942492	0.161620	0.097266
	VARM4	0.000034	0.000049	0.000000	0.005012	0.007823	0.000007	0.081539	0.022825	0.000658

Table 18: Coverage tests for Lower Quantiles of HP3. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_{25}			Q_{50}			Q_{75}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.830988	0.000003	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.413151	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000009	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.156623	0.000000	0.000000	0.002501	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.000043	0.000000	0.000000	0.000010	0.000000	0.000000
2012	ARM3	0.000000	0.000000	0.000000	0.001210	0.000000	0.000000	0.000063	0.000000	0.000000
	VARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000003	0.000000	0.000000
	M4	0.830988	0.000000	0.000000	0.000006	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.975616	0.000000	0.000000	0.000004	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.194493	0.000000	0.000000	0.000249	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.309641	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000036	0.000000	0.000000	0.250018	0.000000	0.000000	0.002743	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.003358	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000542	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.116647	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000039	0.000000	0.000000
	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000002	0.000000	0.000000
	VARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.00898	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.209497	0.000000	0.000000	0.021231	0.000004	0.000000
	ARM4	0.000000	0.000000	0.000000	0.143112	0.000000	0.000000	0.038290	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.143112	0.000000	0.000000	0.085083	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.832815	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.002464	0.000000	0.000000	0.066744	0.000001	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.228484	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000646	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.157394	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2015	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000011	0.000000	0.000000	0.958250	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000082	0.000000	0.000000	0.793540	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000031	0.000000	0.000000	0.564744	0.000002	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.032857	0.009650	0.006893	0.000052	0.000000	0.000000	0.000003	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000082	0.000001	0.000003	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.001968	0.001453	0.000009	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000001	0.000000	0.000000	0.271553	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.101772	0.000000	0.000000	0.228484	0.000000	0.000000	0.522952	0.000000	0.000000
	ARM3	0.000082	0.000000	0.000000	0.003923	0.000000	0.000000	0.132440	0.000000	0.000000
	VARM3	0.000007	0.000000	0.000000	0.104463	0.000000	0.000000	0.418237	0.000000	0.000000
	M4	0.061428	0.000005	0.000000	0.228484	0.000087	0.000010	0.000031	0.000003	0.000000
	ARM4	0.026430	0.000000	0.000000	0.228484	0.000206	0.000011	0.000011	0.000001	0.000000
	VARM4	0.035416	0.000000	0.000000	0.157394	0.003323	0.000098	0.000031	0.000003	0.000000
2017	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000007	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.0509491	0.000807	0.000005	1.000000	0.000000	0.000000	0.108524	0.0239567	0.022321
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.065951	0.000255	0.000481	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.834382	0.829907	0.013538	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000002	0.000000	0.000000	0.673945	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.008854	0.000000	0.000000	0.108524	0.000000	0.000000
2018	ARM3	0.000000	0.000000	0.000000	0.011984	0.000000	0.000000	0.038290	0.000000	0.000000
	VARM3	0.000000	0.000000	0.000000	0.295710	0.000000	0.000000	0.003971	0.000000	0.000000
	M4	0.369477	0.000836	0.000018	0.601155	0.111012	0.002769	0.005683	0.003815	0.003118
	ARM4	0.951900	0.023730	0.008274	0.250018	0.046772	0.005665	0.000149	0.000002	0.000001
	VARM4	0.136865	0.034958	0.003335	1.000000	0.004233	0.000448	0.003971	0.000004	0.000000

Table 19: Coverage tests for Middle Quantiles of HP3. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_{95}			Q_{98}			Q_{99}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000420	0.001309	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.030591	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.030591	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
	VARM2	0.000000	0.000000	0.030591	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
	EGARCHst	0.001886	0.000005	0.000000	0.021708	0.000001	0.000000	0.473190	0.113910	0.000000
	M3	0.000000	0.000001	0.055849	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
2012	ARM3	0.000000	0.000001	0.055789	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
	VARM3	0.000244	0.000176	0.018630	0.000146	0.000737	0.622249	0.007389	0.027654	0.939256
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.026103	0.000120	0.000615	0.603163	0.006681	0.025263	0.934247
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.050591	0.003063	0.012432	0.783078	0.097586	0.252834	0.988034
	VARM2	0.000001	0.000005	0.088280	0.000120	0.000615	0.603163	0.006681	0.025263	0.934247
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000171	0.000798	0.022039	0.003063	0.012432	0.793918	0.006681	0.025263	0.934247
	ARM3	0.000037	0.000193	0.044777	0.003063	0.012432	0.789655	0.006681	0.025263	0.934247
	VARM3	0.002044	0.007509	0.050295	0.000120	0.000615	0.603163	0.006681	0.025263	0.934247
	M4	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000008	0.000003	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000002	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.026369	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.026369	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	VARM2	0.000000	0.000000	0.048856	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000667	0.000000	0.000000	0.068612	0.008689	0.000000	0.724226	0.042852	0.000000
2015	ARM3	0.014147	0.000003	0.000000	0.068612	0.008689	0.000000	0.342459	0.630242	0.000000
	VARM3	0.000007	0.000002	0.000007	0.019025	0.063250	0.000121	0.342459	0.630242	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.952019	0.000001	0.000000	0.054567	0.000000	0.000000	0.257954	0.110944	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.026369	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.117696	0.256373	0.000000
	ARM2	0.000000	0.000000	0.026369	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	VARM2	0.000000	0.000000	0.026369	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000007	0.000038	0.157226	0.003122	0.012651	0.797083	0.006756	0.025518	0.934814
	ARM3	0.000007	0.000038	0.157750	0.003122	0.012651	0.796869	0.009483	0.254687	0.992637
	VARM3	0.005822	0.000670	0.003326	0.003122	0.012651	0.797131	0.006756	0.025518	0.934814
	M4	0.000683	0.000086	0.000000	0.011592	0.002554	0.000000	0.48057	0.004252	0.000000
	ARM4	0.0002710	0.0000167	0.000000	0.000088	0.000029	0.000000	0.000000	0.000033	0.000000
	VARM4	0.000013	0.000001	0.000000	0.000003	0.000000	0.000000	0.000000	0.000000	0.000000
2017	M1	0.180970	0.008472	0.000004	0.175408	0.386258	0.013229	0.097586	0.252834	0.981391
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.026103	0.000120	0.000615	0.603163	0.006681	0.025263
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.860310	0.941947	0.095691
	M2Ser	0.000000	0.000000	0.000000	0.005012	0.009360	0.000000	0.781960	0.097586	0.252834
	ARM2	0.000006	0.000002	0.000324	0.003063	0.012432	0.781960	0.097586	0.252834	0.988763
	VARM2	0.000006	0.000002	0.000330	0.003063	0.012432	0.789155	0.097586	0.252834	0.991279
	EGARCHst	0.000001	0.000000	0.000000	0.000009	0.000019	0.000000	0.000139	0.000144	0.000000
	M3	0.005637	0.007916	0.231153	0.003063	0.012432	0.778046	0.097586	0.252834	0.976308
2018	ARM3	0.029983	0.050783	0.506537	0.003063	0.012432	0.779610	0.097586	0.252834	0.978651
	VARM3	0.059399	0.104759	0.517833	0.003063	0.012432	0.782689	0.097586	0.252834	0.983390
	M4	0.000718	0.000107	0.000000	0.000092	0.0000338	0.000005	0.000529	0.001638	0.000000
	ARM4	0.000161	0.000053	0.000000	0.000030	0.000054	0.000000	0.000034	0.000049	0.000000
	VARM4	0.000001	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.026103	0.000120	0.000615	0.603163	0.006681	0.025263
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.860310</td		

Years	Models	Q_1			Q_2			Q_5		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.362478	0.653153	0.999860	0.077043	0.204213	0.982854	0.131899	0.228489	0.747415
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.755217	0.928583	0.000151	0.046625	0.029370	0.001323	0.100637	0.000106	0.000000
	VARM2	0.755217	0.928583	0.000153	0.176038	0.046094	0.004131	0.038200	0.000166	0.000000
	EGARCHst	0.000001	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.822449	0.088426	0.002268	0.176038	0.003981	0.000002	0.100637	0.000814	0.000002
2012	ARM3	0.238900	0.107340	0.040831	0.176038	0.003981	0.000002	0.154972	0.000648	0.000000
	VARM3	0.473190	0.113910	0.017902	0.093767	0.039186	0.001123	0.100637	0.000814	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.720414	0.042658	0.000001	0.358469	0.002189	0.000000	0.127009	0.000053	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.720414	0.042658	0.000000	0.802449	0.000608	0.000000	0.017599	0.000000	0.000000
	VARM2	0.720414	0.042658	0.000000	0.342897	0.002194	0.000000	0.005374	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.048753	0.000148	0.000000	0.055625	0.000000	0.000000	0.000073	0.000000	0.000000
	ARM3	0.048753	0.000148	0.000000	0.109291	0.000006	0.000000	0.001447	0.000004	0.000000
	VARM3	0.018014	0.000096	0.000000	0.005012	0.000001	0.000000	0.000006	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.002117	0.001101	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.006756	0.025518	0.934814	0.003122	0.12651	0.717610	0.060904	0.000002	0.000000
	VARM2	0.006756	0.025518	0.934814	0.019025	0.001033	0.000000	0.060904	0.000002	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.342459	0.010290	0.000000	0.068612	0.008689	0.000000	0.678752	0.007428	0.000474
2015	ARM3	0.342459	0.010290	0.000000	0.068612	0.008689	0.000001	0.761472	0.009267	0.032507
	VARM3	0.342459	0.010290	0.000000	0.068612	0.008689	0.000001	0.857958	0.004694	0.001115
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000001	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000515	0.001687	0.000000	0.001938	0.003608	0.000000	0.382296	0.249397	0.046475
	M2st	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000000	0.000000	0.049059
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000001	0.000005	0.092642
	VARM2	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000001	0.000005	0.092690
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.006756	0.025518	0.934814	0.003122	0.12651	0.794360	0.006667	0.002814	0.126549
	ARM3	0.006756	0.025518	0.934814	0.003122	0.12651	0.794842	0.005822	0.019076	0.113599
	VARM3	0.006756	0.025518	0.934814	0.003122	0.12651	0.794574	0.184689	0.266163	0.255936
	M4	0.117696	0.083839	0.000061	0.054567	0.004209	0.000000	0.29350	0.000897	0.000021
	ARM4	0.017711	0.002318	0.000000	0.000729	0.000079	0.000000	0.009509	0.008970	0.001904
	VARM4	0.000002	0.000001	0.000000	0.000003	0.000000	0.000000	0.000000	0.000001	0.000000
2017	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000003	0.000000	0.000000	0.009882	0.000807	0.000000
	M2st	0.006681	0.025263	0.934247	0.000120	0.000615	0.603163	0.000006	0.000036	0.155612
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.097586	0.252834	0.992890	0.067621	0.183640	0.968604	0.180970	0.000038	0.000008
	VARM2	0.097586	0.252834	0.992952	0.175408	0.035337	0.005011	0.107839	0.003364	0.011703
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.097586	0.252834	0.989362	0.018707	0.062327	0.912695	0.573304	0.783853	0.013882
2018	ARM3	0.097586	0.252834	0.990810	0.018707	0.062327	0.914305	0.414556	0.622290	0.120020
	VARM3	0.006681	0.025263	0.934247	0.003063	0.012432	0.785032	0.942492	0.981293	0.000045
	M4	0.048753	0.003191	0.946866	0.342897	0.039009	0.009791	0.687712	0.618925	0.184010
	ARM4	0.006054	0.001414	0.000000	0.200579	0.040330	0.018546	0.191077	0.077298	0.090349
	VARM4	0.006054	0.001414	0.000000	0.026521	0.002827	0.000000	0.526512	0.010649	0.000098

Table 21: Coverage tests for Lower Quantiles of HP12. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_{25}			Q_{50}			Q_{75}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.124615	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.194493	0.000000	0.000000	0.004950	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.371055	0.000001	0.000000	0.022693	0.000000	0.000000	0.000997	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.063756	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.002501	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.830988	0.000000	0.000000	0.006855	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.347979	0.000000	0.000000	0.003537	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.003537	0.002722	0.000000	0.076328	0.095002	0.000000
	M3	0.124688	0.000000	0.000000	0.152811	0.000000	0.000000	0.000001	0.000000	0.000000
2012	ARM3	0.058685	0.000000	0.000000	0.711017	0.000000	0.000000	0.000001	0.000000	0.000000
	VARM3	0.024857	0.000000	0.000000	0.491384	0.000000	0.000000	0.000069	0.000000	0.000000
	M4	0.003241	0.000000	0.000000	0.368170	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.002214	0.000000	0.000000	0.185609	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000997	0.000000	0.000000	0.266242	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.008854	0.000001	0.000020	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000556	0.000000	0.000000	0.752801	0.000017	0.000000	0.015538	0.000001	0.000000
	M2st	0.015538	0.000022	0.000002	0.006475	0.000000	0.000013	0.000003	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.011984	0.000000	0.000023	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.021296	0.000004	0.000001	0.000000	0.000000	0.000000
	ARM2	0.011241	0.000002	0.000000	0.834382	0.000002	0.000000	0.000036	0.000000	0.000000
	VARM2	0.011241	0.000004	0.000000	0.530459	0.000104	0.000000	0.000036	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.173952	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.002743	0.000001	0.000000	0.008854	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.085083	0.000012	0.000000	0.006475	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.050541	0.000482	0.000000	0.001160	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000022	0.000000	0.000000	0.753801	0.000272	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000008	0.000000	0.000000	0.530459	0.000041	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.916739	0.002957	0.004808	0.000000	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.793540	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.637565	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.958256	0.000000	0.000000	0.001676	0.000000	0.000000
	M2st	0.000019	0.000000	0.000000	0.714061	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.793540	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000011	0.000000	0.000000	0.104463	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000001	0.000000	0.000000	0.271553	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.228484	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.002164	0.000000	0.000000	0.000292	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.001676	0.000000	0.000000	0.000193	0.000000	0.000000	0.000000	0.000000	0.000000
2015	VARM3	0.000082	0.000000	0.000000	0.005452	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000001	0.000000	0.000000	0.432309	0.000004	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000001	0.000000	0.000000	0.319867	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.496173	0.000011	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000292	0.000001	0.000002	0.000019	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.489975	0.000008	0.000000	0.000052	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.061428	0.000003	0.000000	0.000646	0.000005	0.000036	0.001566	0.000007	0.000211
	M2jsu	0.000000	0.000000	0.000000	0.000126	0.000002	0.000011	0.000000	0.000000	0.000000
	ARM2	0.103177	0.000000	0.000000	0.0002793	0.000038	0.000000	0.000000	0.000000	0.000000
2016	VARM2	0.079241	0.000000	0.000000	0.000008	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.319867	0.000000	0.000000	0.010249	0.000000	0.000000
	M3	0.489975	0.000008	0.000000	0.000052	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.651793	0.000000	0.000000	0.000052	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.262430	0.000000	0.000000	0.005452	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.026430	0.000001	0.000000	0.190510	0.000000	0.000000	0.000002	0.000000	0.000000
	ARM4	0.001676	0.000000	0.000000	0.041022	0.000000	0.000000	0.000004	0.000000	0.000000
	VARM4	0.001676	0.000000	0.000000	0.319867	0.000000	0.000000	0.000002	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.157394	0.008526	0.001411	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2017	QReg	0.489975	0.018289	0.000039	0.714061	0.717797	0.012691	0.003580	0.006926	0.000003
	M2st	0.000007	0.000035	0.000000	0.052583	0.042368	0.001929	0.785058	0.753309	0.000176
	M2jsu	0.000000	0.000000	0.000000	0.083902	0.058584	0.000400	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.001968	0.008298	0.013985	0.000000	0.000000	0.000000
	ARM2	0.446381	0.079471	0.024716	0.958256	0.054159	0.000097	0.376452	0.002840	0.000055
	VARM2	0.693254	0.492014	0.024584	0.373476	0.023611	0.000007	0.522952	0.008893	0.000466
	EGARCHst	0.000000	0.000000	0.000000	0.319867	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.353223	0.033368	0.000082	0.190510	0.000005	0.000000	0.002464	0.000000	0.000000
	ARM3	0.243885	0.001565	0.000000	0.875223	0.000000	0.000000	0.000749	0.000000	0.000000
	VARM3	0.000031	0.000000	0.000000	0.496173	0.000000	0.000000	0.002464	0.000000	0.000000
2018	M4	0.832815	0.205723	0.175900	0.958256	0.289652	0.042589	0.019498	0.000129	0.000002
	ARM4	0.651793	0.018469	0.050663	0.373476	0.146708	0.009501	0.005143	0.000031	0.000001
	VARM4	0.243885	0.001194	0.000195	0.637565	0.000133	0.000000	0.000042	0.019498	0.000073
	M1	0.000002	0.000000	0.000000	0.016056	0.000062	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.762107	0.005903	0.000034	0.094183	0.000000	0.000000	0.000056	0.000010	0.000000
	M2st	0.000340	0.000000	0.000000	0.016056	0.000273	0.000000	0.136365	0.001989	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.059663	0.001201	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.003358	0.004402	0.008863	0.000000</td		

Years	Models	Q_{95}			Q_{98}			Q_{99}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.131899	0.228489	0.189578	0.003639	0.014543	0.811099	0.007389	0.027654	0.939256
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000213	0.000001	0.000000	0.176038	0.000000	0.000000	0.238900	0.107340	0.000000
	VARM2	0.000213	0.000000	0.000000	0.093767	0.000000	0.000000	0.238900	0.107340	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.009491	0.018702	0.000000	0.238900	0.450950	0.000008
	M3	0.00018	0.000000	0.000000	0.000195	0.000001	0.000000	0.015421	0.042117	0.000000
2012	ARM3	0.00008	0.000000	0.000000	0.001519	0.000000	0.000000	0.106849	0.079249	0.000000
	VARM3	0.000939	0.000000	0.000000	0.009491	0.000260	0.000000	0.001520	0.000755	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.867459	0.006110	0.000071	0.358469	0.002189	0.000000	0.720414	0.042658	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.002832	0.000001	0.000000	0.109291	0.000269	0.000000	0.119098	0.000001	0.000000
	VARM2	0.009882	0.000005	0.000000	0.342897	0.000109	0.000000	0.119098	0.000001	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.277538	0.000461	0.000000	0.904210	0.016419	0.000000	0.504858	0.002671	0.000000
	ARM3	0.389106	0.000299	0.000000	0.904210	0.016419	0.000000	0.860310	0.010119	0.000000
	VARM3	0.000161	0.000004	0.000000	0.802449	0.000016	0.000000	0.504858	0.002671	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000069	0.000289	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000178	0.000828	0.134085	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.184689	0.085040	0.003581	0.03122	0.012651	0.740891	0.006756	0.025518	0.934814
	VARM2	0.030817	0.052103	0.033297	0.03122	0.012651	0.792868	0.006756	0.025518	0.934814
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.005157	0.000812	0.000001	0.686312	0.185837	0.004056	0.006756	0.025518	0.934814
2015	ARM3	0.001381	0.000478	0.000000	0.177586	0.385466	0.034533	0.006756	0.025518	0.934814
	VARM3	0.000683	0.000104	0.000000	0.190925	0.063250	0.833024	0.006756	0.025518	0.934814
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000001	0.000003	0.000000	0.000000	0.000000	0.000000	0.000002	0.000004	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000013	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000007	0.000038	0.143988	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000039	0.000201	0.245507	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	VARM2	0.000039	0.000201	0.245598	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.014147	0.039241	0.518372	0.003122	0.012651	0.796835	0.006756	0.025518	0.934814
	ARM3	0.014147	0.039241	0.518013	0.003122	0.012651	0.797138	0.006756	0.025518	0.934814
	VARM3	0.000667	0.002767	0.214394	0.000123	0.000627	0.605281	0.006756	0.025518	0.934814
	M4	0.000013	0.000004	0.000000	0.000001	0.000003	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000005	0.000000	0.000000	0.000029	0.000052	0.000000	0.000000	0.000002	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000001	0.000000
2017	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000345	0.001066	0.000000	0.000092	0.000338	0.000000	0.000000	0.000034	0.000106
	M2st	0.000171	0.000798	0.047675	0.000120	0.000615	0.603163	0.006681	0.025263	0.934247
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000008	0.000015	0.000000
	ARM2	0.414556	0.236078	0.020227	0.175408	0.381984	0.995920	0.006681	0.025263	0.934247
2018	VARM2	0.180970	0.083389	0.006599	0.175408	0.381984	0.995911	0.006681	0.025263	0.934247
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.030336	0.017291	0.000011	0.003063	0.012432	0.794777	0.006681	0.025263	0.934247
	ARM3	0.030336	0.005067	0.000001	0.003063	0.012432	0.794100	0.007586	0.0252834	0.992455
	VARM3	0.127009	0.023772	0.000907	0.018707	0.062327	0.919986	0.006681	0.025263	0.934247
	M4	0.000073	0.000171	0.000000	0.001997	0.000870	0.000000	0.006054	0.011359	0.000426
	ARM4	0.000073	0.000171	0.000000	0.000009	0.000045	0.000000	0.000139	0.0000545	0.000000

Table 23: Coverage tests for Upper Quantiles of HP12. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_1			Q_2			Q_5		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.105957	0.269935	0.993641	0.003639	0.014543	0.811919	0.007528	0.010491	0.337196
	M2jsu	0.000420	0.000318	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.105957	0.269935	0.992711	0.195891	0.414024	0.061076	0.017930	0.027863	0.045510
	ARM2	0.105957	0.269935	0.992861	0.195891	0.414024	0.054235	0.216771	0.013229	0.001150
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.362478	0.653153	0.999797	0.077043	0.204213	0.978866	0.074224	0.023396	0.002994
	ARM3	0.822449	0.088426	0.000000	0.957655	0.290410	0.000011	0.216771	0.099144	0.016069
2012	VARM3	0.755217	0.928583	0.999995	0.657748	0.817914	0.999631	0.970981	0.572167	0.507888
	M4	0.000000	0.000000	1.000000	0.000000	0.000000	1.000000	0.000000	0.000000	1.000000
	ARM4	0.000000	0.000000	1.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.340027	0.627392	0.660684	0.904210	0.000004	0.000000	0.942492	0.000001	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	ARM2	0.0066681	0.025263	0.934247	0.067621	0.000035	0.000000	0.000643	0.000000	0.000006
	VARM2	0.0066681	0.025263	0.934247	0.018707	0.001015	0.000000	0.005637	0.000022	0.000089
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.260387	0.000059	0.000000	0.904210	0.000004	0.000000	0.526512	0.001788	0.000001
	ARM3	0.720414	0.000174	0.000000	0.175408	0.000413	0.000000	0.191077	0.000650	0.000000
	VARM3	0.720414	0.000174	0.000000	0.611055	0.007128	0.000028	0.277538	0.000333	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2014	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.006756	0.025518	0.934814	0.019025	0.0063250	0.903307	0.110310	0.000274	0.000163
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.006756	0.025518	0.934814	0.003122	0.012651	0.795806	0.184689	0.011080	0.000180
	VARM2	0.006756	0.025518	0.934814	0.003122	0.012651	0.794222	0.421216	0.005021	0.000782
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.342459	0.630242	0.999735	0.910088	0.016521	0.000432	0.952019	0.000308	0.000002
	ARM3	0.098483	0.254687	0.992894	0.177586	0.385466	0.991246	0.421216	0.005021	0.001317
2015	VARM3	0.098483	0.254687	0.993119	0.362159	0.095224	0.313273	0.761472	0.018236	0.000042
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.017711	0.047812	0.000008	0.000729	0.003300	0.000033	0.000005	0.000023	0.000001
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000029	0.000052	0.000000	0.005157	0.000812	0.000000
	M2st	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000000	0.000000	0.050568
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	ARM2	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000000	0.000000	0.026569
	VARM2	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.000000	0.000000	0.051838
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.001381	0.000006	0.000000
	ARM3	0.006756	0.025518	0.934814	0.003122	0.012651	0.772641	0.000000	0.000000	0.000000
	VARM3	0.006756	0.025518	0.934814	0.000123	0.000627	0.605281	0.123862	0.000001	0.000000
	M4	0.048057	0.118387	0.000007	0.338885	0.477187	0.066598	0.678752	0.913119	0.000021
	ARM4	0.005937	0.017125	0.000029	0.107486	0.181850	0.001892	0.952019	0.391156	0.001540
	VARM4	0.001823	0.004809	0.000000	0.011592	0.037040	0.000000	0.123862	0.019919	0.000001
	M1	0.000002	0.000000	0.000001	0.000000	0.000000	0.000000	0.000032	0.000007	0.000000
2017	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000045	0.000001	0.000000
	M2st	0.006681	0.025263	0.934247	0.000120	0.000615	0.603163	0.000001	0.000005	0.092709
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.097586	0.252834	0.992478	0.003063	0.012432	0.795178	0.000006	0.000002	0.012444
	VARM2	0.097586	0.252834	0.992478	0.003063	0.012432	0.795202	0.000001	0.000003	0.092210
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.006681	0.025263	0.934247	0.000120	0.000615	0.603163	0.000001	0.000005	0.088045
	ARM3	0.006681	0.025263	0.934247	0.000120	0.000615	0.603163	0.000643	0.000620	0.192393
2018	VARM3	0.006681	0.025263	0.934247	0.000120	0.000615	0.603163	0.000037	0.000193	0.055334
	M4	0.504858	0.746917	0.757009	0.904210	0.865782	0.470460	0.277538	0.077067	0.000008
	ARM4	0.048753	0.119886	0.019862	0.544678	0.662973	0.228152	0.127009	0.072384	0.000159
	VARM4	0.000139	0.000545	0.000000	0.000270	0.001313	0.000000	0.000718	0.000348	0.000000

Table 24: Coverage tests for Lower Quantiles of HP19. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_{25}			Q_{50}			Q_{75}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.033491	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.018233	0.000000	0.000000	0.223334	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.003312	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.560413	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.001210	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.057060	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2012	ARM3	0.000189	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.371055	0.000003	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000276	0.000000	0.000000	0.633821	0.000000	0.000000	0.000001	0.000000	0.000000
	ARM4	0.098125	0.000000	0.000000	0.957791	0.000000	0.000000	0.000042	0.000000	0.000000
	VARM4	0.058685	0.000000	0.000000	0.185609	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000149	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.008039	0.000000	0.000000	0.530459	0.000000	0.000000	0.021231	0.000000	0.000000
	M2st	0.360730	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.402892	0.000002	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.046803	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000012	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000003	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.170675	0.000000	0.000000
	M3	0.008039	0.000000	0.000000	0.036356	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.001873	0.000000	0.000000	0.094183	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.000362	0.000000	0.000000	0.173952	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.753801	0.000000	0.000000	0.005683	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.916739	0.000001	0.000016	0.001873	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.916739	0.000000	0.000000	0.000058	0.000000	0.000000
2014	M1	0.000000	0.000000	0.000000	0.010209	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000002	0.000000	0.000000	0.432309	0.000006	0.000000	0.000000	0.000000	0.000000
	M2st	0.257869	0.000773	0.001197	0.2024238	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.128831	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.958256	0.000013	0.000000	0.000000	0.000000	0.000000
	ARM2	0.522952	0.000005	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.522952	0.000001	0.000000	0.000052	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.001127	0.000000	0.000000
	M3	0.001127	0.000000	0.000000	0.190510	0.000000	0.000000	0.000000	0.000000	0.000000
2015	ARM3	0.003580	0.000000	0.000000	0.104463	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.003580	0.000000	0.000000	0.190510	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.000031	0.000000	0.000000	0.319867	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000051	0.000000	0.000000	0.271553	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.001127	0.000000	0.000000	0.128831	0.000002	0.000000	0.000000	0.000000	0.000000
	M1	0.000001	0.000000	0.000000	0.003923	0.000000	0.000001	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.079521	0.000000	0.000000	0.793540	0.000000	0.000000	0.295113	0.000000	0.000000
	M2st	0.000992	0.000001	0.000002	0.001968	0.000000	0.000000	0.003580	0.000004	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000033	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.005452	0.015702	0.050337	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000005	0.000000	0.000000	0.243885	0.000001	0.000000
	M3	0.035416	0.000000	0.000000	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM3	0.128779	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.001676	0.000000	0.000000	0.000126	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.023767	0.000035	0.000002	0.000126	0.000003	0.000000	0.000001	0.000000	0.000000
	ARM4	0.167642	0.145795	0.000038	0.000292	0.000042	0.000000	0.000001	0.000000	0.000000
	VARM4	0.879714	0.001767	0.000023	0.000646	0.000661	0.000000	0.000001	0.000000	0.000000
2016	M1	0.056165	0.000185	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.056583	0.000408	0.000002	0.011984	0.000004	0.000000	0.097193	0.000000	0.000011
	M2st	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.834382	0.01399	0.000080	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000027	0.000076	0.008786	0.000000	0.000000	0.000011
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM2	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000340	0.000000	0.000000
	M3	0.000022	0.000000	0.000001	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM3	0.007523	0.002866	0.000004	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M4	0.299639	0.002765	0.000013	0.601155	0.000010	0.000000	0.015538	0.000000	0.000000
	ARM4	0.671323	0.003434	0.000001	0.143112	0.000000	0.000000	0.000556	0.000000	0.000000
	VARM4	0.309641	0.000049	0.000073	0.173952	0.000016	0.000008	0.003971	0.000003	0.000000

Table 25: Coverage tests for Middle Quantiles of HP19. P-values of Coverage Tests: UC, CC and DC are respectively Kupec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests. Only EGARCH-st have been reported since they provided the same values as those of EGARCH-sst.

Years	Models	Q_{95}			Q_{98}			Q_{99}		
		UC	CC	DC	UC	CC	DC	UC	CC	DC
2011	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000453	0.000000	0.000000	0.392781	0.102259	0.001235	0.105957	0.269935	0.991296
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.001520	0.000000	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000420	0.000000	0.000000
	EGARCHst	0.074224	0.002021	0.000620	0.392781	0.102259	0.001184	0.822449	0.088426	0.001801
	M3	0.003671	0.000001	0.000000	0.957655	0.000409	0.000000	0.473190	0.000019	0.000000
2012	ARM3	0.000213	0.000016	0.000000	0.749709	0.354551	0.014937	0.362478	0.653153	0.000000
	VARM3	0.000939	0.000109	0.000000	0.093767	0.161552	0.003770	0.473190	0.720013	0.000001
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000	0.000000	0.000000	1.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000073	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000345	0.000000	0.000000	0.011879	0.000001	0.000000	0.001864	0.000000	0.000000
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2013	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000753	0.000000	0.000000	0.048753	0.000002	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000092	0.000000	0.000000	0.000139	0.000000	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.017599	0.000000	1.000000	0.802449	0.810113	1.000000	0.097586	0.252834	1.000000
	ARM3	0.001447	0.000001	1.000000	0.611055	0.794848	1.000000	0.097586	0.252834	1.000000
	VARM3	0.000032	0.000000	0.000000	0.200579	0.273041	0.987808	0.504858	0.746917	0.999937
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000002	0.000000	1.000000	0.000000	0.000000	1.000000	0.000000	0.000000	1.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	1.000000
2014	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2st	0.000000	0.000000	0.000000	0.197757	0.048298	0.002166	0.098483	0.254687	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000515	0.000370	0.000000
	VARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000135	0.000018	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.005937	0.012184	0.000015
2015	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.048057	0.051865	0.001088
	VARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.0001823	0.000858	0.000000	0.000000
	M4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	VARM4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.005157	0.000037	0.000000	0.000000	0.000003	0.000001	0.000000	0.000000	0.000000
	M2st	0.079311	0.004414	0.000000	0.362159	0.095224	0.002892	0.006756	0.025518	0.934814
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2016	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.501312	0.000019	0.000000
	VARM2	0.000000	0.000000	0.000000	0.338885	0.048191	0.000000	0.342459	0.010290	0.000000
	EGARCHst	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000002	0.000009	0.000000
	M3	0.000000	0.000000	0.000000	0.338885	0.048191	0.000000	0.342459	0.630242	0.976142
	ARM3	0.000000	0.000000	0.000000	0.000260	0.000000	0.000000	0.856117	0.940933	0.001518
	VARM3	0.000153	0.000000	0.000000	0.003122	0.012651	0.796569	0.006756	0.025518	0.934814
	M4	0.000005	0.000003	0.000000	0.000003	0.000003	0.000001	0.000000	0.000000	0.000000
	ARM4	0.000005	0.000007	0.000000	0.000003	0.000003	0.000014	0.000000	0.000000	0.000000
	VARM4	0.000013	0.000012	0.000000	0.000003	0.000003	0.000000	0.000000	0.000000	0.000000
2017	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.526512	0.061445	0.000726	0.000753	0.000555	0.000000	0.000139	0.000019	0.000000
	M2st	0.414556	0.036122	0.177217	0.611055	0.007128	0.000004	0.006681	0.025263	0.934247
	M2jsu	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M2Ser	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000001	0.000000
	ARM2	0.000000	0.000000	0.000000	0.026521	0.000327	0.000000	0.720414	0.914928	0.999998
	VARM2	0.389106	0.083226	0.000001	0.904210	0.000377	0.000000	0.860310	0.010119	0.000000
	EGARCHst	0.000032	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M3	0.191077	0.000073	0.000000	0.067621	0.008562	0.010455	0.006681	0.025263	0.934247
2018	ARM3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.007586	0.252834	0.993023
	VARM3	0.127009	0.000000	0.000000	0.018707	0.062327	0.886865	0.006681	0.025263	0.934247
	M4	0.005374	0.000028	0.000000	0.001997	0.000125	0.000000	0.001864	0.000054	0.000000
	ARM4	0.000345	0.000065	0.000000	0.000009	0.000001	0.000000	0.000008	0.000000	0.000000
	VARM4	0.000032	0.000002	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	M1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	ARM1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
	QReg	0.526512	0.061445	0.000726	0.000753	0.000555	0.000000	0.000139	0.000019	0.000000
	M2st	0.414556	0.036122	0.177217	0.611055	0.007128	0.000004	0.006681		

Table 27: Coverage tests for HP8. P-values of Coverage Tests: UC, CC and DC are respectively Kupiec's unconditional coverage and Engle and Manganelli's Dynamic Quantile tests.

		Q_{11}	Q_{21}	Q_{51}	Q_{25}	Q_{50}	Q_{75}	Q_{95}
2011	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.36	0.65	0.66	0.78	0.80	0.87	0.90
	M2su	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARIM2	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	VARM2	0.11	0.27	0.27	0.39	0.66	0.75	0.94
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.11	0.27	0.27	0.39	0.66	0.75	0.94
	VARM3	0.11	0.27	0.27	0.39	0.66	0.75	0.94
2012	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.86	0.86	0.90	0.92	0.98	0.99	0.99
	M2su	0.26	0.00	0.00	0.00	0.00	0.00	0.00
	ARIM2	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	VARM2	0.12	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	VARM3	0.00	0.00	0.00	0.00	0.00	0.00	0.00
2013	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.72	0.92	1.00	0.91	0.28	0.75	0.75
	M2su	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARIM2	0.34	0.63	0.72	0.72	0.07	0.92	0.95
	VARM2	0.50	0.74	0.74	0.89	0.00	0.93	0.93
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.05	0.26	0.38	0.50	0.56	0.56	0.56
	VARM3	0.22	0.36	0.56	0.56	0.00	0.00	0.00
2014	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.34	0.63	0.63	0.68	0.02	0.88	0.88
	M2su	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARIM2	0.34	0.63	0.63	0.72	0.07	0.92	0.95
	VARM2	0.72	0.92	0.92	0.99	0.00	0.93	0.93
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.34	0.63	1.00	0.62	0.12	0.93	0.93
	VARM3	0.34	0.63	1.00	0.62	0.12	0.93	0.93
2015	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2su	0.34	0.63	0.63	0.68	0.02	0.88	0.88
	ARIM2	0.34	0.63	0.63	0.72	0.07	0.92	0.95
	VARM2	0.63	0.82	0.82	0.94	0.07	0.92	0.95
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.10	0.25	0.25	0.39	0.00	0.78	0.80
	VARM3	0.10	0.25	0.25	0.39	0.00	0.78	0.80
2016	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	QReg	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2st	0.34	0.63	0.63	0.68	0.02	0.88	0.88
	M2su	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARIM2	0.34	0.63	0.63	0.72	0.07	0.92	0.95
	VARM2	0.63	0.82	0.82	0.94	0.07	0.92	0.95
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM3	0.01	0.03	0.03	0.03	0.00	0.00	0.00
	VARM3	0.01	0.03	0.03	0.03	0.00	0.00	0.00

Table 28: Coverage tests for HP16. P-values of Coverage Tests: UC, CC and DC are respectively Kupiec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests.

		Q_1	Q_2	Q_3	Q_4	Q_{25}	Q_{50}	Q_{75}	Q_{98}
2011	M1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	Q1 _{0.95}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.95}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.999}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9995}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9999}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	VARM2	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
2012	EGARCHet	0.65	1.00	0.08	0.20	0.00	0.00	0.00	0.00
	EGARCHst	0.24	0.45	0.00	0.31	0.00	0.00	0.00	0.00
	M1	0.36	0.45	0.00	0.20	0.00	0.00	0.00	0.00
	ARM3	0.36	0.45	0.00	0.20	0.00	0.00	0.00	0.00
	ARM4	0.36	0.45	0.00	0.20	0.00	0.00	0.00	0.00
	VARM3	0.76	0.93	0.00	0.20	0.04	0.00	0.00	0.00
	VARM4	0.76	0.93	0.00	0.20	0.04	0.00	0.00	0.00
	M2 _{0.95}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
2013	EGARCHet	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M1	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	ARM1	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	Q1 _{0.95}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.95}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.999}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9995}	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00
2014	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	ARM1	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	Q1 _{0.95}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.95}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.999}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9995}	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
2015	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M1	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	ARM1	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	Q1 _{0.95}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.95}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.999}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9995}	0.34	0.41	0.00	0.00	0.00	0.00	0.00	0.00
2016	EGARCHet	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	EGARCHst	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
	M1	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	ARM1	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	Q1 _{0.95}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.95}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.99}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.995}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.999}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00
	M2 _{0.9995}	0.86	0.94	0.00	0.00	0.00	0.00	0.00	0.00

Table 29: Coverage tests for HP23. P-values of Coverage Tests: UC, CC and DC are respectively Kupiec's unconditional coverage, Christoffesen's conditional coverage and Engle and Manganelli's Dynamic Quantile tests.