

# **ARMA-GARCH models via Kalman filter in Gretl**

**(Conditional Heteroscedastic State Space Modelling in Gretl)**

by

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## A Conditional Heteroscedastic State Space Model (CHES-Model):

$$\mathbf{y}_t = \mathbf{A}'_t \mathbf{x}_t + \mathbf{H}'_t \boldsymbol{\xi}_t + \mathbf{w}_t \quad \mathbf{w}_t \sim VWN(\mathbf{R}_t)$$

$$\boldsymbol{\xi}_{t+1} = \mathbf{F}_t \boldsymbol{\xi}_t + \mathbf{v}_t \quad \mathbf{v}_t \sim VWN(\mathbf{Q}_t)$$

$$\mathbf{R}_t = h_t \mathbf{R}$$

$$\mathbf{Q}_t = h_t \mathbf{Q}$$

$$h_t = \omega + \alpha \mathbf{e}'_{t-1} \mathbf{e}_{t-1} + \beta h_{t-1}$$

## Example1: Heteroscedastic local level model

$$y_t = \xi_t + w_t \quad w_t \sim WN(R_t)$$

$$\xi_{t+1} = \xi_t + v_t \quad v_t \sim WN(Q_t)$$

$$R_t = h_t R$$

$$Q_t = h_t Q$$

fixing  $Q = 1 \implies$

$$Q_t = \omega + \alpha e_{t-1}^2 + \beta Q_{t-1}$$

$$R_t = R \cdot Q_t = Q_t / \lambda$$

$\lambda$  is the signal-noise ratio

## Example2: ARMA(1,1)-GARCH(1,1) Model

$$y_t = \mu + \xi_t + \theta_1 \xi_{t-1} \quad \varepsilon_t \sim WN(h_t)$$
$$\xi_t = \phi_1 \xi_{t-1} + \varepsilon_t \quad h_t = \omega + \alpha \varepsilon_{t-1}^2 + \beta h_{t-1}$$

$$y_t = \mathbf{A} + \mathbf{H}' \boldsymbol{\xi}_t$$
$$\boldsymbol{\xi}_t = \mathbf{F} \boldsymbol{\xi}_{t-1} + \mathbf{v}_t$$
$$\mathbf{Q}_t = h_t \mathbf{Q}$$
$$\mathbf{A} = [\mu] \quad \mathbf{H} = \begin{bmatrix} 1 \\ \theta_1 \end{bmatrix} \quad \boldsymbol{\xi}_t = \begin{bmatrix} \xi_t \\ \xi_{t-1} \end{bmatrix}$$
$$\mathbf{F} = \begin{bmatrix} \phi_1 & 0 \\ 1 & 0 \end{bmatrix} \quad \mathbf{v}_t = \begin{bmatrix} \varepsilon_t \\ 0 \end{bmatrix} \quad \mathbf{Q} = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$$

## The Kalman filter in Gretl

- The Kalman filter is a filter that allows to forecast time series from State Space Models
- The Kalman filter is defined in Gretl by a block of commands defining the matrices of the State Space Model:

**kalman**

```
obsy y
obsymat H
statemat F
statevar Qt ; modify_q(&Qt, ...)
...
```

**end kalman**

- If a matrix is time-varying, a function modifying that matrix has to be included

## The function `modify_q`

```
function void modify_q(matrix *Qt, scalar *ht, matrix Q,  
scalar omega, scalar alpha, scalar beta)  
  
    if $kalman_t=1  
        ht = omega / (1-alpha-beta)  
    else  
        ht = omega + alpha*$kalman_uhat^2 + beta*ht  
    endif  
    Qt = ht*Q  
  
end function
```

## MLE of the model parameters

```
mle ll = ERR ? NA : $kalman_llt
...
Qt = omega/(1-alpha-beta)*Q
...
params ... omega alpha beta
end mle
```

**Note** It is sufficient to include a convenient, not necessary true, relation between  $Q_t$  and its parameters, because the right value is recalculated by `modify_q`

## A comparative study

- 1) 300 observations was simulated according to the following ARMA-GARCH model:

$$(1 - 0.75B)(y_t - 1.2) = (1 + 0.5B)\varepsilon_t$$

$$\varepsilon_t \sim WN(h_t) \quad h_t = 0.4 + 0.07\varepsilon_{t-1}^2 + 0.85h_{t-1}$$

- 2) The model was estimated either using the Kalman filter and the conditional log-likelihood



# Results

	<i>Kalman filter</i>	<i>Cond. likelihood</i>
<b>param</b>	<b>estimate sign</b>	<b>estimate sign</b>
mu	1.201 ***	1.168 **
phi	0.647 ***	0.660 ***
theta	0.387 ***	0.408 ***
omega	0.247	0.423 ***
alpha	0.086 **	0.112 ***
beta	0.859 ***	0.798 ***
log.ver	-642.8	-1'306.8
AIC	1297.5	2'625.5
BIC	1319.8	2'647.8

## Conclusions

- The Kalman Filter ensures exact maximum likelihood estimates
- Its implementation in Gretl is possible but:
  - in case of non-gaussian disturbances, Kalman filter provides QML estimates;
  - it is quite complex to perform garch(p, q) with  $p, q > 1$
  - the estimate algorithm is quite slow

Thank you  
for your attention!