

Winter Term 2019/2020

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# Multivariate Time Series Analysis

## Solution Exercise Sheet 8

### 1 Exercise 1: Forecast Intervals and Distributional Assumptions

- a) Which key assumption about the innovations  $a_t$  is made in the lecture to derive the distribution of the forecast errors  $e_T(h)$ ?

*Solution:*

$$a_t \stackrel{\text{i.i.d.}}{\sim} N(0, \Sigma_a)$$

- i.i.d  $\Rightarrow$  no autocorrelation
- $N$   $\Rightarrow$  normal distributed
- $\Sigma_a \Rightarrow$  heteroskedasticity

- b) Assume we knew all parameters / coefficients and let  $\Sigma_a$  be the identity matrix  $I_{3 \times 3}$ . Based on the assumption from a), derive the distribution of  $e_T(1)$  for any stationary VAR( $p$ ).

*Solution:*

$$\begin{aligned} e_T(1) &= z_{T+1} - z_T(1) \\ &= a_{T+1} \end{aligned}$$

holds for any VAR( $p$ ) since

$$z_T(1) = \mathbb{E}(z_{T+1} | z_T, \dots, z_0)$$

$$\Rightarrow e_t(1) = a_{T+1} \sim N(0, I_{3 \times 3})$$

- c) Derive the confidence ellipsoid associated to b) for  $\alpha = 5\%$ . What is the fraction forecast errors that lie inside the ellipsoid?

*Solution:*

ellipsoid:

$$\left\{ z \in \mathbb{R}^3 : (z_T(1) - z)' \Sigma_e^{-1}(1) (z_T(1) - z) \leq \chi_{3,1-\alpha}^2 \right\}$$

By defining  $z_T(1) - z =: \epsilon$  and using that  $\Sigma_e(1) = \Sigma_a = I_{3 \times 3}$  the ellipsoid is:  $\left\{ \epsilon \in \mathbb{R}^3 : \epsilon' \epsilon \leq \chi_{3,1-\alpha}^2 \right\}$ . For  $\alpha = 5\%$ , 95% of the observed forecast errors are expected to fall inside the confidence ellipsoid.

- d) Run a simulation in 'R': Draw the forecast error  $e_T(1)$  defined in a) and b) with  $K = 3$ . Check if it is located inside or outside the confidence ellipsoid derived in c). Use  $N = 10000$  repetitions in total and conclude whether the confidence ellipsoid is appropriate.

*Solution:*

Just check if  $e_T(1)' e_T(1) \leq \chi_{3,0.95}^2$  and compute  $\frac{1}{N} \sum_{i=1}^N \mathbb{1} \left( e_T(1)' e_T(1) \leq \chi_{3,0.95}^2 \right)$ .

```
N <- 10000 # number of repetitions
K <- 3 # dimension of VAR
# drawing a_t from iid N(0,I)
gauss <- mvrnorm(n = N, mu = c(0,0,0), Sigma = diag(K))
# computing e'e for all draws in one take equals diag(E'E')
msfe_gauss <- diag(gauss %*% t(gauss))
# this is only a one-sided test since we have squared each error!
limit <- qchisq(p = 0.95, df=3, lower.tail=TRUE)
sum(msfe_gauss < limit) / N
```

```
## [1] 0.9523
```

- e) Repeat the simulation from above but this time assume  $a_t$  to be drawn from a uniform distribution.  $\Sigma_a = I_{3 \times 3}$  remains. How reliable is the confidence ellipsoid in this case?  
Hint: Set  $\pm \frac{\sqrt{12}}{2}$  as lower / upper bound for unit variance.

*Solution:*

```
# variance = 1 again, Kurtosis is < 3 for this one
unif <- matrix(data = runif(n = N * 3, min = -sqrt(12)/2,
                           max = sqrt(12)/2), nrow = N, ncol = 3)
msfe_unif <- rowSums(unif^2)
sum(msfe_unif < limit) / N
```

```
## [1] 0.9984
```

It is too conservative.

- f) Repeat the simulation drawing innovations from a  $t$ -distribution with 2 degrees of freedom and conclude.

*Solution:*

```
# variance = 1 by default, kurtosis > 3 and this hurts a lot
t2 <- matrix(data = rt(n = N * 3, df = 2), nrow = N, ncol = 3)
msfe_t2 <- rowSums(t2^2)
sum(msfe_t2 < limit) / N
```

```
## [1] 0.6439
```

Too liberal, the ellipsoid is not appropriate.