

## **ECOM073 Midterm test**

Wednesday, 27 March, 9:00am-10:00am (60 minutes)

**Place: PL-301 (Peter Landin teaching rooms)**

### **Midterm 2024**

Wednesday, 27 March, 9:00-10:00 am

- 1) Basic definitions: stationarity, white noise, i.i.d and others
- 2) Summary statistics: mean, variance, skewness, kurtosis, Jarque Bera test
- 3) Testing for absence of correlation
- 4) AR(p), MA(q) models, selection of order p, q
- 5) Checking the fit of the model, and significance of parameter estimates
- 6) AIC and BIC information criterions

**Test covers: Lecture 2-5, Problem Sets 2-5**

#### **Examples for preparation:**

Problem Set 2: 2.1

Problem Set 3: 3.1, 3.2,

Problem Set 4: 4.3, 4.5

Problem Set 5: 5.2

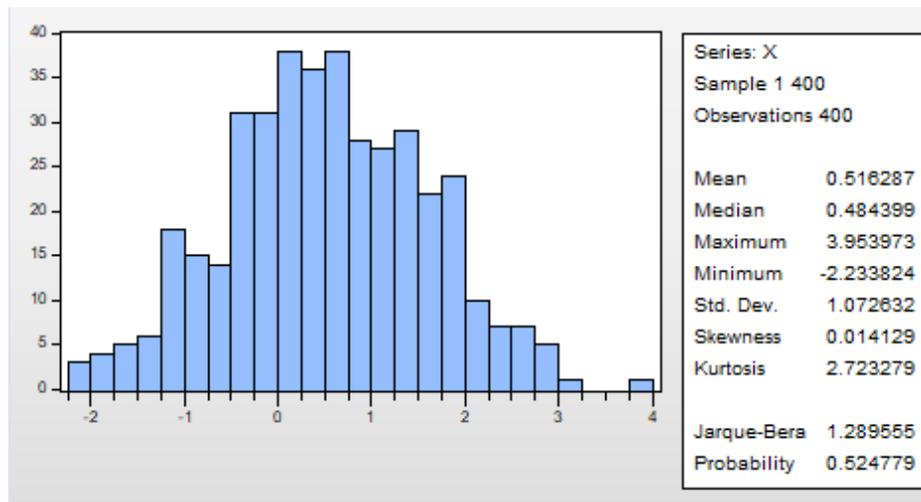
**See solutions of Quizzes 2-4 (QMPLUS, Week 7)**

## Solution Quiz 3

### Question 3.

The researcher computed summary statistics of the time series  $X$  using a sample containing  $N = 100$  observations, see the output below.

1. Test whether skewness  $S(X)$  equals 0.
2. Test whether kurtosis  $K(X)$  equals 3
3. Test whether  $X$  has normal distribution



### Solution of Question 3.

1. We test the hypothesis

$H_0: S(X) = 0$  against alternative  $H_1: S(X) \neq 0$ .

at 5% significance level.

We construct the test statistics:

$$t = \frac{\hat{S}(X)}{\sqrt{6/N}} = \frac{0.014129}{\sqrt{6/400}} = 0.11536$$

Under the null hypothesis,  $t \sim N(0, 1)$  is normally distributed.

Rule: we reject  $H_0$  at 5% significance level, if

$$|t| \geq 2.$$

In our case  $|t| = 0.11536 < 2$ . Hence, the test shows that there no evidence in the data to reject the null hypothesis of zero skewness  $S(X) = 0$ .

2. Next we test at 5% significance level the hypothesis:

$H_0: K(X) - 3 = 0$  against alternative  $H_1: K(X) \neq 3$ .

We use the test statistics:

$$t = \frac{\hat{K}(X) - 3}{\sqrt{24/N}} = \frac{2.723 - 3}{\sqrt{24/400}} = -1.1308.$$

By theory, under null hypothesis,  $t \sim N(0, 1)$  is normally distributed. Therefore the testing rule is similar as for testing skewness:

Rule: reject  $H_0$  at 5% significance level, if

$$|t| \geq 2.$$

In our case  $|t| = 1.1308 < 2$ . Hence, the test does not rejects the null hypothesis of zero kurtosis  $K(X) = 3$ .

3. Jargue-Bera test is used to test the hypothesis:

$H_0: S(X) = 0$  and  $K(X) - 3 = 0$  ("normal distribution")

against alternative

$H_1: S(X) \neq 0$  or  $K(X) - 3 \neq 0$  ("distribution is not normal").

Normal distribution has  $S(X) = 0$  and  $K(X) = 3$ . Thus, in case of normal distribution, test will not reject  $H_0$ .

Test will reject  $H_0$  if either skewness is not 0 or if kurtosis is not 3. That will indicate that distribution is not normal.

Since the  $p = 0.5247$  value of Jargue-Bera test is larger than 0.05 we do not reject the asymptotic normality at significance level 5%.

## Solution Quiz 3

**Question 1.** a) Suppose that

$$X_t = \varepsilon_t + t, \quad t = 1, 2, \dots$$

where  $\varepsilon_t$  is a white noise sequence with zero mean and variance  $E\varepsilon_t^2 = 1$ . Investigate whether time series  $X_t$  is covariance stationary.

b) Suppose that

$$X_t = t\varepsilon_t, \quad t = 1, 2, \dots$$

where  $\varepsilon_t$  is a white noise sequence with zero mean and variance  $E\varepsilon_t^2 = 1$ . Investigate whether time series  $X_t$  is covariance stationary.

### Solution of Question 1.

Time series  $X_t$  is a covariance stationary time series if it satisfies three properties:

- $EX_t = \mu$  for all  $t$  (does not depend on  $t$ );
- $Var(X_t) = \sigma^2$  for all  $t$  (does not depend on  $t$ );
- $Cov(X_t, X_{t-k}) = \gamma_k$  - covariance function depends only on the lag  $k$  and does not depend on  $t$ .

a) We have

$$EX_t = E[\varepsilon_t + t] = E[\varepsilon_t] + t = 0$$

since by assumption  $E[\varepsilon_t] = 0$ . We see that mean  $EX_t$  changes with  $t$ . Therefore, this time series is not covariance stationary.

b) We have

$$EX_t = E[t\varepsilon_t] = tE[\varepsilon_t] = 0$$

since by assumption  $E[\varepsilon_t] = 0$ . Hence, mean  $EX_t$  does not change with  $t$ .

Next we compute the variance:

$$\text{Var}(X_t) = E(X_t - EX_t)^2 = EX_t^2 = E[(t\varepsilon_t)^2] = t^2 E[\varepsilon_t^2] = t^2.$$

since by assumption  $E[\varepsilon_t^2] = 1$ . We see that the variance  $\text{Var}(x_t)$  changes with  $t$ . Therefore, this time series is not covariance stationary.

**Question 2.** Explain why the following sequence

$$\rho_1 = 0.8, \quad \rho_2 = 0.5, \quad \rho_3 = \rho_1 + \rho_2, \quad \rho_4 = \rho_1 + \rho_2 + \rho_3, \dots$$

cannot be the auto-correlation function of a covariance stationary sequence.

**Solution of Question 2.**

Correlation function  $\rho_k$  at lag  $k$  of a covariance stationary time series  $X_t$  satisfies three properties:

- $\rho_0 = 1$
- $\rho_k = \rho_{-k}$  for any lag  $k = 1, 2, 3, \dots$
- $|\rho_k| \leq 1$  for any  $k$ .

We have that

$$\rho_3 = \rho_1 + \rho_2 = 0.8 + 0.5 = 1.3 > 1.$$

Therefore  $\rho_k$  cannot be correlation function of a covariance stationary time series.

**Question 3.** Using the following EViews correlogram of time series  $X_t$ , determine whether  $x_t$  is a white noise time series (series of uncorrelated random variables).

Correlogram of Y					
Date: 29/11/20 Time: 10:53					
Sample: 1 400					
Included observations: 400					
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob
		1 -0.023	-0.023	0.2196	0.639
		2 -0.032	-0.033	0.6458	0.724
		3 -0.036	-0.037	1.1600	0.763
		4 -0.057	-0.060	2.4792	0.648
		5 0.050	0.045	3.4877	0.625
		6 -0.076	-0.080	5.8704	0.438
		7 0.067	0.063	7.7087	0.359
		8 -0.009	-0.012	7.7392	0.459
		9 0.012	0.016	7.7940	0.555
		10 -0.041	-0.049	8.4897	0.581

### Brief solution of question 3.

(i) (**Testing for correlation using ACF.**) Time series is a white noise if it is serially uncorrelated, that is  $\rho_k = 0$  for  $k \geq 1$ . Hence, to test for white noise, we test the hypotheses

$H_0 : \rho_k = 0$  against alternative  $H_1 : \rho_k \neq 0$   
at each lag  $k = 1, 2, \dots$  at significance level 5%.

**Rule:** ACF  $\rho_k$  at lag  $k$  is significantly different from zero at 5% significance level if  $|\hat{\rho}_k| > 2/\sqrt{N}$ , where  $N$  is the number of observations.

If  $|\hat{\rho}_k| \leq 2/\sqrt{N}$ , then ACF at lag  $k$  is not significantly different from 0.

If time series is white noise, then we do not reject  $H_0$  for any  $k = 1, 2, \dots$

(ii) **Ljung-Box test.** This test can be also used to test for zero correlation. We select  $m = 1, 2, \dots$  and test the hypothesis

$H_0 : \rho_1 = \dots \rho_m = 0$  against alternative

$H_1 : \rho_j \neq 0$  for some  $j = 1, \dots, m$ .

We reject the  $H_0$  at 5% significance level, if  $p$ -value satisfies  $p < 0.05$ . If time series is white noise, then we do not reject  $H_0$  for any  $m = 1, 2, \dots$

(iii) We have  $2/\sqrt{N} = 2/\sqrt{400} = 0.1$ . Since  $|\rho_1| = 0.023 < 0.1$ ,  $|\rho_2| = 0.032 < 0.1$ , ..., and so on. We find that  $\rho_k$  is not significant at any lag  $k = 1, 2, \dots$  at 5% significance level, because  $|\rho_k| \leq 2/\sqrt{N} = 2/\sqrt{400} = 0.1$ . So, this time series is a white noise.

From the correlogram we see that  $p$ -values of Ljung Box test satisfy  $p > 0.05$  for all  $m = 1, \dots$ . So  $H_0$  is not rejected at any  $m$  and this time series is a white noise.

## Solution Quiz 4

### Question 1.

Using the following Eviews output determine order  $p$  of AR( $p$ ) model and order  $q$  of MA( $q$ ) model you would fit to the data.

Correlogram of Y						
Date: 04/10/20 Time: 09:39						
Sample: 1 625						
Included observations: 625						
Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob	
		1	-0.487	-0.487	149.22	0.000
		2	0.289	0.067	201.67	0.000
		3	0.012	0.232	201.77	0.000
		4	-0.016	0.074	201.94	0.000
		5	0.062	0.015	204.35	0.000
		6	-0.013	0.000	204.47	0.000
		7	-0.019	-0.051	204.70	0.000
		8	0.039	-0.001	205.68	0.000
		9	0.007	0.059	205.72	0.000
		10	-0.047	-0.033	207.13	0.000

Write down the model you would fit to the data.

Comment on how would you check the fit of your model to the data.

### Brief solution of Question 1

(i) To select the order  $p$  for AR( $p$ ) model, we use the sample PACF function. We test the hypothesis

$$H_0 : \rho_k = 0 \text{ against alternative } H_1 : \rho_k \neq 0$$

at lags  $k = 1, 2, \dots$  at significance level 5%, where  $\rho_k$  is the PACF function.

PACF  $\hat{\rho}_k$  at lag  $k$  is significantly different from 0 at 5% significance level, if  $|\hat{\rho}_k| > 2/\sqrt{N}$ , where  $N$  is the number of observations.

If  $|\hat{\rho}_k| \leq 2/\sqrt{N}$ , then PACF at lag  $k$  is not significantly different from 0.

**Rule:** we select for  $p$  the largest lag  $k$  at which the PACF is significant.

This rule can be used because PACF of the AR( $p$ ) model becomes 0 for  $k > p$ .

(ii) To select the order  $q$  of MA( $q$ ) model, we use the sample ACF function. We test the hypothesis

$$H_0 : \rho_k = 0 \text{ against alternative } H_1 : \rho_k \neq 0$$

at lags  $k = 1, 2, \dots$  at significance level 5%, where  $\rho_k$  is ACF function.

**Rule:** ACF  $\rho_k$  is significant at lag  $k$  at 5% significance level, if  $|\hat{\rho}_k| > 2/\sqrt{N}$ , where  $N$  is the number of observations.

If  $|\hat{\rho}_k| \leq 2/\sqrt{N}$ , then ACF at lag  $k$  is not significantly different from 0.

We select for  $q$  the largest lag  $k$  at which the ACF is significant.

(iii) We have  $2/\sqrt{N} = 2/\sqrt{625} = 0.08$ . The PACF is significant only at lag 1 and 3. Hence we would fit AR(3) model.

The ACF shows significant correlation at the lags 1 and 2. Hence we would fit MA(2) model.

From the two models AR(3) and MA(2) we select a simpler model MA(2) with smaller number of parameters which should to be fitted to the data.

(iv) According to above, we can fit MA(2) model  $Y_t = \phi_0 + \varepsilon_t + \theta_1 \varepsilon_{t-1} + \theta_2 \varepsilon_{t-2}$  where  $\varepsilon_t$  is white noise.

(v) This MA(2) model fits the data if residuals  $\hat{\varepsilon}_t$  are uncorrelated. We could use the correlogram of residuals to test whether residuals are uncorrelated.

**Question 2.**

Using AIC information criterion values obtained fitting an AR( $p$ ) model, select the order  $p$  of an AR model you would fit to the data:

$p$	0	1	2	3	4	5	6
AIC	3	-2.3	-2	-1.1	0.6	1.7	1.8

Write down equation of your AR( $p$ ) model.

**Solution of Question 2.** Using AIC information criterion we select the model which minimizes the AIC value. In this case the minimum value  $-2.3$  corresponds to  $p = 1$ , so AR(1) model should be fitted to the data.