

## Exam in Time Series Analysis, 6 credits

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Exam time:	8-12
Allowed:	Pocket calculator. Text book (Cryer & Chan: "Time Series Analysis- with applications in R"), notes allowed.
Examinator:	Lotta Hallberg.
Assisting teacher:	Per Sidén.
Grades:	Grades: Maximum is 20 points. A=19-20 points B=17-18.5 points C=14-16.5 points D=12-13.5 points E=10-11.5 points F=0-9.5 points

**- Provide a detailed report that shows motivation of the results.**

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- For the processes below, assume  $e_t$  is zero mean white noise with variance  $\sigma_e^2 = 1$ .
  - Show that the process  $Y_t = Y_{t-1} + e_t$  is not weakly stationary. 1p.
  - Consider the process  $Y_t = e_t - \frac{1}{4}e_{t-2}$ . Is it invertible? 1p.
  - Suppose  $Y_t = \beta_0 + \beta_1 \cdot t + e_t$  and  $W_t = Y_t - Y_{t-1}$ . For which real values of  $\beta_0$  and  $\beta_1$  is  $Y_t$  weakly stationary? For which real values of  $\beta_0$  and  $\beta_1$  is  $W_t$  weakly stationary? 2p.
- For the stationary process  $Y_t = 2 + \frac{1}{2}Y_{t-1} + e_t - \frac{1}{3}e_{t-1}$ , compute the mean and the covariance function.  $e_t$  is white noise with variance 1. 3p.
- Below you find three model specifications (a,b,c), three simulated time series (X,Y,Z) and three SAC/SPAC functions (1,2,3).  $e_t$  is white noise with unit variance. Your job is to connect each simulated time series with the model that generated it and also with the corresponding SAC/SPAC functions. Motivate your choices. 3p.
  - AR(1),  $(1 + 0.8B) W_t = e_t$ .
  - MA(1),  $W_t = (1 + 0.8B)e_t$ .
  - ARMA(1,1),  $(1 - 0.8B) W_t = (1 - 0.4B) e_t$

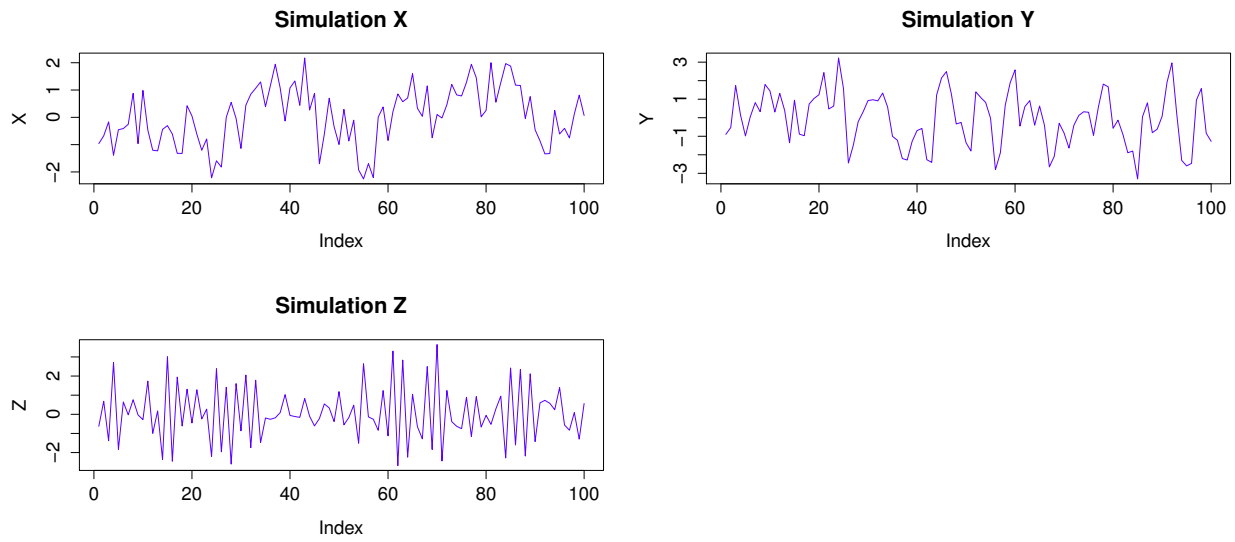


Figure 1: Simulated time series (X,Y,Z).

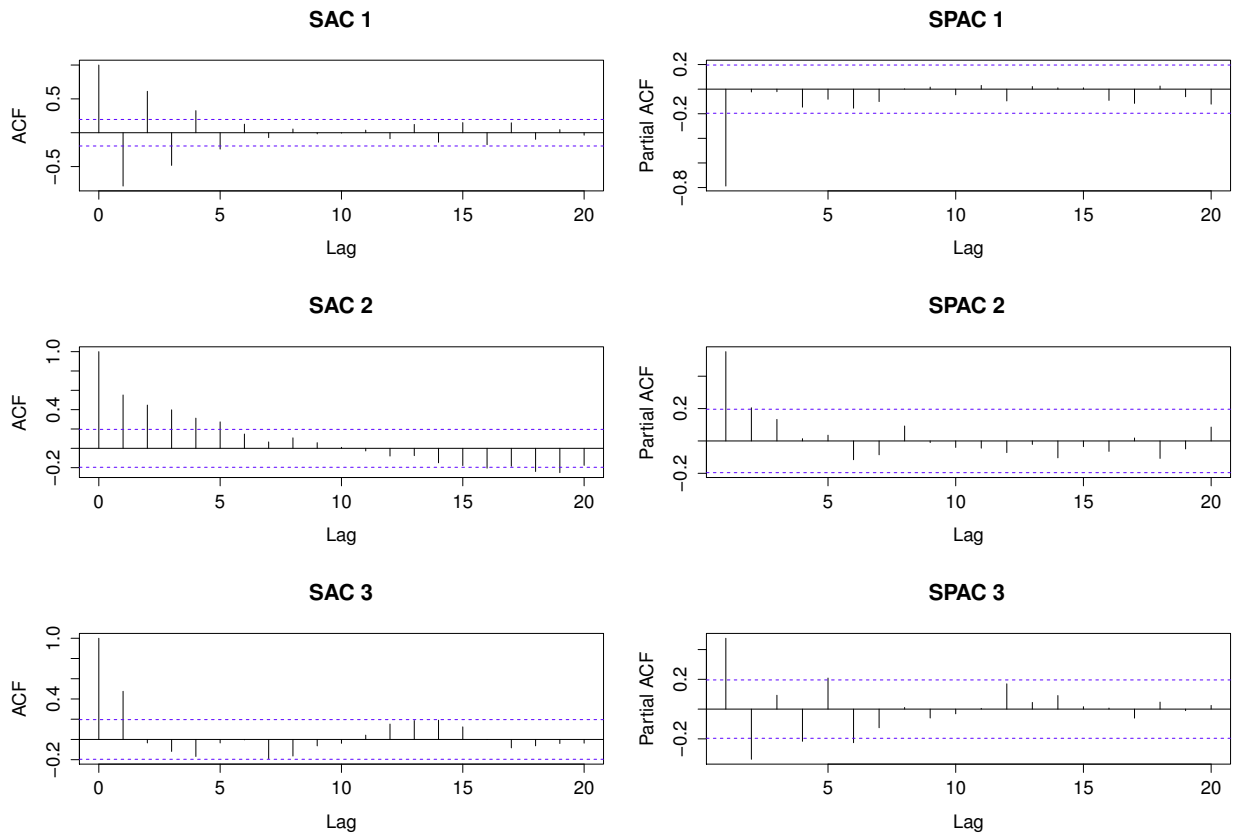


Figure 2: SAC/SPAC graphs (1,2,3).

4. The water-level  $Y_t$  in a lake has been measured once a year for 50 years. The level has been modelled with two different models. In the models below,  $e_t$  is white noise. Calculate the forecasts for  $Y_t$  the next two years. Two years for each of the models. 4p.

Model 1:  $Y_t = \beta_0 + \beta_1 \cdot t + N_t$ , where  $N_t = \phi_1 N_{t-1} + \phi_2 N_{t-2} + e_t$ .

Model 2:  $(1 - B)Y_t = W_t$ , where  $W_t = \rho_1 W_{t-1} + \rho_2 W_{t-2} + e_t$ .

$t$	$Y_t$	Residuals model 1	Residuals model 2
47		0.099	0.554
48	7.31	-0.376	0.159
49	8.43	0.724	1.406
50	6.67	-1.797	-1.686

Parameter	Estimate
$\beta_0$	9.32
$\beta_1$	-0.027
$\phi_1$	0.771
$\phi_2$	-0.239
$\rho_1$	0.035
$\rho_2$	-0.329

5. Consider the model  $Y_t = 4 - \frac{1}{2}Y_{t-1} + e_t + \frac{1}{4}e_{t-1}$ , where  $e_t$  is normally distributed white noise with variance 1. A simulation of the model with 200 values is shown in the graph below.

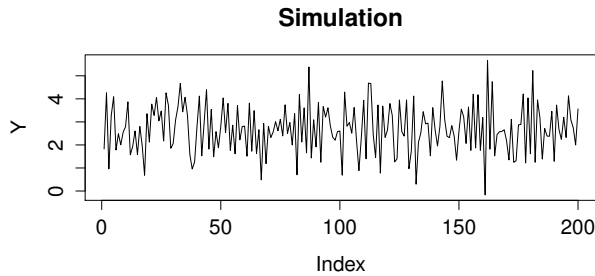


Figure 3: Simulation from the model.

- (a) Express the process as a  $MA(\infty)$ -process. 2p.

We use this simulated series to estimate the parameters. The result is given in the table below:

Parameter	Estimate	SE of estimate
$\mu$	2.699	0.061
$\theta$	-0.330	0.302
$\phi$	-0.531	0.270
$\sigma_e^2$	0.996	

- (b) Estimate the constant  $\delta$  in the process using the fitted values. Use the method of moments. True value is 4. 1p.

- (c) Use the fitted model to calculate 95% forecast-intervals for the time points  $t = 201, 202$  and  $203$ . The last four simulated values are  $Y_{197} = 2.34, Y_{198} = 3.58, Y_{199} = 1.84, Y_{200} = 3.26$ . The last four residuals are  $\hat{e}_{197} = -0.157, \hat{e}_{198} = 0.738, \hat{e}_{199} = -0.635, \hat{e}_{200} = 0.319$ . 3p.

GOOD LUCK!