## MAE 5870 – Análise de Séries temporais Lista #2

Data de entrega: 04/04/2014

- **2.1** For the Johnson & Johnson data, say  $y_t$ , shown in Figure 1.1, let  $x_t = \log(y_t)$ .
- (a) Fit the regression model

$$x_t = \beta t + \alpha_1 Q_1(t) + \alpha_2 Q_2(t) + \alpha_3 Q_3(t) + \alpha_4 Q_4(t) + w_t$$

where  $Q_i(t) = 1$  if time t corresponds to quarter i = 1, 2, 3, 4, and zero otherwise. The  $Q_i(t)$ 's are called indicator variables. We will assume for now that  $w_t$  is a Gaussian white noise sequence. What is the interpretation of the parameters  $\beta$ ,  $\alpha_1$ ,  $\alpha_2$ ,  $\alpha_3$ , and  $\alpha_4$ ? (Detailed code is given in Appendix R on page 574.)

- (b) What happens if you include an intercept term in the model in (a)?
- (c) Graph the data,  $x_t$ , and superimpose the fitted values, say  $\hat{x}_t$ , on the graph. Examine the residuals,  $x_t \hat{x}_t$ , and state your conclusions. Does it appear that the model fits the data well (do the residuals look white)?
  - **2.3** Repeat the following exercise six times and then discuss the results. Generate a random walk with drift, (1.4), of length n = 100 with  $\delta = .01$  and  $\sigma_w = 1$ . Call the data  $x_t$  for t = 1, ..., 100. Fit the regression  $x_t = \beta t + w_t$  using least squares. Plot the data, the mean function (i.e.,  $\mu_t = .01 t$ ) and the fitted line,  $\hat{x}_t = \hat{\beta} t$ , on the same graph. Discuss your results.

- 2.8 The glacial varve record plotted in Figure 2.6 exhibits some nonstationarity that can be improved by transforming to logarithms and some additional nonstationarity that can be corrected by differencing the logarithms.
- (a) Argue that the glacial varves series, say x<sub>t</sub>, exhibits heteroscedasticity by computing the sample variance over the first half and the second half of the data. Argue that the transformation y<sub>t</sub> = log x<sub>t</sub> stabilizes the variance over the series. Plot the histograms of x<sub>t</sub> and y<sub>t</sub> to see whether the approximation to normality is improved by transforming the data.
- (b) Plot the series y<sub>t</sub>. Do any time intervals, of the order 100 years, exist where one can observe behavior comparable to that observed in the global temperature records in Figure 1.2?
- (c) Examine the sample ACF of y<sub>t</sub> and comment.
- (d) Compute the difference u<sub>t</sub> = y<sub>t</sub> − y<sub>t-1</sub>, examine its time plot and sample ACF, and argue that differencing the logged varve data produces a reasonably stationary series. Can you think of a practical interpretation for u<sub>t</sub>? Hint: For |p| close to zero, log(1 + p) ≈ p; let p = (y<sub>t</sub> − y<sub>t-1</sub>)/y<sub>t-1</sub>.
- (e) Based on the sample ACF of the differenced transformed series computed in (c), argue that a generalization of the model given by Example 1.23 might be reasonable. Assume

$$u_t = \mu + w_t - \theta w_{t-1}$$

is stationary when the inputs  $w_t$  are assumed independent with mean 0 and variance  $\sigma_w^2$ . Show that

$$\gamma_u(h) = \begin{cases} \sigma_w^2(1 + \theta^2) & \text{if } h = 0, \\ -\theta \ \sigma_w^2 & \text{if } h = \pm 1, \\ 0 & \text{if } |h| > 1. \end{cases}$$

- (f) Based on part (e), use ρ̂<sub>u</sub>(1) and the estimate of the variance of u<sub>t</sub>, γ̂<sub>u</sub>(0), to derive estimates of θ and σ<sup>2</sup><sub>w</sub>. This is an application of the method of moments from classical statistics, where estimators of the parameters are derived by equating sample moments to theoretical moments.
- 2.9 In this problem, we will explore the periodic nature of S<sub>t</sub>, the SOI series displayed in Figure 1.5.
- (a) Detrend the series by fitting a regression of S<sub>t</sub> on time t. Is there a significant trend in the sea surface temperature? Comment.
- (b) Calculate the periodogram for the detrended series obtained in part (a). Identify the frequencies of the two main peaks (with an obvious one at the frequency of one cycle every 12 months). What is the probable El Niño cycle indicated by the minor peak?
  - 2.11 Consider the two weekly time series oil and gas. The oil series is in dollars per barrel, while the gas series is in cents per gallon; see Appendix R for details.
  - (a) Plot the data on the same graph. Which of the simulated series displayed in §1.3 do these series most resemble? Do you believe the series are stationary (explain your answer)?
  - (b) In economics, it is often the percentage change in price (termed growth rate or return), rather than the absolute price change, that is important. Argue that a transformation of the form y<sub>t</sub> = ∇ log x<sub>t</sub> might be applied to the data, where x<sub>t</sub> is the oil or gas price series [see the hint in Problem 2.8(d)].
  - (c) Transform the data as described in part (b), plot the data on the same graph, look at the sample ACFs of the transformed data, and comment. [Hint: poil = diff(log(oil)) and pgas = diff(log(gas)).]
  - (d) Plot the CCF of the transformed data and comment The small, but significant values when gas leads oil might be considered as feedback. [Hint: ccf(poil, pgas) will have poil leading for negative lag values.]