

Forecasting & Predictive Analytics

ESSEC MSc in Management

ESSEC-CentraleSupélec MSc in Data Sciences & Business Analytics

Guillaume Chevillon

EXAMINATION

March 29, 2017

4.30pm-7.00pm

Calculators are authorized

Please answer the four exercises below.

You can always assume one question answered and proceed to the next

Univariate Time Series

1. Answer the following questions.
 - (a) Precisely provide the definition of covariance (or weak) stationarity. Why is stationarity a useful property?
 - (b) In which of the following models are the $\{y_t\}$ stationary, assuming $\{\epsilon_t\}$ is a mean-zero white noise process? If the answer depends on extra conditions, explain the conditions required. In all cases, explain your answers.
 - i. $\Delta y_t = -0.2y_{t-1} + \epsilon_t$;
 - ii. $y_t = \phi_0 + \phi_1 y_{t-1} + \phi_2 y_{t-2} + \epsilon_t + \theta \epsilon_{t-1}$;
 - iii. $y_t = \phi_0 + 0.1x_{t-1} + \epsilon_t$, $x_t = x_{t-1} + \epsilon_t$;
 - iv. $y_t = 0.8y_{t-1} + \epsilon_t$;
 - v. $y_t = y_{t-1} + \epsilon_t - \epsilon_{t-1}$.
 - vi. $y_t - y_{t-2} = \epsilon_t$

2. Consider the AR(2) process defined as

$$y_t = \phi_1 y_{t-1} + \phi_2 y_{t-2} + \epsilon_t$$

- (a) Rewrite the model with Δy_t on the left-hand side and y_{t-1} and Δy_{t-1} on the right-hand side.
- (b) What restrictions are needed on ϕ_1 and ϕ_2 for this model to collapse to an AR(1) in the first differences Δy_t ?
- (c) When the model collapses to an AR(1) for Δy_t , what does this tell you about the properties of y_t ?
- (d) Explain and discuss the Augmented Dickey-Fuller test, including the null and alternative hypotheses, the methodology and properties.
- (e) We consider forecasting y_{t+h} using information up to time t .
 - i. Derive an explicit expression for the 1-step and 2-step ahead forecast errors $e_{t+1|t}$ and $e_{t+2|t}$ for y_t .
 - ii. What is the autocorrelation function of a time-series of forecast errors, the processes $\{e_{t+1|t}\}$ and $\{e_{t+2|t}\}$ observed for $t \geq 1$? (in words or graphically, no need to prove it mathematically).

Multivariate models

3. Consider the VAR(1) (remember that at all times, you can assume that you have proved previous questions)

$$\begin{bmatrix} x_t \\ y_t \end{bmatrix} = \begin{bmatrix} 0.4 & 0.3 \\ 0.8 & 0.6 \end{bmatrix} \begin{bmatrix} x_{t-1} \\ y_{t-1} \end{bmatrix} + \begin{bmatrix} \epsilon_t \\ \eta_t \end{bmatrix}$$

where

$$\begin{bmatrix} \epsilon_t \\ \eta_t \end{bmatrix} \stackrel{i.i.d}{\sim} \mathbf{N} \left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} \sigma_\epsilon^2 & \sigma_{\epsilon\eta} \\ \sigma_{\epsilon\eta} & \sigma_\eta^2 \end{bmatrix} \right)$$

- Verify that the vector process $(x_t, y_t)'$ is non-stationary (hint: one eigenvalue is trivial)
- Find the two values of β such that $x_t - \beta y_t$ follows an AR(1). What are these two values, and comment on the properties of the resulting AR(1) processes (careful, one of them is trivial).
- Define the concept of cointegration. Do x_t and y_t cointegrate here?
- What assumption do you need to make if you want to compute the impulse response function

$$\frac{\partial x_{t+k}}{\partial \epsilon_t}$$

as a function of k ? (do not compute the function).

EMPIRICAL ANALYSIS

- We study the empirical properties of the sales and advertizing dataset we studied in the course. We denote by a prefix L, the logarithm and by D, the difference, so DLSales is the first difference in the logarithm of Sales. Figures 1, 2 and 3 present the data. Figure 4 reports the autocorrelation and partial autocorrelation functions of the logarithm of Advertizing and Sales and their first differences.

- Comment on the stationarity of the time series presented Figures 1 to 4.
- We want to model LSales and LAdvert via ARIMA(p, d, q) models. What values do you suggest for either variable based on Figures 1 to 4.
- We estimate an AR(2) model for DLSales (standard errors in parentheses below the parameter estimates).

$$\text{DLSales}_t = \underset{(0.0328)}{0.0015} + \underset{(0.138)}{0.00491} \text{DLSales}_{t-1} - \underset{(0.136)}{0.287} \text{DLSales}_{t-2} + \epsilon_t$$

Comment on the quality of this equation: would you simplify the model by reducing the number of parameters?

- We now add a dummy variable for 1936, D1936,

$$\begin{aligned} \text{DLSales}_t = & \underset{(0.0275)}{0.0229} - \underset{(0.123)}{0.216} \text{DLSales}_{t-1} \\ & - \underset{(0.112)}{0.278} \text{DLSales}_{t-2} - \underset{(0.211)}{1.02} \text{D1936}_t + \epsilon_t \end{aligned}$$

Comment on the role of the dummy variable. (The model results are drawn on Figure 5).

- (e) Imagine you are in 1936 and want to use the models above to forecast LAdvert in 1937 and 1938. Propose a method for doing so making all the assumptions you may require. You may use that the values of LAdvert for 1931-1936 are

Year	LAdvert
1931	6.89
1932	6.95
1933	7.28
1934	7.31
1935	6.69
1936	5.82

- (f) We now propose to use the following model for forecasting two years ahead,

$$\begin{aligned}
 \text{D2LAdvert}_t = & \underset{(0.119)}{0.597} \text{D2LAdvert}_{t-1} - \underset{(0.117)}{0.624} \text{D2LAdvert}_{t-2} \\
 & + \underset{(0.114)}{0.375} \text{D2LAdvert}_{t-3} + \underset{(0.0313)}{0.0223} - \underset{(0.231)}{1.08} \text{dumm1936}_t
 \end{aligned}$$

where $\text{D2LAdvert}_t = \text{LAdvert}_t - \text{LAdvert}_{t-2}$. (The model results are drawn on Figure 6). What forecasts does this new model generate?

- (g) Discuss the difference and relative benefits from Iterated and Direct multistep forecasting methods (both from a theoretical and empirical point of view).
- (h) We now estimate a VAR(1) model as follows:

$$\begin{aligned}
 \text{DLAdvert}_t = & \underset{(0.0272)}{0.0193} - \underset{(0.152)}{0.466} \text{DLAdvert}_{t-1} \\
 & + \underset{(0.274)}{0.746} \text{DLSales}_{t-1} - \underset{(0.209)}{1.06} \text{D1936}_t + \epsilon_t \\
 \text{DLSales}_t = & \underset{(0.0162)}{0.00975} - \underset{(0.0909)}{0.0523} \text{DLAdvert}_{t-1} \\
 & + \underset{(0.164)}{0.364} \text{DLSales}_{t-1} - \underset{(0.125)}{0.306} \text{D1936}_t + \eta_t
 \end{aligned}$$

Comment on the Granger Causality properties of the model by carefully defining the concept.

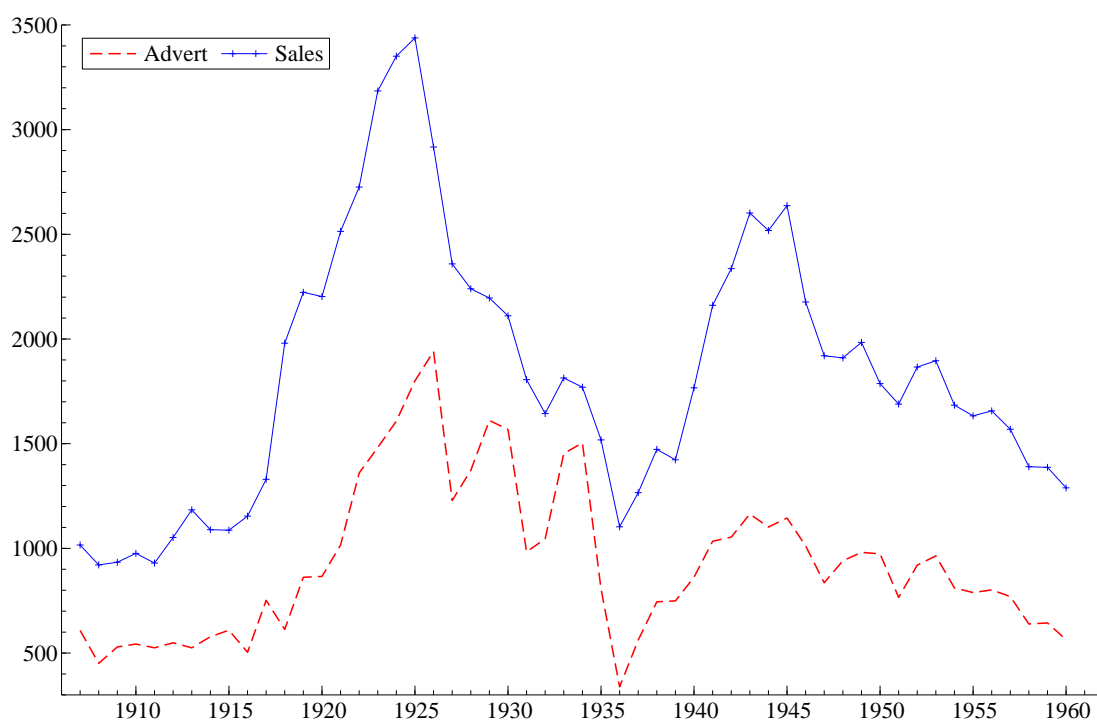


Figure 1:

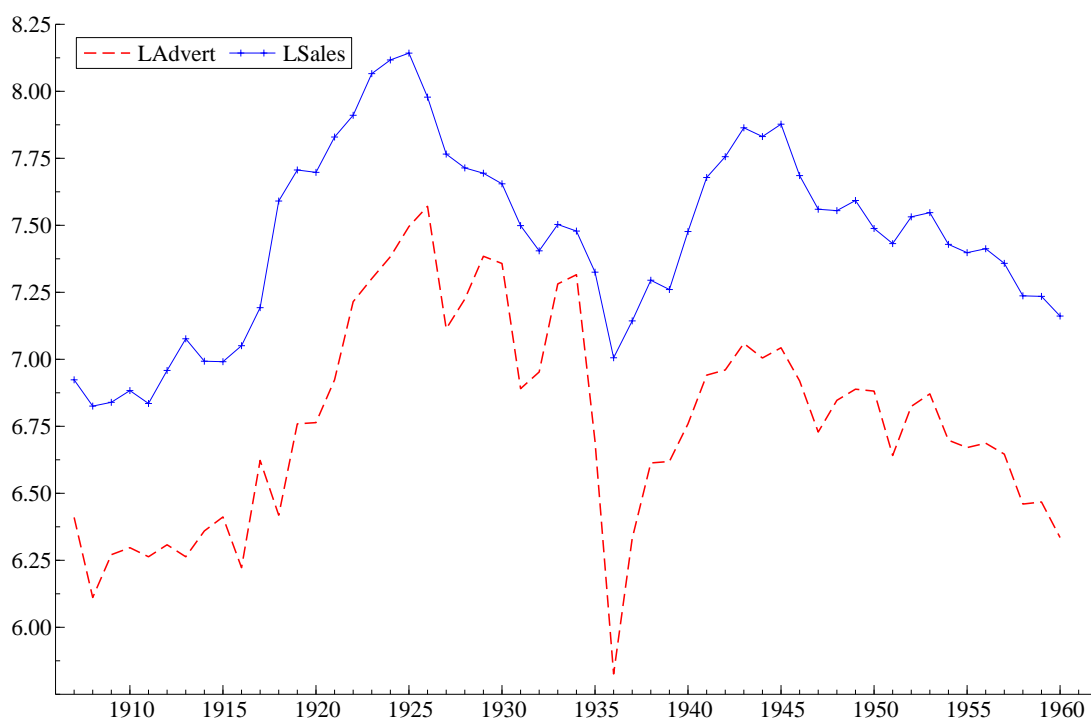


Figure 2:

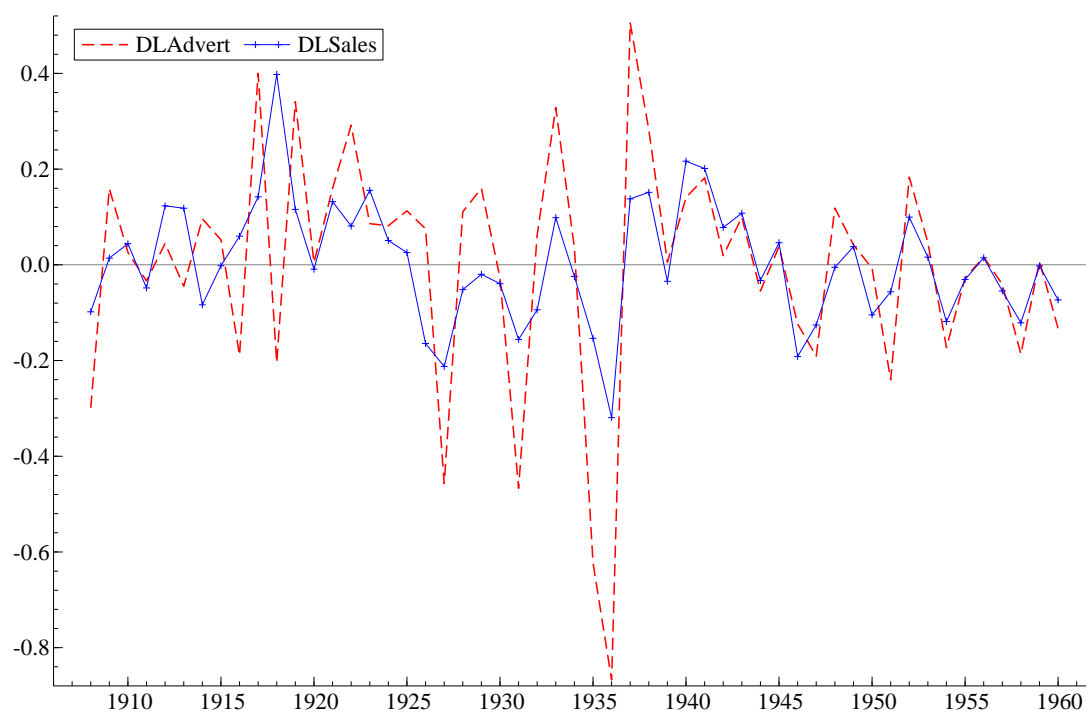


Figure 3:

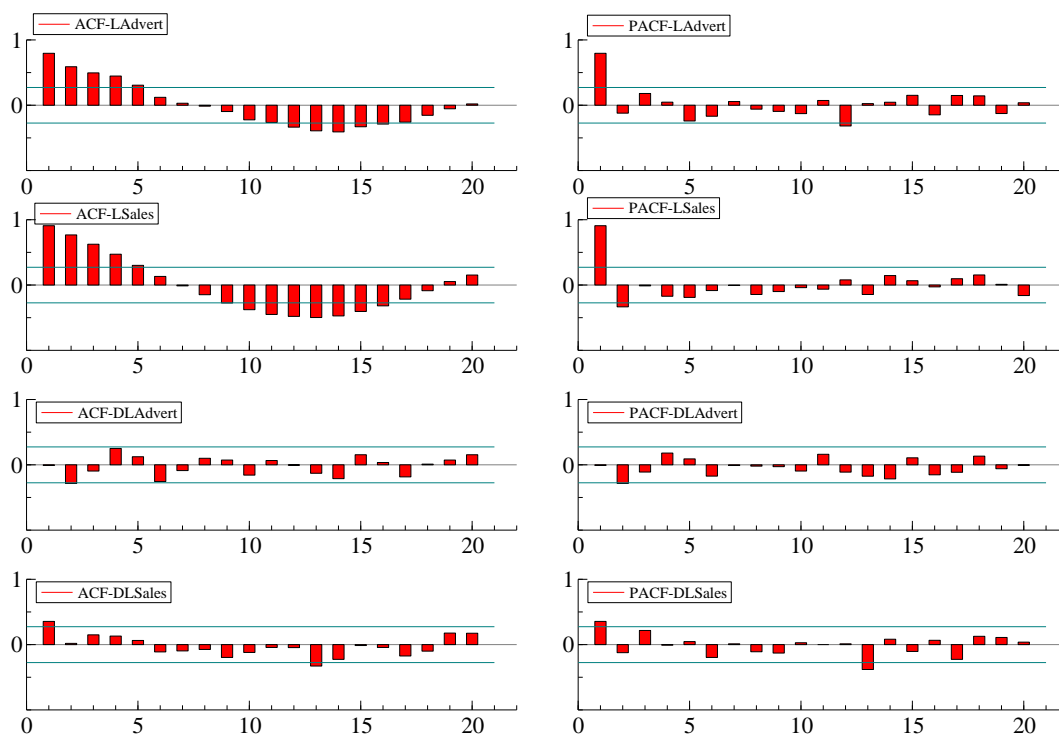


Figure 4:

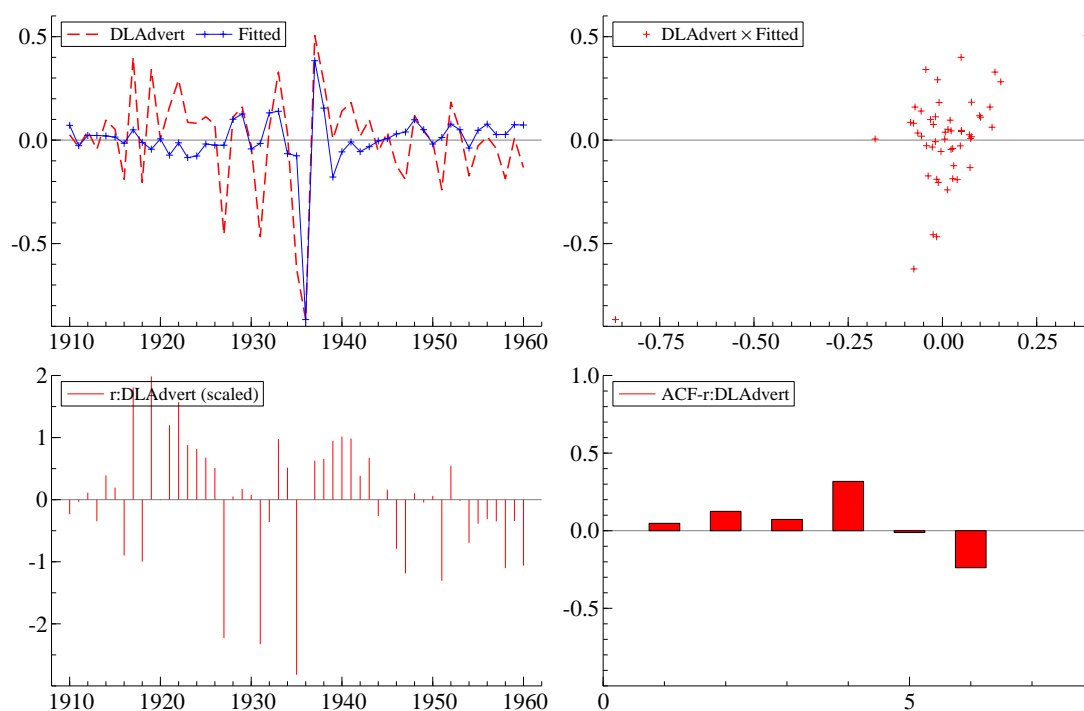


Figure 5:

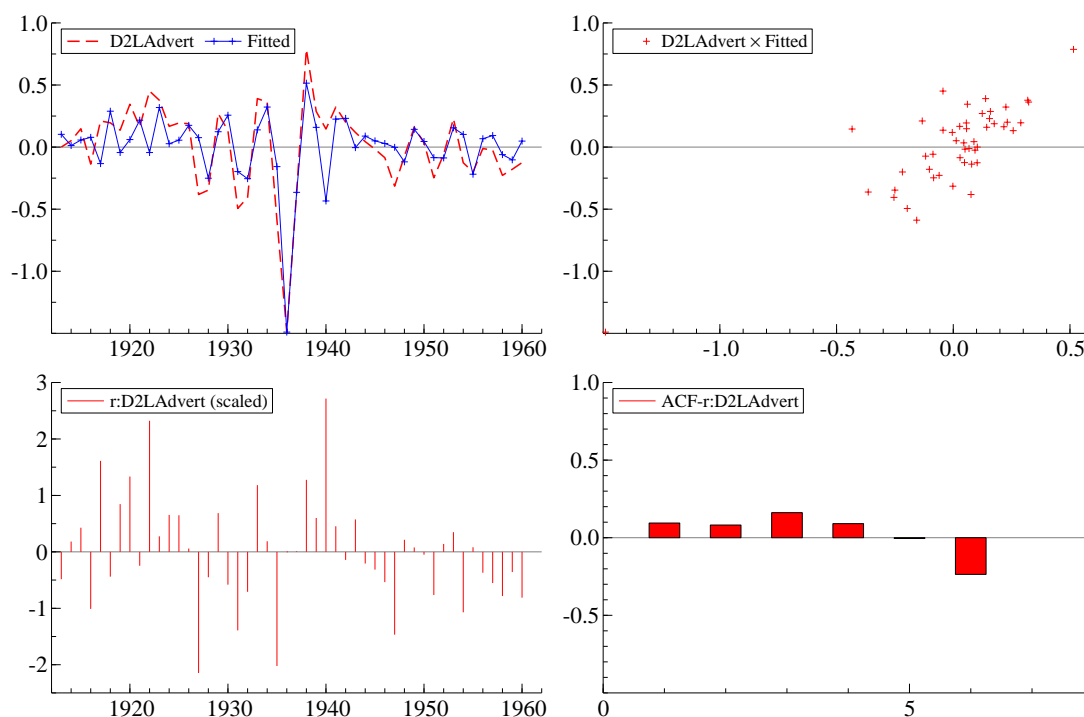


Figure 6: