

# Empirical Methods in Finance

## Homework 7

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### Problem 1: Applying the Box-Jenkins methodology<sup>1</sup>

In PPIFGS.xls you will find quarterly data for the Producer Price Index. Our goal is to develop a quarterly model for the PPI, so we can come up with forecasts. Our boss needs forecasts of inflation, because she wants to hedge inflation exposure. There is not a single ‘correct’ answer to this problem. Well-trained econometricians can end up choosing different specifications even though they are confronted with the same sample. However, there definitely are some wrong answers.

1. We look for a covariance-stationary version of this series. Using the entire sample, make a graph with four subplots:
  - (a) Plot the PPI in levels.
  - (b) Plot  $\Delta PPI$
  - (c) Plot  $\log PPI$
  - (d) Plot  $\Delta \log PPI$ .
2. Which version of the series looks covariance-stationary to you and why? Let’s call the covariance stationary version  $y_t = f(PPI_t)$ .

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<sup>1</sup>In Matlab, there is an **Econometrics Toolbox** and a series of functions : ‘arima, estimate, forecast, infer, simulate, lbqtest’ that can help you solve this problem. Alternatively, you can download Kevin Sheppard’s **MFE toolbox**, which is freely available. You can just Google this and find it. In *R* there is a package called ‘MTS’ for *Multivariate Time Series*, by Ruey Tsay. This is a very useful package, that we will also use when estimating time-varying volatility models.

3. Plot the ACF of  $y_t$  for 12 quarters. What do you conclude? If the ACF converges very slowly, re-think whether  $y_t$  really is covariance stationary.
4. Plot the PACF of  $y_t$  for 12 quarters. What do you conclude?
5. On the basis of the ACF and PACF, select four different ARMA model specifications.
  - (a) Using the entire sample, estimate each one of these. Report the coefficient estimates and standard errors. Check for stationarity of the parameter estimates.
  - (b) Plot the residuals. (Note: the residuals will have conditional heteroskedasticity or ‘GARCH effects’. We will talk about this in Lecture 12. However, in well-specified models, the residuals should not be autocorrelated.)
  - (c) Report the Q-statistic for the residuals for 8 and 12 quarters, as well as the AIC and BIC. Select a preferred model on the basis of these diagnostics. Explain your choice.
6. Re-estimate the 4 models using only data up to the end of 2005 and compute the MSPE (mean squared prediction error) on the remainder of the sample for one-quarter ahead forecasts:

$$\frac{1}{H} \sum_{t=1}^H v_t^2$$

where  $H$  is the length of the hold-out sample, and  $v_i$  is the one-step ahead prediction error. Also report the MSPE assuming there is no predictability in  $y_t$ , i.e. assuming  $y_t$  follows a random walk. What do you conclude?

## Problem 2: Year-on-year quarterly data and ARMA dynamics

Assume the true quarterly log market earnings follow:

$$\begin{aligned} e_t &= e_{t-1} + x_t, \\ x_t &= \phi x_{t-1} + \varepsilon_t, \end{aligned}$$

where  $\text{var}(\varepsilon_t) = \sigma_\varepsilon^2 = 1$  and  $\varepsilon_t$  is i.i.d. over time  $t$ .

The earnings data you are given is year-on-year earnings growth, which in logs is:

$$y_t \equiv e_t - e_{t-4}.$$

1. Assume  $\phi = 0$ . Derive autocovariances of order 0 through 5 for  $y_t$ . I.e.,  $cov(y_t, y_{t-j})$  for  $j = 0, \dots, 5$ .
2. Assume  $\phi = 0$ . Determine the number of AR lags and MA lags you need in the ARMA(p,q) process for  $y_t$ . Give the associated AR and MA coefficients.
3. Optional: assume  $1 > \phi > 0$ . Repeat 1. and 2. under this assumption.