

Exercise 1

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1. (a) Open the file `maunaloa.wf1`; this is a famous data set used in machine learning. Make a time series plot.
(b) Estimate a linear trend by regressing the `co2` series on an intercept and the variable `time`.
(c) Plot the data, together with the estimated linear trend.
(d) Produce a forecast for 2005M1, first manually using the fitted model

$$\hat{Y}_t = \hat{\beta}_0 + \hat{\beta}_1 t,$$

then using EViews.

- (e) Repeat Questions 1b through 1d, but using a quadratic trend.
(f) Repeat Questions 1b through 1d, but using an exponential trend.
2. (a) Compute the 3rd order moving average of the `co2` series for 1964M6 by hand.
(b) Estimate the trend with a 12 month moving average (12 months are necessary to cover a full cycle). Then plot the resulting trend estimate and the data together in a time series plot.
3. (a) Estimate a model with a linear trend and 12 monthly dummies (and no intercept) for the `co2` series. Then, produce an (in-sample) forecast for 2004M12, both by hand and using EViews. Also create an actual-fitted-residual plot.
(b) Same, but include an intercept and remove the last dummy.

Exercise 2

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1. (a) Open the file `simulations.xlsx`. The sheet “White Noise” simulates $T = 1000$ observations from a (Gaussian) white noise process; i.e., 1000 uncorrelated mean-zero normals. By repeatedly pressing F9, you can draw new random numbers. Describe your observations.
(b) Similarly, the sheet “Random Walk” simulates $T = 1000$ observations from a (Gaussian) random walk. Describe your observations.
2. (a) Open the file `sp500.wp1`. Generate a new series `logsp500` containing the log prices, and a series `r` containing the continuously compounded returns. Make a time series plot for each, and a histogram of the returns. Describe your findings.
(b) Use the skewness and kurtosis given in the histogram to manually conduct a Jarque-Bera test.
(c) Generate a correlogram of the returns and interpret it.
(d) Test whether the first 10 autocorrelations are jointly significant at the 5% level.
(e) Generate a correlogram of the log prices and interpret it.
3. (a) Show that for the random walk $Y_t = Y_{t-1} + U_t$, where U_t is white noise and Y_0 some constant,

$$Y_t = Y_0 + U_1 + U_2 + \cdots + U_t = Y_0 + \sum_{s=1}^t U_s.$$

- (b) Building on the result from the previous question, show that

$$\mathbb{E}[Y_t] = Y_0, \quad \text{and} \\ \text{var}(Y_t) = \sigma^2 t.$$

Exercise 3

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1. (a) Open the file `simulations.xlsx`. The sheet “AR(1)” simulates $T = 1000$ observations from an AR(1) process. Play around with α and $-1 < \phi_1 < 1$ and describe your observations.
(b) Also try setting $\phi_1 = 1$ and describe the effect of α .
(c) The file `simulated_data.wfl` contains three series simulated using the same spreadsheet, `simulation.xlsx`, one each for an AR(1), an MA(1), and an ARMA(1, 1) process. The AR and ARMA processes use $\phi_1 = 0.7$, and the MA and ARMA processes use $\theta_1 = 0.7$. Describe your observations.
2. (a) Use the Box-Jenkins approach to model year-on-year real GDP growth in the file `realgdpch.wfl`.
(b) Produce a forecast for 2022Q3 and 2022Q4, both manually and using EViews.
3. (a) Obtain the mean and variance of a random walk with drift.
(b) Show that the random walk with drift is integrated of order 1.
(c) Derive the expression for the variance of a stationary AR(1) given in the slides.
(d) Find the mean, variance, and ACF of an MA(1).
(e) **Optional:** Find the ACF of a stationary AR(1).

Exercise 4

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1. Open the file `simulations.xlsx`. Use the sheets “AR(1)” (with ϕ_1 set to 1) to simulate a random walk with drift, and the sheet “Linear Trend” to simulate a trend-stationary process. Play with the parameters and describe your observations.
2.
 - (a) The file `tbill.wf1` contains monthly data for the 3-month T-Bill rate. Plot them, study the correlogram, and conduct a unit root test.
 - (b) Model the first difference of the T-Bill rate as an ARMA process, hence modelling the T-Bill rate as an ARIMA process.
 - (c) Forecast the T-Bill rate for 2022M11 and 2022M12 based on the model you found in the previous question.
3.
 - (a) The file `ibm_capm.wf` contains data for the S&P500, IBM stock, and the 3-month T-Bill rate. Use it to estimate the CAPM β of IBM, by regressing the excess returns of IBM on the excess returns of the market.
 - (b) Use the Durbin-Watson test to test for first-order autocorrelation in the residuals.
 - (c) Use the Breusch-Pagan test to test for autocorrelation up to order 5 in the residuals.
 - (d) Re-estimate the regression using HAC standard errors.
4.
 - (a) Show that for both

$$\begin{aligned} Y_{1,t} &= \delta t + U_{1,t} \quad \text{and} \\ Y_{2,t} &= \delta + Y_{2,t-1} + U_{2,t}. \end{aligned}$$

we have $\mathbb{E}[\Delta Y_{i,t}] = 0$.

- (b) Derive the ADF regression for an AR(2) process.

Exercise 5

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1.
 - (a) In the file `sp500`, construct the returns, produce a correlogram of the squared residuals, and interpret it.
 - (b) Perform an ARCH-LM test by regressing the returns on an intercept.
 - (c) Compute the historical volatility and plot it.
 - (d) Compute the EWMA volatility and plot it.
 - (e) Find a suitable GARCH/TGARCH/EGARCH model. Start with a GARCH(1, 1) or an ARCH(6) model, and determine whether it needs to be adjusted.
 - (f) Make a plot of the volatility estimates that your model generates, and of the NIC.
 - (g) Forecast the volatility for $T + 1$.

2.
 - (a) Show that

$$\hat{\sigma}_{t+1,EWMA}^2 = \lambda \hat{\sigma}_{t,EWMA}^2 + (1 - \lambda)r_t^2, \quad 0 < \lambda < 1.$$

Exercise 6

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1. (a) In the file `sp500.wf1`, compute the historical 1% VaR using `@quantile`. Note that this uses the entire sample, rather than the last m returns.
- (b) Determine the Normal VaR, both using EViews and manually, based on the mean return of 0.000390 and the volatility 0.010946.
- (c) Determine the VaR based on a GARCH(1, 1) model with Normal innovations, and with standardized t innovations.
- (d) Produce a manual VaR forecast for 10/27/2022 based on the GARCH model with t innovations, using $\sigma_t = 0.014875$.
- (e) Make a plot with your VaR estimates overlaid on the negative log returns.
- (f) Test your VaR forecasts for correct unconditional coverage, independence, and correct conditional coverage.

Exercise 7

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1. Until 1971, as part of the Bretton-Woods system of fixed exchange rates, the US dollar was convertible to gold, i.e., it was possible for foreign central banks to redeem US dollars for gold at a fixed rate of 35\$ per troy ounce, so that the price of gold was fixed. In 1971, US president Nixon unilaterally cancelled the direct convertibility, ultimately ending the Bretton-Woods agreement. Gold became a floating asset, and its price increased sharply; in other words, the US\$ was massively devalued. In this exercise, we will analyze the hypothesis that the increasing price (in US\$) of oil is not a consequence of an increased demand for (or a reduced supply of) oil, but rather of a continued devaluation of the US\$. We have at our disposal monthly data from April 1968 to January 2017 (586 observations) on the following variables:

- GOLD, the spot price of one troy ounce of gold in US\$;
 - OIL, the spot price of one barrel of WTI crude oil in US\$.
- (a) Assuming that GOLD is integrated of order one, explain why the hypothesis that the relative price of oil (in troy ounces of gold per barrel) is stationary implies cointegration between $\log(\text{OIL})$ and $\log(\text{GOLD})$.
- (b) Using the file `oil_gold_2017.wfl`, analyze whether this cointegrating relationship can be found in the data, based on the Engle-Granger procedure.

2. Consider the model

$$\begin{aligned} Y_t &= \beta_1 + \beta_2 X_t + U_{1,t} \\ X_t &= X_{t-1} + U_{2,t} \end{aligned}$$

where $\beta_2 \neq 0$, $U_{1,t}, U_{2,t} \stackrel{\text{iid}}{\sim} (0, \sigma^2)$ independently of each other.

- (a) Is X_t stationary?
- (b) Is Y_t stationary?
- (c) Are X_t and Y_t cointegrated? If yes, what is the cointegrating vector?
- (d) Derive the bivariate VECM for Y_t and X_t .