Partial autocorrelation

Partial autocorrelation for exchange rate between GBP and NZD

Read in the quarterly GBP to NZD exchange rate, and save it as a vector called exchange_data:

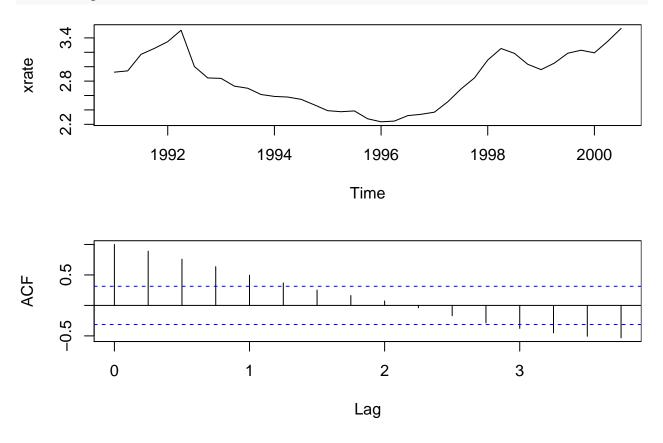
```
www <- "http://asta.math.aau.dk/eng/static/datasets?file=pounds_nz.dat"
exchange_data <- read.table(www, header = TRUE)</pre>
```

Now convert it to a time series object (ts) with the correct starting date (First quarter 1991) and frequency and call it exchange:

```
exchange <- ts(exchange_data, start = 1991, freq = 4)</pre>
```

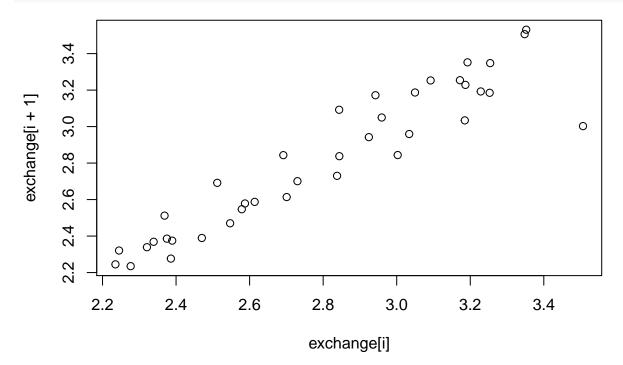
• Plot the series and its autocorrelation function.

par(mfrow = c(2,1), mar = c(5,4,1,0))
plot(exchange)
acf(exchange)

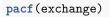


• Make a scatterplot of the series and its lag 1 values.

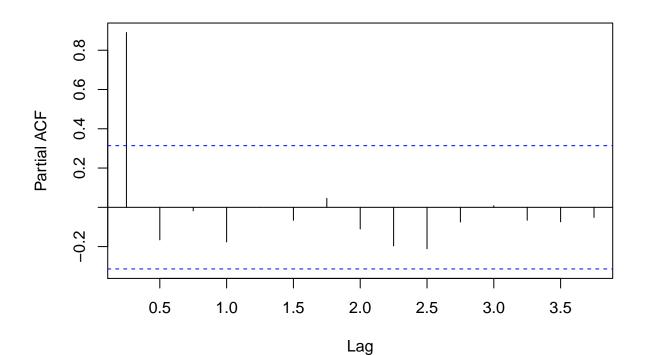
i <- 1:(length(exchange) - 1)
plot(exchange[i], exchange[i+1])</pre>



• Plot the partial autocorrelation function.



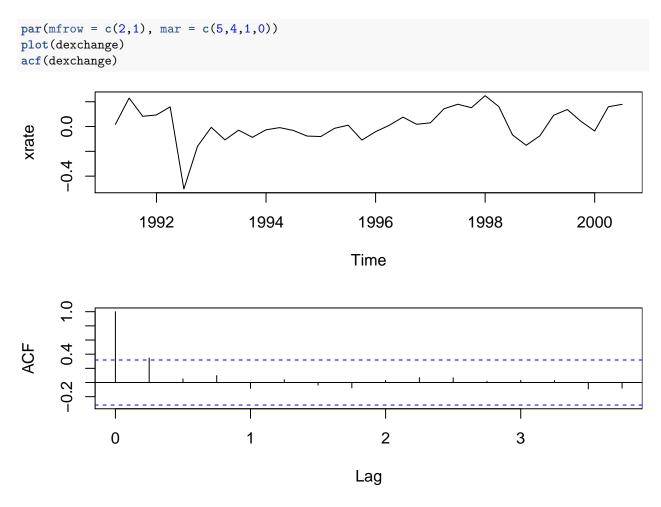




Consider the series of first order differences $y_t = x_{t+1} - x_t$ for $t = 1, \ldots, 38$:

dexchange <- diff(exchange)</pre>

• Plot the series of differences and its autocorrelation function and comment on it.



Random walk

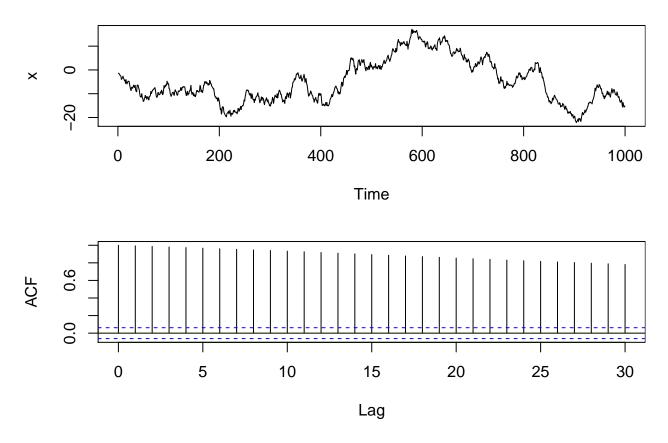
The method of removing trend by differencing can be really effective as seen in the following where we redo the exercise above for an artificial dataset \mathbf{x} :

x <- cumsum(rnorm(1000))</pre>

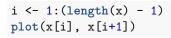
This is called a random walk and we will discuss it more in the next lecture.

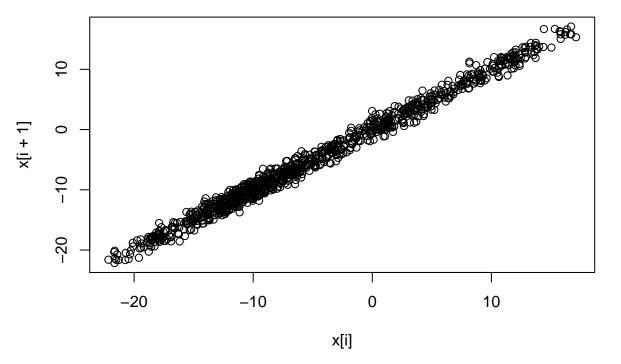
• Plot the series and its autocorrelation function.

par(mfrow = c(2,1), mar = c(5,4,1,0))
ts.plot(x)
acf(x)



• Make a scatterplot of the series and its lag 1 values.

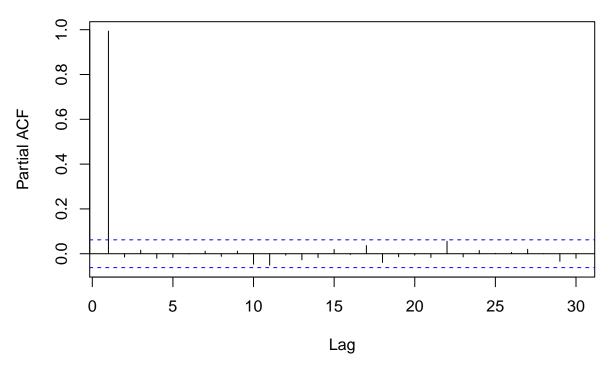




• Plot the partial autocorrelation function.

pacf(x)





Consider the series of first order differences $y_t = x_{t+1} - x_t$ for $t = 1, \ldots, 999$:

$dx \leftarrow diff(x)$

• Plot the series and its autocorrelation function.

par(mfrow = c(2,1), mar = c(5,4,1,0))
ts.plot(dx)
acf(dx)

