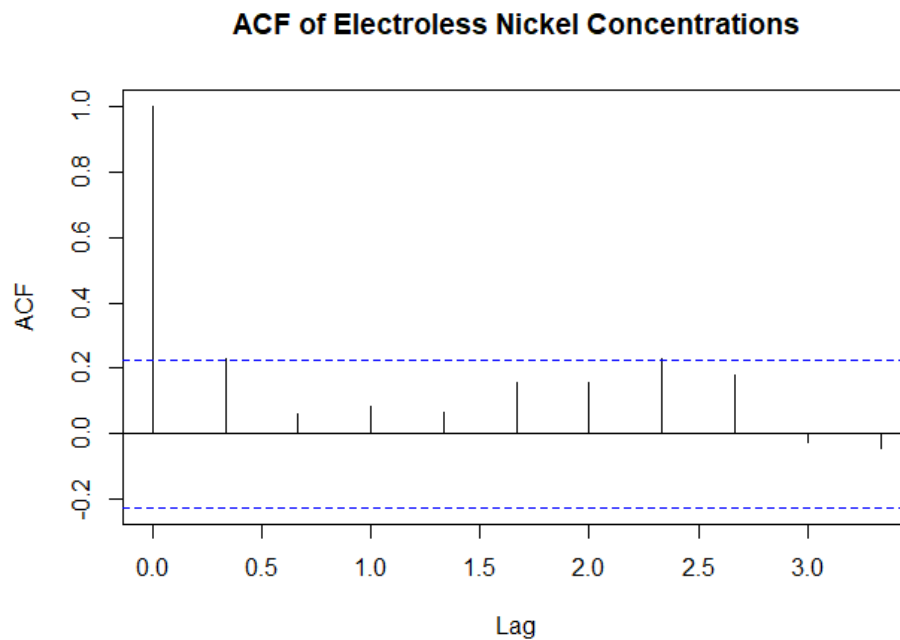


# DATA315 Assignment 3

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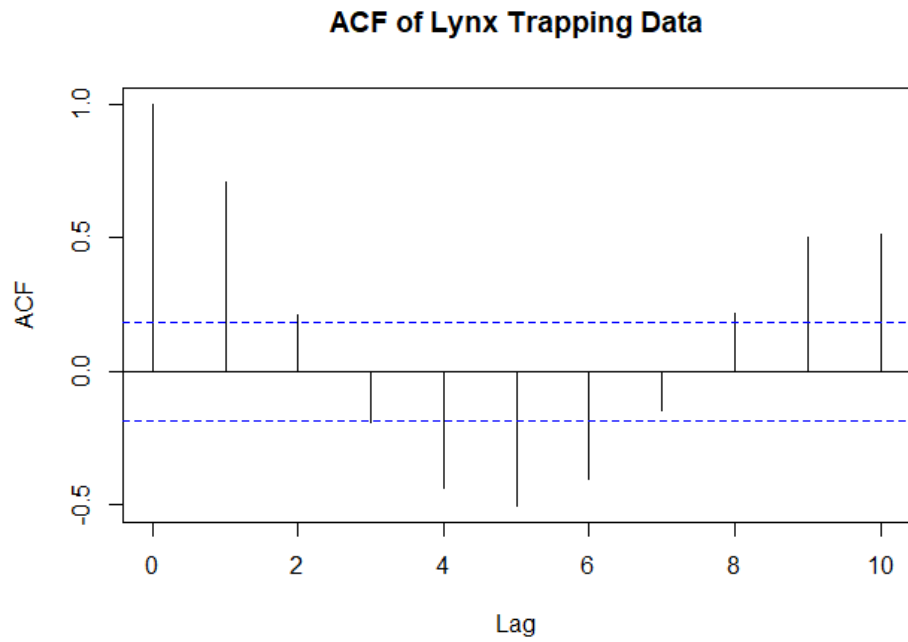
March 16, 2025

```
1. source("nickel.R")
   acf(nickel, lag.max = 10,
       main = "ACF of Electroless Nickel Concentrations")
```



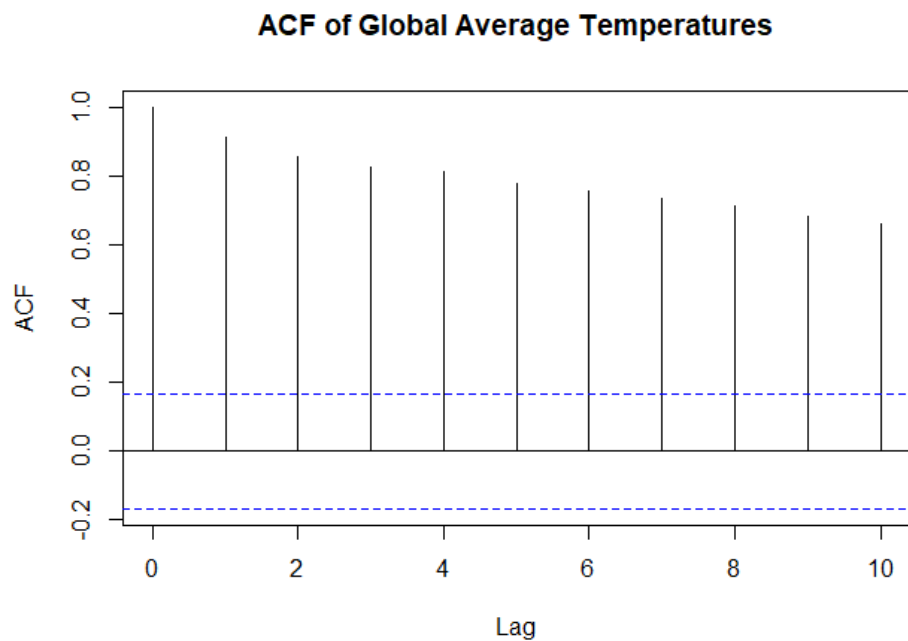
The ACF plot seems to follow an MA(1) process, as significant correlation at lag 1 followed by immediate drop to near zero.

```
2. data(lynx)
   acf(lynx, lag.max = 10, main = "ACF of Lynx Trapping Data")
```



So far, there are no models that fit, because the plot shows a cyclic pattern between predator and prey populations.

```
3. source("Globaltemps.R")
   temps <- ts(temps, start = 1880, end = 2016)
   acf(temps, lag.max = 10,
       main = "ACF of Global Average Temperatures")
```



The ACF plot seems to follow an AR(1) process, as significant correlation at lag 1 followed by gradual decay.

```

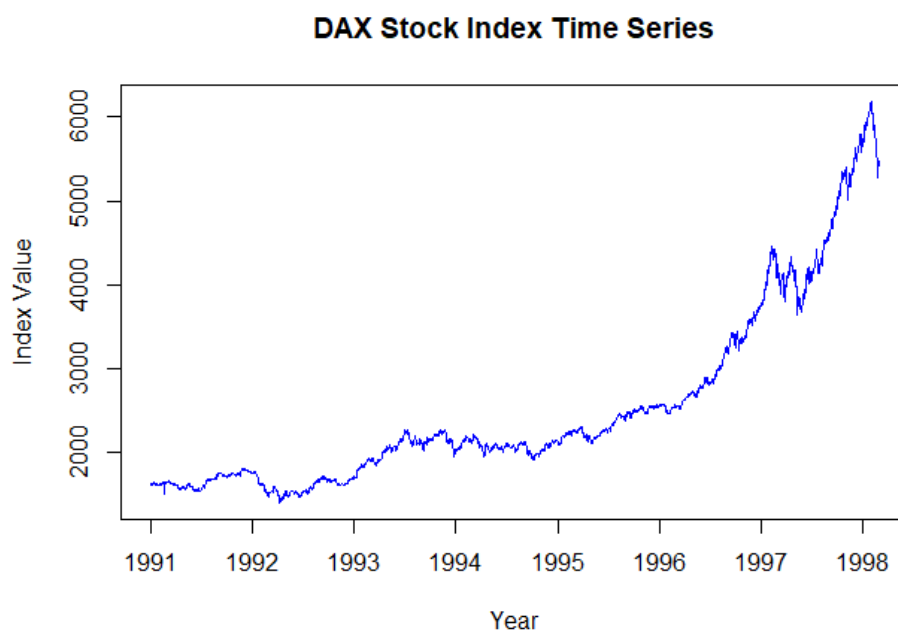
4. data("EuStockMarkets")
   dax <- EuStockMarkets[, 1]
   dax_ts <- ts(dax, start = c(1991, 1), frequency = 260)
   # 260 trading days per year

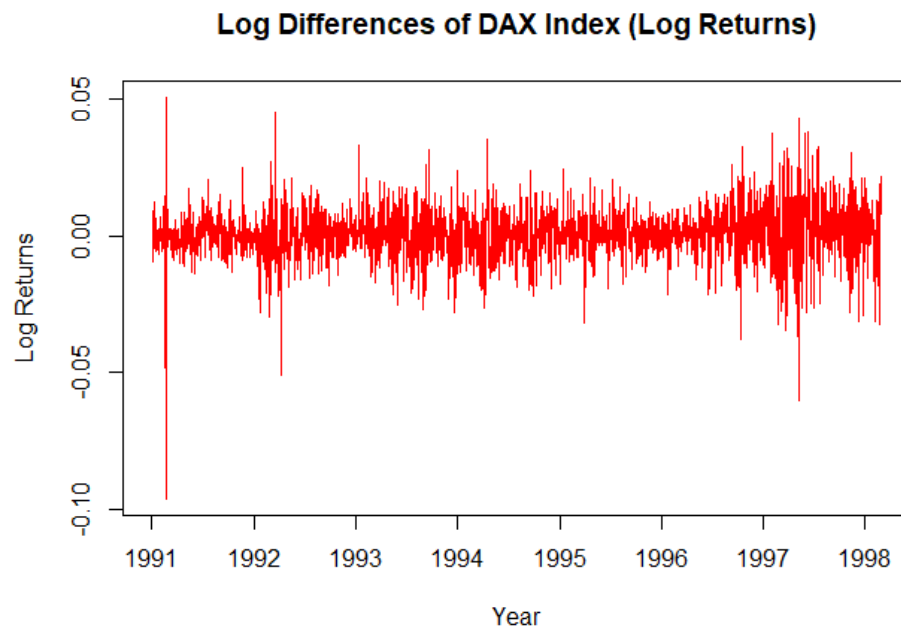
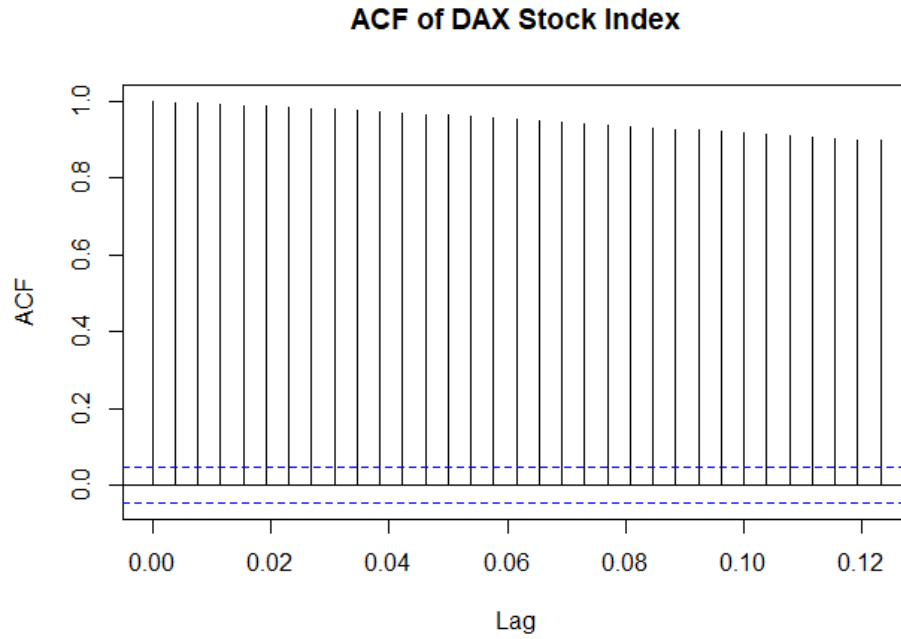
   plot(dax_ts,
        main = "DAX Stock Index Time Series",
        ylab = "Index Value", xlab = "Year",
        col = "blue", type = "l") # Time series plot

   acf(dax_ts, main = "ACF of DAX Stock Index") # ACF plot

   log_dax <- log(dax_ts)           # Take the natural log
   diff_log_dax <- diff(log_dax)     # Compute first differences (log returns)

```





Some observations for the DAX stock index time series plot; visually, we can see that there is a general upward trend with some fluctuations. The ACF plot shows that there is a significant correlation at lag 1, followed by a gradual, slow decay, which means that it may be following other process we have not covered yet. The log returns plot shows that the data revolves around 0, which is a good sign for stationarity.