We clean the memory and charge the libraries

```
rm(list=ls())
 library(fpp2)
 ## Warning: package 'fpp2' was built under R version 4.0.5
 ## Registered S3 method overwritten by 'quantmod':
 ##
      method
 ##
      as.zoo.data.frame zoo
 ## -- Attaching packages ------ fpp2 2.4 --
 ## v ggplot2
                          v fma
                                      2.4
                3.3.5
 ## v forecast 8.16
                         v expsmooth 2.3
 ## Warning: package 'ggplot2' was built under R version 4.0.5
 ## Warning: package 'forecast' was built under R version 4.0.5
 ## Warning: package 'fma' was built under R version 4.0.5
 ## Warning: package 'expsmooth' was built under R version 4.0.5
 ##
We assign a variable with the time series of the train data
 tr_NO2_24h_00 <- read.csv("~/It Academy/Data Science/Aire/tr_NO2_24h_00.csv", row.names=1)</pre>
We execute auto.arima to find the optimal parameters of the model
 mod.tr_NO2_24h_00 <- auto.arima(tr_NO2_24h_00, trace=TRUE, allowdrift=FALSE)</pre>
```

```
##
##
    Fitting models using approximations to speed things up...
##
                                     : 6491.072
##
   ARIMA(2,1,2)
                                     : 6705.866
   ARIMA(0,1,0)
##
##
   ARIMA(1,1,0)
                                     : 6672.009
## ARIMA(0,1,1)
                                     : 6619.224
##
   ARIMA(1,1,2)
                                     : 6498.167
##
   ARIMA(2,1,1)
                                     : 6489.622
   ARIMA(1,1,1)
                                     : 6505.887
##
##
   ARIMA(2,1,0)
                                     : 6614.865
##
   ARIMA(3,1,1)
                                     : 6497.743
##
   ARIMA(3,1,0)
                                     : 6595.524
##
   ARIMA(3,1,2)
                                     : 6494.934
##
   Now re-fitting the best model(s) without approximations...
##
##
##
   ARIMA(2,1,1)
                                     : 6495.288
##
##
    Best model: ARIMA(2,1,1)
```

The best model is ARIMA (2,1,1). We create a model with this parameters and the train data

```
mod.tr_NO2_24h_00 <- Arima(tr_NO2_24h_00[,1], order=c(2,1,1))
```

We evaluate the accuracy of the prediction model, in this case the one step forecasting is executed on the train data

```
accuracy(mod.tr_NO2_24h_00)
```

```
## ME RMSE MAE MPE MAPE MASE
## Training set -0.1270531 10.45401 8.327531 -8.482339 24.96462 0.8945777
## ACF1
## Training set 0.001653211
```

We assign a variable with the time series of the test data

```
te_NO2_24h_00 <- read.csv("~/It Academy/Data Science/Aire/te_NO2_24h_00.csv", row.names=1)
```

We perform a one step forecast in the data test using the model with the fitted parameters from the train data

```
mod.te_NO2_24h_00<- Arima(te_NO2_24h_00[,1], model=mod.tr_NO2_24h_00)
```

We evaluate the accuracy of the model on the test data,

```
accuracy(mod.te_NO2_24h_00)
```

```
## ME RMSE MAE MPE MAPE MASE
## Training set 0.08205777 12.92776 10.48077 -8.869367 26.72146 0.9063232
## ACF1
## Training set 0.01880632
```

We write a csv with the values of the time series forecast of the data test