Time series analysis of Belgium and Euro Area 10-years Government Bond Yields

**Advanced Time Series Analysis - D0M63B** 

#### **Presentation of the Dataset**

- Dataset published on FRED (Federal Reserve Economic Data).
- Belgium and Euro area 10-years Government Bond Yields\*.
- Time series with quarterly frequency (end of period value).
- From the first quarter of 1970 to the last quarter of 1995.
- Critical value 0.05 will be applied for all tests.

#### **Univariate Time Series Analysis**

- Explorative Analysis of Belgian Bond Yields (BE\_ts).
- Creation of the best Arima model for forecasting.

#### **Multivariate Time Series Analysis**

- Analysis of the relationship between the two series
- Cointegration test, ADLM(2) and VAR model

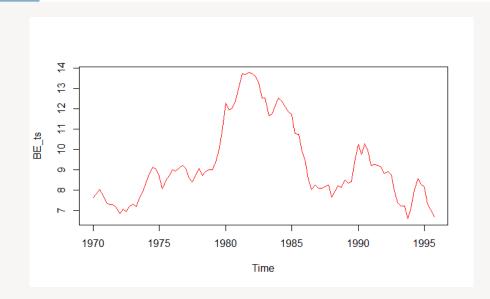


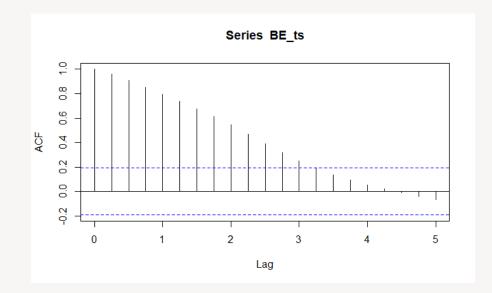
**Student: Federico Soldati - r0924528** 

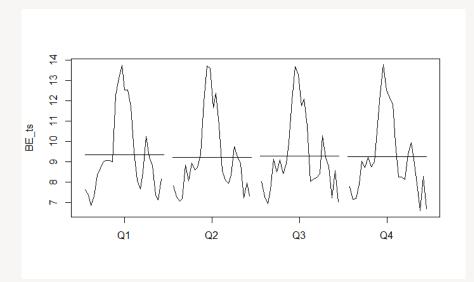
**Professor: Christophe Croux** 

# **Univariate Time Series analysis**

## Belgium bond: BE\_ts





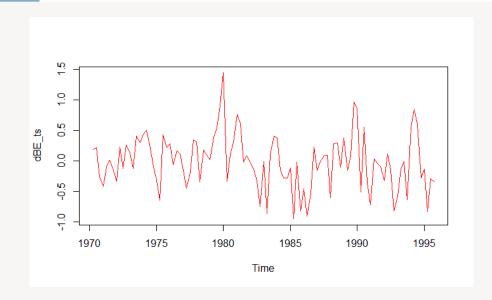


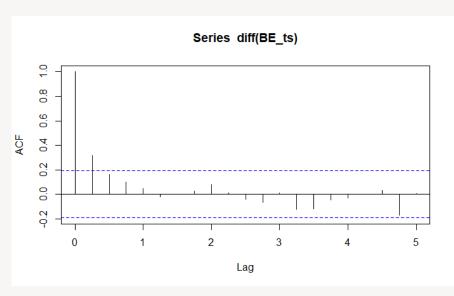
#### **Comments**

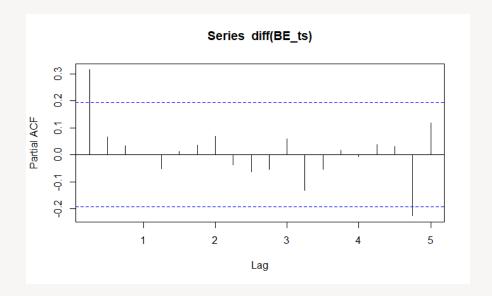
- ADF test: p-value = 0.51, the TS is not stationary.
- Ljung-Box test: p-value = 0, the TS is not white noise.
- No seasonal pattern is visible in the plot by quarter.
- The series shows strong persistency
- To fit a Arima model, is required a stationary series, so a new TS is created using the differences.

# **Univariate Time Series analysis**

## Belgium bond: dBE\_ts





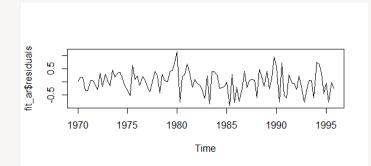


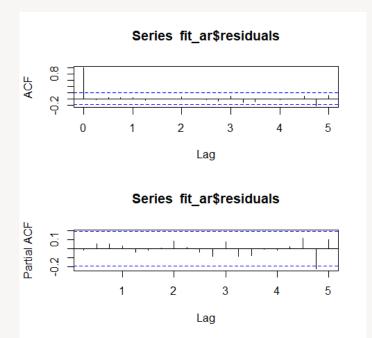
#### **Comments**

- ADF test: p-value = 0, the TS in differences is stationary.
- Ljung-Box test: p-value = 0.11, the TS seems to be white noise.
- A significant correlation can be seen in the ACF and one in the PACF.
- It was chosen to fit and compare two different models, an MA(1) and an AR(1).

## ARIMA(1,1,0)

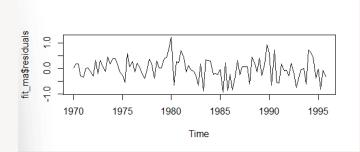
## ARIMA(0,1,1)

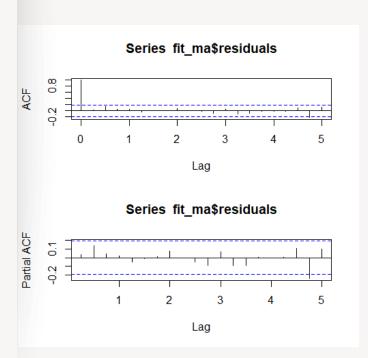




- Ljung-Box test: p-value= 0.9982, the residuals are white noise
- Since we created the CI using 5% significance level, a significant correlation is acceptable.
- Is possible to say that the model is valid
- The AR term is significant

AIC: 115.7366SIC: 121.0254





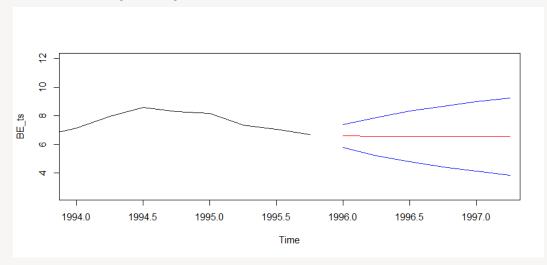
- Ljung-Box test : p-value= 0.9611, the residuals are white noise
- Since we created the CI using 5% significance level, a significant correlation is acceptable.
- Is possible to say that the model is valid
- The MA term is significant

AIC: 117.5694SIC: 122.8581

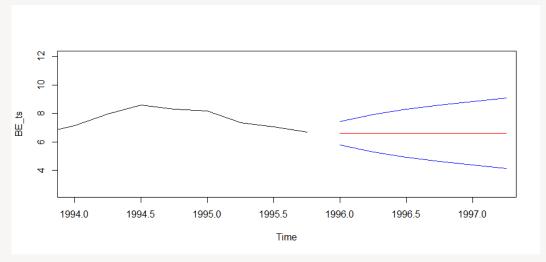
# **Univariate Time Series analysis**

# **Forecasting**

#### **ARIMA(1,1,0)**



**ARIMA(0,1,1)** 



• MAE(AR): 0.4052151

• MAE(MA): 0.405801

• Diebold-Mariano p-value = 0.953

• MSE(AR): 0.2479685

• MSE(MA): 0.2450036

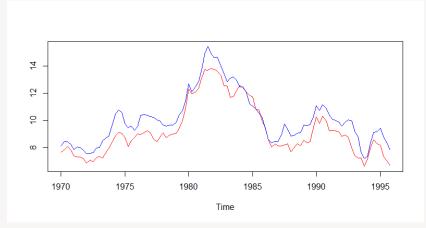
• Diebold-Mariano p-value = 0.7929

### **Comments**

- The forecasts are very similar.
- MAE and MSE are almost the same.
- Diebold-Mariano tests: we reject the null hypothesis in both the cases. So, is possible to say that the forecast performance of the two models is not significantly different.
- Essentially, both models predict only a slight negative change in the next quarter's yields from those of time t. Forecasted values have wide prediction intervals.
- The AR(1) model is preferred because it has lower AIC, SIC, and MAE.

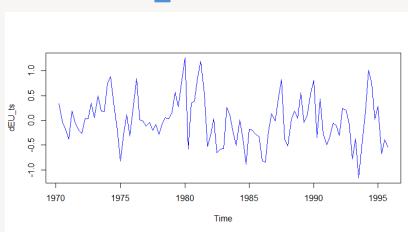
### **Cointegration Test**

### EU\_ts (blue) and BE\_ts (red)



- ADF test: the TS is not stationary.
- dEU\_ts is created using the diff() function.

### dEU\_ts



- ADF test: the TS is now stationary.
- EU\_ts and BE\_ts are both integrated of order 1, it is possible to proceed with a cointegration test.

### **Cointegration Test**

- The test statistics is -3.6296, which is smaller that the Engle-Granger ADF test statistics for one explanatory variable -3.41. Therefore, we reject the H0 hypothesis of no cointegration and conclude that BE\_ts and EU\_ts are cointegrated.
- Since they are cointegrated, an error correction model (ECM) can now be constructed and estimated.
- Automatic lag selection will also be used to select the order of a VAR model.

### **ECM** and VAR selection

```
Call:
lm(formula = dBE_ts \sim dEU_ts + ECT)
Residuals:
    Min
              1Q Median
-0.90731 -0.17119 0.00488 0.19259 0.73247
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
                       0.03004 -0.217 0.828650
(Intercept) -0.00652
            0.66574
                       0.06431 10.351 < 2e-16 ***
dEU ts
            -0.21808
                       0.06005 -3.632 0.000446 ***
ECT
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.3049 on 100 degrees of freedom
Multiple R-squared: 0.5305,
                               Adjusted R-squared: 0.5212
F-statistic: 56.51 on 2 and 100 DF, p-value: < 2.2e-16
```

- 53% of the variance of dBE\_ts is explained by the regressors in the ECM model.
- All the variables are significant (except the intercept)
- Ljung-Box test: p-value=0.07 so H0 is not rejected, the model is valid. However, the decision to reject H0 in the test could be debated given the low p-value.
- The parameter ECT measures the speed of adjustment towards equilibrium.
- This model captures both the short-term and long-term relationship between the two TS in differences, which is significant.

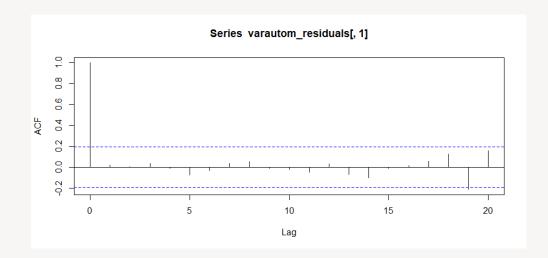
#### **VAR** selection

- The selectvar() function was used with dBE\_ts and dUE\_ts.
- The order of the VAR model selected by Schwarz's information criterion is 1, so a VAR(1) was estimated.

# **VAR(1)**

### VAR(1) for equation dBE

```
VAR Estimation Results:
Endogenous variables: dBE, dEU
Deterministic variables: const
Sample size: 102
Log Likelihood: -84.302
Roots of the characteristic polynomial:
0.4651 0.07444
Call:
VAR(y = data, p = 1, type = "const")
Estimation results for equation dBE:
_____
dBE = dBE.l1 + dEU.l1 + const
      Estimate Std. Error t value Pr(>|t|)
dBE. 11 0.10225
              0.12677
                         0.807
dEU.11 0.29446 0.11859 2.483
                                 0.0147
const -0.01119 0.04073 -0.275
                                 0.7841
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.4112 on 99 degrees of freedom
Multiple R-Squared: 0.1529,
                            Adjusted R-squared: 0.1358
F-statistic: 8.936 on 2 and 99 DF, p-value: 0.0002706
```

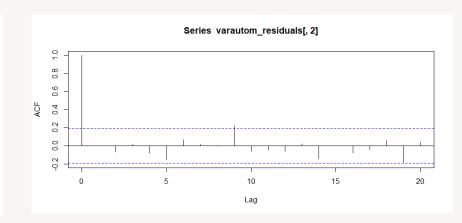


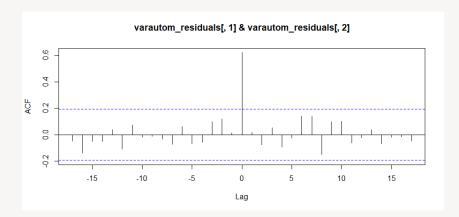
- The regressors are jointly significant, but only one regressor is significant (dEU.l1).
- The model explain 15.29% of the variance of dBE.
- Correlogram: there is only one significant residual.
- Ljung-Box test: p-value = 0.991, the residuals are white noise, the model is valid.

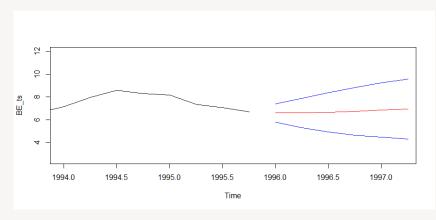
## **VAR(1)**

### VAR(1) for equation dEU

```
Estimation results for equation dEU:
     -----
dEU = dBE.l1 + dEU.l1 + const
       Estimate Std. Error t value Pr(>|t|)
                 0.131701
dEU. 11 0.288438
                 0.123201
                            2.341
                                   0.0212
const -0.005571
                 0.042314 -0.132
                                   0.8955
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.4272 on 99 degrees of freedom
Multiple R-Squared: 0.2033,
                              Adjusted R-squared: 0.1872
F-statistic: 12.63 on 2 and 99 DF. p-value: 1.302e-05
```



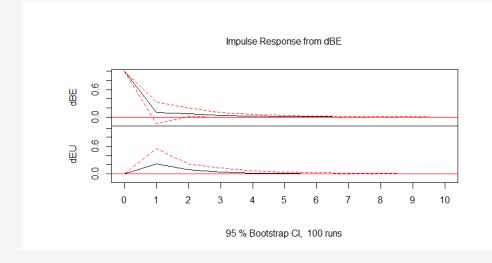


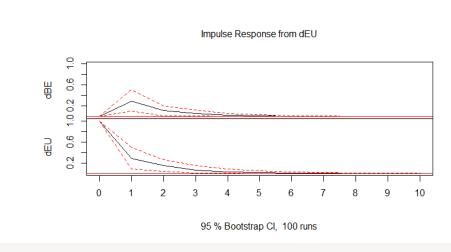


- The regressors are jointly significant.
- Only one regressor is significant (dEU.l1).
- The model explain 20.33% of the variance of dEU.
- Correlogram: there is one significant residual.
- Ljung-Box test: p-value=0.3971, the residuals are white noise.
- It is possible to say that the model is valid.
- Cross-correlogram: A strong significant cross-correlation without lag is observed. There is no problem with contemporaneous correlations in a VAR model, as it does not allow for explicit modeling of contemporaneous interdependence.
- The forecast is created using the var model. The forecast is different from the two created using the Arima models.

### **IRF**

### Impulse Response Functions based on the VAR(1) estimates





- The IRFs provide a simple way to interpret the estimated coefficients of the VAR model.
- Given a unitary impulse in dBE at time t, no significant response is observed.
- Given a unitary impulse in dEU at time t, a significant positive dBE response is observed at time t + 1.
- It is possible to use this information for forecasting dBE, by predicting a response if the value of dEU changes.
- From the VAR model we can infer that the value of the Belgian bond yields (in difference) in t depends on the value of the European bond yields (in difference) in t-1.
- This positive relationship potentially allows investors to predict Belgian bond yields performance using the Euro Area bond yields, letting them adjust their strategies.