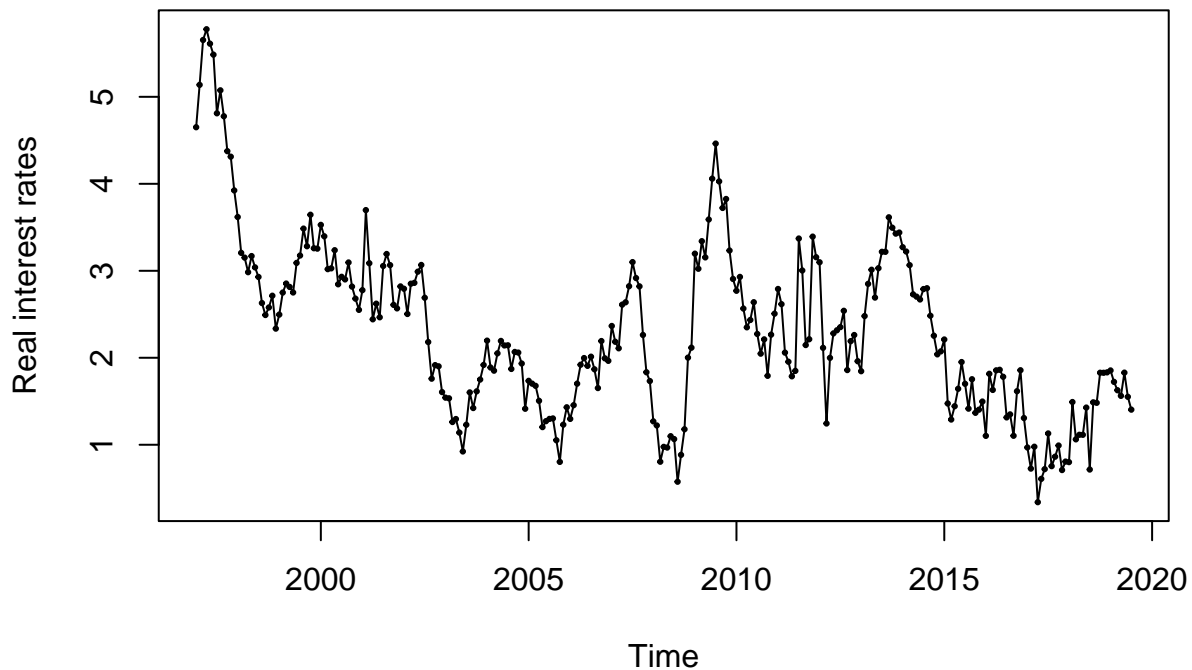


Problem Set 3 - Group 11

2023-04-05

The dataset contains data on *nominal interest rate* for the 10-year Italian government bond and the corresponding *inflation rate* represented by the Harmonised Index of Consumer Prices (HICP). The *real interest rate* can be computed by subtracting the inflation rate from the nominal interest rate. Such data is already provided in the dataset.

10-year Italian Bond Real Interest Rates



From the plot, we can see that the real interest rate appears to be volatile and exhibits some periods of stability, which suggests that an HMM could be a reasonable model. The use of HMMs in modeling financial time series has become increasingly popular in recent years, as they can capture the latent states or regimes that govern the behavior of the time series, and provide a flexible framework for modeling the dynamics of the time series. Moreover, HMMs can handle missing data or irregularly spaced observations, which is often the case in financial time series. Therefore, an HMM could be a reasonable model to capture the latent states or regimes that govern the behavior of the real interest rate in the Italian government bond market.

Model specification:

We specify the model to have a HMM with three states, and Gaussian emission distributions, with state-dependent mean and variance.

```
## Initial state probabilities model
## pr1 pr2 pr3
## 0.333 0.333 0.333
##
## Transition matrix
```

```
##           toS1  toS2  toS3
## fromS1 0.333 0.333 0.333
## fromS2 0.333 0.333 0.333
## fromS3 0.333 0.333 0.333
##
## Response parameters
## Resp 1 : gaussian
##      Re1.(Intercept) Re1.sd
## St1                0      1
## St2                0      1
## St3                0      1
```

The model specifies the starting values of the unknown parameters: the initial state probabilities S_0 , the transition matrix, and the parameters $\theta_1 = (\mu_1, \sigma_1)$; $\theta_2 = (\mu_2, \sigma_2)$.

Computation of the MLEs of the unknown parameters (model fitting):

The model fitting generates the following MLEs for the unknown parameters. We report the output that we get from R.

```
## Initial state probabilities model
## pr1 pr2 pr3
##   1   0   0
##
## Transition matrix
##           toS1  toS2  toS3
## fromS1 0.948 0.000 0.052
## fromS2 0.000 0.935 0.065
## fromS3 0.085 0.104 0.811
##
## Response parameters
## Resp 1 : gaussian
##      Re1.(Intercept) Re1.sd
## St1                3.169 0.737
## St2                1.290 0.351
## St3                2.019 0.207
```

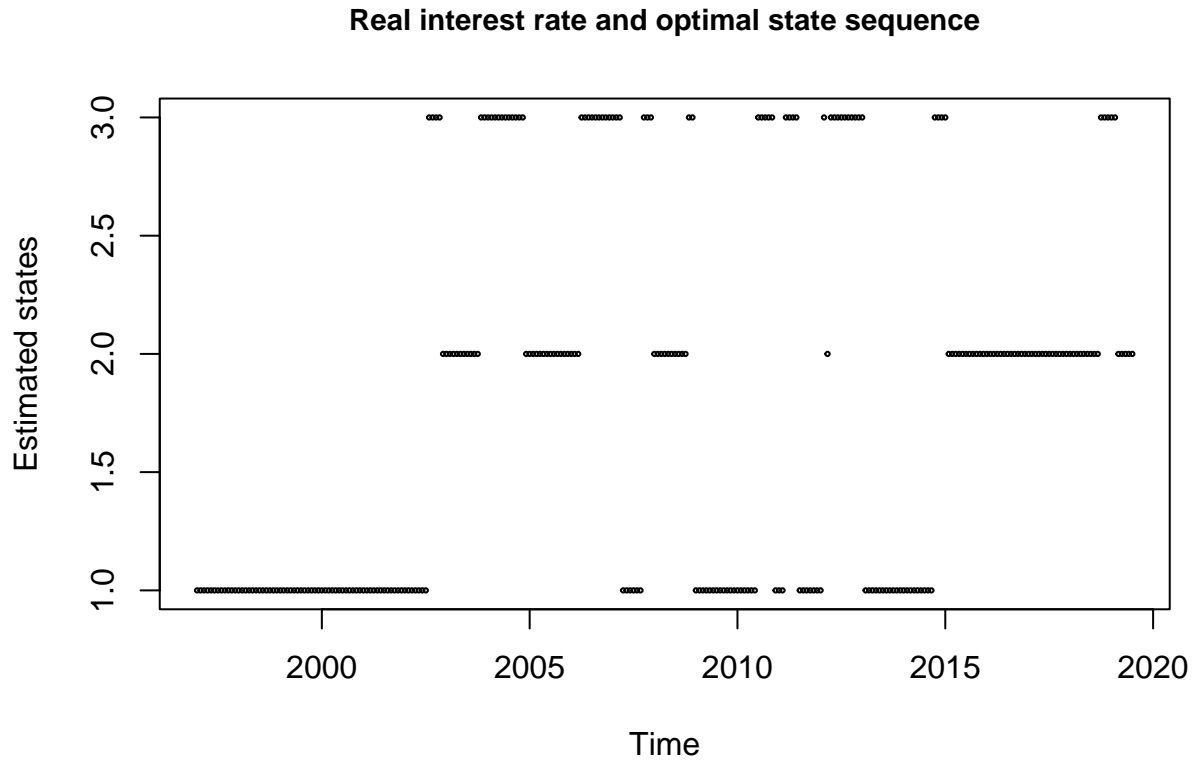
The following table summarizes the **MLEs** and their **associated standard errors** for the initial state probabilities, the transition matrix, and the mean and standard deviation corresponding to the three hidden states.

Coefficient	Standard error
1.000	NA
0.000	NA
0.000	NA
0.948	0.021
0.000	NA
0.052	0.021
0.000	NA
0.935	0.039
0.065	0.039
0.085	0.039
0.104	0.043
0.811	0.054
3.169	0.071
0.737	0.049

1.290	0.067
0.351	0.033
2.019	0.055
0.207	0.033

Decoding:

We plot the data with the estimated **optimal state sequence**.



We plot the data with the **estimated state-dependent means**.

The graph displays the real interest rate over time, with a black line representing the data and blue horizontal segments indicating regime shifts. The y-axis is labeled 'Real interest rate' and ranges from 1 to 5. The x-axis is labeled 'Time' and ranges from 1995 to 2020. The data shows a general downward trend with significant volatility. Regime shifts are marked by blue horizontal lines at various levels: approximately 3.2 (1995-2003), 2.0 (2004-2005), 1.3 (2006-2007), 3.2 (2008-2012), 2.0 (2013-2015), 1.3 (2016-2019), and 2.0 (2020).