

# IRF

CLMENT CARRIER

## FORECAST PERFORMANCE

In this section, using a sample from Q1 1998 to Q4 2009, I forecast HICP from Q1 2010 to Q4 2013 with several models. These models differ thanks to the number of lag used, if they are adaptive or not.

I plot the results and give the RMSE of these forecasts.

Here is the R code :

```
library(knitr); opts_chunk$set(message=FALSE)
```

```
require(lassovar)
require(ggplot2)
require(reshape2)
require(urca)
require(MSBVAR)
library(xtable)
```

```
forecast2<-function(data,lag,horizon,preforecast,adap){
  fore<-matrix(0,nrow=dim(data)[2],ncol=horizon+preforecast)
  fore[,1:(preforecast)]<-t(data[(dim(data)[1]-preforecast+1):dim(data)[1],])
  lv<-lassovar(dat=data,lags=lag,adaptive=adap,trend=TRUE)
  intercept<-as.matrix(lv$coefficients[1,],dim(data)[2],1)
  trend<-as.matrix(lv$coefficients[lag*dim(data)[2]+2,],dim(data)[2],1)
  if(lag==1){
    coeff<-as.matrix(t(lv$coefficients[2:(dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
    for (i in (preforecast+1):(horizon+preforecast)){
      fore[,i]<-intercept+trend*(i+dim(data)[1]-(preforecast))+coeff%*%fore[,i-1]
    }
  } else {
    if(lag==2){
      coeff1<-as.matrix(t(lv$coefficients[2:(dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
      coeff2<-as.matrix(t(lv$coefficients[(dim(data)[2]+2):(2*dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
      for (i in (preforecast+1):(horizon+preforecast)){
        fore[,i]<-intercept+trend*(i+dim(data)[1]-(preforecast))+coeff1%*%fore[,i-1]+coeff2%*%fore[,i-2]
      }
    } else {
      if(lag==3){
        coeff1<-as.matrix(t(lv$coefficients[2:(dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
        coeff2<-as.matrix(t(lv$coefficients[(dim(data)[2]+2):(2*dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
        coeff3<-as.matrix(t(lv$coefficients[(2*dim(data)[2]+2):(3*dim(data)[2]+1),]),dim(data)[2],dim(data)[2])
        for (i in (preforecast+1):(horizon+preforecast)){
          fore[,i]<-intercept+trend*(i+dim(data)[1]-(preforecast))+coeff1%*%fore[,i-1]+coeff2%*%fore[,i-2]+coeff3%*%fore[,i-3]
        }
      }
    }
  }
}
```

```

    coeff3<-as.matrix(t(lv$coefficients[(2*dim(data)[2]+2):(3*dim(data)[2]+1)],),dim(data)[2],dim(data)[1])
    for (i in (preforecast+1):(horizon+preforecast)){
      fore[,i]<-intercept+trend*(i+dim(data)[1]-(preforecast))+coeff1%*%fore[,i-1]+coeff2%*%fore[,i-1]+coeff3%*%fore[,i-1]+coeff4%*%fore[,i-1]
    }
  }
  else {
    coeff1<-as.matrix(t(lv$coefficients[2:(dim(data)[2]+1)],),dim(data)[2],dim(data)[1])
    coeff2<-as.matrix(t(lv$coefficients[(dim(data)[2]+2):(2*dim(data)[2]+1)],),dim(data)[2],dim(data)[1])
    coeff3<-as.matrix(t(lv$coefficients[(2*dim(data)[2]+2):(3*dim(data)[2]+1)],),dim(data)[2],dim(data)[1])
    coeff4<-as.matrix(t(lv$coefficients[(3*dim(data)[2]+2):(4*dim(data)[2]+1)],),dim(data)[2],dim(data)[1])
    for (i in (preforecast+1):(horizon+preforecast)){
      fore[,i]<-intercept+trend*(i+dim(data)[1]-(preforecast))+coeff1%*%fore[,i-1]+coeff2%*%fore[,i-1]+coeff3%*%fore[,i-1]+coeff4%*%fore[,i-1]
    }
  }
}
rownames(fore)<-names(data)
return(t(fore))
}

```

I load and keep the data from Q1 1998 to Q4 2009 :

```

load("vardata2")
data<-subset(vardataframe[117:164,])

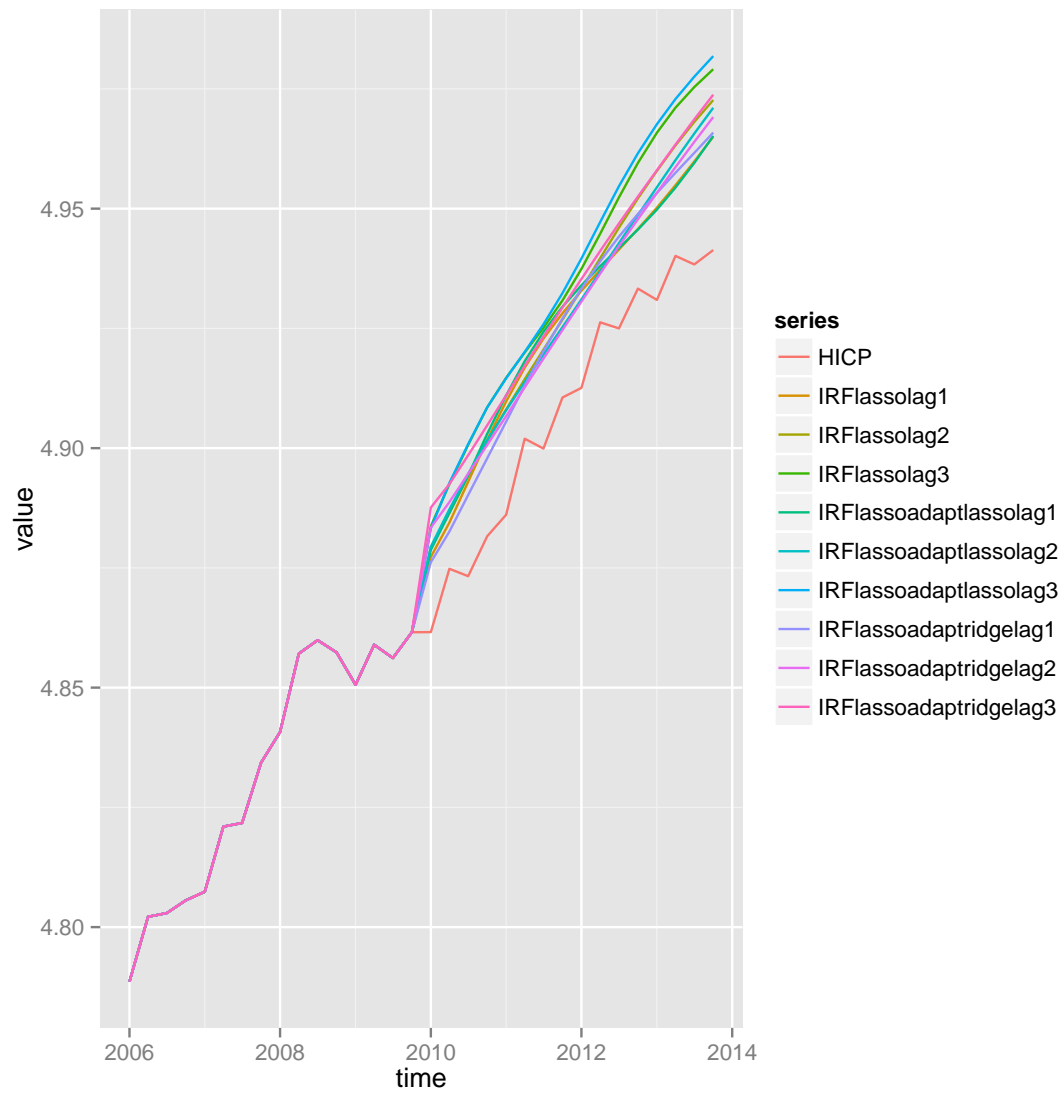
```

```

HICPtrue<-subset(vardataframe[149:180,])["HICP"]
IRFlasolag1<-forecast2(data,1,16,16,"none")[, "HICP"]
IRFlasolag2<-forecast2(data,2,16,16,"none")[, "HICP"]
IRFlasolag3<-forecast2(data,3,16,16,"none")[, "HICP"]
IRFlasoadaptlassolag1<-forecast2(data,1,16,16,"lasso")[, "HICP"]
## initial estimator for the adaptive lasso: lasso
IRFlasoadaptlassolag2<-forecast2(data,2,16,16,"lasso")[, "HICP"]
## initial estimator for the adaptive lasso: lasso
IRFlasoadaptlassolag3<-forecast2(data,3,16,16,"lasso")[, "HICP"]
## initial estimator for the adaptive lasso: lasso
IRFlasoadaptridgelag1<-forecast2(data,1,16,16,"ridge")[, "HICP"]
## initial estimator for the adaptive lasso: ridge
IRFlasoadaptridgelag2<-forecast2(data,2,16,16,"ridge")[, "HICP"]
## initial estimator for the adaptive lasso: ridge
IRFlasoadaptridgelag3<-forecast2(data,3,16,16,"ridge")[, "HICP"]
## initial estimator for the adaptive lasso: ridge
df<-data.frame(HICPtrue,IRFlasolag1,IRFlasolag2,IRFlasolag3,IRFlasoadaptlassolag1,IRFlasoadaptridgelag1,IRFlasoadaptridgelag2,IRFlasoadaptridgelag3)

time<-seq(as.Date("2006/01/01"), as.Date("2013/10/01"), by = "quarter")
df$time<-time
mvar1 <- melt(df, id = 'time', variable.name = 'series')
ggplot(mvar1, aes(time, value, col=series)) + geom_line()

```



I compute the RMSE :

```
df2<-df[17:32,]
RMSE<-NULL
for (i in 2:(length(df)-1)){
  RMSE[i]<-as.matrix(t(df2[,1]-df2[,i])%*(df2[,1]-df2[,i]))/16
}
RMSEmodel<-RMSE[-1]
names(RMSEmodel)<-names(df[2:10])
```

TABLE 1. blabla

Model	lag	adaptive	RMSE
1	1	non	0.000332
2	2	non	0.000446
3	3	non	0.000740
4	1	lasso	0.000361
5	2	lasso	0.000383
6	3	lasso	0.000813
7	1	ridge	0.000323
8	2	ridge	0.000361
9	3	ridge	0.000547