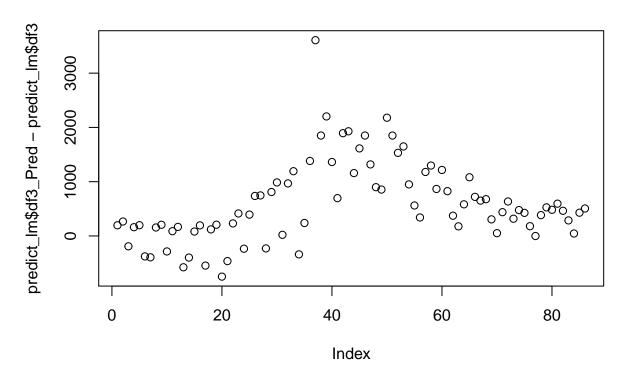
Utilisation de Caret avec un model predictif

```
#Installation de l'ensemble de ces packages
#install.packages("tidyverse")
#install.packages("caret")
#install.packages("mlbench")
#install.packages("lattice")
#install.packages("rpart")
\#install.packages("xlsx")
#tinytex::install_tinytex()
library(here)
## here() starts at C:/Users/manu_/Desktop
library(tidyverse)
## Warning: package 'tidyverse' was built under R version 4.0.3
## -- Attaching packages ------ tidyv
## v ggplot2 3.3.2 v purrr 0.3.4
## v tibble 3.0.3 v dplyr 1.0.2
## v tidyr 1.1.2 v stringr 1.4.0
## v readr 1.3.1 v forcats 0.5.0
## Warning: package 'tidyr' was built under R version 4.0.3
## Warning: package 'forcats' was built under R version 4.0.3
## -- Conflicts ------ tidyverse_c
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
library(caret)
## Warning: package 'caret' was built under R version 4.0.3
## Loading required package: lattice
## Warning: package 'lattice' was built under R version 4.0.3
##
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
##
      lift
```

```
library(readxl)
## Warning: package 'readxl' was built under R version 4.0.3
library(mlbench)
## Warning: package 'mlbench' was built under R version 4.0.3
library(rpart)
## Warning: package 'rpart' was built under R version 4.0.3
library(xlsx)
## Warning: package 'xlsx' was built under R version 4.0.3
set.seed(12345)
# Import des donnees
df <- readxl::read_xlsx(path ="C:/Users/manu_/Desktop/caret.xlsx")</pre>
# Changement du format de la date
date <- as.Date(df$dateRep)</pre>
df3 <- cbind(df,date)</pre>
head(df3)
##
     countriesAndTerritories
                               dateRep day month year cases deaths
## 1
                     France 2020-12-14 14
                                             12 2020 11533
                                                               150 2020-12-14
                     France 2020-12-13 13 12 2020 13947 194 2020-12-13
## 2
## 3
                    France 2020-12-12 12 12 2020 13406 627 2020-12-12
## 4
                    France 2020-12-11 11 12 2020 13750 292 2020-12-11
## 5
                     France 2020-12-10 10 12 2020 14595
                                                               296 2020-12-10
## 6
                     France 2020-12-09 9
                                              12 2020 13713
                                                               831 2020-12-09
# D'abord, on cree notre echantillon d'entrainement en prenant 75%
# du dataset et un echantillon test.
train_data <- df3[df3$date < '2020-09-19',]
test_data <- df3[df3$date > '2020-09-19',]
death <- test_data$deaths</pre>
attach(df3)
## The following object is masked _by_ .GlobalEnv:
##
##
       date
```

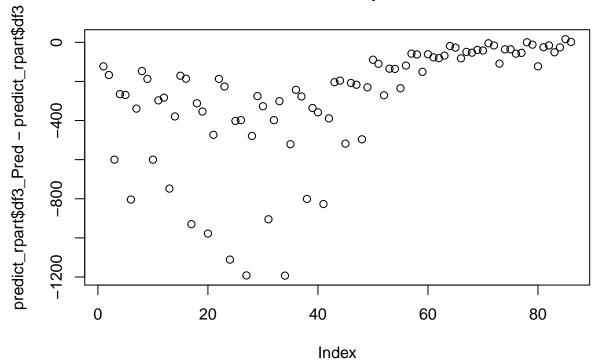
```
# Model 1: Creattion d'un modele de regression lineaire
# entrainer avec l'echantillon d'entrainement
model_lm <- lm(deaths ~ cases + date, data = train_data)</pre>
#Model 2: Utilisation de rpart pour cree un
# arbre de decison avec l'echantillon d'entrainement
model_rpart <- rpart(deaths ~ cases + date, data = train_data)</pre>
# On utlise les deux models pour faire
# une prediction en utilisant l'echantillon test
predict_lm <- predict(model_lm, test_data)</pre>
predict_lm <- data.frame(df3_Pred = predict_lm, df3 = test_data$deaths, date = test_data$date)</pre>
predict_rpart <- predict(model_rpart, test_data)</pre>
predict_rpart <- data.frame(df3_Pred = predict_rpart, df3 = test_data$deaths, date = test_data$date)</pre>
# On verifie les resultats en affichant les
# premieres lignes des dataset de predictions
head(predict_lm)
     df3_Pred df3
## 1 346.1586 150 2020-12-14
## 2 460.2733 194 2020-12-13
## 3 436.0360 627 2020-12-12
## 4 453.2341 292 2020-12-11
## 5 493.8888 296 2020-12-10
## 6 453.6860 831 2020-12-09
head(predict_rpart)
     df3_Pred df3
                        date
## 1 27.04348 150 2020-12-14
## 2 27.04348 194 2020-12-13
## 3 27.04348 627 2020-12-12
## 4 27.04348 292 2020-12-11
## 5 27.04348 296 2020-12-10
## 6 27.04348 831 2020-12-09
# On compare les resultats obtenue a l'aide d'un graph
par(mfrow = c(1, 1))
plot(predict_lm$df3_Pred - predict_lm$df3, main = "Difference lm")
```

Difference Im



plot(predict_rpart\$df3_Pred - predict_rpart\$df3, main = "Difference rpart")

Difference rpart



D'apres les resultats des 2 graphiques, on peut retenir le modele de regression lineaire # car le nuage de points comme une reaction normal d'un virus, c'est a dire sur une premiere # phase croissante qui correspond a la propagation du virus puis une deuxieme phase de pic # des morts et pour finir une phase decroissante de la lethalite.

Nous pouvons voir qu'il y aura eventuelle troisieme vague.

#Bibliographie

https://docs.microsoft.com/fr-fr/sql/machine-learning/tutorials/r-predictive-model-train?view = sql-server-ver15

http://eric.univ-lyon2.fr/~ricco/tanagra/fichiers/fr_Tanagra_package_caret.pdf

https://www.ecdc.europa.eu/en/publications-data/download-todays-data-geographic-distribution-covid-19-cases-worldwide

https://github.com/dsobo/caret-beginner-tutorial

 $https://rstudio-pubs-static.s3.amazonaws.com/253860_05f11cddd938407a9cb3b06d9dc38c9a.html$