

## caret

```
# On télécharge le paquet
install.packages("caret")
```

```
# on charge le paquet
library(caret)

# On utilisera le dataset 'mtcars'
data("cars")
# Utilisation de 'set.seed()'
set.seed(300)

# On imposera la loi de partition des données

trainIndex <- createDataPartition(cars$Price, p = .8,
                                   list = FALSE,
                                   times = 1)

# On observe le contenu des 6 première ligne du tableau
head(cars)
```

```
##      Price Mileage Cylinder Doors Cruise Sound Leather Buick Cadillac Chevy
## 1 22661.05  20105         6     4       1     0       0     1       0     0
## 2 21725.01  13457         6     2       1     1       0     0       0     1
## 3 29142.71  31655         4     2       1     1       1     0       0     0
## 4 30731.94  22479         4     2       1     0       0     0       0     0
## 5 33358.77  17590         4     2       1     1       1     0       0     0
## 6 30315.17  23635         4     2       1     0       0     0       0     0
## Pontiac Saab Saturn convertible coupe hatchback sedan wagon
## 1      0     0     0           0     0           0     1     0
## 2      0     0     0           0     1           0     0     0
## 3      0     1     0           1     0           0     0     0
## 4      0     1     0           1     0           0     0     0
## 5      0     1     0           1     0           0     0     0
## 6      0     1     0           1     0           0     0     0
```

```
# On vérifie le contenu du data set
str(cars)
```

```
## 'data.frame':   804 obs. of  18 variables:
## $ Price      : num  22661 21725 29143 30732 33359 ...
## $ Mileage    : int   20105 13457 31655 22479 17590 23635 17381 27558 25049 17319 ...
## $ Cylinder   : int    6 6 4 4 4 4 4 4 4 4 ...
## $ Doors      : int    4 2 2 2 2 2 2 2 2 4 ...
## $ Cruise     : int    1 1 1 1 1 1 1 1 1 1 ...
## $ Sound      : int    0 1 1 0 1 0 1 0 0 0 ...
```

```
## $ Leather      : int  0 0 1 0 1 0 1 1 0 1 ...
## $ Buick        : int  1 0 0 0 0 0 0 0 0 0 ...
## $ Cadillac     : int  0 0 0 0 0 0 0 0 0 0 ...
## $ Chevy        : int  0 1 0 0 0 0 0 0 0 0 ...
## $ Pontiac      : int  0 0 0 0 0 0 0 0 0 0 ...
## $ Saab         : int  0 0 1 1 1 1 1 1 1 1 ...
## $ Saturn       : int  0 0 0 0 0 0 0 0 0 0 ...
## $ convertible: int  0 0 1 1 1 1 1 1 1 0 ...
## $ coupe        : int  0 1 0 0 0 0 0 0 0 0 ...
## $ hatchback    : int  0 0 0 0 0 0 0 0 0 0 ...
## $ sedan        : int  1 0 0 0 0 0 0 0 0 1 ...
## $ wagon        : int  0 0 0 0 0 0 0 0 0 0 ...
```

```
# On établie les partitions de test et d'entraînement
```

```
carsTrain <- cars[trainIndex,]
carsTest  <- cars[-trainIndex,]
```

```
# zone de test
```

```
fitControl <- trainControl(## 5-fold CV
  method = "repeatedcv",
  number = 5,
  # Répéter 5 fois
  repeats = 5)
```

```
set.seed(300)
```

```
t_before <- Sys.time()
```

```
Fit1 <- train(Price ~ ., data = carsTrain,
  method = "rf", # methode d'arbre de décision random forest(rf), on peut tester "nnet" aus
  trControl = fitControl,
  verbose = FALSE)
```

```
t_after <- Sys.time()
```

```
duree <- t_after - t_before # code pour voir combien de temps ca va durée
print(duree)
```

```
## Time difference of 42.23284 secs
```

```
# on teste le modèle
```

```
Fit1
```

```
## Random Forest
```

```
##
```

```
## 644 samples
```

```
## 17 predictor
```

```
##
```

```
## No pre-processing
```

```
## Resampling: Cross-Validated (5 fold, repeated 5 times)
```

```
## Summary of sample sizes: 516, 514, 515, 516, 515, 516, ...
```

```
## Resampling results across tuning parameters:
```

```
##
```

```
## mtry RMSE      Rsquared  MAE
## 2    3665.833  0.9116957  2705.964
## 9    2204.804  0.9517320  1538.431
```

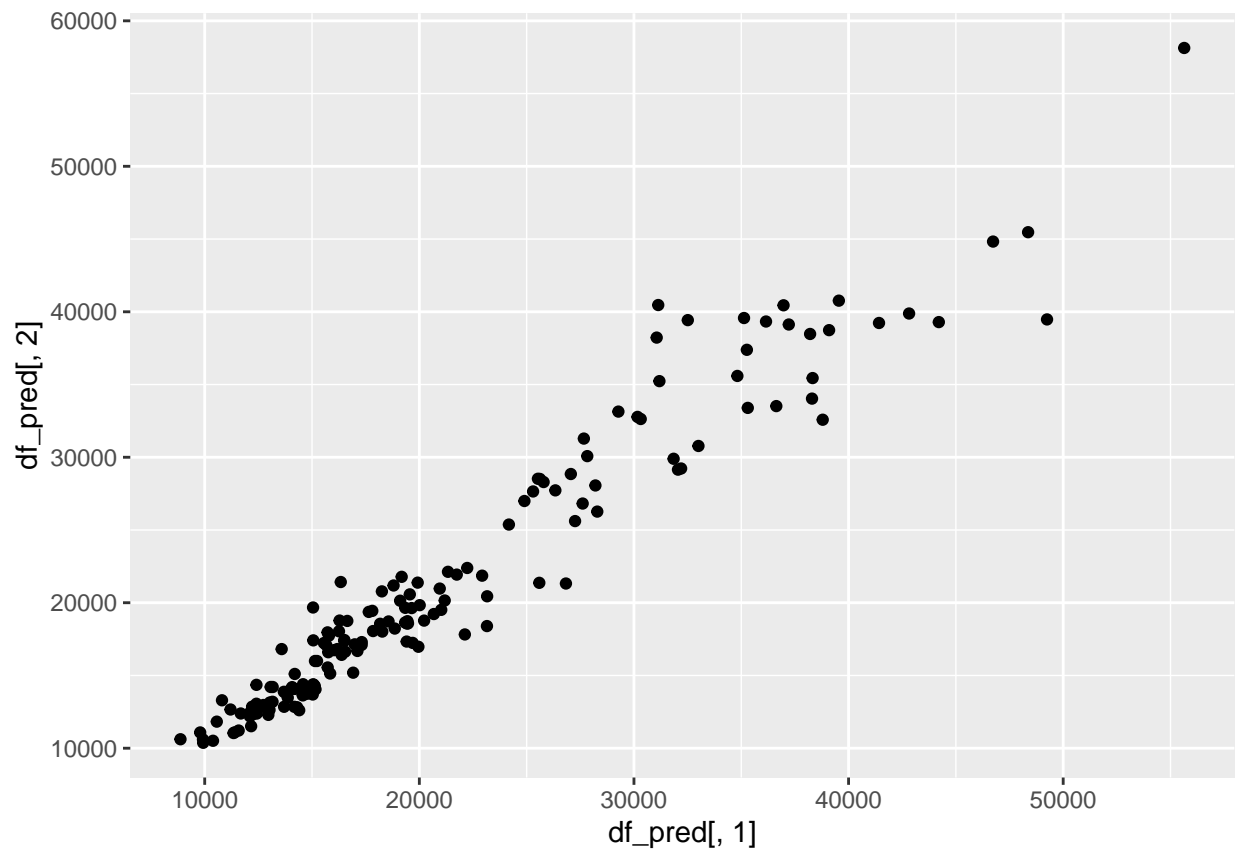
```
## 17 2301.991 0.9471867 1620.119
##
## RMSE was used to select the optimal model using the smallest value.
## The final value used for the model was mtry = 9.

# Définir le modèle de prédiction pour le test
prediction_carsTest <- predict(Fit1, carsTest)
df_pred <- data.frame(carsTest$Price, prediction_carsTest )

ecart <- df_pred[,1] == df_pred[,2]
nombre_reussites <- sum(ecart)
nombre <- nrow(df_pred)
accuracy <- nombre_reussites/nombre * 100

# on utilisera le paquet ggplot2
library(ggplot2)

qplot(df_pred[,1], df_pred[,2])
```



```
# Représentation graphique
ggplot(df_pred) +
  aes(x = carsTest.Price, fill = prediction_carsTest) +
  geom_histogram() +
  scale_fill_hue() +
  labs(x = "Réalité", y = "Prévision", title = "Réalité versus prevision par classe", fill = "Région")
  theme_gray()
```

Réalité versus prevision par classe

