Predictive Analytics Solutions to Hands On Exercises

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Hands on Linear Regression and Random Forests

Load in the data set Servo from package **mlbench** and answer the following questions:

- 1 How would you obtain a random forest with 750 trees to forecast the value of *Class* (it is a numeric variable) solution
- 2 Repeat the previous exercise but now using a linear regression model. solution
- Obtain the predictions of the two previous models for the data used to obtain them. Draw a scatterplot comparing these predictions colution
- A Split the data in train and test sets (80%-20%). Obtain the two previous models on the training data and get their predictions for the test set. Compare the predictions of the models.

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 How would you obtain a random forest with 750 trees to forecast the value of *Class* (it is a numeric variable)

```
library(tidymodels)
data(Servo, package="mlbench")
s <- as_tibble(Servo)

rfSpec <-
  rand_forest(trees = 750) %>%  # the type of model
  set_engine("ranger") %>%  # the implementation to use
  set_mode("regression")  # type of task

rf <-
  rfSpec %>% fit(Class ~ ., data = s)
```

Go back

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Repeat the previous exercise but now using a linear regression model.



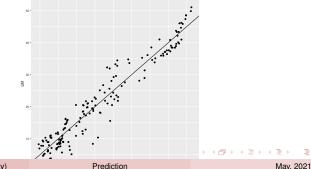
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 Obtain the predictions of the two previous models for the data used to obtain them. Draw a scatterplot comparing these predictions

```
psrf <- predict(rf,s)
pslm <- predict(lm,s)
preds <- psrf %>% rename(RF = .pred) %>%
    bind_cols(pslm) %>% rename(LM = .pred)
library(ggplot2)
ggplot(preds, aes(x=RF,y=LM)) + geom_point() + geom_abline(slope=1, intercept = 0)
```



2

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Split the data in train and test sets (80%-20%). Obtain the two previous models on the training data and get their predictions for the test set. Compare the predictions of the models.

```
spl <- initial split(s, prop = 0.8)</pre>
sTR <- training(spl)</pre>
sTS <- testing(spl)
rf <- rfSpec %>% fit (Class ~ ., sTR)
lm <- lmSpec %>% fit(Class ~ ., sTR)
psrf <- predict (rf, sTS)
pslm <- predict (lm, sTS)
preds <- psrf %>% rename(RF = .pred) %>% bind cols(pslm) %>% rename(LM = .pred)
sTS %>% bind cols(preds) %>% metrics(Class,RF)
## # A tibble: 3 x 3
## .metric .estimator .estimate
## <chr> <chr> <dbl>
## 1 rmse standard 3.77
## 2 rsg standard 0.950
## 3 mae standard 2.69
sTS %>% bind cols(preds) %>% metrics(Class,LM)
## # A tibble 3 x 3
    .metric .estimator .estimate
    <chr> <chr> <chr> <dbl>
                       5.41
         standard
## 1 rmse
## 2
          standard
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                                           Prediction
```