# Prediction in R

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# Install package

#### #install new package

install.packages('randomForest',dependencies = T)
install.packages('rpart',dependencies = T)
install.packages('MASS',dependencies = T)

#### #import your packages before using it

library(randomForest) library(MASS) library(rpart)

#### #check package information

help(rpart) help(randomForest) help(lda)

#### #package information

rpart: recursive partitioning and regression tree

randomForest: classification and regression with random forest.

Ida: linear discriminant analysis

# Load your data

Training data: data that is used for training the model, all decision tree, random forest and Ida model

Testing without label: data without incidents, you need to predict the incidents(bite or not bite).

Testing data: sample solution including the right answer to calculate the error rate of the result.

#### #import training and testing data

lou\_training <- read.csv("~/workspace/data\_literacy/week10/lou\_training.csv")
lou\_testing\_nolabel <- read.csv("~/workspace/data\_literacy/week10/lou\_testing\_nolabel.csv")</pre>

# Train the model

rpart(): train the relation between INCIDENT and other attributes using decision tree model.

randomForest(): train the relation between INCIDENT and other attributes using random forest model.

Ida(): train the relation between INCIDENT and other attributes using linear discriminant analysis model.

#### #train data model

lou.tree <- rpart(INCIDENT ~ PENALTY + VICTIM + LOCATION +COUNTRY + RESULT + GOALS + YEAR,data=lou\_training)

lou.forest <- randomForest(INCIDENT ~ PENALTY + VICTIM + LOCATION +COUNTRY + RESULT + GOALS + YEAR,data=lou\_training)

lou.lda <- lda(INCIDENT ~ PENALTY + VICTIM + LOCATION +COUNTRY + RESULT + GOALS + YEAR,data=lou\_training)

# Prediction

We train data by training data set, in which the incident (bite or not bite) is given. Then we predict the incident type (if Louis bites or not in a game) in the testing data.

# Now use these \*models\* to predict the labels in the testing data (i.e. predict whether or not a bite incident)

pred.tree <- predict(lou.tree,newdata=lou\_testing\_nolabel,type="class")
pred.forest <- predict(lou.forest,newdata=lou\_testing\_nolabel,type="class")
pred.lda <- predict(lou.lda,newdata=lou\_testing\_nolabel)\$class

# Calculate the error

When prediction is done, the error can be calculated by the standard solution( which is not given when you do the project). lou\_testing <- read.csv("~/workspace/data\_literacy/week10/lou\_testing.csv")
incident <- lou\_testing\$INCIDENT</pre>

#### #calculate the ratio of wrong prediction

error.tree <- mean(pred.tree != incident)
error.forest <- mean(pred.forest != incident)
error.lda <- mean(pred.lda != incident)

# Error rate

```
> error.tree <- mean(pred.tree != incident)
> error.forest <- mean(pred.forest != incident)
> error.lda <- mean(pred.lda != incident)
> error.tree
[1] 0.1179487
> error.lda
[1] 0.1179487
> error.forest
[1] 0.03076923
> |
```

Here, the random forest model has the lowest error rate.

In your project, you can use 20% of training data to do test and calculate the error rate when the test solution is not given to you.