# Decision Tree Tutorial Part 1 - IRIS

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### Use C5.0 Algorithm on the IRIS Dataset

### Step 1: Load required packages and libraries

```
library(C50)
data(iris)
np <- ceiling(0.1*nrow(iris))</pre>
## [1] 15
```

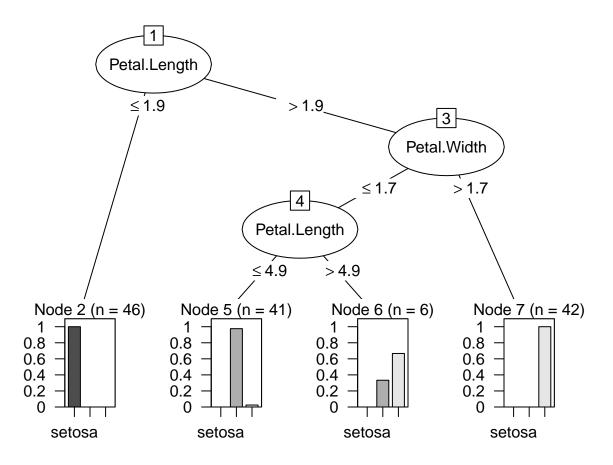
### Step 2: Set Seed and create training and test datasets

```
set.seed(3)
test.index <- sample(1:nrow(iris), np)</pre>
iris.test <- iris[test.index, ]</pre>
iris.train <- iris[-test.index, ]</pre>
```

## Step 3: Run C5.0 Algorithm

```
c <- C5.0Control(subset = FALSE,
                 bands = 0,
                 winnow = FALSE,
                 noGlobalPruning = FALSE,
                 CF = 0.25,
                 minCases = 2,
                 fuzzyThreshold = FALSE,
                 sample = 0,
                 seed = sample.int(4096, size = 1) -1L,
                 earlyStopping = TRUE
iris_treeModel <- C5.0(x = iris.train[, -5], y = iris.train$Species,control =c)
summary(iris_treeModel)
##
## C5.0.default(x = iris.train[, -5], y = iris.train$Species, control = c)
##
##
## C5.0 [Release 2.07 GPL Edition]
                                        Mon Feb 11 17:52:16 2019
##
## Class specified by attribute `outcome'
## Read 135 cases (5 attributes) from undefined.data
## Decision tree:
```

```
##
## Petal.Length <= 1.9: setosa (46)
## Petal.Length > 1.9:
## :...Petal.Width > 1.7: virginica (42)
       Petal.Width <= 1.7:</pre>
##
##
       :...Petal.Length <= 4.9: versicolor (41/1)
           Petal.Length > 4.9: virginica (6/2)
##
##
##
## Evaluation on training data (135 cases):
##
        Decision Tree
##
##
      Size
                Errors
##
              3(2.2%)
##
         4
##
##
##
                           <-classified as
       (a)
             (b)
                    (c)
##
      ----
                           (a): class setosa
##
        46
##
              40
                      2
                           (b): class versicolor
##
                    46
                           (c): class virginica
               1
##
##
##
   Attribute usage:
##
##
   100.00% Petal.Length
     65.93% Petal.Width
##
##
##
## Time: 0.0 secs
plot(iris_treeModel)
```



Step 4: Test Output and Accuracy

```
test.output <- predict(iris_treeModel, iris.test[, -5], type = "class")
n <- length(test.output)
number = 0
for ( i in 1:n){
   if(test.output[i] == iris.test[i, 5])
   {
      number=number+1}
}
test.accuracy = number/n*100
test.accuracy</pre>
```

## [1] 93.33333