

```

type ?new() to see new features/changes/bug fixes.
> # Loading data
> data(iris)
> # Structure
> str(iris)
'data.frame': 150 obs. of 5 variables:
 $ Sepal.Length: num 5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
 $ Sepal.Width : num 3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
 $ Petal.Length: num 1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
 $ Petal.Width : num 0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
 $ Species : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...
> # Splitting data in train and test data
> split <- sample.split(iris, SplitRatio = 0.7)
> split
[1] FALSE TRUE TRUE FALSE TRUE
> train <- subset(iris, split == "TRUE")
> test <- subset(iris, split == "FALSE")
> # Fitting Random Forest to the train dataset
> set.seed(120) # Setting seed
> classifier_RF = randomForest(x = train[-5],
+ y = train$Species,
+ ntree = 500)
> classifier_RF

Call:
randomForest(x = train[-5], y = train$Species, ntree = 500)
Type of random forest: classification
Number of trees: 500
No. of variables tried at each split: 2

OOB estimate of error rate: 3.33%
Confusion matrix:
      setosa versicolor virginica class.error
setosa      30          0          0 0.00000000
versicolor   0         29          1 0.03333333
virginica     0          2         28 0.06666667
> # Predicting the Test set results
> y_pred = predict(classifier_RF, newdata = test[-5])
> # Confusion Matrix
> confusion_mtx = table(test[, 5], y_pred)
> confusion_mtx
      y_pred
      setosa versicolor virginica
setosa      20          0          0
versicolor   0         18          2
virginica     0          0         20
> # Plotting model
> plot(classifier_RF)
> # Importance plot
> importance(classifier_RF)
      MeanDecreaseGini
Sepal.Length      5.772467
Sepal.Width       1.419426
Petal.Length      25.728012
Petal.Width       26.374384
> # Variable importance plot
> varImpPlot(classifier_RF)
>

```

