

Decision Tree example

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R Markdown

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When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
library(caTools)
library(rpart)
```

```
set.seed(1)
split <- sample.split(iris,SplitRatio = 0.7)
split
```

```
## [1] TRUE TRUE FALSE FALSE TRUE
```

```
train <- subset(iris, split = "TRUE")
test <- subset(iris, split = "FALSE")
train
```

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	5.1	3.5	1.4	0.2	setosa
## 2	4.9	3.0	1.4	0.2	setosa
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 5	5.0	3.6	1.4	0.2	setosa
## 6	5.4	3.9	1.7	0.4	setosa
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa

## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa
## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 65	5.6	2.9	3.6	1.3	versicolor
## 66	6.7	3.1	4.4	1.4	versicolor
## 67	5.6	3.0	4.5	1.5	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 70	5.6	2.5	3.9	1.1	versicolor
## 71	5.9	3.2	4.8	1.8	versicolor
## 72	6.1	2.8	4.0	1.3	versicolor

## 73	6.3	2.5	4.9	1.5 versicolor
## 74	6.1	2.8	4.7	1.2 versicolor
## 75	6.4	2.9	4.3	1.3 versicolor
## 76	6.6	3.0	4.4	1.4 versicolor
## 77	6.8	2.8	4.8	1.4 versicolor
## 78	6.7	3.0	5.0	1.7 versicolor
## 79	6.0	2.9	4.5	1.5 versicolor
## 80	5.7	2.6	3.5	1.0 versicolor
## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica
## 111	6.5	3.2	5.1	2.0 virginica
## 112	6.4	2.7	5.3	1.9 virginica
## 113	6.8	3.0	5.5	2.1 virginica
## 114	5.7	2.5	5.0	2.0 virginica
## 115	5.8	2.8	5.1	2.4 virginica
## 116	6.4	3.2	5.3	2.3 virginica
## 117	6.5	3.0	5.5	1.8 virginica
## 118	7.7	3.8	6.7	2.2 virginica
## 119	7.7	2.6	6.9	2.3 virginica
## 120	6.0	2.2	5.0	1.5 virginica
## 121	6.9	3.2	5.7	2.3 virginica
## 122	5.6	2.8	4.9	2.0 virginica
## 123	7.7	2.8	6.7	2.0 virginica
## 124	6.3	2.7	4.9	1.8 virginica
## 125	6.7	3.3	5.7	2.1 virginica
## 126	7.2	3.2	6.0	1.8 virginica

## 127	6.2	2.8	4.8	1.8	virginica
## 128	6.1	3.0	4.9	1.8	virginica
## 129	6.4	2.8	5.6	2.1	virginica
## 130	7.2	3.0	5.8	1.6	virginica
## 131	7.4	2.8	6.1	1.9	virginica
## 132	7.9	3.8	6.4	2.0	virginica
## 133	6.4	2.8	5.6	2.2	virginica
## 134	6.3	2.8	5.1	1.5	virginica
## 135	6.1	2.6	5.6	1.4	virginica
## 136	7.7	3.0	6.1	2.3	virginica
## 137	6.3	3.4	5.6	2.4	virginica
## 138	6.4	3.1	5.5	1.8	virginica
## 139	6.0	3.0	4.8	1.8	virginica
## 140	6.9	3.1	5.4	2.1	virginica
## 141	6.7	3.1	5.6	2.4	virginica
## 142	6.9	3.1	5.1	2.3	virginica
## 143	5.8	2.7	5.1	1.9	virginica
## 144	6.8	3.2	5.9	2.3	virginica
## 145	6.7	3.3	5.7	2.5	virginica
## 146	6.7	3.0	5.2	2.3	virginica
## 147	6.3	2.5	5.0	1.9	virginica
## 148	6.5	3.0	5.2	2.0	virginica
## 149	6.2	3.4	5.4	2.3	virginica
## 150	5.9	3.0	5.1	1.8	virginica

test

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 1	5.1	3.5	1.4	0.2	setosa
## 2	4.9	3.0	1.4	0.2	setosa
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 5	5.0	3.6	1.4	0.2	setosa
## 6	5.4	3.9	1.7	0.4	setosa
## 7	4.6	3.4	1.4	0.3	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 10	4.9	3.1	1.5	0.1	setosa
## 11	5.4	3.7	1.5	0.2	setosa
## 12	4.8	3.4	1.6	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 15	5.8	4.0	1.2	0.2	setosa
## 16	5.7	4.4	1.5	0.4	setosa
## 17	5.4	3.9	1.3	0.4	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 20	5.1	3.8	1.5	0.3	setosa
## 21	5.4	3.4	1.7	0.2	setosa
## 22	5.1	3.7	1.5	0.4	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 25	4.8	3.4	1.9	0.2	setosa
## 26	5.0	3.0	1.6	0.2	setosa

## 27	5.0	3.4	1.6	0.4	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 30	4.7	3.2	1.6	0.2	setosa
## 31	4.8	3.1	1.6	0.2	setosa
## 32	5.4	3.4	1.5	0.4	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 35	4.9	3.1	1.5	0.2	setosa
## 36	5.0	3.2	1.2	0.2	setosa
## 37	5.5	3.5	1.3	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 40	5.1	3.4	1.5	0.2	setosa
## 41	5.0	3.5	1.3	0.3	setosa
## 42	4.5	2.3	1.3	0.3	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 45	5.1	3.8	1.9	0.4	setosa
## 46	4.8	3.0	1.4	0.3	setosa
## 47	5.1	3.8	1.6	0.2	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 50	5.0	3.3	1.4	0.2	setosa
## 51	7.0	3.2	4.7	1.4	versicolor
## 52	6.4	3.2	4.5	1.5	versicolor
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 55	6.5	2.8	4.6	1.5	versicolor
## 56	5.7	2.8	4.5	1.3	versicolor
## 57	6.3	3.3	4.7	1.6	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 60	5.2	2.7	3.9	1.4	versicolor
## 61	5.0	2.0	3.5	1.0	versicolor
## 62	5.9	3.0	4.2	1.5	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 65	5.6	2.9	3.6	1.3	versicolor
## 66	6.7	3.1	4.4	1.4	versicolor
## 67	5.6	3.0	4.5	1.5	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 70	5.6	2.5	3.9	1.1	versicolor
## 71	5.9	3.2	4.8	1.8	versicolor
## 72	6.1	2.8	4.0	1.3	versicolor
## 73	6.3	2.5	4.9	1.5	versicolor
## 74	6.1	2.8	4.7	1.2	versicolor
## 75	6.4	2.9	4.3	1.3	versicolor
## 76	6.6	3.0	4.4	1.4	versicolor
## 77	6.8	2.8	4.8	1.4	versicolor
## 78	6.7	3.0	5.0	1.7	versicolor
## 79	6.0	2.9	4.5	1.5	versicolor
## 80	5.7	2.6	3.5	1.0	versicolor

## 81	5.5	2.4	3.8	1.1 versicolor
## 82	5.5	2.4	3.7	1.0 versicolor
## 83	5.8	2.7	3.9	1.2 versicolor
## 84	6.0	2.7	5.1	1.6 versicolor
## 85	5.4	3.0	4.5	1.5 versicolor
## 86	6.0	3.4	4.5	1.6 versicolor
## 87	6.7	3.1	4.7	1.5 versicolor
## 88	6.3	2.3	4.4	1.3 versicolor
## 89	5.6	3.0	4.1	1.3 versicolor
## 90	5.5	2.5	4.0	1.3 versicolor
## 91	5.5	2.6	4.4	1.2 versicolor
## 92	6.1	3.0	4.6	1.4 versicolor
## 93	5.8	2.6	4.0	1.2 versicolor
## 94	5.0	2.3	3.3	1.0 versicolor
## 95	5.6	2.7	4.2	1.3 versicolor
## 96	5.7	3.0	4.2	1.2 versicolor
## 97	5.7	2.9	4.2	1.3 versicolor
## 98	6.2	2.9	4.3	1.3 versicolor
## 99	5.1	2.5	3.0	1.1 versicolor
## 100	5.7	2.8	4.1	1.3 versicolor
## 101	6.3	3.3	6.0	2.5 virginica
## 102	5.8	2.7	5.1	1.9 virginica
## 103	7.1	3.0	5.9	2.1 virginica
## 104	6.3	2.9	5.6	1.8 virginica
## 105	6.5	3.0	5.8	2.2 virginica
## 106	7.6	3.0	6.6	2.1 virginica
## 107	4.9	2.5	4.5	1.7 virginica
## 108	7.3	2.9	6.3	1.8 virginica
## 109	6.7	2.5	5.8	1.8 virginica
## 110	7.2	3.6	6.1	2.5 virginica
## 111	6.5	3.2	5.1	2.0 virginica
## 112	6.4	2.7	5.3	1.9 virginica
## 113	6.8	3.0	5.5	2.1 virginica
## 114	5.7	2.5	5.0	2.0 virginica
## 115	5.8	2.8	5.1	2.4 virginica
## 116	6.4	3.2	5.3	2.3 virginica
## 117	6.5	3.0	5.5	1.8 virginica
## 118	7.7	3.8	6.7	2.2 virginica
## 119	7.7	2.6	6.9	2.3 virginica
## 120	6.0	2.2	5.0	1.5 virginica
## 121	6.9	3.2	5.7	2.3 virginica
## 122	5.6	2.8	4.9	2.0 virginica
## 123	7.7	2.8	6.7	2.0 virginica
## 124	6.3	2.7	4.9	1.8 virginica
## 125	6.7	3.3	5.7	2.1 virginica
## 126	7.2	3.2	6.0	1.8 virginica
## 127	6.2	2.8	4.8	1.8 virginica
## 128	6.1	3.0	4.9	1.8 virginica
## 129	6.4	2.8	5.6	2.1 virginica
## 130	7.2	3.0	5.8	1.6 virginica
## 131	7.4	2.8	6.1	1.9 virginica
## 132	7.9	3.8	6.4	2.0 virginica
## 133	6.4	2.8	5.6	2.2 virginica
## 134	6.3	2.8	5.1	1.5 virginica

```
## 135      6.1      2.6      5.6      1.4 virginica
## 136      7.7      3.0      6.1      2.3 virginica
## 137      6.3      3.4      5.6      2.4 virginica
## 138      6.4      3.1      5.5      1.8 virginica
## 139      6.0      3.0      4.8      1.8 virginica
## 140      6.9      3.1      5.4      2.1 virginica
## 141      6.7      3.1      5.6      2.4 virginica
## 142      6.9      3.1      5.1      2.3 virginica
## 143      5.8      2.7      5.1      1.9 virginica
## 144      6.8      3.2      5.9      2.3 virginica
## 145      6.7      3.3      5.7      2.5 virginica
## 146      6.7      3.0      5.2      2.3 virginica
## 147      6.3      2.5      5.0      1.9 virginica
## 148      6.5      3.0      5.2      2.0 virginica
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica
```

```
decision_tree_model <- rpart(Species ~., data = train, method = "class")
summary(decision_tree_model)
```

```
## Call:
## rpart(formula = Species ~ ., data = train, method = "class")
##   n= 150
##
##      CP nsplit rel error xerror      xstd
## 1 0.50     0     1.00   1.14 0.05230679
## 2 0.44     1     0.50   0.60 0.06000000
## 3 0.01     2     0.06   0.09 0.02908608
##
## Variable importance
##   Petal.Width Petal.Length Sepal.Length  Sepal.Width
##           34           31           21           14
##
## Node number 1: 150 observations,      complexity param=0.5
##   predicted class=setosa      expected loss=0.6666667  P(node) =1
##   class counts:    50    50    50
##   probabilities: 0.333 0.333 0.333
##   left son=2 (50 obs) right son=3 (100 obs)
##   Primary splits:
##     Petal.Length < 2.45 to the left,  improve=50.00000, (0 missing)
##     Petal.Width < 0.8  to the left,  improve=50.00000, (0 missing)
##     Sepal.Length < 5.45 to the left,  improve=34.16405, (0 missing)
##     Sepal.Width < 3.35 to the right, improve=19.03851, (0 missing)
##   Surrogate splits:
##     Petal.Width < 0.8  to the left,  agree=1.000, adj=1.00, (0 split)
##     Sepal.Length < 5.45 to the left,  agree=0.920, adj=0.76, (0 split)
##     Sepal.Width < 3.35 to the right, agree=0.833, adj=0.50, (0 split)
##
## Node number 2: 50 observations
##   predicted class=setosa      expected loss=0  P(node) =0.3333333
##   class counts:    50     0     0
##   probabilities: 1.000 0.000 0.000
##
## Node number 3: 100 observations,      complexity param=0.44
```

```

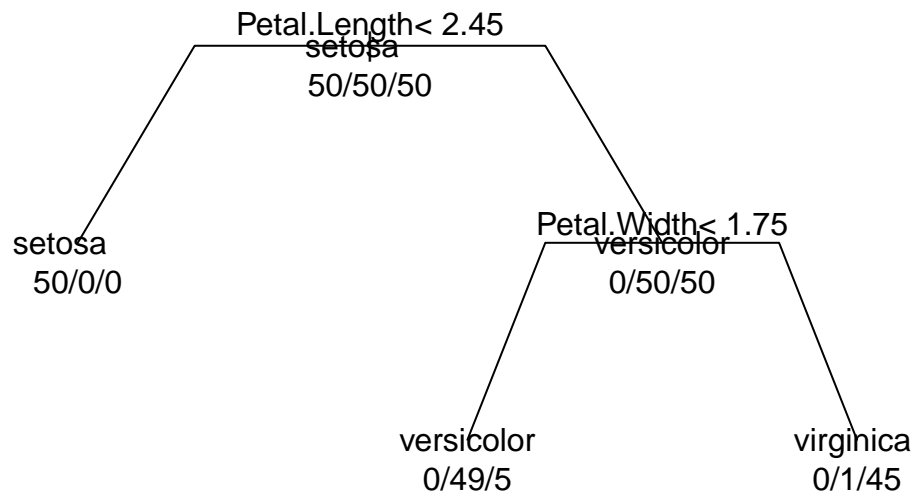
## predicted class=versicolor expected loss=0.5 P(node) =0.6666667
## class counts:      0    50    50
## probabilities: 0.000 0.500 0.500
## left son=6 (54 obs) right son=7 (46 obs)
## Primary splits:
##   Petal.Width < 1.75 to the left, improve=38.969400, (0 missing)
##   Petal.Length < 4.75 to the left, improve=37.353540, (0 missing)
##   Sepal.Length < 6.15 to the left, improve=10.686870, (0 missing)
##   Sepal.Width < 2.45 to the left, improve= 3.555556, (0 missing)
## Surrogate splits:
##   Petal.Length < 4.75 to the left, agree=0.91, adj=0.804, (0 split)
##   Sepal.Length < 6.15 to the left, agree=0.73, adj=0.413, (0 split)
##   Sepal.Width < 2.95 to the left, agree=0.67, adj=0.283, (0 split)
##
## Node number 6: 54 observations
## predicted class=versicolor expected loss=0.09259259 P(node) =0.36
## class counts:      0    49    5
## probabilities: 0.000 0.907 0.093
##
## Node number 7: 46 observations
## predicted class=virginica expected loss=0.02173913 P(node) =0.3066667
## class counts:      0     1    45
## probabilities: 0.000 0.022 0.978

```

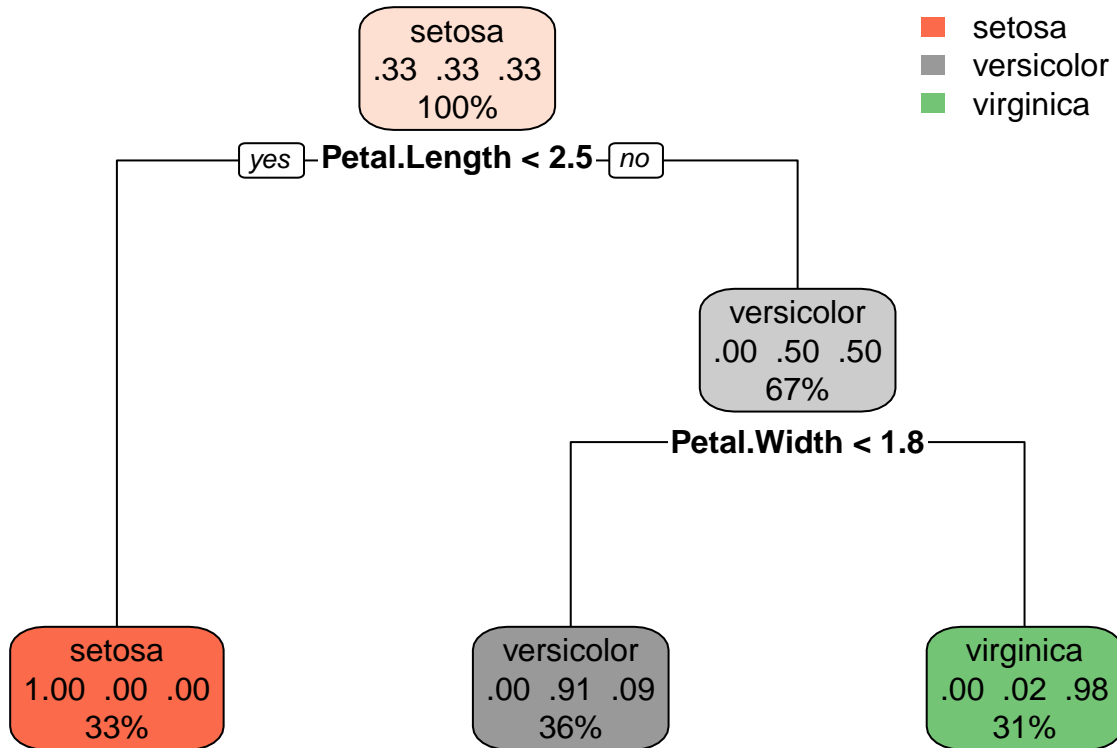
```

plot(decision_tree_model,uniform = TRUE , branch = 0.6, margin = 0.1)
text(decision_tree_model, all = TRUE, use.n=TRUE)

```




```
library(rpart.plot)
rpart.plot(decision_tree_model)
```



```
#predict
test$Species_predicted<-predict(decision_tree_model,newdata=test,type="class")
table(test$Species, test$Species_predicted)
```

```
##
##          setosa versicolor virginica
## setosa          50           0         0
## versicolor       0          49         1
## virginica        0           5        45
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

```
confusionMatrix(table(test$Species, test$Species_predicted))
```

```
## Confusion Matrix and Statistics
```

```
##
##
##           setosa versicolor virginica
##   setosa      50         0         0
##   versicolor   0        49         1
##   virginica    0         5        45
##
## Overall Statistics
##
##           Accuracy : 0.96
##           95% CI : (0.915, 0.9852)
##   No Information Rate : 0.36
##   P-Value [Acc > NIR] : < 2.2e-16
##
##           Kappa : 0.94
##
##   McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##           Class: setosa Class: versicolor Class: virginica
## Sensitivity           1.0000           0.9074           0.9783
## Specificity           1.0000           0.9896           0.9519
## Pos Pred Value        1.0000           0.9800           0.9000
## Neg Pred Value        1.0000           0.9500           0.9900
## Prevalence            0.3333           0.3600           0.3067
## Detection Rate        0.3333           0.3267           0.3000
## Detection Prevalence  0.3333           0.3333           0.3333
## Balanced Accuracy     1.0000           0.9485           0.9651
```

Pruning

```
?printcp
```

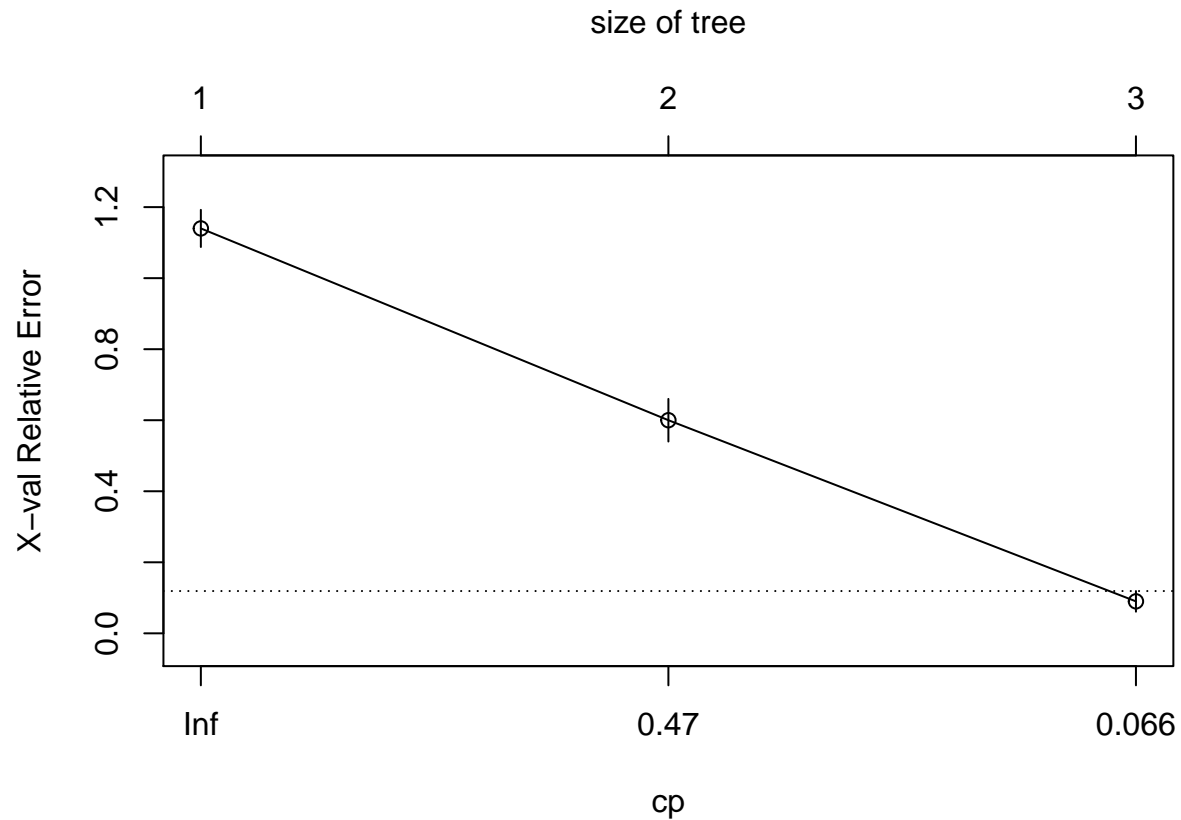
```
## starting httpd help server ... done
```

```
printcp(decision_tree_model)
```

```
##
## Classification tree:
## rpart(formula = Species ~ ., data = train, method = "class")
##
## Variables actually used in tree construction:
## [1] Petal.Length Petal.Width
##
## Root node error: 100/150 = 0.66667
##
## n= 150
##
##      CP nsplit rel error xerror      xstd
```

```
## 1 0.50      0      1.00   1.14 0.052307
## 2 0.44      1      0.50   0.60 0.060000
## 3 0.01      2      0.06   0.09 0.029086
```

```
plotcp(decision_tree_model)
```



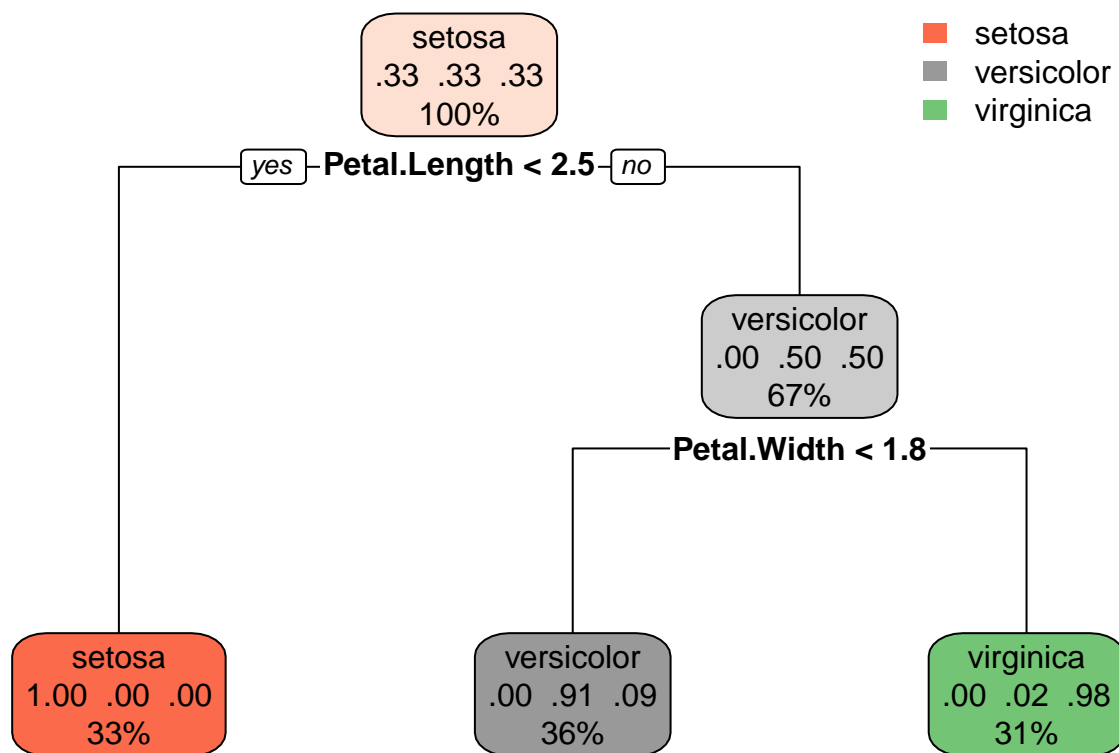
```
min(decision_tree_model$cptable[, "xerror"])
```

```
## [1] 0.09
```

```
which.min(decision_tree_model$cptable[, "xerror"])
```

```
## 3
## 3
```

```
cpmin <- decision_tree_model$cptable[3, "CP"]
decision_tree_pruned = prune(decision_tree_model, cp = cpmin)
rpart.plot(decision_tree_pruned)
```



#predict based on test data

```
test$Species_predicted<-predict(decison_tree_pruned,newdata=test,type="class")
table(test$Species, test$Species_predicted)
```

```
##
##          setosa versicolor virginica
## setosa      50          0          0
## versicolor   0          49          1
## virginica    0           5          45
```

```
confusionMatrix(table(test$Species, test$Species_predicted))
```

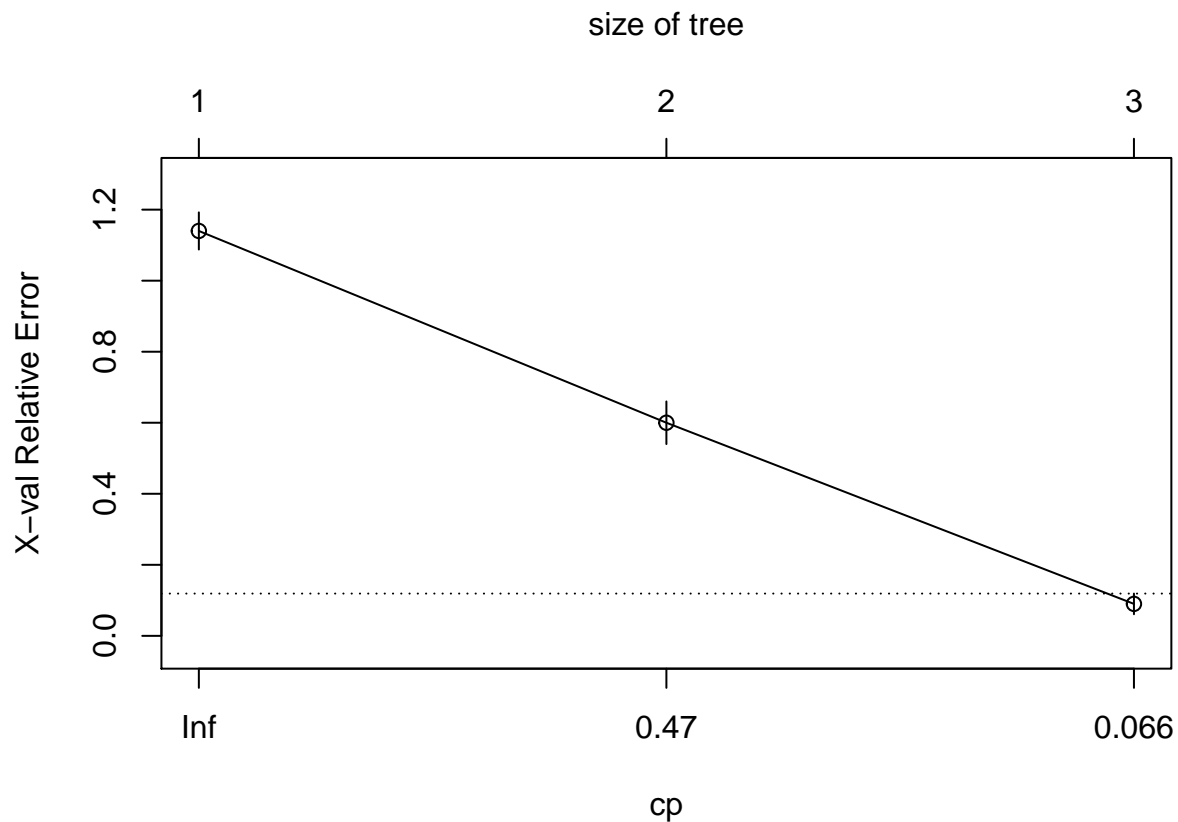
Confusion Matrix and Statistics

```
##
##          setosa versicolor virginica
## setosa      50          0          0
## versicolor   0          49          1
## virginica    0           5          45
##
## Overall Statistics
##
##          Accuracy : 0.96
##          95% CI : (0.915, 0.9852)
##          No Information Rate : 0.36
```

```
##      P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.94
##
## McNemar's Test P-Value : NA
##
## Statistics by Class:
##
##              Class: setosa Class: versicolor Class: virginica
## Sensitivity          1.0000          0.9074          0.9783
## Specificity          1.0000          0.9896          0.9519
## Pos Pred Value       1.0000          0.9800          0.9000
## Neg Pred Value       1.0000          0.9500          0.9900
## Prevalence           0.3333          0.3600          0.3067
## Detection Rate       0.3333          0.3267          0.3000
## Detection Prevalence 0.3333          0.3333          0.3333
## Balanced Accuracy     1.0000          0.9485          0.9651
```

```
#we see minimum error when N = 3
```

```
plotcp(decision_tree_model)
```



```
#find the cp value for which corss validation is minimum
```

```
min(decision_tree_model$cptable[, "xerror"])
```

```
## [1] 0.09
```

```
which.min(decision_tree_model$cptable[, "xerror"])
```

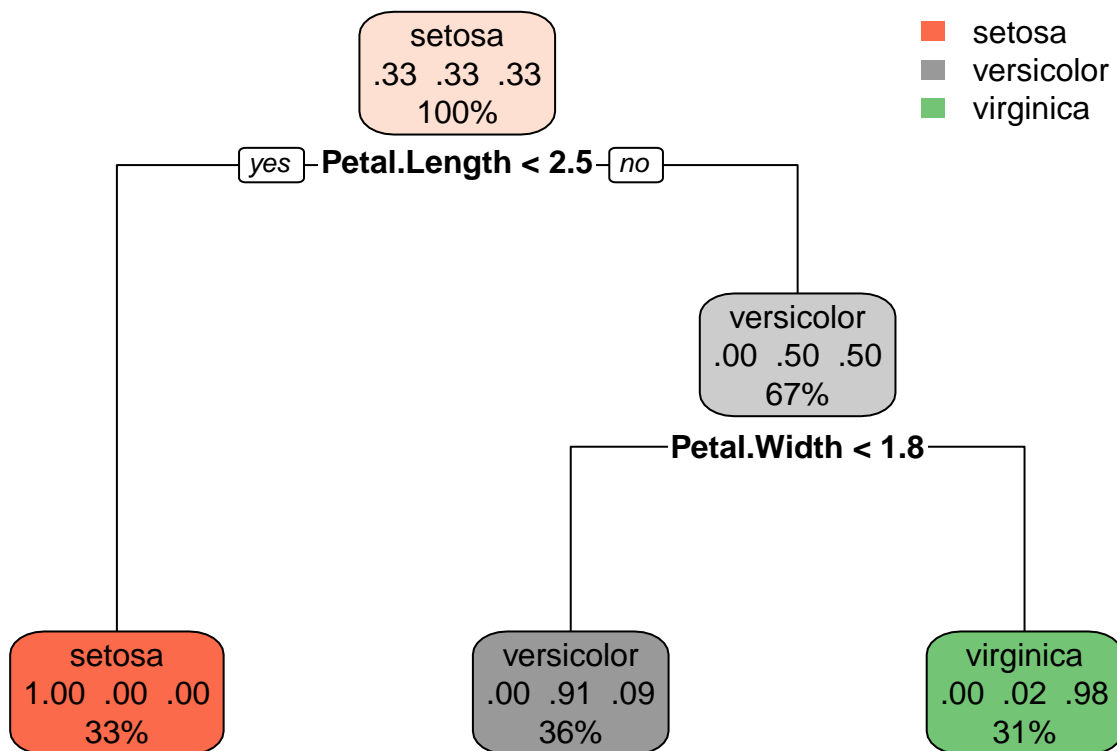
```
## 3
```

```
## 3
```

```
cpmin <- decision_tree_model$cptable[3, "CP"]
```

```
#prune the tree by setting cp value as Cpmin
```

```
decision_tree_pruned = prune(decision_tree_model, cp = cpmin)
rpart.plot(decision_tree_pruned)
```



```
#predict based on test data
```

```
test$Species_predicted<-predict(decision_tree_pruned, newdata=test, type="class")
table(test$Species, test$Species_predicted)
```

```
##
##      setosa versicolor virginica
## setosa      50         0         0
## versicolor   0        49         1
## virginica    0         5        45
```

```
confusionMatrix(table(test$Species, test$Species_predicted))
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##
```

```
##           setosa versicolor virginica
## setosa         50          0          0
## versicolor      0          49         1
## virginica       0           5        45
```

```
##
```

```
## Overall Statistics
```

```
##
```

```
##           Accuracy : 0.96
```

```
##           95% CI : (0.915, 0.9852)
```

```
## No Information Rate : 0.36
```

```
## P-Value [Acc > NIR] : < 2.2e-16
```

```
##
```

```
##           Kappa : 0.94
```

```
##
```

```
## McNemar's Test P-Value : NA
```

```
##
```

```
## Statistics by Class:
```

```
##
```

```
##           Class: setosa Class: versicolor Class: virginica
## Sensitivity           1.0000           0.9074           0.9783
## Specificity           1.0000           0.9896           0.9519
## Pos Pred Value        1.0000           0.9800           0.9000
## Neg Pred Value        1.0000           0.9500           0.9900
## Prevalence            0.3333           0.3600           0.3067
## Detection Rate        0.3333           0.3267           0.3000
## Detection Prevalence  0.3333           0.3333           0.3333
## Balanced Accuracy      1.0000           0.9485           0.9651
```