

KNN examples

Abhirup Sen

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```
data(iris)
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 ...
```

#step 1 : Split the data into train and test

```
library(caTools)
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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 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
## 150	5.9	3.0	5.1	1.8	virginica

test

##	Sepal.Length	Sepal.Width	Petal.Length	Petal.Width	Species
## 3	4.7	3.2	1.3	0.2	setosa
## 4	4.6	3.1	1.5	0.2	setosa
## 8	5.0	3.4	1.5	0.2	setosa
## 9	4.4	2.9	1.4	0.2	setosa
## 13	4.8	3.0	1.4	0.1	setosa
## 14	4.3	3.0	1.1	0.1	setosa
## 18	5.1	3.5	1.4	0.3	setosa
## 19	5.7	3.8	1.7	0.3	setosa
## 23	4.6	3.6	1.0	0.2	setosa
## 24	5.1	3.3	1.7	0.5	setosa
## 28	5.2	3.5	1.5	0.2	setosa
## 29	5.2	3.4	1.4	0.2	setosa
## 33	5.2	4.1	1.5	0.1	setosa
## 34	5.5	4.2	1.4	0.2	setosa
## 38	4.9	3.6	1.4	0.1	setosa
## 39	4.4	3.0	1.3	0.2	setosa
## 43	4.4	3.2	1.3	0.2	setosa
## 44	5.0	3.5	1.6	0.6	setosa
## 48	4.6	3.2	1.4	0.2	setosa
## 49	5.3	3.7	1.5	0.2	setosa
## 53	6.9	3.1	4.9	1.5	versicolor
## 54	5.5	2.3	4.0	1.3	versicolor
## 58	4.9	2.4	3.3	1.0	versicolor
## 59	6.6	2.9	4.6	1.3	versicolor
## 63	6.0	2.2	4.0	1.0	versicolor
## 64	6.1	2.9	4.7	1.4	versicolor
## 68	5.8	2.7	4.1	1.0	versicolor
## 69	6.2	2.2	4.5	1.5	versicolor
## 73	6.3	2.5	4.9	1.5	versicolor
## 74	6.1	2.8	4.7	1.2	versicolor

```
## 78      6.7      3.0      5.0      1.7 versicolor
## 79      6.0      2.9      4.5      1.5 versicolor
## 83      5.8      2.7      3.9      1.2 versicolor
## 84      6.0      2.7      5.1      1.6 versicolor
## 88      6.3      2.3      4.4      1.3 versicolor
## 89      5.6      3.0      4.1      1.3 versicolor
## 93      5.8      2.6      4.0      1.2 versicolor
## 94      5.0      2.3      3.3      1.0 versicolor
## 98      6.2      2.9      4.3      1.3 versicolor
## 99      5.1      2.5      3.0      1.1 versicolor
## 103     7.1      3.0      5.9      2.1 virginica
## 104     6.3      2.9      5.6      1.8 virginica
## 108     7.3      2.9      6.3      1.8 virginica
## 109     6.7      2.5      5.8      1.8 virginica
## 113     6.8      3.0      5.5      2.1 virginica
## 114     5.7      2.5      5.0      2.0 virginica
## 118     7.7      3.8      6.7      2.2 virginica
## 119     7.7      2.6      6.9      2.3 virginica
## 123     7.7      2.8      6.7      2.0 virginica
## 124     6.3      2.7      4.9      1.8 virginica
## 128     6.1      3.0      4.9      1.8 virginica
## 129     6.4      2.8      5.6      2.1 virginica
## 133     6.4      2.8      5.6      2.2 virginica
## 134     6.3      2.8      5.1      1.5 virginica
## 138     6.4      3.1      5.5      1.8 virginica
## 139     6.0      3.0      4.8      1.8 virginica
## 143     5.8      2.7      5.1      1.9 virginica
## 144     6.8      3.2      5.9      2.3 virginica
## 148     6.5      3.0      5.2      2.0 virginica
## 149     6.2      3.4      5.4      2.3 virginica
```

```
train$Species
```

```
## [1] setosa      setosa      setosa      setosa      setosa      setosa
## [7] setosa      setosa      setosa      setosa      setosa      setosa
## [13] setosa      setosa      setosa      setosa      setosa      setosa
## [19] setosa      setosa      setosa      setosa      setosa      setosa
## [25] setosa      setosa      setosa      setosa      setosa      setosa
## [31] versicolor  versicolor  versicolor  versicolor  versicolor  versicolor
## [37] versicolor  versicolor  versicolor  versicolor  versicolor  versicolor
## [43] versicolor  versicolor  versicolor  versicolor  versicolor  versicolor
## [49] versicolor  versicolor  versicolor  versicolor  versicolor  versicolor
## [55] versicolor  versicolor  versicolor  versicolor  versicolor  versicolor
## [61] virginica   virginica   virginica   virginica   virginica   virginica
## [67] virginica   virginica   virginica   virginica   virginica   virginica
## [73] virginica   virginica   virginica   virginica   virginica   virginica
## [79] virginica   virginica   virginica   virginica   virginica   virginica
## [85] virginica   virginica   virginica   virginica   virginica   virginica
## Levels: setosa versicolor virginica
```

```
#step 2: Feature Scaling
```

```
train_scale <- scale(train[,1:4])
test_scale <- scale(test[,1:4])
```

#step 3 : Fitting _KNN to the training dataset and predicting the test set

```
library(class)
?knn
```

starting httpd help server ... done

```
test$Species_predicted <- knn(train = train_scale,
                               test = test_scale,
                               cl = train$Species,
                               k=1)
table(test$Species,test$Species_predicted)
```

```
##
##           setosa versicolor virginica
## setosa           20           0           0
## versicolor        0          17           3
## virginica         0           2          18
```

#step 4 : Model evaluation - choose the right k

#calculate the error

```
misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=',1-misClassError))
```

```
## [1] "Accuracy= 0.916666666666667"
```

#K = 3

```
test$Species_predicted <- knn(train = train_scale,
                               test = test_scale,
                               cl = train$Species,
                               k=3)
```

```
misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=',1-misClassError))
```

```
## [1] "Accuracy= 0.883333333333333"
```

```
test$Species_predicted <- knn(train = train_scale,
                               test = test_scale,
                               cl = train$Species,
                               k=5)
```

```
misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=',1-misClassError))
```

```
## [1] "Accuracy= 0.883333333333333"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=7)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.9"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=10)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.933333333333333"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=13)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.916666666666667"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=15)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.933333333333333"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=17)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.933333333333333"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=19)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.916666666666667"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=21)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.9"
```

```
test$Species_predicted <- knn(train = train_scale,
                              test = test_scale,
                              cl = train$Species,
                              k=20)

misClassError <- mean(test$Species_predicted != test$Species)
print(paste('Accuracy=', 1-misClassError))
```

```
## [1] "Accuracy= 0.9"
```