

## Data-sets used

[Digit Dataset](#)

[Iris Dataset](#)

## Logistic Regression

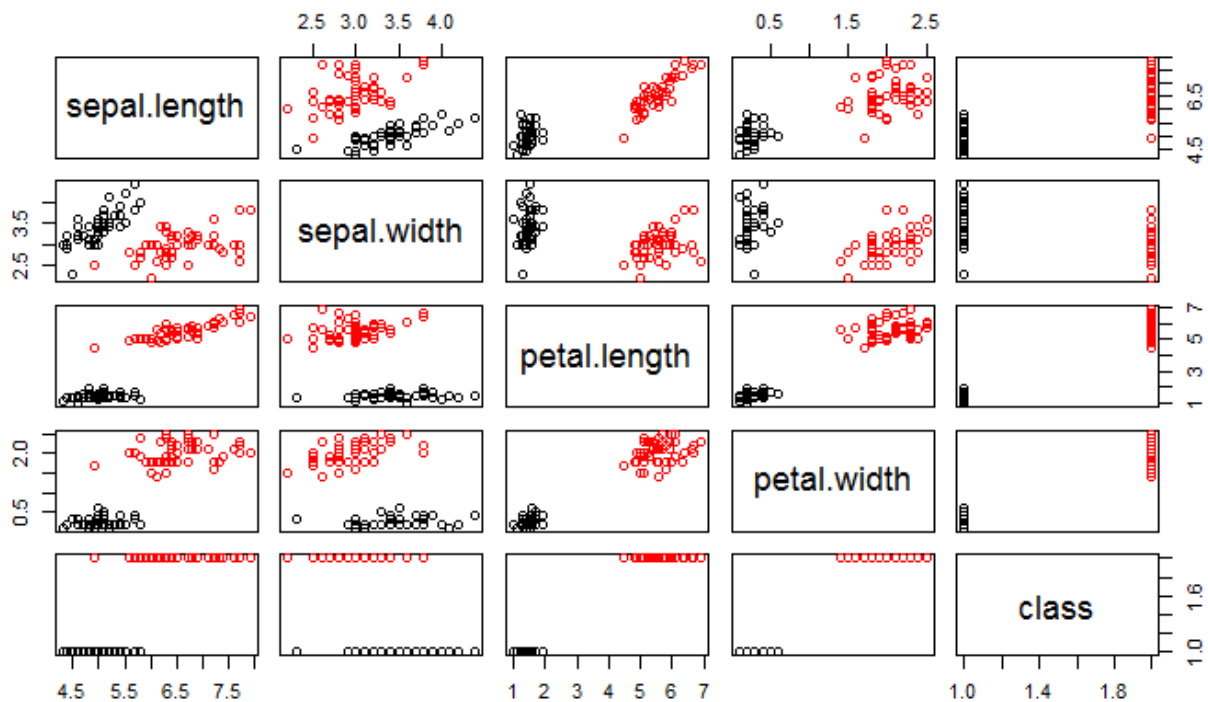
In this we perform logistic regression on our Datasets

Iris Dataset

1. Implement the logistic regression algorithm for two-class discrimination.

**Data-set Iris**

**Plot Data**



The two class seems to be pretty much separate. Let's implement logistic regression on it.

We are going to run the logistic regression for 500 steps and with learning rate 0.001

We stop if there is no much change in theta values (0.00001 difference)

## Coefficients

```
$`Iris-setosa`
                29
sepal.length  0.5648780713
sepal.width   1.3350987960
petal.length -2.0004793421
petal.width  -1.0101234148
theta_0       0.3182478349

$Iris-virginica`
                29
sepal.length -0.5656285414
sepal.width  -1.3344881979
petal.length  2.0003279285
petal.width   1.0113073135
theta_0      -0.3166929130
```

As we can see from above sepal width seems to be good predictor of Iris-setosa and petal length seems to be good predictor of Iris-Virginia

## Evaluation

```
$confusion_matrix
      truelabel
prediction  Iris-setosa Iris-virginica
Iris-setosa         13             0
Iris-virginica       0             12
```

```

$accuaracy
[1] 1

$percision
  Iris-setosa Iris-virginica
            1             1

$recall
  Iris-setosa Iris-virginica
            1             1

$fmeasure
  Iris-setosa Iris-virginica
            1             1

```

As there is a clear separation of variables in our dataset we have 100% in all evaluations

### **5-fold Cross Validation**

Let's do 5 fold cross validation to test our model further

```

$avgAccuracy
[1] 1

$avgPercision
  Iris-setosa Iris-virginica
            1             1

$avgRecall
  Iris-setosa Iris-virginica
            1             1

$avgFmeasure
  Iris-setosa Iris-virginica
            1             1

```

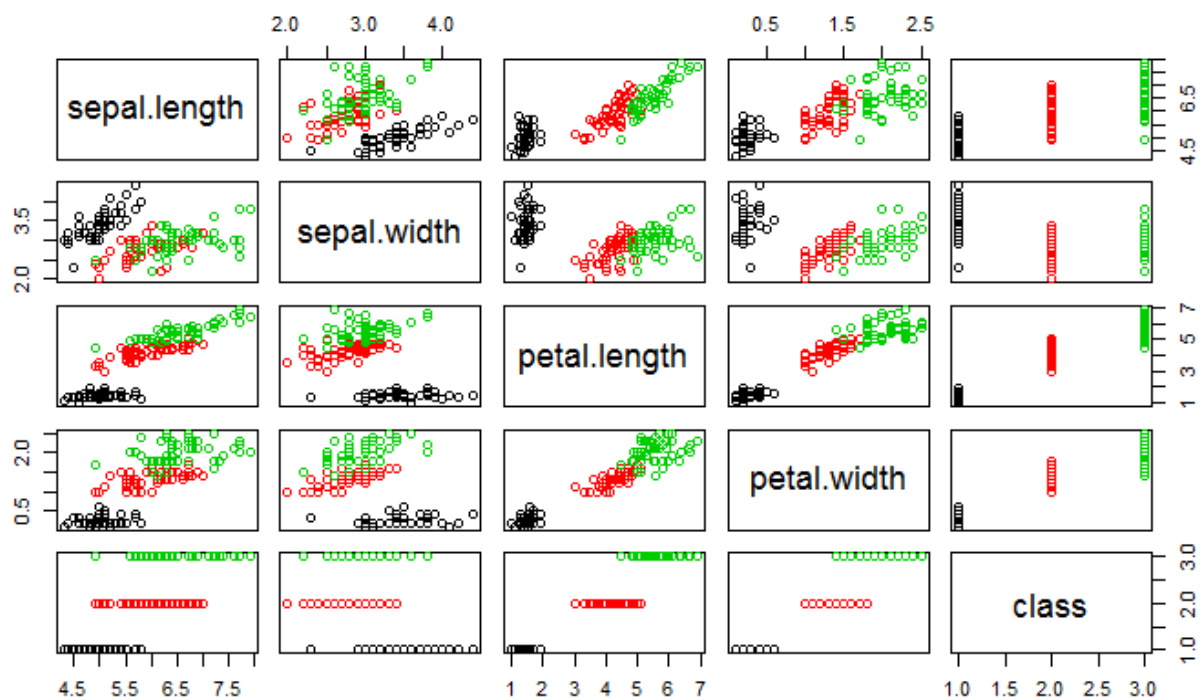
So in our cross validation also we get same result as before

3. Implement the logistic regression algorithm for K-class discrimination.

## Data-set Iris

Number of classes 3

## Plot Data-set



We are going to run the logistic regression for 500 steps and with learning rate 0.001

We stop if there is no much change in theta values (0.00001 difference)

## Coefficients

```
$`Iris-setosa`  
43  
sepal.length 0.4488581428  
sepal.width  1.5986167818  
petal.length -2.4740212555  
petal.width  -1.1305198608  
theta_0      0.2841501169  
  
$`Iris-versicolor`  
43  
sepal.length 0.4039267901  
sepal.width  -1.3706302477  
petal.length 0.4775172896  
petal.width  -1.0507511957  
theta_0      0.5108465848  
  
$`Iris-virginica`  
43  
sepal.length -1.6163881221  
sepal.width  -1.5176042741  
petal.length 2.3840254426  
petal.width  2.1787602488  
theta_0      -0.8352562783
```

## Evaluation

```
$confusion_matrix
      truelabel
prediction Iris-setosa Iris-versicolor Iris-virginica
Iris-setosa      13           0           0
Iris-versicolor   0          10           0
Iris-virginica    0           1          14
```

```
$accuracy
[1] 0.9736842105
```

```
$precision
      Iris-setosa Iris-versicolor Iris-virginica
1.000000000000   0.9090909091   1.000000000000
```

```
$recall
      Iris-setosa Iris-versicolor Iris-virginica
1.000000000000   1.000000000000   0.933333333333
```

```
$fmeasure
      Iris-setosa Iris-versicolor Iris-virginica
1.000000000000   0.9523809524   0.9655172414
```

## 5-Fold Cross Validation

```
$avgAccuracy
[1] 0.9597701149
```

```
$avgPrecision
      Iris-setosa Iris-versicolor Iris-virginica
1.00           0.88           1.00
```

```
$avgRecall
      Iris-setosa Iris-versicolor Iris-virginica
1.000000000000   1.000000000000   0.896969697
```

```
$avgFmeasure
      Iris-setosa Iris-versicolor Iris-virginica
1.000000000000   0.9345029240   0.9445887446
```

