

# Report on Breast Cancer Wisconsin (Prognostic) Data Set

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## DataSet

Dataset is from UCI Machine Learning Repository - Breast Cancer Wisconsin (Prognostic) Data Set. This dataset contains three files :

- 1) Wisconsin Breast Cancer Database
- 2) Wisconsin Diagnostic Breast Cancer (WDBC)
- 3) Wisconsin Prognostic Breast Cancer (WPBC)

This dataset is available online at <https://archive.ics.uci.edu/ml/datasets/Breast+Cancer+Wisconsin+%28Prognostic%29>.

Reading the dataset can be done similar to that of the CSV as the fields are “,” separated and newline contains new records. Data is clean so not much of the preprocessing is required. File contain no headers.

```
WDBC_data=read.csv("wdbc.data",head=FALSE,sep = ",")
WPBC_data=read.csv("wpbc.data",head=FALSE,sep = ",")
bcw_data=read.csv("breast-cancer-wisconsin.data",head=FALSE,sep = ",")
```

Loading required libraries :

```
library(caret)
```

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

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

```
library(class)
```

## Preprocessing the data

Using just the first dataset

```
bcw_data<-bcw_data[-(which(bcw_data$V7=="?")),]
```

## Classification on Wisconsin Breast Cancer Database

Target Variable - Class ( 2 for Benign Tumor and 4 for Malignant Tumor )

```
summary(bcw_data)
```

```
##           V1           V2           V3           V4
## Min.      : 63375   Min.    : 1.000   Min.    : 1.000   Min.    : 1.000
## 1st Qu.: 877617   1st Qu.: 2.000   1st Qu.: 1.000   1st Qu.: 1.000
## Median : 1171795   Median : 4.000   Median : 1.000   Median : 1.000
## Mean    : 1076720   Mean     : 4.442   Mean    : 3.151   Mean    : 3.215
## 3rd Qu.: 1238705   3rd Qu.: 6.000   3rd Qu.: 5.000   3rd Qu.: 5.000
## Max.    :13454352   Max.     :10.000   Max.    :10.000   Max.    :10.000
##
##           V5           V6           V7           V8
## Min.      : 1.00   Min.    : 1.000   1         :402   Min.    : 1.000
## 1st Qu.: 1.00   1st Qu.: 2.000   10        :132   1st Qu.: 2.000
## Median : 1.00   Median : 2.000   2         : 30   Median : 3.000
## Mean    : 2.83   Mean     : 3.234   5         : 30   Mean    : 3.445
## 3rd Qu.: 4.00   3rd Qu.: 4.000   3         : 28   3rd Qu.: 5.000
## Max.    :10.00   Max.     :10.000   8         : 21   Max.    :10.000
##
##                               (Other): 40
##           V9           V10          V11
## Min.      : 1.00   Min.    : 1.000   Min.     :2.0
## 1st Qu.: 1.00   1st Qu.: 1.000   1st Qu.:2.0
## Median : 1.00   Median : 1.000   Median :2.0
## Mean    : 2.87   Mean     : 1.603   Mean     :2.7
## 3rd Qu.: 4.00   3rd Qu.: 1.000   3rd Qu.:4.0
## Max.    :10.00   Max.     :10.000   Max.     :4.0
##
```

Data is skewed.

Preparing data for training a classifier

```
### dimension of dataset
dim(bcw_data)
```

```
## [1] 683 11
```

```
inTrain<-createDataPartition(y=bcw_data$V11,p=0.75,list=FALSE)
training<-bcw_data[inTrain,]
testing<-bcw_data[-inTrain,]
### dimension of the training set
dim(training)
```

```
## [1] 513 11
```

```
### dimension of testing set
dim(testing)
```

```
## [1] 170 11
```

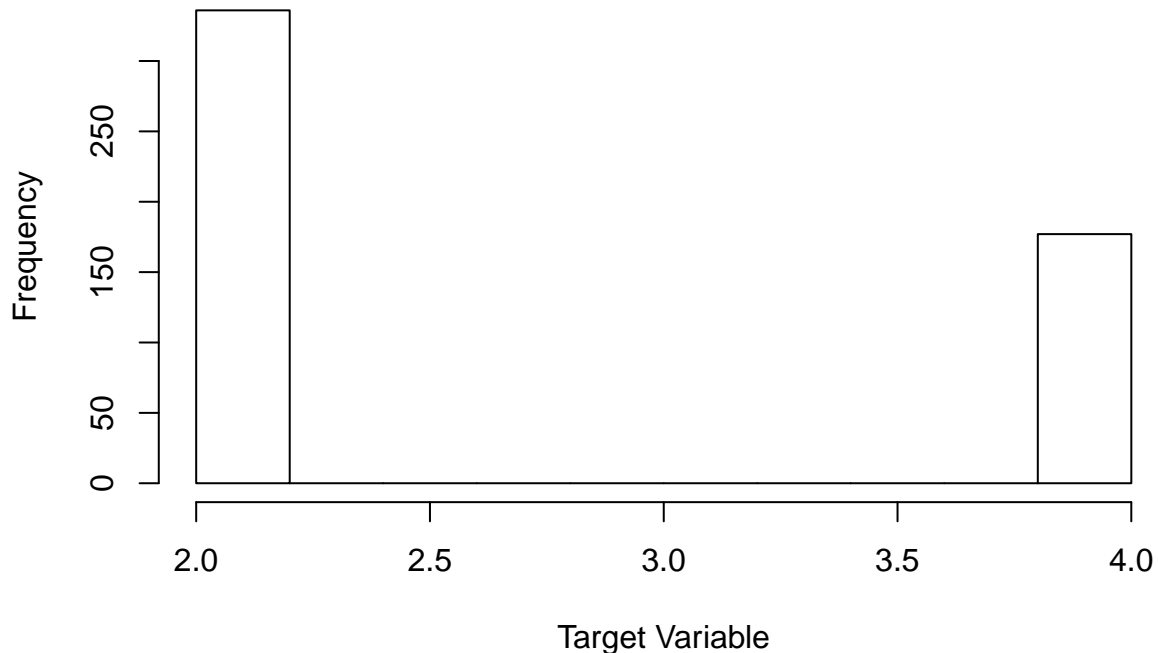
Summary shows that data is skewed but when we explore the target variable we see that the class is well balanced.

```
summary(training)
```

```
##          V1          V2          V3          V4
## Min.   : 76389   Min.   : 1.000   Min.   : 1.000   Min.   : 1.000
## 1st Qu.: 857774   1st Qu.: 2.000   1st Qu.: 1.000   1st Qu.: 1.000
## Median : 1165926   Median : 4.000   Median : 1.000   Median : 1.000
## Mean   : 1063260   Mean    : 4.431   Mean    : 3.099   Mean    : 3.136
## 3rd Qu.: 1231387   3rd Qu.: 6.000   3rd Qu.: 5.000   3rd Qu.: 5.000
## Max.   :13454352   Max.    :10.000   Max.    :10.000   Max.    :10.000
##
##          V5          V6          V7          V8
## Min.   : 1.000   Min.   : 1.000   1      :296   Min.   : 1.000
## 1st Qu.: 1.000   1st Qu.: 2.000   10     :103   1st Qu.: 2.000
## Median : 1.000   Median : 2.000   2      : 22   Median : 3.000
## Mean   : 2.877   Mean    : 3.193   3      : 22   Mean    : 3.454
## 3rd Qu.: 4.000   3rd Qu.: 4.000   5      : 22   3rd Qu.: 4.000
## Max.   :10.000   Max.    :10.000   4      : 17   Max.    :10.000
##
##                                (Other): 31
##          V9          V10         V11
## Min.   : 1.000   Min.   : 1.000   Min.   :2.00
## 1st Qu.: 1.000   1st Qu.: 1.000   1st Qu.:2.00
## Median : 1.000   Median : 1.000   Median :2.00
## Mean   : 2.865   Mean    : 1.561   Mean    :2.69
## 3rd Qu.: 4.000   3rd Qu.: 1.000   3rd Qu.:4.00
## Max.   :10.000   Max.    :10.000   Max.    :4.00
##
```

```
hist(training$V11,main = "Class Distribution" , xlab = "Target Variable" )
```

## Class Distribution



##

Techniques that are available for Classification

1. Quadratic Discriminant Analysis
2. Linear Discriminant Analysis
3. K - Nearest Neighbour - 1
4. K - Nearest Neighbour - 3
5. Logistic Regression

Logistic Regression and KNN was tried as they do not assume the gaussian distribution of the data. Variables being skewed this condition was not satisfied

```
set.seed(1)
knn.pred1=knn(training[,2:10],testing[,2:10],training[,11],k=1)
p=table(knn.pred1,testing[,11])
print(p)
```

```
##
## knn.pred1  2   4
##           2 105  6
##           4   3 56
```

```
accuracy=(p[1,1]+p[2,2])/length(testing$V11)
accuracy
```

```
## [1] 0.9470588
```

```
knn.pred2=knn(training[,2:10],testing[,2:10],training[,11],k=3)
p=table(knn.pred2,testing[,11])
print(p)
```

```
##
## knn.pred2  2   4
##           2 105  4
##           4   3 58
```

```
accuracy=(p[1,1]+p[2,2])/length(testing$V11)
accuracy
```

```
## [1] 0.9588235
```

K - Nearest Neighbour was selected as the classifier was erring more on the False Positive side on the False Negative Side.

## Clustering

Techniques available:

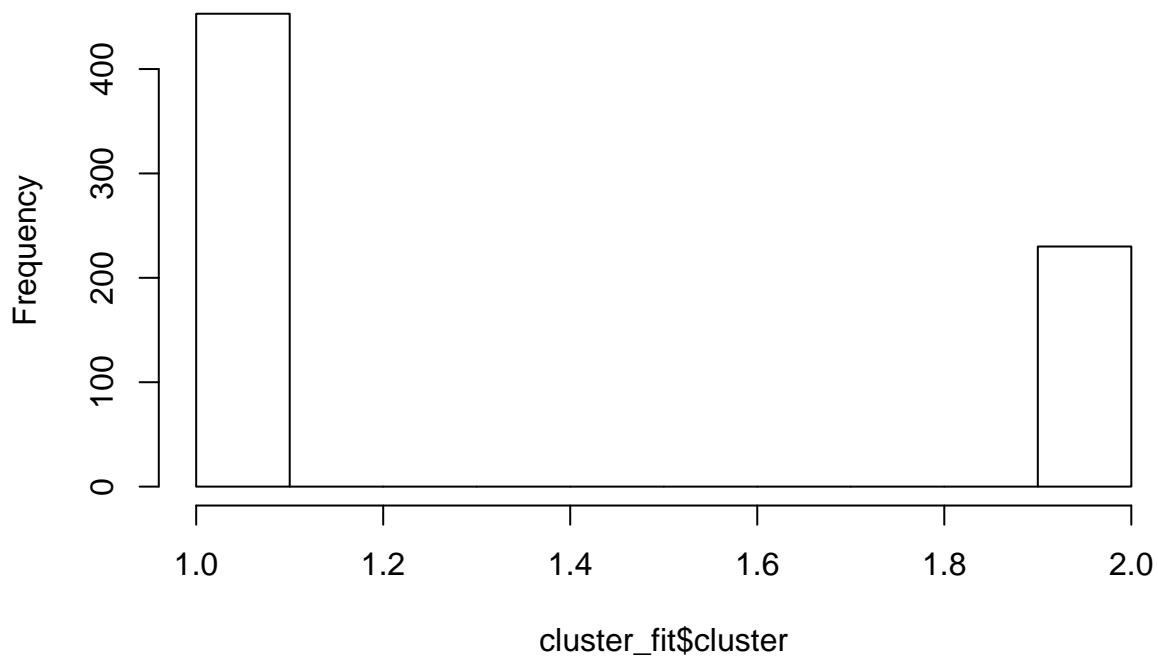
1. Principle Component Analysis
2. K - Means Algorithm
  - Hartigan-Wong
  - Lloyd
  - Forgy
  - MacQueen

Not used PCA as the number of feature was much less than the number of data points and k - means can better handle skewed data.

Data points were spanning over small range of values so Variability was also not an issue

```
cluster_fit=kmeans(bcw_data[,2:10],2,iter.max = 100)
hist(cluster_fit$cluster)
```

**Histogram of cluster\_fit\$cluster**



```
res=cluster_fit$cluster
p=table(ifelse(res==1,2,4),bcw_data[,11])
print(p)
```

```
##
##      2   4
##    2 435 18
##    4   9 221
```

```
accuracy=(p[1,1]+p[2,2])/length(cluster_fit$cluster)
accuracy
```

```
## [1] 0.9604685
```

## Regression

```
summary(WPBC_data)
```

```
##      V1      V2      V3      V4
## Min.   : 8423 N:151 Min.   : 1.00 Min.   :10.95
## 1st Qu.: 855745 R: 47 1st Qu.: 14.00 1st Qu.:15.05
## Median : 886339      Median : 39.50 Median :17.29
## Mean   :1990469      Mean   : 46.73 Mean   :17.41
## 3rd Qu.: 927996      3rd Qu.: 72.75 3rd Qu.:19.58
## Max.   :9411300      Max.   :125.00 Max.   :27.22
##
##      V5      V6      V7      V8
## Min.   :10.38 Min.   : 71.90 Min.   : 361.6 Min.   :0.07497
## 1st Qu.:19.41 1st Qu.: 98.16 1st Qu.: 702.5 1st Qu.:0.09390
## Median :21.75 Median :113.70 Median : 929.1 Median :0.10190
## Mean   :22.28 Mean   :114.86 Mean   : 970.0 Mean   :0.10268
## 3rd Qu.:24.66 3rd Qu.:129.65 3rd Qu.:1193.5 3rd Qu.:0.11098
## Max.   :39.28 Max.   :182.10 Max.   :2250.0 Max.   :0.14470
##
##      V9      V10      V11      V12
## Min.   :0.04605 Min.   :0.02398 Min.   :0.02031 Min.   :0.1308
## 1st Qu.:0.11020 1st Qu.:0.10685 1st Qu.:0.06367 1st Qu.:0.1741
## Median :0.13175 Median :0.15135 Median :0.08607 Median :0.1893
## Mean   :0.14265 Mean   :0.15624 Mean   :0.08678 Mean   :0.1928
## 3rd Qu.:0.17220 3rd Qu.:0.20050 3rd Qu.:0.10393 3rd Qu.:0.2093
## Max.   :0.31140 Max.   :0.42680 Max.   :0.20120 Max.   :0.3040
##
##      V13      V14      V15      V16
## Min.   :0.05025 Min.   :0.1938 Min.   :0.3621 Min.   : 1.153
## 1st Qu.:0.05672 1st Qu.:0.3882 1st Qu.:0.9213 1st Qu.: 2.743
## Median :0.06171 Median :0.5333 Median :1.1685 Median : 3.767
## Mean   :0.06271 Mean   :0.6033 Mean   :1.2645 Mean   : 4.255
## 3rd Qu.:0.06671 3rd Qu.:0.7509 3rd Qu.:1.4632 3rd Qu.: 5.213
## Max.   :0.09744 Max.   :1.8190 Max.   :3.5030 Max.   :13.280
##
##      V17      V18      V19      V20
## Min.   : 13.99 Min.   :0.002667 Min.   :0.007347 Min.   :0.01094
## 1st Qu.: 35.37 1st Qu.:0.005001 1st Qu.:0.019803 1st Qu.:0.02681
## Median : 58.45 Median :0.006193 Median :0.027880 Median :0.03691
## Mean   : 70.23 Mean   :0.006762 Mean   :0.031199 Mean   :0.04075
```

```
## 3rd Qu.: 92.48 3rd Qu.:0.007973 3rd Qu.:0.038335 3rd Qu.:0.04897
## Max. :316.00 Max. :0.031130 Max. :0.135400 Max. :0.14380
##
## V21 V22 V23 V24
## Min. :0.005174 Min. :0.007882 Min. :0.001087 Min. :12.84
## 1st Qu.:0.011423 1st Qu.:0.014795 1st Qu.:0.002748 1st Qu.:17.63
## Median :0.014175 Median :0.017905 Median :0.003719 Median :20.52
## Mean :0.015099 Mean :0.020555 Mean :0.003987 Mean :21.02
## 3rd Qu.:0.017665 3rd Qu.:0.022880 3rd Qu.:0.004630 3rd Qu.:23.73
## Max. :0.039270 Max. :0.060410 Max. :0.012560 Max. :35.13
##
## V25 V26 V27 V28
## Min. :16.67 Min. : 85.1 Min. : 508.1 Min. :0.08191
## 1st Qu.:26.21 1st Qu.:118.1 1st Qu.: 947.3 1st Qu.:0.12932
## Median :30.14 Median :136.5 Median :1295.0 Median :0.14185
## Mean :30.14 Mean :140.3 Mean :1405.0 Mean :0.14392
## 3rd Qu.:33.55 3rd Qu.:159.9 3rd Qu.:1694.2 3rd Qu.:0.15488
## Max. :49.54 Max. :232.2 Max. :3903.0 Max. :0.22260
##
## V29 V30 V31 V32
## Min. :0.05131 Min. :0.02398 Min. :0.02899 Min. :0.1565
## 1st Qu.:0.24870 1st Qu.:0.32215 1st Qu.:0.15265 1st Qu.:0.2759
## Median :0.35130 Median :0.40235 Median :0.17925 Median :0.3103
## Mean :0.36510 Mean :0.43669 Mean :0.17878 Mean :0.3234
## 3rd Qu.:0.42368 3rd Qu.:0.54105 3rd Qu.:0.20713 3rd Qu.:0.3588
## Max. :1.05800 Max. :1.17000 Max. :0.29030 Max. :0.6638
##
## V33 V34 V35
## Min. :0.05504 Min. : 0.400 0 :87
## 1st Qu.:0.07658 1st Qu.: 1.500 1 :35
## Median :0.08689 Median : 2.500 2 :17
## Mean :0.09083 Mean : 2.847 4 :10
## 3rd Qu.:0.10138 3rd Qu.: 3.500 13 : 6
## Max. :0.20750 Max. :10.000 7 : 6
## (Other):37
```

Preparing data for regression :

```
inTrain1<-createDataPartition(y=WPBC_data$V3,p=0.75,list=FALSE)
training<-WPBC_data[inTrain,]
testing<-WPBC_data[-inTrain,]
```

Fitting the model 1 with pca :

```
dat <- training[,2:34]

lmFit1 <- train(V3~., dat, method = "lm", preProcess=c("pca"),
               trControl = trainControl(method = "cv"))
lmFit1.pred<-predict(lmFit1,testing[2:34])
sqrt(sum((lmFit1.pred-testing[,3])^2)/94)
```

```
## [1] 22.79267
```

Fitting the model 2 without pca:

```
lmFit2 <- train(V3~., dat, method = "lm",  
               trControl = trainControl(method = "cv"))  
lmFit2.pred<-predict(lmFit2,testing[2:34])  
sqrt(sum((lmFit2.pred-testing[,3])^2)/94)
```

```
## [1] 23.51304
```

As the data has many correlated features PCA is used to reduce the number of features for better and improved prediction but the results does not support the argument

Cross validation was used to better generalise the error

So fit that give less standard error was selected as they give better result.