

## Random-forest

Data Set Information: 2126 fetal cardiocotograms (CTGs) were automatically processed and the respective diagnostic features measured. The CTGs were also classified by three expert obstetricians and a consensus classification label assigned to each of them. NSP: 1. Normal, 2. Suspected, 3. pathology

```
data <-  
read.csv("file:///C:/Users/badal/Desktop/dataset_/Cardiotocographic.csv")  
head(data)
```

##	LB	AC	FM	UC	DL	DS	DP	ASTV	MSTV	ALTV
## 1	120	0.000000000	0	0.000000000	0	0.000000000	73	0.5	43	
## 2	132	0.006379585	0	0.006379585	0	0.000000000	17	2.1	0	
## 3	133	0.003322259	0	0.008305648	0	0.000000000	16	2.1	0	
## 4	134	0.002560819	0	0.007682458	0	0.000000000	16	2.4	0	
## 5	132	0.006514658	0	0.008143322	0	0.000000000	16	2.4	0	
## 6	134	0.001049318	0	0.010493179	0	0.002098636	26	5.9	0	

```
##      Width Min Max Nmax Nzeros Mode Mean Median Variance Tendency NSP  
## 1      64  62 126   2      0  120  137   121      73         1    2  
## 2     130  68 198   6      1  141  136   140      12         0    1  
## 3     130  68 198   5      1  141  135   138      13         0    1  
## 4     117  53 170  11      0  137  134   137      13         1    1  
## 5     117  53 170   9      0  137  136   138      11         1    1  
## 6     150  50 200   5      3   76  107   107     170         0    3
```

```
str(data)
```

```
## 'data.frame':   2126 obs. of  22 variables:  
## $ LB          : int  120 132 133 134 132 134 134 122 122 122 ...  
## $ AC          : num  0 0.00638 0.00332 0.00256 0.00651 ...  
## $ FM          : num  0 0 0 0 0 0 0 0 0 0 ...  
## $ UC          : num  0 0.00638 0.00831 0.00768 0.00814 ...  
## $ DL          : num  0 0.00319 0.00332 0.00256 0 ...  
## $ DS          : num  0 0 0 0 0 0 0 0 0 0 ...  
## $ DP          : num  0 0 0 0 0 ...  
## $ ASTV        : int  73 17 16 16 16 26 29 83 84 86 ...  
## $ MSTV        : num  0.5 2.1 2.1 2.4 2.4 5.9 6.3 0.5 0.5 0.3 ...  
## $ ALTV        : int  43 0 0 0 0 0 0 6 5 6 ...  
## $ MLTV        : num  2.4 10.4 13.4 23 19.9 0 0 15.6 13.6 10.6 ...  
## $ Width       : int  64 130 130 117 117 150 150 68 68 68 ...
```

```
## $ Min      : int  62 68 68 53 53 50 50 62 62 62 ...
## $ Max      : int 126 198 198 170 170 200 200 130 130 130 ...
## $ Nmax     : int  2 6 5 11 9 5 6 0 0 1 ...
## $ Nzeros   : int  0 1 1 0 0 3 3 0 0 0 ...
## $ Mode     : int 120 141 141 137 137 76 71 122 122 122 ...
## $ Mean     : int 137 136 135 134 136 107 107 122 122 122 ...
## $ Median   : int 121 140 138 137 138 107 106 123 123 123 ...
## $ Variance : int  73 12 13 13 11 170 215 3 3 1 ...
## $ Tendency : int  1 0 0 1 1 0 0 1 1 1 ...
## $ NSP      : int  2 1 1 1 1 3 3 3 3 3 ...
```

```
data$NSP <- as.factor(data$NSP)
```

```
summary(data)
```

```
##          LB          AC          FM          UC
## Min.    :106.0   Min.    :0.000000   Min.    :0.000000   Min.    :0.000000
## 1st Qu.:126.0   1st Qu.:0.000000   1st Qu.:0.000000   1st Qu.:0.001876
## Median :133.0   Median :0.001630   Median :0.000000   Median :0.004482
## Mean    :133.3   Mean    :0.003170   Mean    :0.009474   Mean    :0.004357
## 3rd Qu.:140.0   3rd Qu.:0.005631   3rd Qu.:0.002512   3rd Qu.:0.006525
## Max.    :160.0   Max.    :0.019284   Max.    :0.480634   Max.    :0.014925
##          DL          DS          DP          ASTV
## Min.    :0.000000   Min.    :0.000e+00   Min.    :0.0000000   Min.    :12.00
## 1st Qu.:0.000000   1st Qu.:0.000e+00   1st Qu.:0.0000000   1st Qu.:32.00
## Median :0.000000   Median :0.000e+00   Median :0.0000000   Median :49.00
## Mean    :0.001885   Mean    :3.585e-06   Mean    :0.0001566   Mean    :46.99
## 3rd Qu.:0.003264   3rd Qu.:0.000e+00   3rd Qu.:0.0000000   3rd Qu.:61.00
## Max.    :0.015385   Max.    :1.353e-03   Max.    :0.0053476   Max.    :87.00
##          MSTV          ALTV          MLTV          Width
## Min.    :0.200   Min.    : 0.000   Min.    : 0.000   Min.    : 3.00
## 1st Qu.:0.700   1st Qu.: 0.000   1st Qu.: 4.600   1st Qu.: 37.00
## Median :1.200   Median : 0.000   Median : 7.400   Median : 67.50
## Mean    :1.333   Mean    : 9.847   Mean    : 8.188   Mean    : 70.45
## 3rd Qu.:1.700   3rd Qu.:11.000   3rd Qu.:10.800   3rd Qu.:100.00
## Max.    :7.000   Max.    :91.000   Max.    :50.700   Max.    :180.00
##          Min          Max          Nmax          Nzeros
## Min.    : 50.00   Min.    :122   Min.    : 0.000   Min.    : 0.0000
## 1st Qu.: 67.00   1st Qu.:152   1st Qu.: 2.000   1st Qu.: 0.0000
## Median : 93.00   Median :162   Median : 3.000   Median : 0.0000
## Mean    : 93.58   Mean    :164   Mean    : 4.068   Mean    : 0.3236
## 3rd Qu.:120.00   3rd Qu.:174   3rd Qu.: 6.000   3rd Qu.: 0.0000
## Max.    :159.00   Max.    :238   Max.    :18.000   Max.    :10.0000
##          Mode          Mean          Median          Variance
## Min.    : 60.0   Min.    : 73.0   Min.    : 77.0   Min.    : 0.00
## 1st Qu.:129.0   1st Qu.:125.0   1st Qu.:129.0   1st Qu.: 2.00
## Median :139.0   Median :136.0   Median :139.0   Median : 7.00
## Mean    :137.5   Mean    :134.6   Mean    :138.1   Mean    :18.81
## 3rd Qu.:148.0   3rd Qu.:145.0   3rd Qu.:148.0   3rd Qu.:24.00
## Max.    :187.0   Max.    :182.0   Max.    :186.0   Max.    :269.00
```

```
##      Tendency      NSP
##  Min.    :-1.0000    1:1655
##  1st Qu.: 0.0000    2: 295
##  Median : 0.0000    3: 176
##  Mean    : 0.3203
##  3rd Qu.: 1.0000
##  Max.    : 1.0000
```

```
table(data$NSP)
```

```
##
##      1      2      3
## 1655  295  176
```

partition data into training and validation sets

```
set.seed(1234)
index <- sample(2, nrow(data), replace = T, prob = c(0.70, 0.30))
train <- data[index==1,]
validate <- data[index==2,]
```

Random forest model:

```
#install.packages("randomForest")
library(randomForest)

## Warning: package 'randomForest' was built under R version 3.6.3
## randomForest 4.6-14

## Type rfNews() to see new features/changes/bug fixes.

set.seed(111)
rf<-randomForest(NSP~., data = train)
rf

##
## Call:
## randomForest(formula = NSP ~ ., data = train)
##              Type of random forest: classification
##              Number of trees: 500
## No. of variables tried at each split: 4
##
##              OOB estimate of  error rate: 5.84%
## Confusion matrix:
##      1  2  3 class.error
## 1 1175  17   3  0.01673640
## 2   51 144   6  0.28358209
## 3    6   6 115  0.09448819

summary(rf) #attributes of rf
```

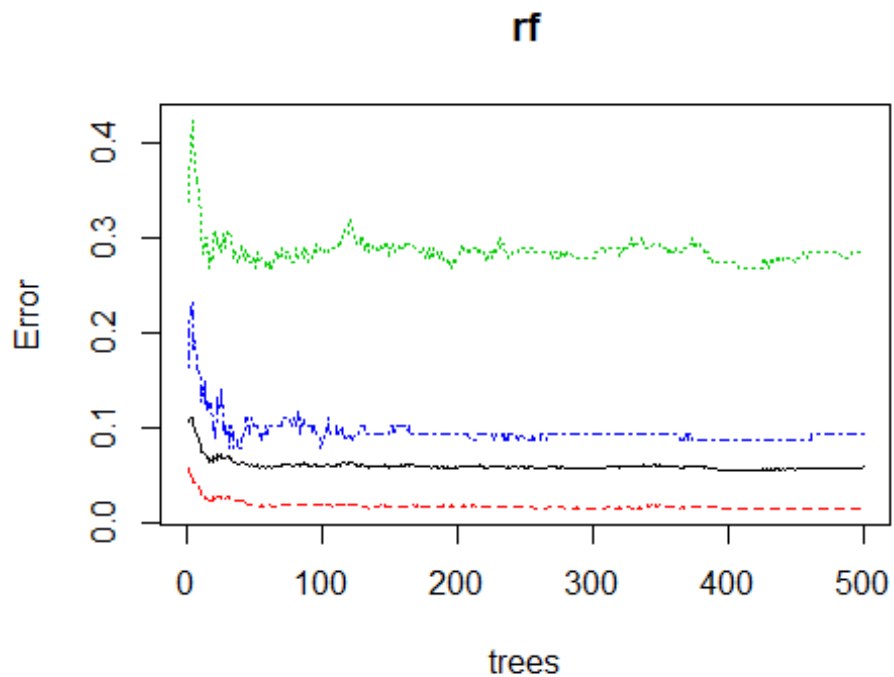
```
##           Length Class  Mode
## call           3  -none-  call
## type           1  -none- character
## predicted     1523  factor numeric
## err.rate      2000  -none- numeric
## confusion      12  -none- numeric
## votes         4569  matrix numeric
## oob.times      1523  -none- numeric
## classes        3  -none- character
## importance      21  -none- numeric
## importanceSD     0  -none-  NULL
## localImportance  0  -none-  NULL
## proximity       0  -none-  NULL
## ntree           1  -none- numeric
## mtry            1  -none- numeric
## forest          14  -none-  list
## y              1523  factor numeric
## test           0  -none-  NULL
## inbag           0  -none-  NULL
## terms           3   terms  call
```

```
rf$confusion
```

```
##      1  2  3 class.error
## 1 1175 17  3  0.01673640
## 2   51 144  6  0.28358209
## 3    6  6 115  0.09448819
```

Error rate

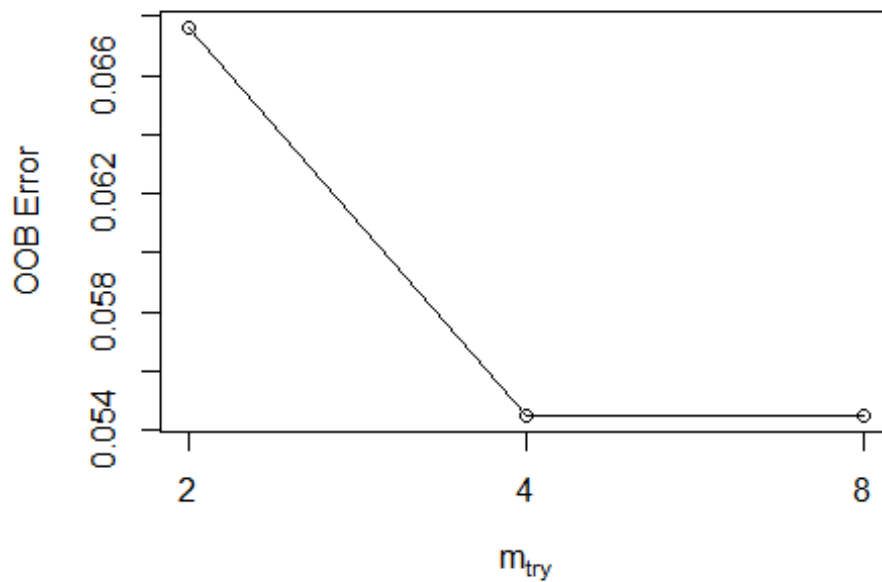
```
plot(rf)
```



Tune random forest model

```
tuneRF(train[, -22], train$NSP,
        stepFactor = 0.5,
        plot = TRUE,
        ntreeTry = 300,
        trace = TRUE,
        improve = 0.05
)

## mtry = 4   OOB error = 5.45%
## Searching left ...
## mtry = 8   OOB error = 5.45%
## 0 0.05
## Searching right ...
## mtry = 2   OOB error = 6.76%
## -0.2409639 0.05
```



```
##      mtry  OOBError
## 2.00B    2 0.06762968
## 4.00B    4 0.05449770
## 8.00B    8 0.05449770

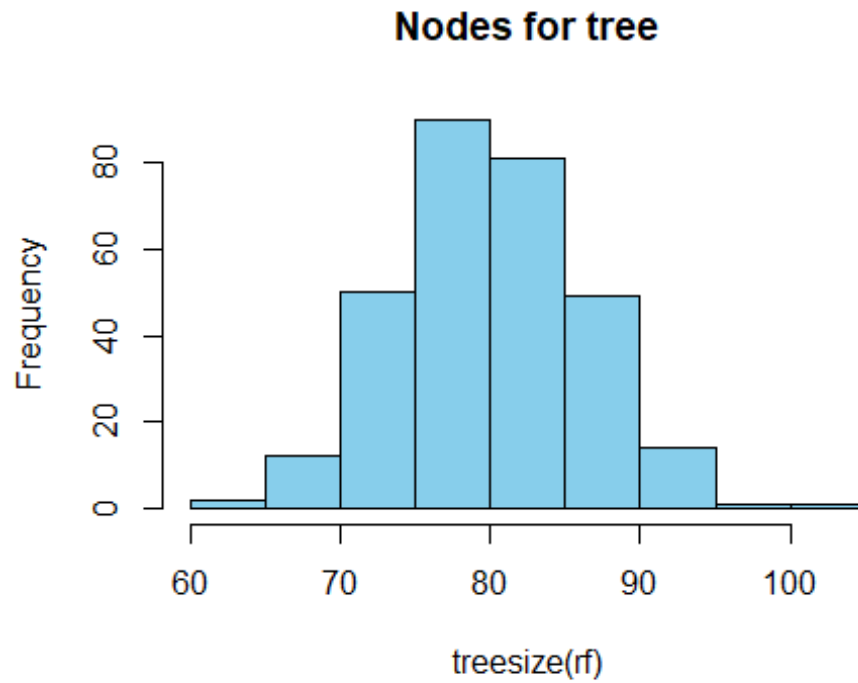
set.seed(222)
rf<-randomForest(NSP~., data = train,
                  ntree = 300,
                  mtry = 8,
                  importance = TRUE,
                  proximity = TRUE)

rf

##
## Call:
## randomForest(formula = NSP ~ ., data = train, ntree = 300, mtry = 8,
## importance = TRUE, proximity = TRUE)
##              Type of random forest: classification
##              Number of trees: 300
## No. of variables tried at each split: 8
##
##              OOB estimate of  error rate: 5.58%
## Confusion matrix:
##      1   2   3 class.error
## 1 1172  18   5 0.01924686
## 2   51 147   3 0.26865672
## 3    5   3 119 0.06299213
```

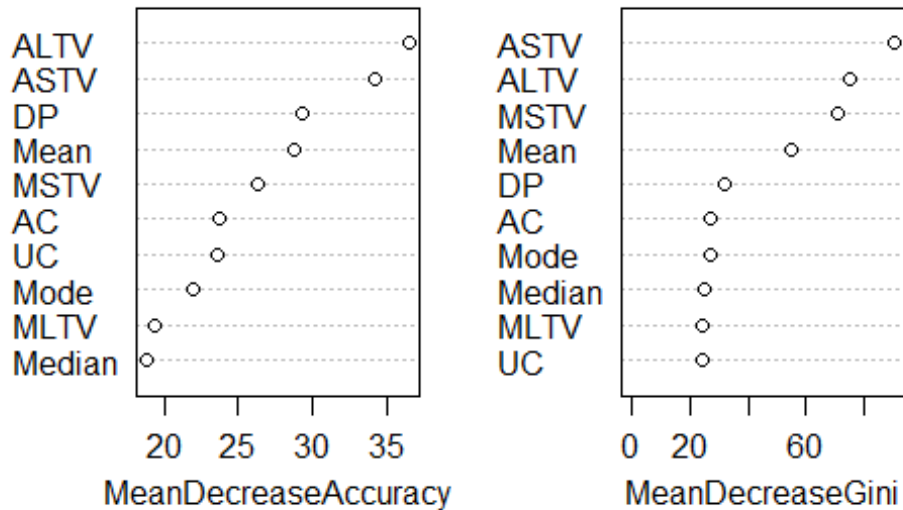
Number of nodes for the tree

```
hist(treesize(rf), main = "Nodes for tree",  
     col = "skyblue")
```



```
varImpPlot(rf, sort = TRUE, n.var = 10, main = "top-10 features", cex = 1)
```

## top-10 features



graph 1 : tests how worse the model performs without each variable. graph 2: tells us how pure the node are at the end of the tree without each variable.

quantative values.

`importance(rf)` #gives Gini Index (priority of variable)

##	1	2	3	MeanDecreaseAccuracy
MeanDecreaseGini				
## LB	17.132286	7.16274976	7.915169	18.805473
18.02488392				
## AC	21.100620	17.98906880	11.200648	23.698048
27.22863814				
## FM	9.744602	10.39940650	2.186584	13.131800
9.73630145				
## UC	12.705099	17.20023502	16.931399	23.630715
24.11442412				
## DL	3.083279	1.35286211	5.486728	6.098206
2.98148979				
## DS	1.001671	0.00000000	1.001671	1.418872
0.04737006				
## DP	27.312479	8.34349989	16.750946	29.359711
31.80676618				
## ASTV	18.921616	32.72507442	31.466362	34.225644
90.53128264				
## MSTV	14.854489	23.26088999	21.324683	26.363290
70.92520421				
## ALTV	23.305467	26.53479445	33.902552	36.558538



```

74.86578357
## MLTV      12.360262 12.71562783  9.990416      19.410874
24.36258107
## Width     13.333146  4.41577766  5.059051      14.839750
12.88031946
## Min       10.621741  5.63112208  8.668200      15.052170
14.10036320
## Max       11.920657  5.81771565  6.494584      15.013933
13.25917462
## Nmax      8.490765  2.61552171  4.776950      9.589162
8.94732671
## Nzeros    3.498580  1.86340874  2.545108      4.551117
2.00771437
## Mode      17.060443  8.53445890 10.136993      21.992184
26.93020145
## Mean      23.503174 11.56692165 19.502985      28.809405
55.15406327
## Median    15.417586 10.14251823 10.577731      18.873297
24.60305914
## Variance  11.838876  1.90823463  6.554857      12.023446
10.53096562
## Tendency  4.870128 -0.01828876  4.359204      6.385997
2.71653751

```

To find out which predictor variables are actually used in the random forest.

```
varUsed(rf)
```

```

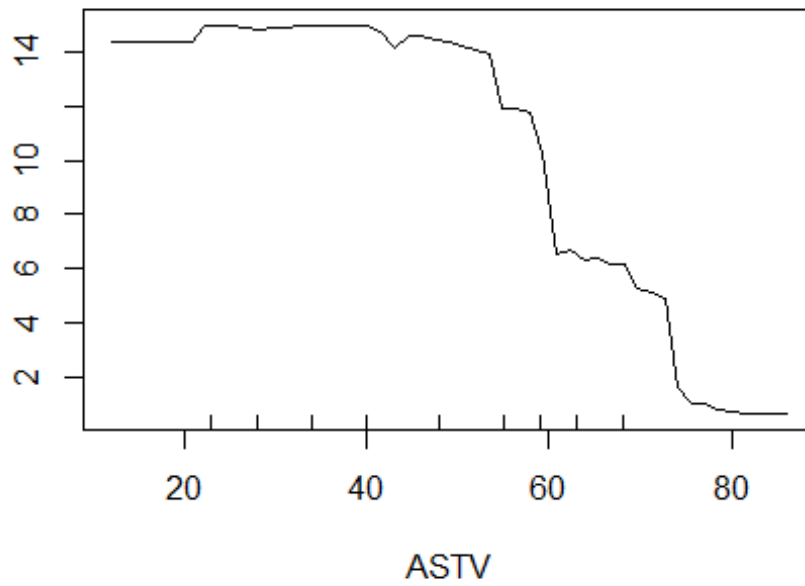
## [1] 1275 1058 1011 1691 331 4 696 2203 1220 2151 1659 1248 1374 1201
856
## [16] 278 1384 1589 1300 953 351

```

Partial dependance plot.

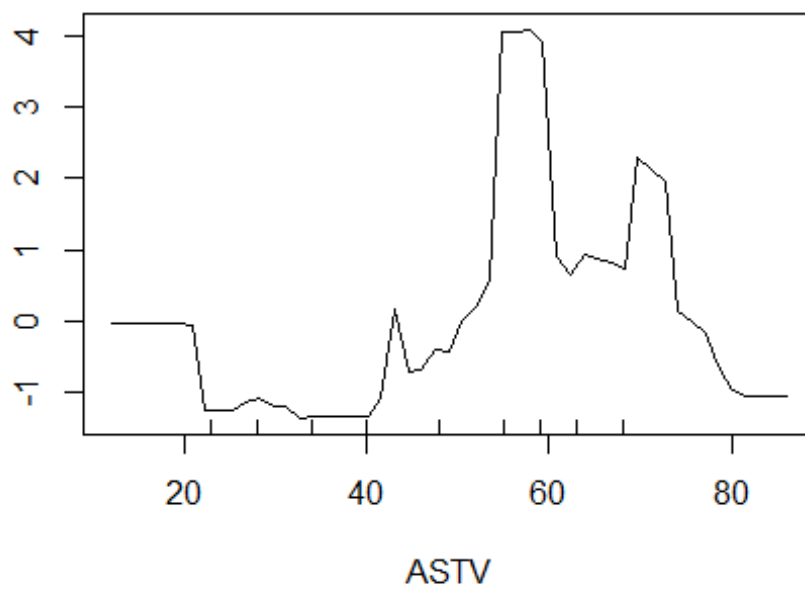
```
partialPlot(rf, train, ASTV, "1")
```

**Partial Dependence on ASTV**



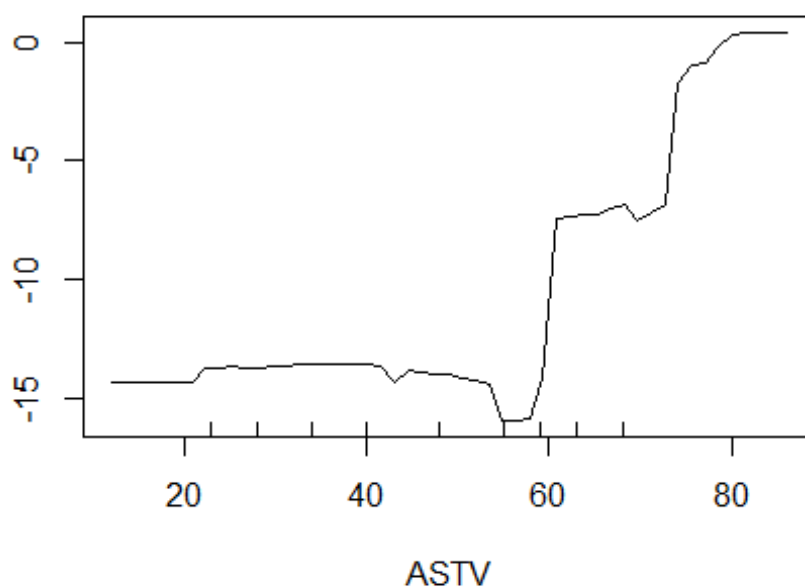
```
partialPlot(rf, train, ASTV, "2")
```

**Partial Dependence on ASTV**



```
partialPlot(rf, train, ASTV, "3")
```

## Partial Dependence on ASTV



Extract Single Tree

from the forest:

```
getTree(rf,1,labelVar = TRUE)
```

##	left daughter	right daughter	split var	split point	status	prediction
## 1	2	3	ALTV	1.650000e+01	1	<NA>
## 2	4	5	LB	1.485000e+02	1	<NA>
## 3	6	7	ASTV	5.950000e+01	1	<NA>
## 4	8	9	DP	1.563723e-03	1	<NA>
## 5	10	11	Nmax	5.000000e-01	1	<NA>
## 6	12	13	ALTV	5.150000e+01	1	<NA>
## 7	14	15	FM	2.843826e-03	1	<NA>
## 8	16	17	Mode	9.250000e+01	1	<NA>
## 9	18	19	Mean	1.100000e+02	1	<NA>
## 10	0	0	<NA>	0.000000e+00	-1	2
## 11	20	21	AC	2.934742e-03	1	<NA>
## 12	22	23	Mode	1.610000e+02	1	<NA>
## 13	24	25	MLTV	8.300000e+00	1	<NA>
## 14	26	27	Max	1.545000e+02	1	<NA>
## 15	28	29	MLTV	4.050000e+00	1	<NA>
## 16	0	0	<NA>	0.000000e+00	-1	3
## 17	30	31	ASTV	7.750000e+01	1	<NA>
## 18	32	33	Nmax	2.500000e+00	1	<NA>
## 19	34	35	UC	1.858616e-03	1	<NA>
## 20	36	37	Max	1.780000e+02	1	<NA>
## 21	0	0	<NA>	0.000000e+00	-1	1
## 22	38	39	MLTV	1.060000e+01	1	<NA>

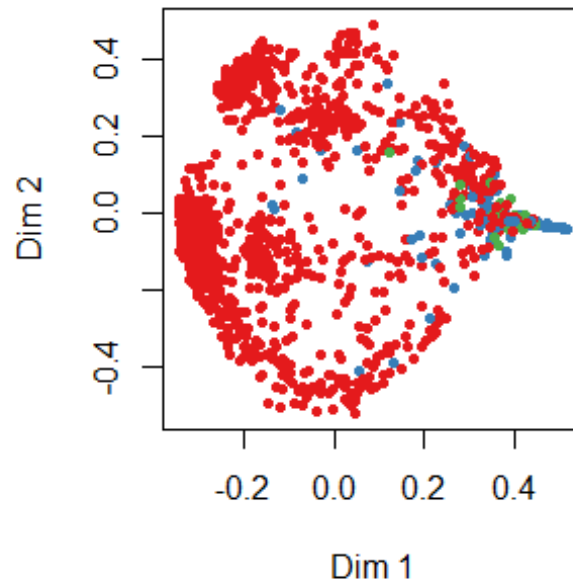
## 23	40	41	Width	9.600000e+01	1	<NA>
## 24	42	43	Median	1.595000e+02	1	<NA>
## 25	0	0	<NA>	0.000000e+00	-1	2
## 26	44	45	MSTV	3.500000e-01	1	<NA>
## 27	46	47	MLTV	6.950000e+00	1	<NA>
## 28	48	49	UC	3.690524e-03	1	<NA>
## 29	0	0	<NA>	0.000000e+00	-1	2
## 30	50	51	MSTV	5.500000e-01	1	<NA>
## 31	52	53	ALTV	4.500000e+00	1	<NA>
## 32	0	0	<NA>	0.000000e+00	-1	2
## 33	0	0	<NA>	0.000000e+00	-1	3
## 34	0	0	<NA>	0.000000e+00	-1	3
## 35	54	55	Mode	1.220000e+02	1	<NA>
## 36	0	0	<NA>	0.000000e+00	-1	1
## 37	0	0	<NA>	0.000000e+00	-1	2
## 38	56	57	Median	1.545000e+02	1	<NA>
## 39	58	59	Min	9.400000e+01	1	<NA>
## 40	0	0	<NA>	0.000000e+00	-1	1
## 41	60	61	MLTV	1.210000e+01	1	<NA>
## 42	62	63	Max	1.595000e+02	1	<NA>
## 43	0	0	<NA>	0.000000e+00	-1	1
## 44	64	65	UC	4.542465e-03	1	<NA>
## 45	66	67	UC	4.549297e-03	1	<NA>
## 46	68	69	ASTV	8.050000e+01	1	<NA>
## 47	0	0	<NA>	0.000000e+00	-1	2
## 48	0	0	<NA>	0.000000e+00	-1	3
## 49	0	0	<NA>	0.000000e+00	-1	1
## 50	70	71	ALTV	7.500000e+00	1	<NA>
## 51	72	73	Min	1.440000e+02	1	<NA>
## 52	0	0	<NA>	0.000000e+00	-1	2
## 53	0	0	<NA>	0.000000e+00	-1	3
## 54	74	75	MLTV	2.250000e+00	1	<NA>
## 55	76	77	Nmax	1.000000e+01	1	<NA>
## 56	0	0	<NA>	0.000000e+00	-1	1
## 57	78	79	LB	1.430000e+02	1	<NA>
## 58	0	0	<NA>	0.000000e+00	-1	1
## 59	80	81	Max	1.550000e+02	1	<NA>
## 60	0	0	<NA>	0.000000e+00	-1	1
## 61	0	0	<NA>	0.000000e+00	-1	2
## 62	82	83	ALTV	5.950000e+01	1	<NA>
## 63	0	0	<NA>	0.000000e+00	-1	2
## 64	84	85	UC	1.101009e-03	1	<NA>
## 65	0	0	<NA>	0.000000e+00	-1	1
## 66	86	87	Width	7.000000e+01	1	<NA>
## 67	88	89	UC	7.059175e-03	1	<NA>
## 68	90	91	ALTV	6.300000e+01	1	<NA>
## 69	0	0	<NA>	0.000000e+00	-1	3
## 70	92	93	Width	7.800000e+01	1	<NA>
## 71	94	95	Mean	1.395000e+02	1	<NA>
## 72	96	97	DP	1.473432e-03	1	<NA>

## 73	0	0	<NA> 0.000000e+00	-1	2
## 74	0	0	<NA> 0.000000e+00	-1	3
## 75	0	0	<NA> 0.000000e+00	-1	2
## 76	98	99	MSTV 2.250000e+00	1	<NA>
## 77	0	0	<NA> 0.000000e+00	-1	2
## 78	0	0	<NA> 0.000000e+00	-1	2
## 79	100	101	Width 2.800000e+01	1	<NA>
## 80	0	0	<NA> 0.000000e+00	-1	2
## 81	0	0	<NA> 0.000000e+00	-1	1
## 82	0	0	<NA> 0.000000e+00	-1	1
## 83	0	0	<NA> 0.000000e+00	-1	2
## 84	0	0	<NA> 0.000000e+00	-1	3
## 85	102	103	Max 1.330000e+02	1	<NA>
## 86	0	0	<NA> 0.000000e+00	-1	2
## 87	0	0	<NA> 0.000000e+00	-1	3
## 88	0	0	<NA> 0.000000e+00	-1	1
## 89	0	0	<NA> 0.000000e+00	-1	3
## 90	104	105	AC 1.363310e-03	1	<NA>
## 91	0	0	<NA> 0.000000e+00	-1	3
## 92	106	107	Nzeros 5.000000e-01	1	<NA>
## 93	0	0	<NA> 0.000000e+00	-1	2
## 94	108	109	Min 9.900000e+01	1	<NA>
## 95	110	111	Min 1.355000e+02	1	<NA>
## 96	112	113	DL 1.172872e-02	1	<NA>
## 97	114	115	Min 6.200000e+01	1	<NA>
## 98	0	0	<NA> 0.000000e+00	-1	1
## 99	116	117	Min 5.200000e+01	1	<NA>
## 100	118	119	LB 1.470000e+02	1	<NA>
## 101	0	0	<NA> 0.000000e+00	-1	1
## 102	0	0	<NA> 0.000000e+00	-1	3
## 103	0	0	<NA> 0.000000e+00	-1	2
## 104	0	0	<NA> 0.000000e+00	-1	2
## 105	0	0	<NA> 0.000000e+00	-1	1
## 106	0	0	<NA> 0.000000e+00	-1	1
## 107	120	121	Mean 1.435000e+02	1	<NA>
## 108	0	0	<NA> 0.000000e+00	-1	2
## 109	0	0	<NA> 0.000000e+00	-1	1
## 110	0	0	<NA> 0.000000e+00	-1	2
## 111	0	0	<NA> 0.000000e+00	-1	1
## 112	122	123	Mean 1.485000e+02	1	<NA>
## 113	124	125	ASTV 5.850000e+01	1	<NA>
## 114	0	0	<NA> 0.000000e+00	-1	1
## 115	126	127	UC 6.175553e-03	1	<NA>
## 116	0	0	<NA> 0.000000e+00	-1	3
## 117	0	0	<NA> 0.000000e+00	-1	1
## 118	128	129	ALTV 2.650000e+01	1	<NA>
## 119	0	0	<NA> 0.000000e+00	-1	1
## 120	0	0	<NA> 0.000000e+00	-1	2
## 121	0	0	<NA> 0.000000e+00	-1	1
## 122	130	131	Mode 1.030000e+02	1	<NA>

## 123	132	133	Max 1.645000e+02	1	<NA>
## 124	0	0	<NA> 0.000000e+00	-1	1
## 125	0	0	<NA> 0.000000e+00	-1	3
## 126	0	0	<NA> 0.000000e+00	-1	2
## 127	0	0	<NA> 0.000000e+00	-1	3
## 128	0	0	<NA> 0.000000e+00	-1	1
## 129	0	0	<NA> 0.000000e+00	-1	1
## 130	134	135	Variance 1.050000e+02	1	<NA>
## 131	136	137	Max 1.365000e+02	1	<NA>
## 132	138	139	Min 9.100000e+01	1	<NA>
## 133	140	141	Min 1.390000e+02	1	<NA>
## 134	0	0	<NA> 0.000000e+00	-1	1
## 135	0	0	<NA> 0.000000e+00	-1	2
## 136	142	143	LB 1.225000e+02	1	<NA>
## 137	144	145	Mode 1.535000e+02	1	<NA>
## 138	0	0	<NA> 0.000000e+00	-1	2
## 139	146	147	MSTV 7.500000e-01	1	<NA>
## 140	148	149	MLTV 1.380000e+01	1	<NA>
## 141	150	151	Nmax 2.500000e+00	1	<NA>
## 142	0	0	<NA> 0.000000e+00	-1	1
## 143	152	153	Mode 1.240000e+02	1	<NA>
## 144	154	155	FM 3.216645e-01	1	<NA>
## 145	156	157	Variance 1.300000e+01	1	<NA>
## 146	0	0	<NA> 0.000000e+00	-1	1
## 147	158	159	Max 1.605000e+02	1	<NA>
## 148	160	161	AC 6.172840e-04	1	<NA>
## 149	162	163	MSTV 1.400000e+00	1	<NA>
## 150	0	0	<NA> 0.000000e+00	-1	1
## 151	0	0	<NA> 0.000000e+00	-1	2
## 152	0	0	<NA> 0.000000e+00	-1	1
## 153	0	0	<NA> 0.000000e+00	-1	2
## 154	0	0	<NA> 0.000000e+00	-1	1
## 155	164	165	MLTV 1.455000e+01	1	<NA>
## 156	0	0	<NA> 0.000000e+00	-1	2
## 157	0	0	<NA> 0.000000e+00	-1	1
## 158	0	0	<NA> 0.000000e+00	-1	1
## 159	0	0	<NA> 0.000000e+00	-1	2
## 160	166	167	ASTV 4.450000e+01	1	<NA>
## 161	0	0	<NA> 0.000000e+00	-1	1
## 162	0	0	<NA> 0.000000e+00	-1	2
## 163	168	169	UC 5.153023e-03	1	<NA>
## 164	0	0	<NA> 0.000000e+00	-1	1
## 165	0	0	<NA> 0.000000e+00	-1	2
## 166	0	0	<NA> 0.000000e+00	-1	2
## 167	0	0	<NA> 0.000000e+00	-1	1
## 168	0	0	<NA> 0.000000e+00	-1	1
## 169	170	171	UC 5.629126e-03	1	<NA>
## 170	0	0	<NA> 0.000000e+00	-1	2
## 171	0	0	<NA> 0.000000e+00	-1	1

Multi-Dimension scaling plot of proximity Matrix

```
MDSplot(rf, train$NSP)
```



Prediction and confusion Matrix

```
library(caret)
```

```
## Warning: package 'caret' was built under R version 3.6.3
```

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

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

```
## Warning: package 'ggplot2' was built under R version 3.6.3
```

```
##
```

```
## Attaching package: 'ggplot2'
```

```
## The following object is masked from 'package:randomForest':
```

```
##
```

```
##      margin
```

```
prd <- predict(rf, train)
```

```
head(prd,20) #predicted
```

```
##  1  2  3  4  6  7  8  9 10 11 12 13 15 17 18 19 20 21 22 23
```

```
##  2  1  1  1  3  3  3  3  3  2  2  1  1  1  2  1  1  3  1  3
```

```
## Levels: 1 2 3
```

```
head(train$NSP, 20) #actual
## [1] 2 1 1 1 3 3 3 3 3 2 2 1 1 1 2 1 1 3 1 3
## Levels: 1 2 3
```

prediction & Confusion Matrix on train data

```
confusionMatrix(prd, train$NSP)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    1    2    3
##           1 1195    1    0
##           2    0  200    0
##           3    0    0  127
##
## Overall Statistics
##
##               Accuracy : 0.9993
##               95% CI : (0.9963, 1)
##       No Information Rate : 0.7846
##       P-Value [Acc > NIR] : < 2.2e-16
##
##               Kappa : 0.9982
##
##  Mcnemar's Test P-Value : NA
##
## Statistics by Class:
##
##              Class: 1 Class: 2 Class: 3
## Sensitivity          1.0000   0.9950  1.00000
## Specificity          0.9970   1.0000  1.00000
## Pos Pred Value       0.9992   1.0000  1.00000
## Neg Pred Value       1.0000   0.9992  1.00000
## Prevalence           0.7846   0.1320  0.08339
## Detection Rate       0.7846   0.1313  0.08339
## Detection Prevalence 0.7853   0.1313  0.08339
## Balanced Accuracy    0.9985   0.9975  1.00000
```

prediction & Confusion Matrix on test data

```
prd2 <- predict(rf, validate)
confusionMatrix(prd2, validate$NSP)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    1    2    3
##           1  457   19    5
##           2    2   73    2
```



```

##          3    1    2   42
##
## Overall Statistics
##
##          Accuracy : 0.9486
##          95% CI : (0.9278, 0.9648)
##      No Information Rate : 0.7629
##      P-Value [Acc > NIR] : < 2.2e-16
##
##          Kappa : 0.8594
##
##  McNemar's Test P-Value : 0.0009261
##
## Statistics by Class:
##
##          Class: 1 Class: 2 Class: 3
## Sensitivity      0.9935   0.7766   0.85714
## Specificity      0.8322   0.9921   0.99458
## Pos Pred Value   0.9501   0.9481   0.93333
## Neg Pred Value   0.9754   0.9601   0.98746
## Prevalence       0.7629   0.1559   0.08126
## Detection Rate   0.7579   0.1211   0.06965
## Detection Prevalence 0.7977   0.1277   0.07463
## Balanced Accuracy 0.9128   0.8844   0.92586

```