

# R\_DECISION\_TREES\_RANDOM\_FOREST

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## Importing libraries and dataset

```
library(ggplot2)
```

```
library(ISLR)
```

```
head(College)
```

```
## Private Apps Accept Enroll Top10perc
```

```
Top25perc
```

```
## Abilene Christian University Yes 1660 1232 721 23
```

```
52
```

```
## Adelphi University Yes 2186 1924 512 16
```

```
29
```

```
## Adrian College Yes 1428 1097 336 22
```

```
50
```

```
## Agnes Scott College Yes 417 349 137 60
```

```
89
```

```
## Alaska Pacific University Yes 193 146 55 16
```

```
44
```

```
## Albertson College Yes 587 479 158 38
```

```
62
```

```
## F.Undergrad P.Undergrad Outstate
```

```
Room.Board Books
```

```
## Abilene Christian University 2885 537 7440
```

```
3300 450
```

```
## Adelphi University 2683 1227 12280
```

```
6450 750
```

```
## Adrian College 1036 99 11250
```

```
3750 400
```

```
## Agnes Scott College 510 63 12960
```

```
5450 450
```

```
## Alaska Pacific University 249 869 7560
```

```
4120 800
```

```
## Albertson College 678 41 13500
```

```
3335 500
```

```
## Personal PhD Terminal S.F.Ratio
```

```
perc.alumni Expend
```

```
## Abilene Christian University 2200 70 78 18.1
```

```
12 7041
```

```
## Adelphi University 1500 29 30 12.2
```

```
16 10527
```

```
## Adrian College 1165 53 66 12.9
```

```

30  8735
## Agnes Scott College      875  92    97    7.7
37 19016
## Alaska Pacific University 1500  76    72   11.9
2  10922
## Albertson College        675  67    73    9.4
11  9727
##                               Grad.Rate
## Abilene Christian University    60
## Adelphi University             56
## Adrian College                 54
## Agnes Scott College            59
## Alaska Pacific University      15
## Albertson College              55

df = College

```

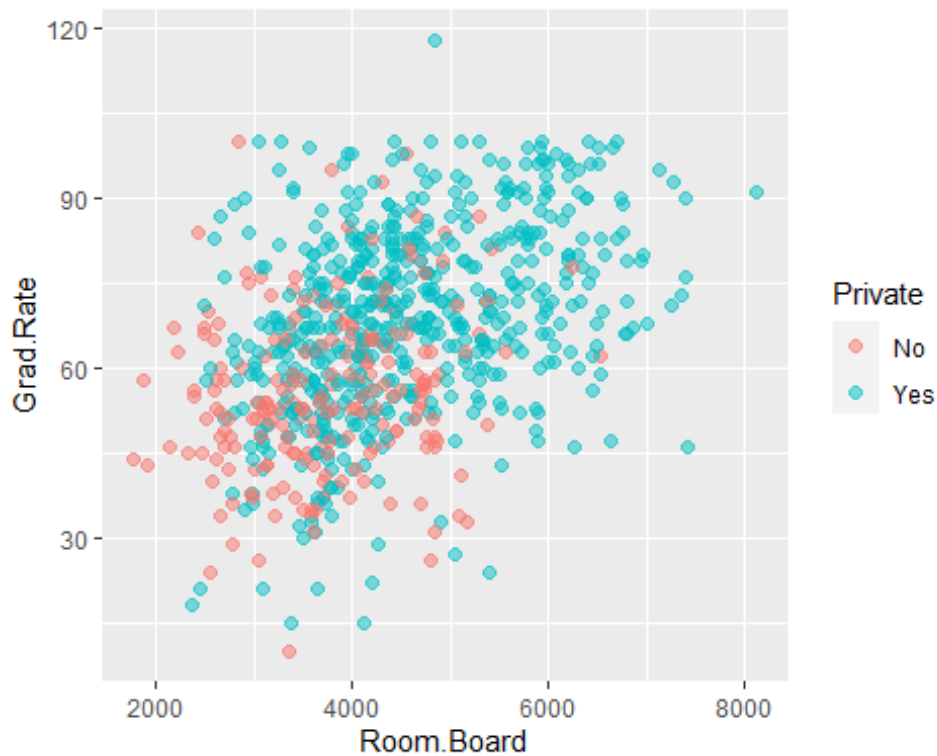
## Exploratory Data Analysis

### Correlation plot

```

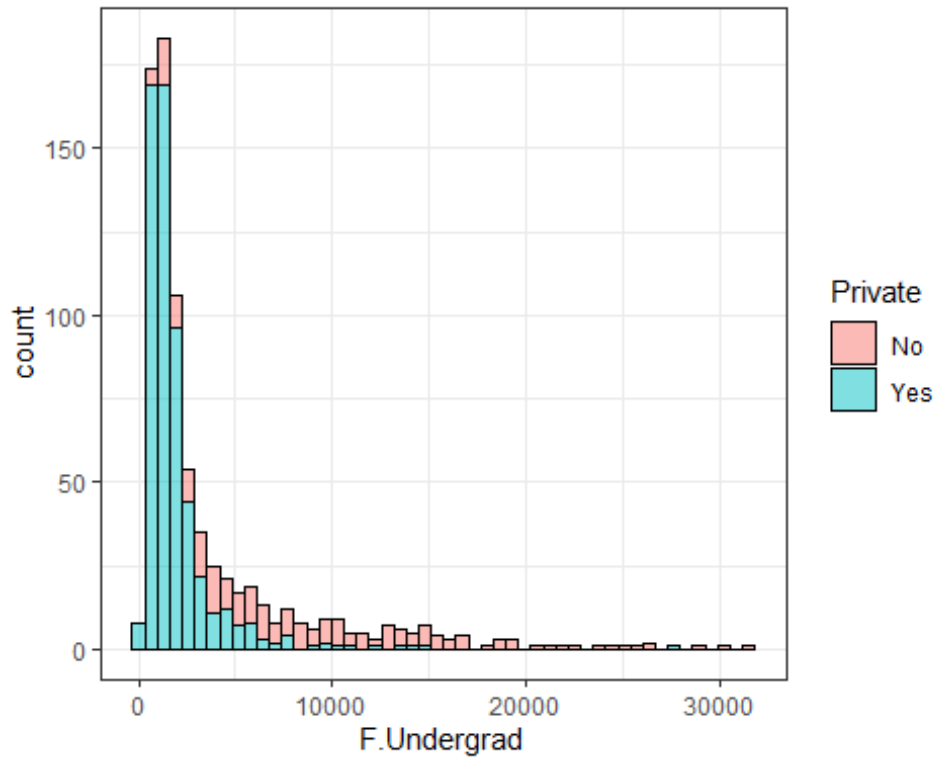
ggplot(df, aes(Room.Board, Grad.Rate)) + geom_point(aes(color = Private),
size = 2, alpha = 0.5)

```



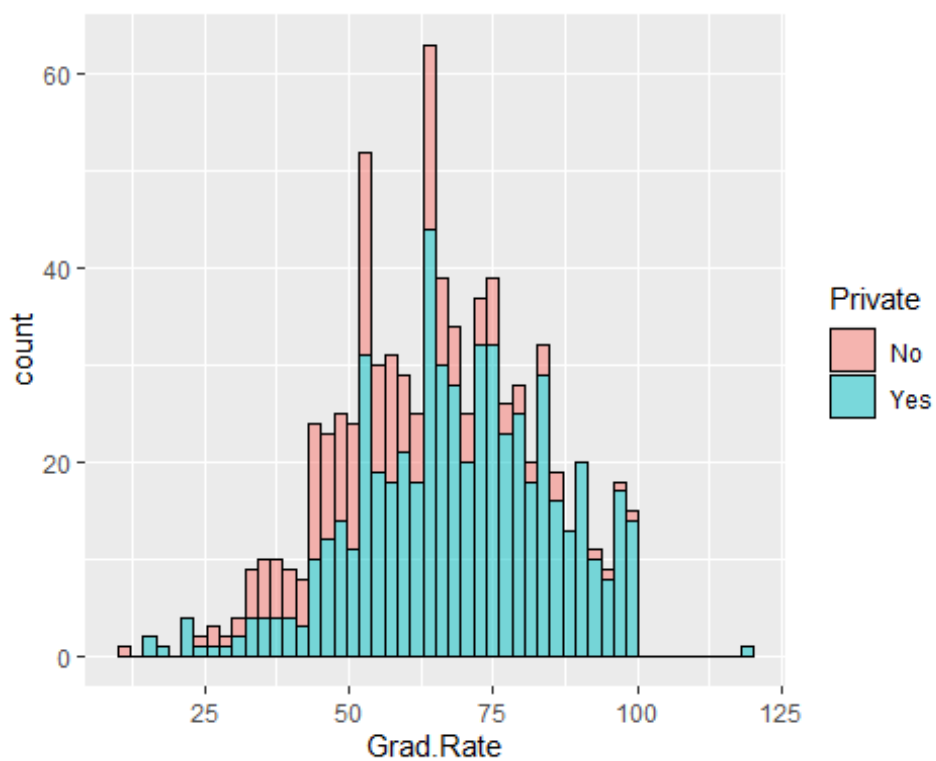
## Histogram plot

```
ggplot(df, aes(F.Undergrad)) + geom_histogram(aes(fill = Private),  
color = "black", bins = 50,  
alpha = 0.5) + theme_bw()
```



We noticed an outlier, so we are going to change that value.

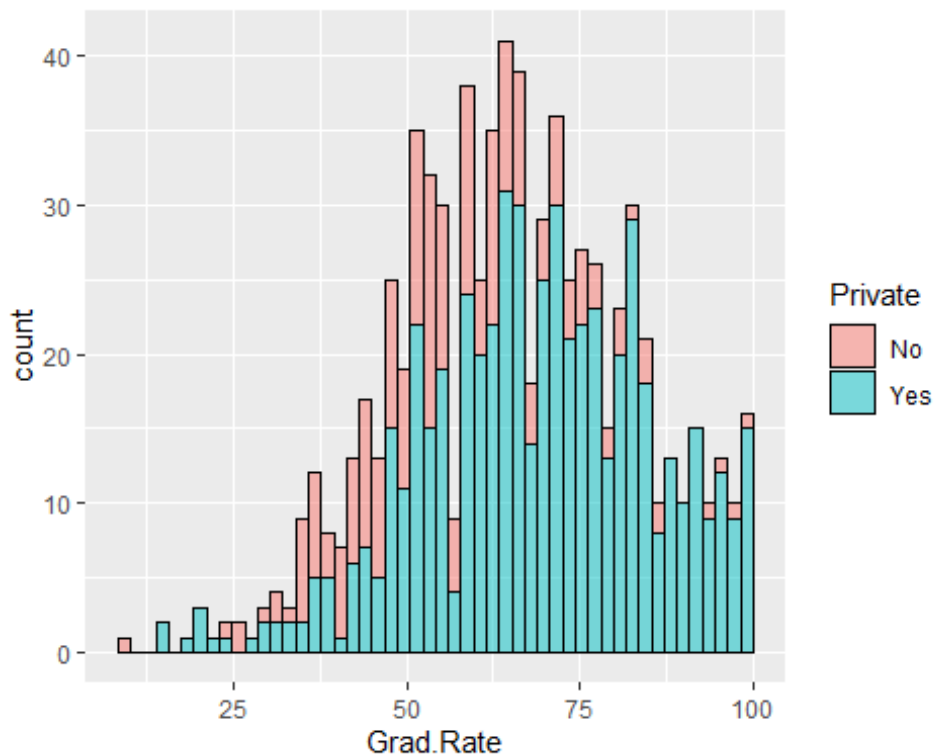
```
ggplot(df, aes(Grad.Rate)) + geom_histogram(aes(fill = Private),  
                                             color = "black",  
                                             bins = 50,  
                                             alpha = 0.5)
```



## Confirming that the outlier is changed

```
df["Cazenovia College", "Grad.Rate"] = 100
```

```
ggplot(df, aes(Grad.Rate)) + geom_histogram(aes(fill = Private),  
                                             color = "black",  
                                             bins = 50,  
                                             alpha = 0.5)
```



## Creating the Decision Tree model

Splitting the dataset to start with the ML model.

```
library(caTools)  
sample = sample.split(df$Private, SplitRatio= 0.7 )  
dfTrain = subset(df,sample == T)  
dfTest = subset(df,sample == F)
```

Training the Decision Tree model

```
library(rpart)  
tree = rpart(Private~., method = "class", data = dfTrain)  
summary(tree)
```

```
## Call:
```

```
## rpart(formula = Private ~ ., data = dfTrain, method = "class")
```

```
## n= 544
```

```
##
```

```

##          CP nsplit rel error      xerror      xstd
## 1 0.45945946      0 1.0000000 1.0000000 0.07013217
## 2 0.25000000      1 0.5405405 0.6554054 0.06032238
## 3 0.03378378      2 0.2905405 0.4189189 0.05007929
## 4 0.01801802      3 0.2567568 0.4459459 0.05145466
## 5 0.01000000      7 0.1756757 0.4459459 0.05145466
##
## Variable importance
## F.Undergrad      Enroll      Accept      Apps      Outstate
P.Undergrad
##          17          15          13          11          10
8
## S.F.Ratio  Room.Board      Expend  Grad.Rate  Top10perc
Terminal
##          7          6          5          4          3
1
##          PhD
##          1
##
## Node number 1: 544 observations,      complexity param=0.4594595
## predicted class=Yes expected loss=0.2720588 P(node) =1
## class counts: 148 396
## probabilities: 0.272 0.728
## left son=2 (176 obs) right son=3 (368 obs)
## Primary splits:
## F.Undergrad < 2995      to the right, improve=92.28086, (0
missing)
## Outstate < 7960      to the left, improve=90.46910, (0
missing)
## Enroll < 754.5      to the right, improve=73.60392, (0
missing)
## P.Undergrad < 896      to the right, improve=63.18460, (0
missing)
## S.F.Ratio < 14.55      to the right, improve=56.98459, (0
missing)
## Surrogate splits:
## Enroll < 754.5      to the right, agree=0.960, adj=0.875, (0
split)
## Accept < 2040.5      to the right, agree=0.926, adj=0.773, (0
split)
## Apps < 2995      to the right, agree=0.892, adj=0.665, (0
split)
## P.Undergrad < 870      to the right, agree=0.831, adj=0.477, (0
split)
## S.F.Ratio < 17.85      to the right, agree=0.752, adj=0.233, (0
split)
##
## Node number 2: 176 observations,      complexity param=0.25
## predicted class=No expected loss=0.3068182 P(node) =0.3235294
## class counts: 122 54

```

```

## probabilities: 0.693 0.307
## left son=4 (117 obs) right son=5 (59 obs)
## Primary splits:
## Outstate < 10218.5 to the left, improve=45.58072, (0 missing)
## Room.Board < 5460.5 to the left, improve=42.82834, (0 missing)
## Expend < 16636 to the left, improve=22.91525, (0 missing)
## S.F.Ratio < 14.85 to the right, improve=21.41266, (0 missing)
## Grad.Rate < 79.5 to the left, improve=18.86120, (0 missing)
## Surrogate splits:
## Room.Board < 5460.5 to the left, agree=0.881, adj=0.644, (0
split)
## Expend < 9436 to the left, agree=0.858, adj=0.576, (0
split)
## S.F.Ratio < 14.55 to the right, agree=0.824, adj=0.475, (0
split)
## Grad.Rate < 72.5 to the left, agree=0.824, adj=0.475, (0
split)
## Top10perc < 52.5 to the left, agree=0.784, adj=0.356, (0
split)
##
## Node number 3: 368 observations, complexity param=0.01801802
## predicted class=Yes expected loss=0.07065217 P(node) =0.6764706
## class counts: 26 342
## probabilities: 0.071 0.929
## left son=6 (67 obs) right son=7 (301 obs)
## Primary splits:
## Outstate < 7960 to the left, improve=13.546690, (0
missing)
## Room.Board < 2790 to the left, improve= 7.299453, (0
missing)
## Expend < 6717.5 to the left, improve= 5.366087, (0
missing)
## S.F.Ratio < 14.35 to the right, improve= 5.093944, (0
missing)
## Grad.Rate < 52.5 to the left, improve= 4.687918, (0
missing)
## Surrogate splits:
## Room.Board < 3072.5 to the left, agree=0.867, adj=0.269, (0
split)
## Expend < 6206.5 to the left, agree=0.853, adj=0.194, (0
split)
## perc.alumni < 8.5 to the left, agree=0.845, adj=0.149, (0
split)
## Grad.Rate < 45.5 to the left, agree=0.840, adj=0.119, (0
split)
## Top25perc < 13.5 to the left, agree=0.826, adj=0.045, (0
split)
##
## Node number 4: 117 observations
## predicted class=No expected loss=0.05128205 P(node) =0.2150735

```

```

##      class counts:   111     6
##      probabilities: 0.949 0.051
##
## Node number 5: 59 observations,      complexity param=0.03378378
## predicted class=Yes expected loss=0.1864407 P(node) =0.1084559
##      class counts:    11    48
##      probabilities: 0.186 0.814
##      left son=10 (13 obs) right son=11 (46 obs)
##      Primary splits:
##          Accept      < 5934      to the right, improve=8.533757, (0
missing)
##          F.Undergrad < 9781.5   to the right, improve=8.533757, (0
missing)
##          Enroll      < 2477.5   to the right, improve=7.909669, (0
missing)
##          Apps        < 14155.5  to the right, improve=7.145558, (0
missing)
##          Room.Board  < 5406      to the left,  improve=5.949587, (0
missing)
##      Surrogate splits:
##          F.Undergrad < 9781.5   to the right, agree=0.966, adj=0.846, (0
split)
##          Enroll      < 1985      to the right, agree=0.949, adj=0.769, (0
split)
##          P.Undergrad < 1653      to the right, agree=0.898, adj=0.538, (0
split)
##          Apps        < 14155.5  to the right, agree=0.864, adj=0.385, (0
split)
##          Outstate    < 10674     to the left,  agree=0.814, adj=0.154, (0
split)
##
## Node number 6: 67 observations,      complexity param=0.01801802
## predicted class=Yes expected loss=0.358209 P(node) =0.1231618
##      class counts:    24    43
##      probabilities: 0.358 0.642
##      left son=12 (41 obs) right son=13 (26 obs)
##      Primary splits:
##          F.Undergrad < 955       to the right, improve=5.010473, (0
missing)
##          Terminal    < 90        to the right, improve=4.852580, (0
missing)
##          Top10perc   < 7.5       to the left,  improve=4.588426, (0
missing)
##          Outstate    < 4905      to the left,  improve=3.600842, (0
missing)
##          P.Undergrad < 186.5     to the right, improve=3.570256, (0
missing)
##      Surrogate splits:
##          Enroll      < 220       to the right, agree=0.866, adj=0.654, (0
split)

```



```

##      Accept    < 402.5    to the right, agree=0.776, adj=0.423, (0
split)
##      Apps      < 451      to the right, agree=0.746, adj=0.346, (0
split)
##      PhD       < 47       to the right, agree=0.731, adj=0.308, (0
split)
##      Terminal < 61.5     to the right, agree=0.716, adj=0.269, (0
split)
##
## Node number 7: 301 observations
## predicted class=Yes expected loss=0.006644518 P(node) =0.5533088
## class counts:      2    299
## probabilities: 0.007 0.993
##
## Node number 10: 13 observations
## predicted class=No  expected loss=0.3076923 P(node) =0.02389706
## class counts:      9     4
## probabilities: 0.692 0.308
##
## Node number 11: 46 observations
## predicted class=Yes expected loss=0.04347826 P(node) =0.08455882
## class counts:      2    44
## probabilities: 0.043 0.957
##
## Node number 12: 41 observations, complexity param=0.01801802
## predicted class=No  expected loss=0.4878049 P(node) =0.07536765
## class counts:      21    20
## probabilities: 0.512 0.488
## left son=24 (8 obs) right son=25 (33 obs)
## Primary splits:
##      Terminal    < 85.5    to the right, improve=4.730229, (0
missing)
##      Top10perc   < 17      to the left, improve=4.416376, (0
missing)
##      Top25perc   < 53.5    to the left, improve=3.773519, (0
missing)
##      P.Undergrad < 114.5    to the right, improve=3.710027, (0
missing)
##      Grad.Rate   < 55      to the left, improve=3.573929, (0
missing)
## Surrogate splits:
##      PhD         < 78.5    to the right, agree=0.927, adj=0.625, (0
split)
##      Top10perc   < 5.5     to the left, agree=0.854, adj=0.250, (0
split)
##      Room.Board  < 4665    to the right, agree=0.854, adj=0.250, (0
split)
##      P.Undergrad < 1426.5  to the right, agree=0.829, adj=0.125, (0
split)
##

```

```

## Node number 13: 26 observations
##   predicted class=Yes   expected loss=0.1153846   P(node) =0.04779412
##   class counts:      3    23
##   probabilities: 0.115 0.885
##
## Node number 24: 8 observations
##   predicted class=No   expected loss=0   P(node) =0.01470588
##   class counts:      8    0
##   probabilities: 1.000 0.000
##
## Node number 25: 33 observations,   complexity param=0.01801802
##   predicted class=Yes   expected loss=0.3939394   P(node) =0.06066176
##   class counts:      13    20
##   probabilities: 0.394 0.606
##   left son=50 (14 obs) right son=51 (19 obs)
##   Primary splits:
##       Top10perc  < 17      to the left,   improve=3.013215, (0
missing)
##       Expend     < 6623.5  to the left,   improve=2.647282, (0
missing)
##       Grad.Rate  < 59.5    to the left,   improve=2.647282, (0
missing)
##       Top25perc  < 53.5    to the left,   improve=2.472960, (0
missing)
##       P.Undergrad < 114.5  to the right,  improve=1.979798, (0
missing)
##   Surrogate splits:
##       Top25perc  < 42.5    to the left,   agree=0.909, adj=0.786, (0
split)
##       Expend     < 5661.5  to the left,   agree=0.758, adj=0.429, (0
split)
##       Grad.Rate  < 58.5    to the left,   agree=0.758, adj=0.429, (0
split)
##       Room.Board < 2805    to the left,   agree=0.697, adj=0.286, (0
split)
##       Enroll     < 723     to the right,  agree=0.667, adj=0.214, (0
split)
##
## Node number 50: 14 observations
##   predicted class=No   expected loss=0.3571429   P(node) =0.02573529
##   class counts:      9    5
##   probabilities: 0.643 0.357
##
## Node number 51: 19 observations
##   predicted class=Yes   expected loss=0.2105263   P(node) =0.03492647
##   class counts:      4    15
##   probabilities: 0.211 0.789

```

## Making Predictions

```
treePreds = predict(tree, dfTest)
head(treePreds)
```

```
##                No      Yes
## Agnes Scott College 0.006644518 0.9933555
## Albertson College   0.006644518 0.9933555
## Amherst College     0.006644518 0.9933555
## Aquinas College     0.006644518 0.9933555
## Augustana College   0.006644518 0.9933555
## Baker University    0.006644518 0.9933555
```

## Creating the confusion matrix

*First we need to make a dummy variable*

```
treePreds = as.data.frame(treePreds)
```

```
joiner = function(x){
  if (x >= 0.5){
    return("Yes")
  } else{
    return("No")
  }
}
```

```
treePreds$Private = sapply(treePreds$Yes, joiner)
head(treePreds)
```

```
##                No      Yes Private
## Agnes Scott College 0.006644518 0.9933555    Yes
## Albertson College   0.006644518 0.9933555    Yes
## Amherst College     0.006644518 0.9933555    Yes
## Aquinas College     0.006644518 0.9933555    Yes
## Augustana College   0.006644518 0.9933555    Yes
## Baker University    0.006644518 0.9933555    Yes
```

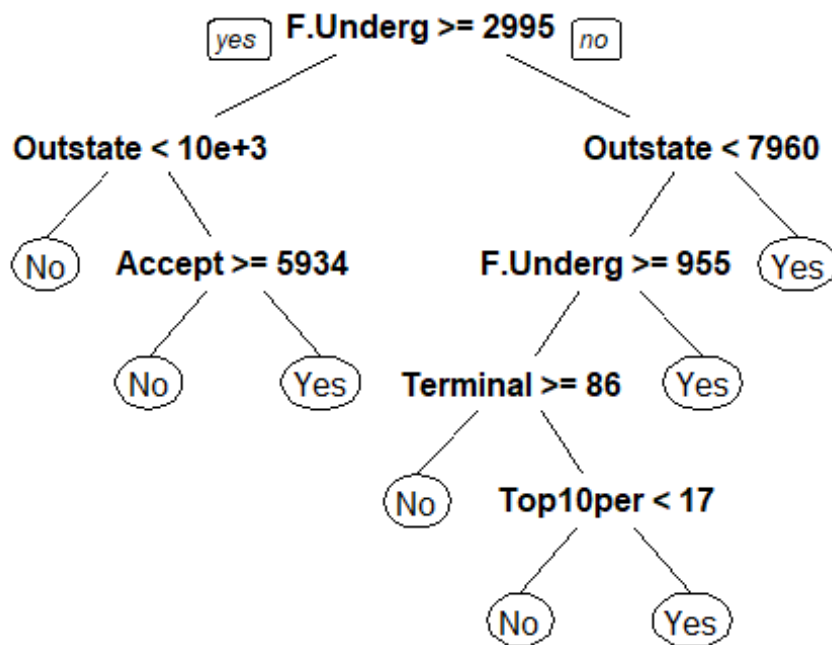
## Confusion Matrix

```
table(treePreds$Private, dfTest$Private)
```

```
##
##      No Yes
## No    60  9
## Yes   4 160
```

## Plotting the model

```
#install.packages("rpart.plot")
library(rpart.plot)
prp(tree)
```



## Creating a Random Forest model

```

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

## randomForest 4.7-1.1

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

##
## Attaching package: 'randomForest'

## The following object is masked from 'package:ggplot2':
##
##   margin

rfModel = randomForest(Private ~., data = dfTrain, importance = T )

rfModel$confusion

##      No Yes class.error
## No  124  24  0.16216216
## Yes  14 382  0.03535354
  
```

### More info about the configuration of the Random Forest configuration

rfModel\$importance

##		No	Yes	MeanDecreaseAccuracy
##	MeanDecreaseGini			
##	Apps	0.019644949	0.009420944	0.0121952782
8.504058				
##	Accept	0.022989110	0.012614294	0.0154104886
11.183018				
##	Enroll	0.043154967	0.030628605	0.0338984148
22.698295				
##	Top10perc	0.009255763	0.003593597	0.0051316264
4.717788				
##	Top25perc	0.006804887	0.003006713	0.0040544520
3.771558				
##	F.Undergrad	0.146791497	0.070376957	0.0909161310
41.555166				
##	P.Undergrad	0.039214644	0.008148831	0.0164449899
17.119321				
##	Outstate	0.151277928	0.061442054	0.0854828819
44.750812				
##	Room.Board	0.020907343	0.018122874	0.0188477754
11.817341				
##	Books	-0.002248360	0.000480029	-0.0002285399
2.286619				
##	Personal	0.002545971	0.002065830	0.0022172322
3.824844				
##	PhD	0.012257379	0.005647215	0.0074292547
4.351807				
##	Terminal	0.005512467	0.007153085	0.0066905877
3.937940				
##	S.F.Ratio	0.031537815	0.006891188	0.0134351017
12.338404				
##	perc.alumni	0.030090264	0.004197012	0.0112169506
6.141727				
##	Expend	0.017987767	0.014976355	0.0157928524
9.435223				
##	Grad.Rate	0.011403514	0.007071716	0.0082546110
6.541857				

### Predictions and confusion Matrix

```
rfPreds = predict(rfModel, dfTest)
table(rfPreds, dfTest$Private)
```

##			
##	rfPreds	No	Yes
##	No	59	4
##	Yes	5	165