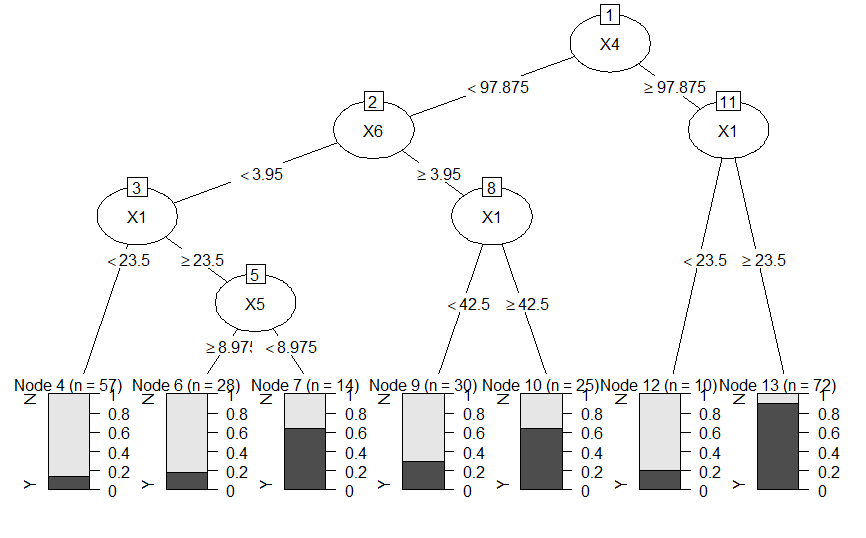
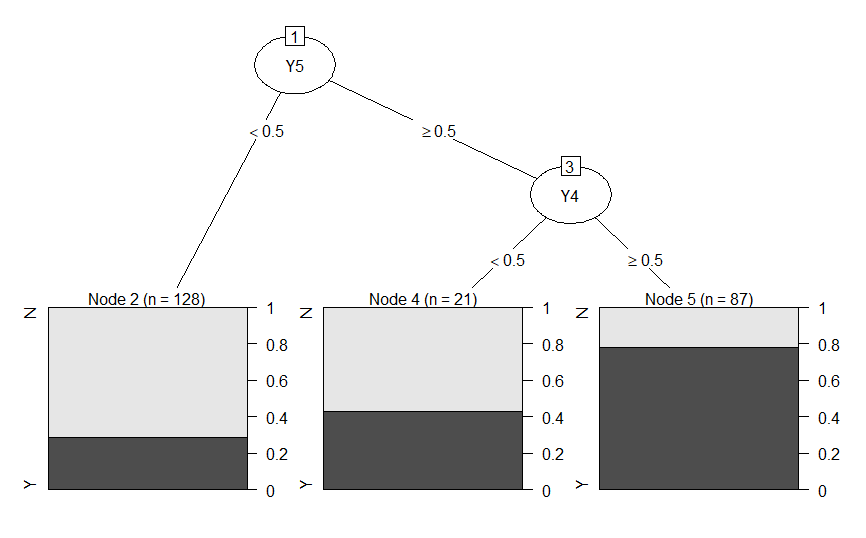
**Project Report for CA 1**

By Arambakam Mukesh - 19301497



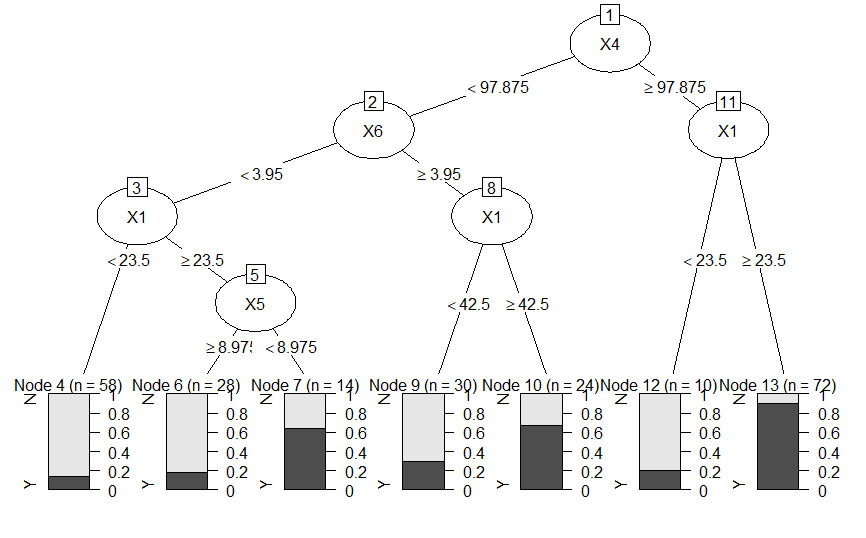
Plot 1 - DT over only X

Plot 1 represents the Decision Tree over the entire data set but with the Predictors though X1-X7. This DT predicts with an accuracy of 80% (0.8).



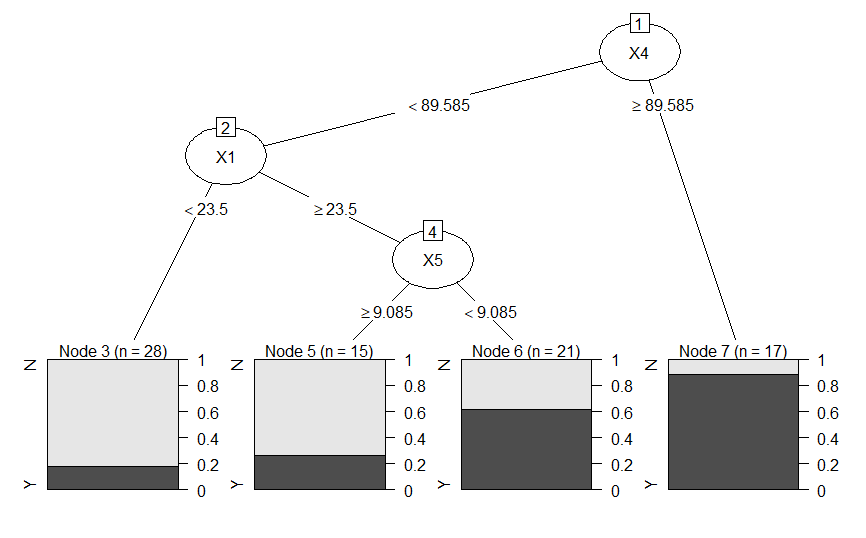
Plot 2 - DT over only Y

Plot 2 represents the Decision Tree over the entire data set but with the Predictors though Y1-Y7. This DT predicts with an accuracy of 80% (0.8).



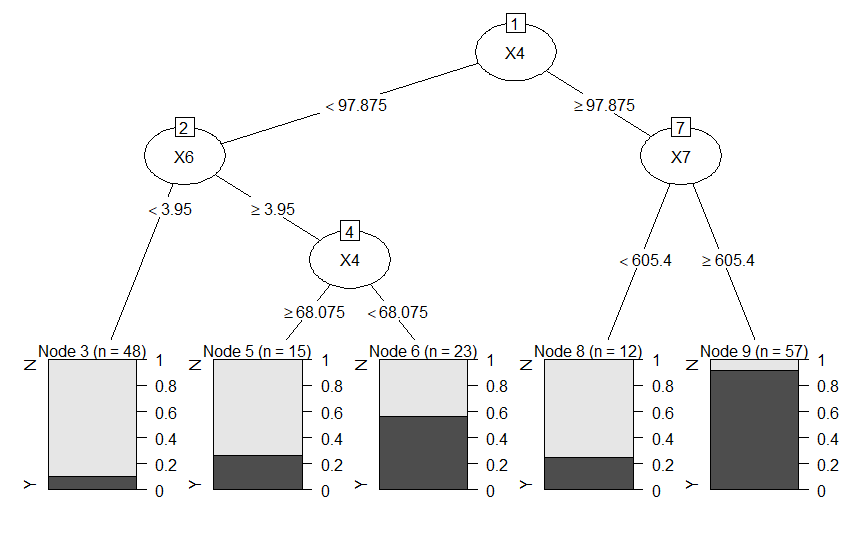
Plot 3 - DT over X and Y

Plot 3 represents the Decision Tree over the entire data set but with the Predictors though X1-X7 and Y1-Y7. This DT predicts with an accuracy of 80% (0.8).



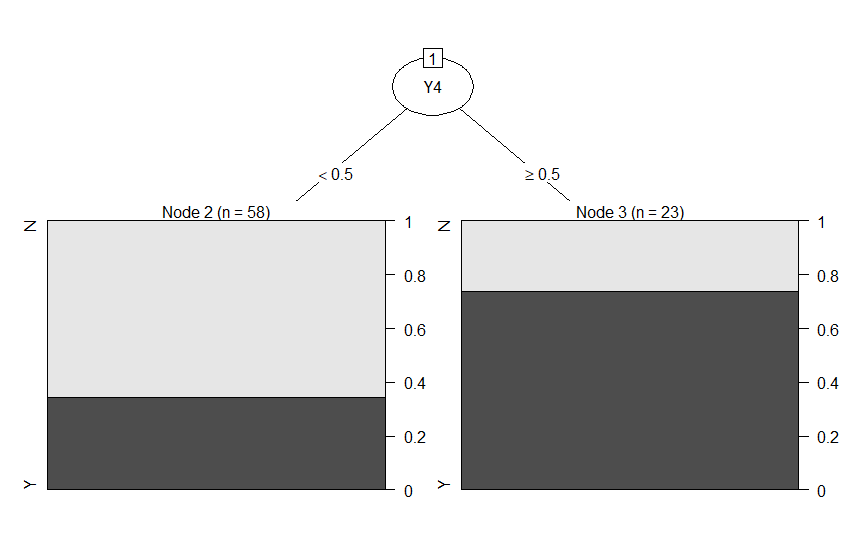
Plot 4 - DT over X with Group 0

Plot 4 represents the Decision Tree over Group 0 set but with the Predictors though X1-X7. This DT predicts with an accuracy of 77% (0.77).



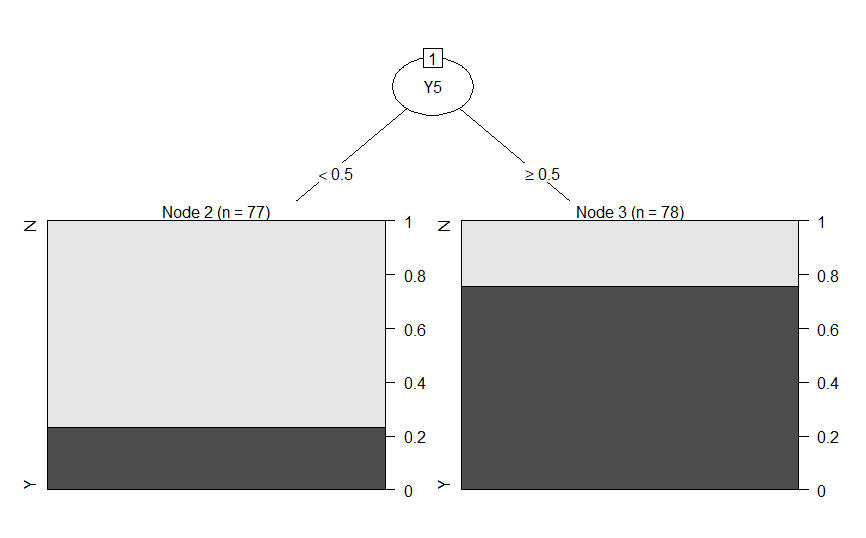
Plot 5 - DT over X with Group 1

Plot 5 represents the Decision Tree over Group 1 set but with the Predictors though X1-X7. This DT predicts with an accuracy of 78% (0.78).



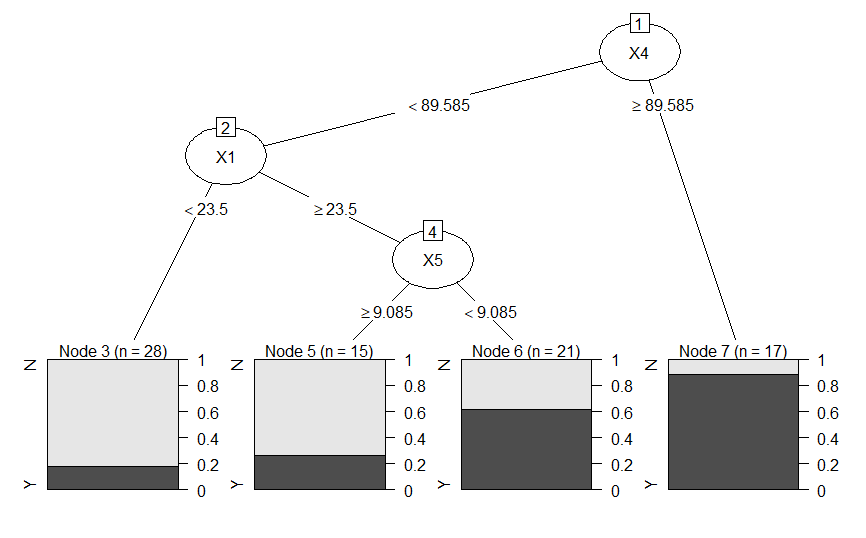
Plot 6 - DT over Y with Group 0

Plot 6 represents the Decision Tree over Group 0 set but with the Predictors though Y1-Y7. This DT predicts with an accuracy of 72% (0.72).



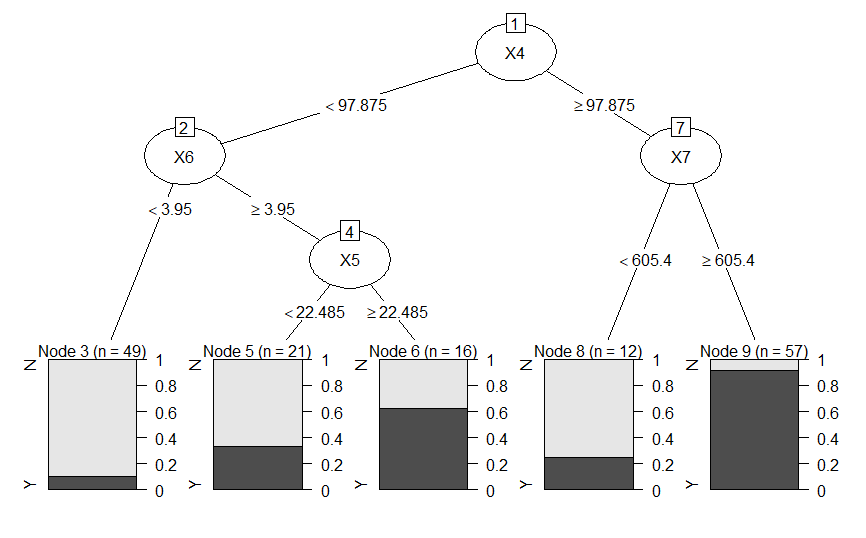
Plot 7 - DT over Y with Group 1

Plot 7 represents the Decision Tree over Group 1 set but with the Predictors though Y1-Y7. This DT predicts with an accuracy of 75% (0.75).



Plot 8 - DT over X and Y with Group 0

Plot 8 represents the Decision Tree over Group 0 set but with the Predictors though X1-X7 and Y1-Y7. This DT predicts with an accuracy of 77% (0.77).



Plot 9 - DT over X and Y with Group 1

Plot 9 represents the Decision Tree over Group 1 set but with the Predictors though X1-X7 and Y1-Y7. This DT predicts with an accuracy of 80% (0.8).

**Conclusion:**

The best Decision Tree generated is the DT generated over **Entire Dataset with the Predictors X1-X7** as it has the highest **accuracy of 80%** - see **Plot 1** for the DT. The below is the Decision Tree’s summary, indicating the splits:

n= 236

node), split, n, loss, yval, (yprob)

\* denotes terminal node

1) root 236 114 N (0.51694915 0.48305085)

2) X4< 97.875 154 47 N (0.69480519 0.30519481)

4) X6< 3.95 99 22 N (0.77777778 0.22222222)

8) X1< 23.5 57 8 N (0.85964912 0.14035088) \*

9) X1>=23.5 42 14 N (0.66666667 0.33333333)

18) X5>=8.975 28 5 N (0.82142857 0.17857143) \*

19) X5< 8.975 14 5 Y (0.35714286 0.64285714) \*

5) X6>=3.95 55 25 N (0.54545455 0.45454545)

10) X1< 42.5 30 9 N (0.70000000 0.30000000) \*

11) X1>=42.5 25 9 Y (0.36000000 0.64000000) \*

3) X4>=97.875 82 15 Y (0.18292683 0.81707317)

6) X1< 23.5 10 2 N (0.80000000 0.20000000) \*

7) X1>=23.5 72 7 Y (0.09722222 0.90277778) \*

Though the DT’s in Plot 3 and Plot 9 also have an accuracy of 80% - Plot 1 is better because it gives a consistent accuracy of 80% when tested with different `rpart` configurations like minsplit, minbucket and maxdepth, more over it has a **Low Variance** between different decision tree when run with different seed values for the data and different rpart configuration. We can also clearly say that X’s are the features with **High Information Gain** compared to Y, and also splitting the data into Group’s do not show a significant improvement accuracy.

The code for this can be found on my GitHub, please find the link to the repo below.

<https://github.com/mukeshmk/r-project>