

Lab 2

Ayesha Gamage-(ayega981)/Muditha Cherangani(mudch175)

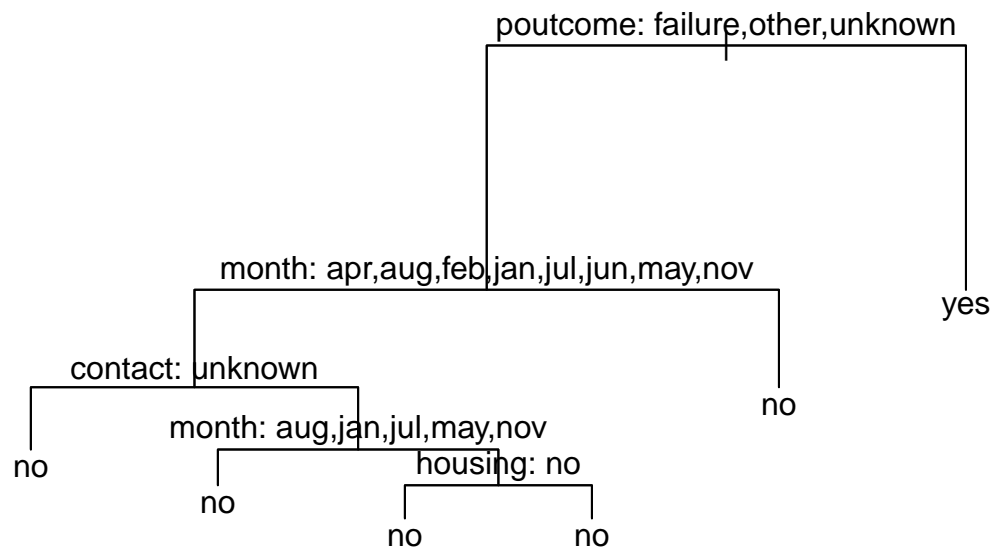
2023-12-05

Assignment 2. Decision trees and logistic regression for bank marketing

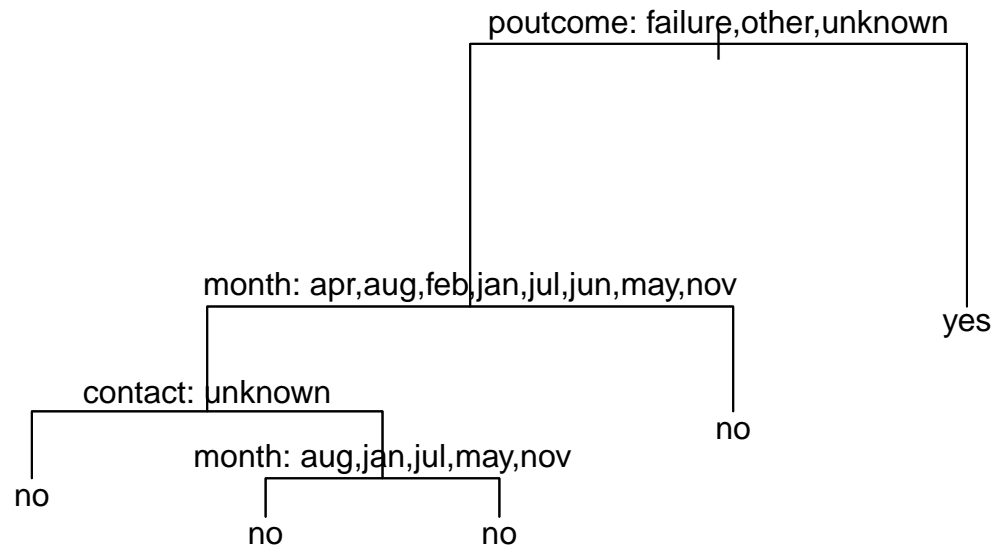
```
## Loading required package: ggplot2
```

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

a. Decision Tree with default settings



b. Decision Tree with smallest allowed node size equal to 7000.



c. Decision trees minimum deviance to 0.0005.



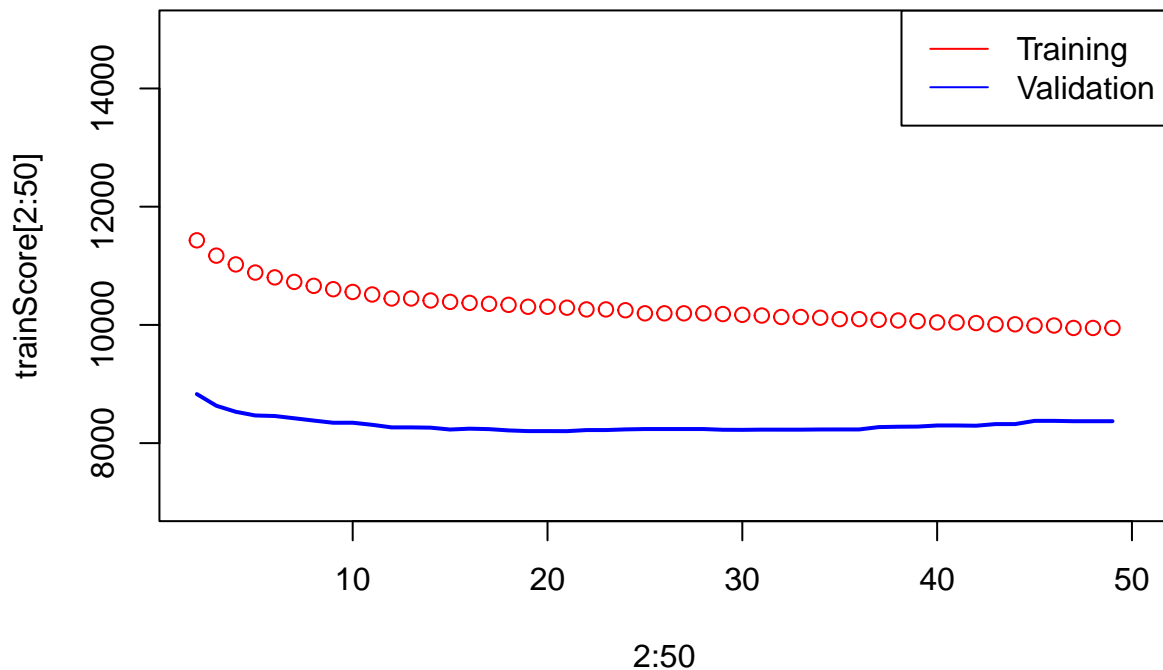
3. Using training and validation sets to choose the optimal tree depth

Table 1: Misclassification Rate

	Train_error	Valid_error
Default	0.1048441	0.1092679
Nodesize_7000	0.1048441	0.1092679
Mindev_0.0005	0.0940058	0.1119221

According to the misclassification rates above ,the Default and Nodesize_7000 models can be regarded as the best ones. In these two models, the Valid_errors are small than the Mindev_0.0005 model and Mindev_0.0005 model's Valid_errors larger.

Deviances of training and validation data with number of leaves



Examine the plot, Deviance are decrease when numbers of leaves increases. But after about 21 it has little increase. This is because as the number of leaves increase, the tree model becomes more and more complex. In here we can take optimal leave when Deviance is minimum.

```
## Optimal number of leaves 21
```

4. Confusion matrix

```
##           Reference
## Prediction    no   yes
##           no 11822  157
##           yes 1309  276
```

```
##
## Accuracy  0.8919198
```

```
## F1 score  0.9416169
```

Accuracy of the optimal model is about 0.9. This model has good predictability.

```
##
## Classification tree:
## tree::tree(formula = as.factor(y) ~ ., data = test, control = tree.control(nobs = n,
##   mincut = optimal_leaves))
## Variables actually used in tree construction:
```

```
## [1] "poutcome" "month"      "contact"  "housing"
## Number of terminal nodes: 6
## Residual mean deviance: 0.6096 = 8265 / 13560
## Misclassification error rate: 0.1081 = 1466 / 13564
```

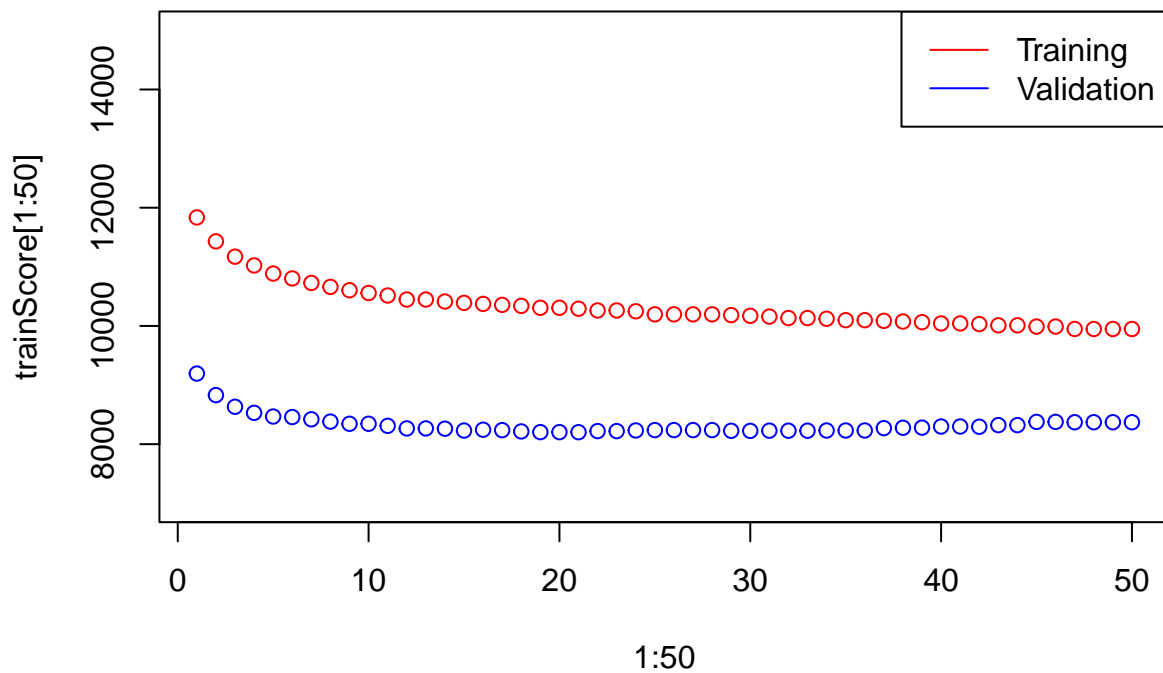
According to this summary “poutcome”, “month”, “contact” and “housing” variables are the most important.

5.A decision tree classification

```
##           Reference
## Prediction    no  yes
##           no 11979  0
##           yes 1585  0
```

```
##
## Accuracy 0.8831466
```

```
## F1 score 0.9379478
```



```
## Optimal leave is, 21
```

4.

```
##           Reference
## Prediction    no   yes
##           no 11822 1309
##           yes   157  276
```

```
## Accuracy
## 0.8919198
```

```
##           F1
## 0.9416169
```

5.

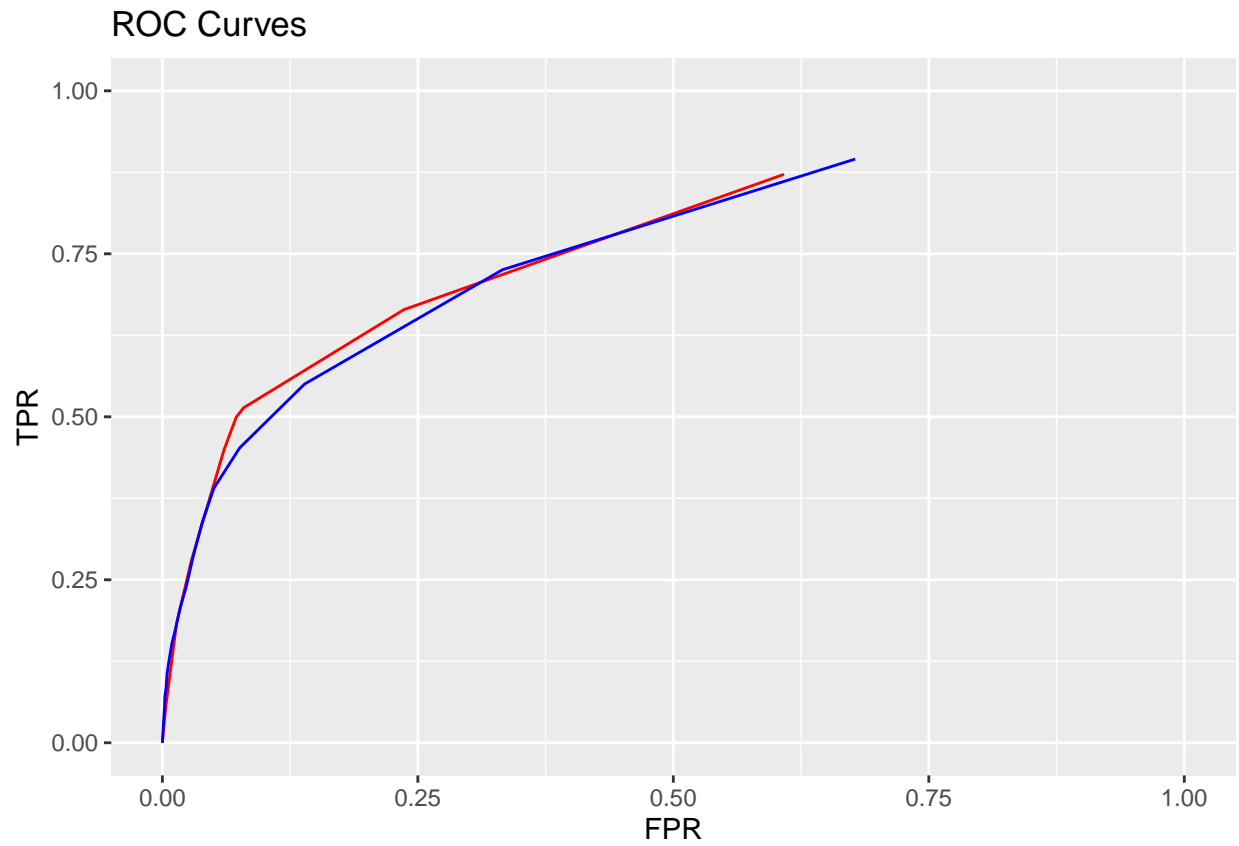
```
##           Reference
## Prediction    no   yes
##           no 11189  861
##           yes   790  724
```

```
## Accuracy
## 0.8919198
```

```
##           F1
## 0.9416169
```

According to the data above we can see that F1 scores and Accuracy approximately equal. But in the confusion matrix, increased the predicted value of yes. We can conclude this model is better.

6. Optimal tree and a Logistic regression



When comparing two plots for the models, we can find that the AUC of the tree model (red curve) is larger than that of glm (blue curve). Here we can conclude the tree model is better.