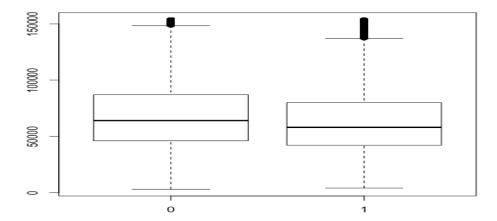
The variable selection process:

- 1. Transform all the columns into numeric or factor. Otherwise we cannot do any regression analysis on the data.
- 2. See the NAs in each column. If a large part of the column is NAs that means lots of information is missing in this variable, so I delete those that have more than 50000,or 5% of total, NA records.
- 3. Next is EDA. As logistic regression does not require the independent variables to be normal, I focus on solving the problem of "complete separation", because this means our model is useless, at least for Inference. So I build boxplots by Loan Status.



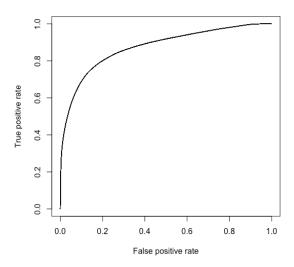
By inspecting the plots, we can know some of the variables will properly cause complete separation. I also draw histograms for all variables.

In the process, I found that annual income and dti have extreme values; I think this may come from those who are un-honest. This will harm the model because it is not real data. So I take 95% of both variables. It's hard to tell whether other variables are abnormal, so I did not do the same thing to them.

- 4. Next is about correlation. Some of them are highly correlated, thus by inspecting correlation matrix, I only leave one of the highly correlated columns. It contains most information.
- 5. Then we come to model building part. In logistic regression, AIC is a better criteria when comparing models and selecting variables. So I use stepAIC(Forward) to choose variables..
- 6. After we get the model, we will look at summary() and anova(). Some

coefficients are not significant, so we can discard them.

- 7. we need to do analysis.
 - A. VIF(). If the vif value is very high(>8), then we need to discard that variable, since it's collinear with others.
 - B. Next is outlierTest. We need to detect the outliers, and remove them from the data set. An then we should do regression again.
 - C. Mcfadden test.
 - D. hoslem.test
 - E. durbinWatsonTest() this is not that useful because this test requires that the variables are normally distributed.
 - F. ROC plot: (I got 0.87 from the final model)



"As a rule of thumb, a model with good predictive ability should have an AUC closer to 1 (1 is ideal) than to 0.5."