

Multiple linear regression and binary logistic regression models

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Overview

This assignment is about to explore, to analyze and to model a data set containing approximately 8000 records representing a customer at an auto insurance company. Each record has two response variables. The first response variable, TARGET_FLAG, is a 1 or a 0. A "1" means that the person was in a car crash. A zero means that the person was not in a car crash. The second response variable is TARGET_AMT. This value is zero if the person did not crash their car. But if they did crash their car, this number will be a value greater than zero. Your objective is to build multiple linear regression and binary logistic regression models on the df1ing data to predict the probability that a person will crash their car and also the amount of money it will cost if the person does crash their car. You can only use the variables given to you (or variables that you derive from the variables provided).

Data exploration

Below is a short description of the variables of interest in the data set:

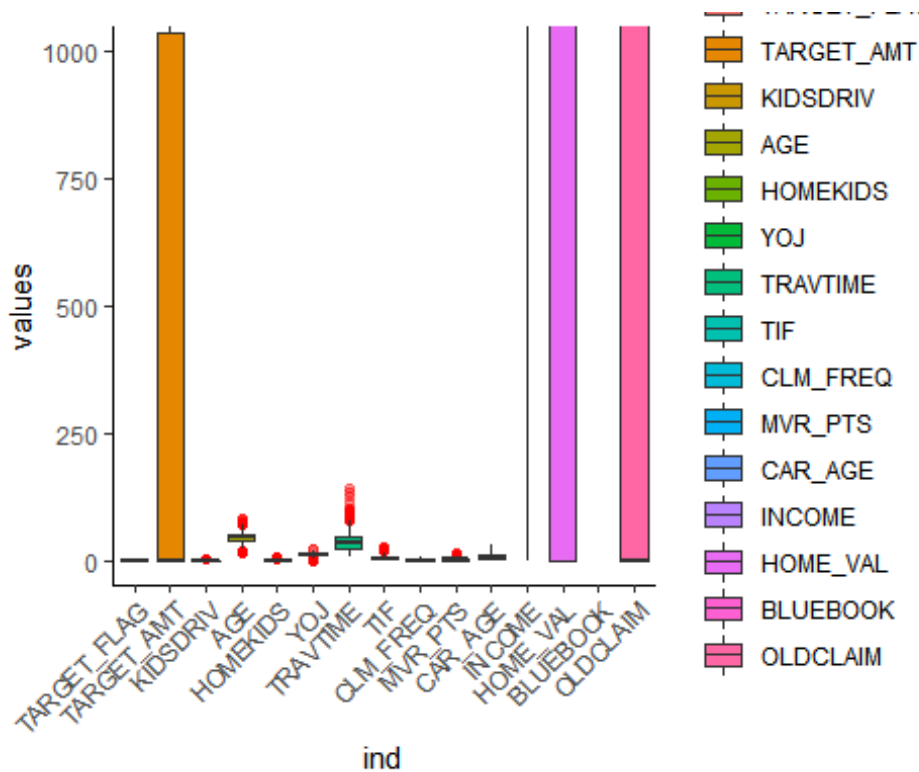
We need to make some cleaning of data such as removing the index, and special characters on certain variables. Below is the head of the data...

Summary table

Visualization

Outliers

The following diagram shows the outliers for all the variables(numerical), both dependent and independent.



Correlation

Below is the correlation between the numerical variables...

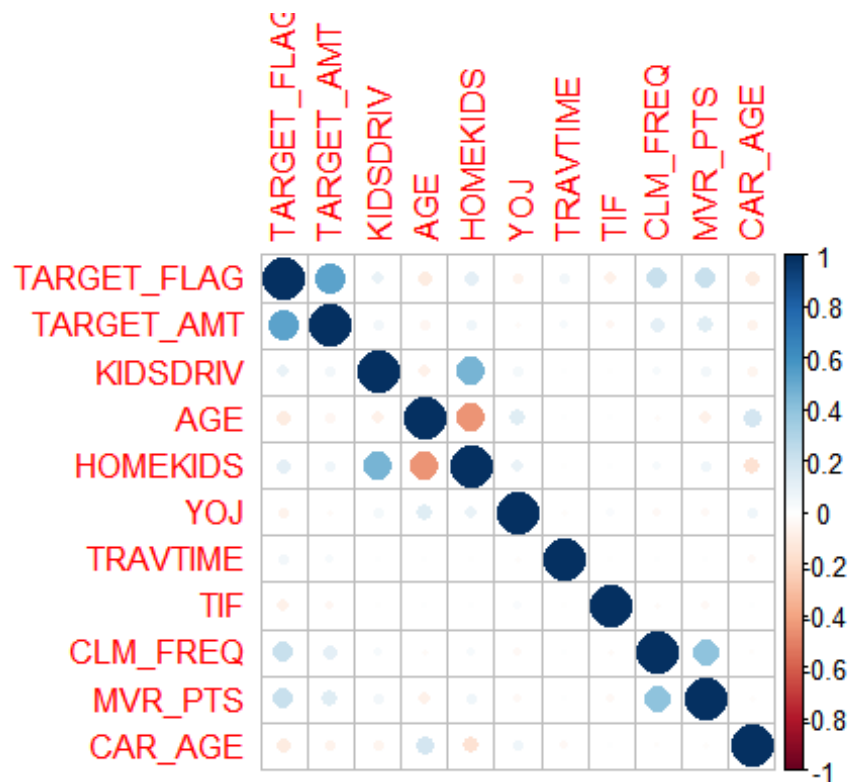
##	TARGET_FLAG	TARGET_AMT	KIDSDRIV	AGE	HOMEKIDS
## TARGET_FLAG	1.00000000	0.53992929	0.097295404	-0.103649398	0.115481537
## TARGET_AMT	0.53992929	1.00000000	0.056857228	-0.046745810	0.068857861
## KIDSDRIV	0.09729540	0.05685723	1.000000000	-0.072361631	0.463046635
## AGE	-0.10364940	-0.04674581	-0.072361631	1.000000000	-0.442383841
## HOMEKIDS	0.11548154	0.06885786	0.463046635	-0.442383841	1.000000000
## YOJ	-0.06658661	-0.02144631	0.048112090	0.139566052	0.090416449
## TRAVTIME	0.05149130	0.03270817	0.008979590	0.004555303	-0.007787772
## TIF	-0.07718644	-0.04525925	-0.003423442	0.002871951	0.004673246
## CLM_FREQ	0.22338169	0.11515694	0.035087170	-0.026312189	0.030695809
## MVR_PTS	0.22526236	0.13770829	0.055019621	-0.073523273	0.062776101
## CAR_AGE	-0.10435770	-0.06283345	-0.055877063	0.182184524	-0.156534495
##	YOJ	TRAVTIME	TIF	CLM_FREQ	
MVR_PTS					

```

## TARGET_FLAG -0.06658661  0.051491295 -0.077186438  0.223381685
0.225262361
## TARGET_AMT -0.02144631  0.032708168 -0.045259254  0.115156936
0.137708292
## KIDSDRIV  0.04811209  0.008979590 -0.003423442  0.035087170
0.055019621
## AGE  0.13956605  0.004555303  0.002871951 -0.026312189 -
0.073523273
## HOMEKIDS  0.09041645 -0.007787772  0.004673246  0.030695809
0.062776101
## YOJ  1.00000000 -0.015762889  0.029302946 -0.030658029 -
0.039172617
## TRAVTIME -0.01576289  1.000000000 -0.009343232  0.009306981
0.009937566
## TIF  0.02930295 -0.009343232  1.000000000 -0.024972898 -
0.037174513
## CLM_FREQ -0.03065803  0.009306981 -0.024972898  1.000000000
0.400121265
## MVR_PTS -0.03917262  0.009937566 -0.037174513  0.400121265
1.000000000
## CAR_AGE  0.06122969 -0.037055196  0.009125709 -0.011538390 -
0.019363647
##          CAR_AGE
## TARGET_FLAG -0.104357704
## TARGET_AMT -0.062833451
## KIDSDRIV -0.055877063
## AGE  0.182184524
## HOMEKIDS -0.156534495
## YOJ  0.061229694
## TRAVTIME -0.037055196
## TIF  0.009125709
## CLM_FREQ -0.011538390
## MVR_PTS -0.019363647
## CAR_AGE  1.000000000

```

Here's the correlation matrix visualization...



There is a little bit of correlation between HomeKids and Kidsdriv, and MVR_PTS and CLM_FREQ

Data Preparation

Missing values

Let explore the number of missing values in each variable

```
##      CAR_AGE      HOME_VAL      YOJ      INCOME      AGE      URBANICITY
##      510      464      454      445      6      0
##      CAR_TYPE      JOB      EDUCATION      SEX      MSTATUS      OLDCLAIM
##      0      0      0      0      0      0
##      BLUEBOOK      MVR_PTS      REVOKED      CLM_FREQ      RED_CAR      TIF
##      0      0      0      0      0      0
##      CAR_USE      TRAVTIME      PARENT1      HOMEKIDS      KIDSDRIV      TARGET_AMT
##      0      0      0      0      0      0
## TARGET_FLAG
##      0

## [1] 1879
```

We will use mean and BoxCox transformation for data imputation, but we need to transform little the data for BoxCox transformation... BoxCox needs variable to have value greater or equal to 1

BoxCox imputation prep

Prepare for BoxCox imputation...

```
##          TARGET_FLAG          TARGET_AMT          KIDSDRIV
##          "0"          " 0.00000"          "0"
##          AGE          HOMEKIDS          YOJ
##          "16"          "0"          " 0"
##          PARENT1          TRAVTIME          CAR_USE
##          "No"          " 5"          "Commercial"
##          TIF          RED_CAR          CLM_FREQ
##          " 1"          "no"          "0"
##          REVOKED          MVR_PTS          CAR_AGE
##          "No"          " 0"          NA
##          INCOME          HOME_VAL          BLUEBOOK
##          " 0"          " 0"          " 1500"
##          OLDCLAIM          MSTATUS          SEX
##          " 0"          "No"          "F"
##          EDUCATION          JOB          CAR_TYPE
##          "<High School"          ""          "Minivan"
##          URBANICITY
##          "Highly Rural/ Rural"

## $TIF
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   1.00   1.00   4.00   5.36   7.00   25.00
##
## Largest/Smallest: 25
## Sample Skewness: 0.89
##
## Estimated Lambda: 0.2
##
##
## $BLUEBOOK
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   1500   9315  14440  15720  20820  69740
##
## Largest/Smallest: 46.5
## Sample Skewness: 0.802
##
```

```

## Estimated Lambda: 0.5
##
##
## $TRAVTIME
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   5.00   23.00   33.00   33.55   44.00   142.00
##
## Largest/Smallest: 28.4
## Sample Skewness: 0.452
##
## Estimated Lambda: 0.7
##
##
## $AGE
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   16.00   39.00   45.00   44.81   51.00   81.00
##
## Largest/Smallest: 5.06
## Sample Skewness: -0.0247
##
## Estimated Lambda: 1
## With fudge factor, no transformation is applied
##
##
## $CAR_AGE
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##   1.00    2.00    9.00    9.33   13.00   29.00
##
## Largest/Smallest: 29
## Sample Skewness: 0.282
##
## Estimated Lambda: 0.5
##
##
## $Y0J

```

```
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      1.00   10.00   12.00   11.55   14.00   24.00
##
## Largest/Smallest: 24
## Sample Skewness: -1.26
##
## Estimated Lambda: 1.6
##
##
## $INCOME
## Box-Cox Transformation
##
## 7650 data points used to estimate Lambda
##
## Input data summary:
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##       1   29701   54029   61450   83302   367031
##
## Largest/Smallest: 367000
## Sample Skewness: 1.25
##
## Estimated Lambda: 0.4
```

Build Models

Multiple Linear Regression Models

Model 1: Mean full model

Building the first model using all the variables...

```
##
## Call:
## lm(formula = TARGET_AMT ~ ., data = df_lm)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8934  -3199  -1537    493   99141
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.025e+03  2.095e+03   1.444 0.148807
## KIDSDRIV     -1.167e+02  3.378e+02  -0.346 0.729696
## AGE           1.272e+01  2.236e+01   0.569 0.569617
## HOMEKIDS      2.288e+02  2.169e+02   1.055 0.291669
```

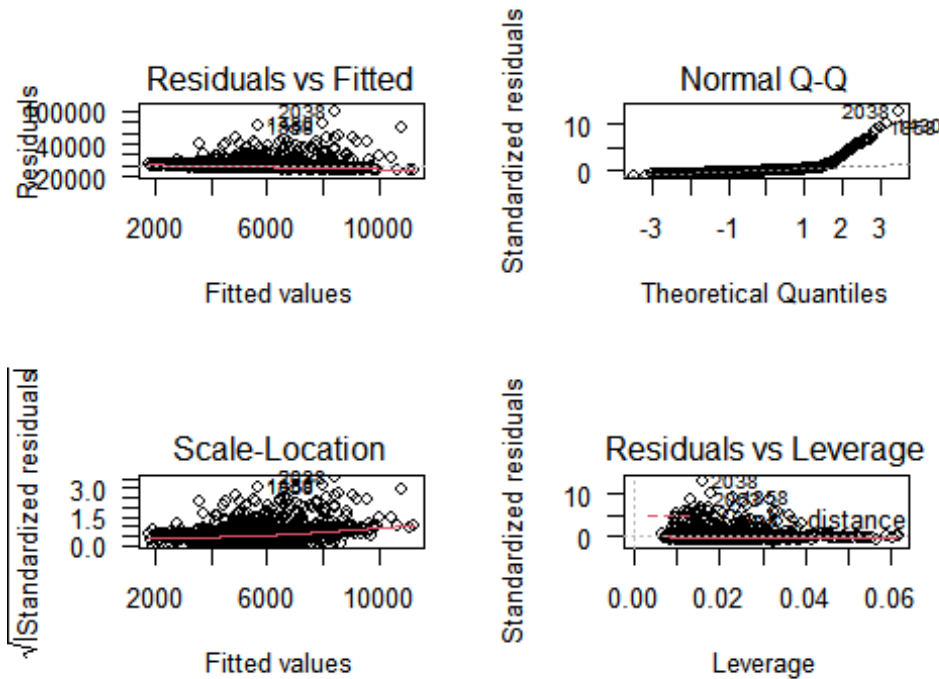
```

## YOJ                2.451e+01  5.173e+01  0.474 0.635662
## PARENT1Yes         1.406e+02  6.166e+02  0.228 0.819623
## TRAVTIME           3.617e+00  1.169e+01  0.309 0.757114
## CAR_USEPrivate     -5.479e+02  5.487e+02 -0.999 0.318094
## TIF                -3.882e+01  4.482e+01 -0.866 0.386583
## RED_CARyes         -2.473e+02  5.214e+02 -0.474 0.635302
## CLM_FREQ          -1.463e+02  1.662e+02 -0.880 0.378836
## REVOKEDYes        -1.192e+03  5.427e+02 -2.197 0.028162 *
## MVR_PTS            1.088e+02  7.231e+01  1.504 0.132743
## CAR_AGE            -9.230e+01  4.617e+01 -1.999 0.045712 *
## INCOME             -9.670e-03  7.073e-03 -1.367 0.171729
## HOME_VAL           1.671e-03  2.112e-03  0.791 0.429070
## BLUEBOOK           1.227e-01  3.196e-02  3.840 0.000127 ***
## OLDCLAIM           3.257e-02  2.364e-02  1.378 0.168373
## MSTATUSYes        -8.493e+02  5.218e+02 -1.628 0.103786
## SEXM               1.397e+03  6.880e+02  2.030 0.042451 *
## EDUCATIONBachelors 2.182e+02  6.777e+02  0.322 0.747551
## EDUCATIONHigh School -5.850e+02  5.389e+02 -1.086 0.277780
## EDUCATIONMasters    9.636e+02  1.152e+03  0.836 0.403142
## EDUCATIONPhD        2.460e+03  1.396e+03  1.762 0.078207 .
## JOBBlue Collar      6.499e+02  1.205e+03  0.539 0.589660
## JOBClerical         3.778e+02  1.263e+03  0.299 0.764818
## JOBDoctor          -2.073e+03  1.812e+03 -1.144 0.252835
## JOBHome Maker       1.686e+02  1.333e+03  0.126 0.899392
## JOBLawyer           4.172e+02  1.075e+03  0.388 0.697950
## JOBManager         -7.283e+02  1.128e+03 -0.646 0.518587
## JOBProfessional     1.389e+03  1.181e+03  1.176 0.239802
## JOBStudent          1.484e+02  1.355e+03  0.110 0.912816
## CAR_TYPEPanel Truck -5.584e+02  1.004e+03 -0.556 0.578258
## CAR_TYPEPickup      -2.292e+02  6.279e+02 -0.365 0.715141
## CAR_TYPESports Car  1.074e+03  7.867e+02  1.365 0.172386
## CAR_TYPESUV          9.386e+02  6.968e+02  1.347 0.178104
## CAR_TYPEVan         3.322e+02  8.104e+02  0.410 0.681889
## URBANICITYHighly Urban/ Urban 1.499e+02  7.962e+02  0.188 0.850687
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7817 on 1973 degrees of freedom
## (142 observations deleted due to missingness)
## Multiple R-squared:  0.03181,    Adjusted R-squared:  0.01366
## F-statistic: 1.752 on 37 and 1973 DF,  p-value: 0.003512

```

Visualization...

lm(TARGET_AMT ~ .)



Model 2: Stepwise

```
## Start: AIC=36091.11
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
## CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +
## CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
## SEX + EDUCATION + JOB + CAR_TYPE + URBANICITY
##
##           Df Sum of Sq      RSS   AIC
## - JOB       8 469943135 1.2103e+11 36083
## - CAR_TYPE   5 253319413 1.2082e+11 36085
## - URBANICITY 1  2165878 1.2057e+11 36089
## - PARENT1    1  3178268 1.2057e+11 36089
## - TRAVTIME   1  5846477 1.2057e+11 36089
## - KIDSDRIV   1  7297697 1.2057e+11 36089
## - YOJ        1 13720342 1.2058e+11 36089
## - RED_CAR    1 13749572 1.2058e+11 36089
## - AGE        1 19763912 1.2058e+11 36089
## - HOME_VAL   1 38228288 1.2060e+11 36090
## - EDUCATION  4 405041171 1.2097e+11 36090
## - TIF        1 45829844 1.2061e+11 36090
## - CLM_FREQ   1 47346961 1.2061e+11 36090
## - CAR_USE    1 60940121 1.2062e+11 36090
## - HOMEKIDS   1 67981141 1.2063e+11 36090
## - INCOME     1 114217654 1.2068e+11 36091
## - OLDCLAIM   1 116027467 1.2068e+11 36091
## <none>                1.2056e+11 36091
```

```

## - MVR_PTS      1 138223654 1.2070e+11 36091
## - MSTATUS      1 161861282 1.2073e+11 36092
## - CAR_AGE      1 244258656 1.2081e+11 36093
## - SEX          1 251910571 1.2082e+11 36093
## - REVOKED      1 294856563 1.2086e+11 36094
## - BLUEBOOK     1 901120949 1.2146e+11 36104
##
## Step:  AIC=36082.94
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
##      SEX + EDUCATION + CAR_TYPE + URBANICITY
##
##           Df Sum of Sq      RSS   AIC
## - CAR_TYPE      5 236966467 1.2127e+11 36077
## - EDUCATION      4 308229999 1.2134e+11 36080
## - URBANICITY     1   965617 1.2103e+11 36081
## - PARENT1        1   5312998 1.2104e+11 36081
## - TRAVTIME        1   8714537 1.2104e+11 36081
## - KIDSDRIV        1   8789882 1.2104e+11 36081
## - RED_CAR         1  17718790 1.2105e+11 36081
## - AGE             1  22400164 1.2106e+11 36081
## - TIF             1  35799007 1.2107e+11 36082
## - YOJ             1  43615339 1.2108e+11 36082
## - CLM_FREQ        1  46060690 1.2108e+11 36082
## - HOMEKIDS        1   50786527 1.2108e+11 36082
## - HOME_VAL        1   55459128 1.2109e+11 36082
## - OLDCLAIM        1   93964699 1.2113e+11 36082
## - CAR_USE         1 109054463 1.2114e+11 36083
## - INCOME          1 112628408 1.2115e+11 36083
## <none>                1.2103e+11 36083
## - MVR_PTS        1 164427107 1.2120e+11 36084
## - MSTATUS        1 165584306 1.2120e+11 36084
## - CAR_AGE        1 240391833 1.2127e+11 36085
## - SEX            1 248609327 1.2128e+11 36085
## - REVOKED        1 249404349 1.2128e+11 36085
## - BLUEBOOK       1 892963211 1.2193e+11 36096
##
## Step:  AIC=36076.87
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
##      SEX + EDUCATION + URBANICITY
##
##           Df Sum of Sq      RSS   AIC
## - EDUCATION      4 293028995 1.2156e+11 36074
## - URBANICITY     1   856675 1.2127e+11 36075
## - PARENT1        1   6956166 1.2128e+11 36075
## - TRAVTIME        1  10004549 1.2128e+11 36075
## - KIDSDRIV        1  15091560 1.2129e+11 36075

```

```

## - RED_CAR      1  22955445 1.2129e+11 36075
## - TIF          1  31683557 1.2130e+11 36075
## - YOJ          1  40394006 1.2131e+11 36076
## - CLM_FREQ     1  48329600 1.2132e+11 36076
## - AGE          1  49960692 1.2132e+11 36076
## - HOMEKIDS     1  55516846 1.2133e+11 36076
## - HOME_VAL     1  58416578 1.2133e+11 36076
## - CAR_USE      1  62849650 1.2133e+11 36076
## - OLDCLAIM     1  94247726 1.2136e+11 36076
## - INCOME       1 102728766 1.2137e+11 36077
## <none>                1.2127e+11 36077
## - SEX          1 126458571 1.2140e+11 36077
## - MSTATUS      1 155747969 1.2143e+11 36077
## - MVR_PTS      1 166354304 1.2144e+11 36078
## - REVOKED      1 246739099 1.2152e+11 36079
## - CAR_AGE      1 255944523 1.2153e+11 36079
## - BLUEBOOK     1 1191468027 1.2246e+11 36095
##
## Step:  AIC=36073.72
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
##      SEX + URBANICITY
##
##              Df  Sum of Sq      RSS    AIC
## - URBANICITY  1    1842097 1.2157e+11 36072
## - PARENT1     1     5746780 1.2157e+11 36072
## - TRAVTIME    1    11362492 1.2158e+11 36072
## - KIDSDRIV    1    12710883 1.2158e+11 36072
## - CAR_USE     1    18809855 1.2158e+11 36072
## - RED_CAR     1    19154959 1.2158e+11 36072
## - YOJ         1    19249674 1.2158e+11 36072
## - INCOME      1    30009939 1.2159e+11 36072
## - TIF         1    33478524 1.2160e+11 36072
## - CLM_FREQ    1    43826138 1.2161e+11 36072
## - HOMEKIDS    1    56286122 1.2162e+11 36073
## - HOME_VAL    1    64738692 1.2163e+11 36073
## - AGE         1    68188686 1.2163e+11 36073
## - OLDCLAIM    1    89443723 1.2165e+11 36073
## - CAR_AGE     1   106343391 1.2167e+11 36073
## <none>                1.2156e+11 36074
## - SEX         1   149903559 1.2171e+11 36074
## - MSTATUS     1   182520593 1.2175e+11 36075
## - MVR_PTS     1   183291459 1.2175e+11 36075
## - REVOKED     1   245461948 1.2181e+11 36076
## - BLUEBOOK    1 1297078713 1.2286e+11 36093
##
## Step:  AIC=36071.75
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +

```

```

##      CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
##      SEX
##
##      Df  Sum of Sq      RSS    AIC
## - PARENT1  1    5693754 1.2157e+11 36070
## - TRAVTIME  1   10766761 1.2158e+11 36070
## - KIDSDRIV  1   12739425 1.2158e+11 36070
## - CAR_USE   1   18874900 1.2158e+11 36070
## - RED_CAR   1   19054285 1.2158e+11 36070
## - YOJ       1   19438380 1.2158e+11 36070
## - INCOME    1   29169043 1.2159e+11 36070
## - TIF       1   33102736 1.2160e+11 36070
## - CLM_FREQ  1   42851795 1.2161e+11 36070
## - HOMEKIDS  1   55811739 1.2162e+11 36071
## - HOME_VAL  1   64662586 1.2163e+11 36071
## - AGE       1   67542493 1.2163e+11 36071
## - OLDCLAIM  1   88398537 1.2165e+11 36071
## - CAR_AGE   1  105826822 1.2167e+11 36072
## <none>             1.2157e+11 36072
## - SEX       1  148981133 1.2171e+11 36072
## - MSTATUS   1  181291557 1.2175e+11 36073
## - MVR_PTS   1  185108458 1.2175e+11 36073
## - REVOKED   1  243957432 1.2181e+11 36074
## - BLUEBOOK  1 1302879099 1.2287e+11 36091
##
## Step:  AIC=36069.85
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + TRAVTIME + CAR_USE +
##      TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE +
##      INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##      Df  Sum of Sq      RSS    AIC
## - TRAVTIME  1   10583155 1.2158e+11 36068
## - KIDSDRIV  1   12026440 1.2158e+11 36068
## - CAR_USE   1   18142211 1.2159e+11 36068
## - RED_CAR   1   19040049 1.2159e+11 36068
## - YOJ       1   19348550 1.2159e+11 36068
## - INCOME    1   29818379 1.2160e+11 36068
## - TIF       1   33585592 1.2160e+11 36068
## - CLM_FREQ  1   43050354 1.2161e+11 36069
## - AGE       1   64223029 1.2164e+11 36069
## - HOME_VAL  1   65317881 1.2164e+11 36069
## - OLDCLAIM  1   88804959 1.2166e+11 36069
## - HOMEKIDS  1   90944370 1.2166e+11 36069
## - CAR_AGE   1  105327287 1.2168e+11 36070
## <none>             1.2157e+11 36070
## - SEX       1  147371685 1.2172e+11 36070
## - MVR_PTS   1  187176286 1.2176e+11 36071
## - REVOKED   1  245828287 1.2182e+11 36072
## - MSTATUS   1  312899566 1.2188e+11 36073
## - BLUEBOOK  1 1303707132 1.2287e+11 36089

```

```

##
## Step: AIC=36068.02
## TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + CAR_USE + TIF +
## RED_CAR + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE + INCOME +
## HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##          Df Sum of Sq      RSS      AIC
## - KIDSDRIV 1  12717116 1.2159e+11 36066
## - CAR_USE   1  15900397 1.2160e+11 36066
## - RED_CAR   1  18818620 1.2160e+11 36066
## - YOJ       1  19663593 1.2160e+11 36066
## - INCOME    1  28657692 1.2161e+11 36066
## - TIF       1  34495010 1.2162e+11 36067
## - CLM_FREQ  1  41430585 1.2162e+11 36067
## - HOME_VAL  1  64787510 1.2165e+11 36067
## - AGE       1  65838971 1.2165e+11 36067
## - OLDCLAIM  1  86104148 1.2167e+11 36067
## - HOMEKIDS  1  90316554 1.2167e+11 36068
## - CAR_AGE   1 103997897 1.2169e+11 36068
## <none>      1.2158e+11 36068
## - SEX       1 148338358 1.2173e+11 36068
## - MVR_PTS   1 188617256 1.2177e+11 36069
## - REVOKED   1 243714639 1.2183e+11 36070
## - MSTATUS   1 309146076 1.2189e+11 36071
## - BLUEBOOK  11309357711 1.2289e+11 36088
##
## Step: AIC=36066.23
## TARGET_AMT ~ AGE + HOMEKIDS + YOJ + CAR_USE + TIF + RED_CAR +
## CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE + INCOME + HOME_VAL +
## BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##          Df Sum of Sq      RSS      AIC
## - CAR_USE   1  17681725 1.2161e+11 36065
## - RED_CAR   1  18235722 1.2161e+11 36065
## - YOJ       1  20575176 1.2162e+11 36065
## - INCOME    1  31431047 1.2163e+11 36065
## - TIF       1  34132616 1.2163e+11 36065
## - CLM_FREQ  1  42623742 1.2164e+11 36065
## - AGE       1  56669663 1.2165e+11 36065
## - HOME_VAL  1  64015480 1.2166e+11 36065
## - HOMEKIDS  1  79558260 1.2167e+11 36066
## - OLDCLAIM  1  87579344 1.2168e+11 36066
## - CAR_AGE   1 102810967 1.2170e+11 36066
## <none>      1.2159e+11 36066
## - SEX       1 145934102 1.2174e+11 36067
## - MVR_PTS   1 188017744 1.2178e+11 36067
## - REVOKED   1 250237302 1.2184e+11 36068
## - MSTATUS   1 307680961 1.2190e+11 36069
## - BLUEBOOK  11300629748 1.2290e+11 36086
##

```

```
## Step: AIC=36064.53
```

```
## TARGET_AMT ~ AGE + HOMEKIDS + YOJ + TIF + RED_CAR + CLM_FREQ +  
## REVOKED + MVR_PTS + CAR_AGE + INCOME + HOME_VAL + BLUEBOOK +  
## OLDCLAIM + MSTATUS + SEX
```

```
##  
##          Df Sum of Sq      RSS   AIC  
## - RED_CAR  1  18131672 1.2163e+11 36063  
## - YOJ      1  18515875 1.2163e+11 36063  
## - INCOME   1  28249484 1.2164e+11 36063  
## - TIF      1  30953795 1.2164e+11 36063  
## - CLM_FREQ 1  40604710 1.2165e+11 36063  
## - AGE      1  51155104 1.2166e+11 36063  
## - HOME_VAL 1  63614498 1.2168e+11 36064  
## - HOMEKIDS 1  75609382 1.2169e+11 36064  
## - OLDCLAIM 1  85221119 1.2170e+11 36064  
## - CAR_AGE  1 113267333 1.2173e+11 36064  
## <none>          1.2161e+11 36065  
## - SEX       1 185303260 1.2180e+11 36066  
## - MVR_PTS   1 197769471 1.2181e+11 36066  
## - REVOKED   1 241532321 1.2185e+11 36067  
## - MSTATUS   1 301138160 1.2191e+11 36068  
## - BLUEBOOK 1 1430529112 1.2304e+11 36086  
##
```

```
## Step: AIC=36062.83
```

```
## TARGET_AMT ~ AGE + HOMEKIDS + YOJ + TIF + CLM_FREQ + REVOKED +  
## MVR_PTS + CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM +  
## MSTATUS + SEX
```

```
##  
##          Df Sum of Sq      RSS   AIC  
## - YOJ      1  20029097 1.2165e+11 36061  
## - INCOME   1  28847085 1.2166e+11 36061  
## - TIF      1  30758533 1.2166e+11 36061  
## - CLM_FREQ 1  39971637 1.2167e+11 36061  
## - AGE      1  53552250 1.2168e+11 36062  
## - HOME_VAL 1  64187132 1.2169e+11 36062  
## - HOMEKIDS 1  75837247 1.2171e+11 36062  
## - OLDCLAIM 1  82877633 1.2171e+11 36062  
## - CAR_AGE  1 115339012 1.2175e+11 36063  
## <none>          1.2163e+11 36063  
## - MVR_PTS   1 194860176 1.2183e+11 36064  
## - SEX       1 208618206 1.2184e+11 36064  
## - REVOKED   1 238490978 1.2187e+11 36065  
## - MSTATUS   1 302094286 1.2193e+11 36066  
## - BLUEBOOK 1 1428900494 1.2306e+11 36084  
##
```

```
## Step: AIC=36061.16
```

```
## TARGET_AMT ~ AGE + HOMEKIDS + TIF + CLM_FREQ + REVOKED + MVR_PTS +  
## CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +  
## SEX
```

```
##
```

```

##           Df Sum of Sq      RSS   AIC
## - INCOME    1  19115469 1.2167e+11 36059
## - TIF        1  29924316 1.2168e+11 36060
## - CLM_FREQ   1  41351255 1.2169e+11 36060
## - AGE        1  61834474 1.2171e+11 36060
## - HOME_VAL   1  73217972 1.2172e+11 36060
## - OLDCLAIM   1  83923138 1.2173e+11 36061
## - HOMEKIDS   1  89680817 1.2174e+11 36061
## <none>                1.2165e+11 36061
## - CAR_AGE    1  127930245 1.2178e+11 36061
## - MVR_PTS    1  193196473 1.2184e+11 36062
## - SEX        1  213453598 1.2186e+11 36063
## - REVOKED    1  234184361 1.2188e+11 36063
## - MSTATUS    1  294639752 1.2195e+11 36064
## - BLUEBOOK   1 1432597492 1.2308e+11 36083
##
## Step:   AIC=36059.47
## TARGET_AMT ~ AGE + HOMEKIDS + TIF + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##           Df Sum of Sq      RSS   AIC
## - TIF        1  28608666 1.2170e+11 36058
## - CLM_FREQ   1  41330543 1.2171e+11 36058
## - HOME_VAL   1  54237900 1.2172e+11 36058
## - AGE        1  59356258 1.2173e+11 36058
## - OLDCLAIM   1  84070404 1.2175e+11 36059
## - HOMEKIDS   1  90047387 1.2176e+11 36059
## <none>                1.2167e+11 36059
## - CAR_AGE    1  180542974 1.2185e+11 36060
## - MVR_PTS    1  193870373 1.2186e+11 36061
## - SEX        1  204643732 1.2187e+11 36061
## - REVOKED    1  235054611 1.2190e+11 36061
## - MSTATUS    1  276135658 1.2195e+11 36062
## - BLUEBOOK   1 1467753477 1.2314e+11 36082
##
## Step:   AIC=36057.95
## TARGET_AMT ~ AGE + HOMEKIDS + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##           Df Sum of Sq      RSS   AIC
## - CLM_FREQ   1  44067998 1.2174e+11 36057
## - HOME_VAL   1  55524616 1.2175e+11 36057
## - AGE        1  59659792 1.2176e+11 36057
## - OLDCLAIM   1  84734409 1.2178e+11 36057
## - HOMEKIDS   1  90426149 1.2179e+11 36057
## <none>                1.2170e+11 36058
## - CAR_AGE    1  177870390 1.2188e+11 36059
## - MVR_PTS    1  200822202 1.2190e+11 36059
## - SEX        1  205447327 1.2190e+11 36059
## - REVOKED    1  234411359 1.2193e+11 36060

```

```

## - MSTATUS      1  270025675 1.2197e+11 36060
## - BLUEBOOK     1 1466813096 1.2316e+11 36080
##
## Step:  AIC=36056.67
## TARGET_AMT ~ AGE + HOMEKIDS + REVOKED + MVR_PTS + CAR_AGE + HOME_VAL +
##      BLUEBOOK + OLDCLAIM + MSTATUS + SEX
##
##              Df  Sum of Sq      RSS    AIC
## - OLDCLAIM    1   48636304 1.2179e+11 36055
## - HOME_VAL    1   55667628 1.2180e+11 36056
## - AGE         1   58174981 1.2180e+11 36056
## - HOMEKIDS    1   91125091 1.2183e+11 36056
## <none>                1.2174e+11 36057
## - MVR_PTS     1  167280814 1.2191e+11 36057
## - CAR_AGE     1  188399330 1.2193e+11 36058
## - REVOKED     1  198124852 1.2194e+11 36058
## - SEX         1  203997706 1.2195e+11 36058
## - MSTATUS     1  269490060 1.2201e+11 36059
## - BLUEBOOK    1 1471542457 1.2321e+11 36079
##
## Step:  AIC=36055.48
## TARGET_AMT ~ AGE + HOMEKIDS + REVOKED + MVR_PTS + CAR_AGE + HOME_VAL +
##      BLUEBOOK + MSTATUS + SEX
##
##              Df  Sum of Sq      RSS    AIC
## - HOME_VAL    1   58740059 1.2185e+11 36054
## - AGE         1   59261910 1.2185e+11 36054
## - HOMEKIDS    1   88684255 1.2188e+11 36055
## <none>                1.2179e+11 36055
## - REVOKED     1  150394814 1.2194e+11 36056
## - CAR_AGE     1  184351389 1.2198e+11 36057
## - SEX         1  197901853 1.2199e+11 36057
## - MVR_PTS     1  212421634 1.2200e+11 36057
## - MSTATUS     1  267708130 1.2206e+11 36058
## - BLUEBOOK    1 1452132145 1.2324e+11 36077
##
## Step:  AIC=36054.45
## TARGET_AMT ~ AGE + HOMEKIDS + REVOKED + MVR_PTS + CAR_AGE + BLUEBOOK +
##      MSTATUS + SEX
##
##              Df  Sum of Sq      RSS    AIC
## - AGE         1   71370786 1.2192e+11 36054
## - HOMEKIDS    1   83747039 1.2193e+11 36054
## <none>                1.2185e+11 36054
## - CAR_AGE     1  154513388 1.2200e+11 36055
## - REVOKED     1  155653552 1.2201e+11 36055
## - MVR_PTS     1  203955706 1.2205e+11 36056
## - SEX         1  208593868 1.2206e+11 36056
## - MSTATUS     1  209106117 1.2206e+11 36056
## - BLUEBOOK    1 1645147417 1.2349e+11 36079

```



```

##
## Step: AIC=36053.62
## TARGET_AMT ~ HOMEKIDS + REVOKED + MVR_PTS + CAR_AGE + BLUEBOOK +
##   MSTATUS + SEX
##
##           Df Sum of Sq      RSS   AIC
## - HOMEKIDS  1  38484030 1.2196e+11 36052
## <none>                                1.2192e+11 36054
## - CAR_AGE   1  131131760 1.2205e+11 36054
## - REVOKED   1  148060934 1.2207e+11 36054
## - MSTATUS   1  175086472 1.2210e+11 36055
## - SEX       1  202608661 1.2212e+11 36055
## - MVR_PTS   1  202828770 1.2212e+11 36055
## - BLUEBOOK  1 1746551592 1.2367e+11 36080
##
## Step: AIC=36052.26
## TARGET_AMT ~ REVOKED + MVR_PTS + CAR_AGE + BLUEBOOK + MSTATUS +
##   SEX
##
##           Df Sum of Sq      RSS   AIC
## <none>                                1.2196e+11 36052
## - CAR_AGE   1  140440225 1.2210e+11 36053
## - REVOKED   1  141378251 1.2210e+11 36053
## - MSTATUS   1  164261705 1.2212e+11 36053
## - SEX       1  188329377 1.2215e+11 36053
## - MVR_PTS   1  210544699 1.2217e+11 36054
## - BLUEBOOK  1 1724209476 1.2368e+11 36078
##
## Call:
## lm(formula = TARGET_AMT ~ REVOKED + MVR_PTS + CAR_AGE + BLUEBOOK +
##   MSTATUS + SEX, data = df_lm)
##
## Residuals:
##   Min       1Q   Median       3Q      Max
## -8263  -3162  -1574    325  100465
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4281.91344   503.98221    8.496 < 2e-16 ***
## REVOKEDYes  -656.16299   430.50600   -1.524  0.1276
## MVR_PTS      125.73815    67.60109    1.860  0.0630 .
## CAR_AGE     -48.79606    32.12165   -1.519  0.1289
## BLUEBOOK     0.11376     0.02137    5.323 1.14e-07 ***
## MSTATUSYes  -573.12750   348.85252   -1.643  0.1006
## SEXM        618.26245   351.45734    1.759  0.0787 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7801 on 2004 degrees of freedom

```

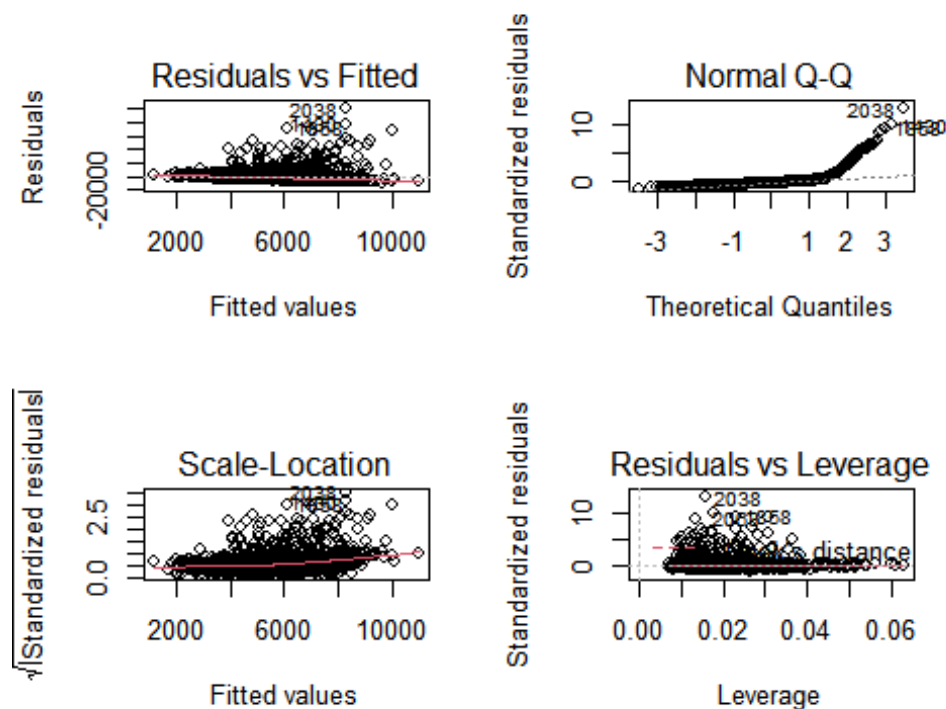
```
## (142 observations deleted due to missingness)
## Multiple R-squared:  0.02061,    Adjusted R-squared:  0.01767
## F-statistic: 7.027 on 6 and 2004 DF,  p-value: 2.043e-07
```

Model 3: BoxCox imputation

```
##
## Call:
## lm(formula = TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + I(YOJ^1.6) +
##     PARENT1 + I(TRAVERSE^0.7) + CAR_USE + I(TIF^0.2) + RED_CAR +
##     CLM_FREQ + REVOKED + MVR_PTS + I(CAR_AGE^0.5) + I(INCOME^0.4) +
##     HOME_VAL + I(BLUEBOOK^0.5) + OLDCLAIM + MSTATUS + SEX + EDUCATION +
##     JOB + CAR_TYPE + URBANICITY, data = df_lm)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8792  -3250  -1499    530   99271
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      1.950e+03  2.476e+03   0.788   0.4310
## KIDSDRIV         -1.437e+02  3.390e+02  -0.424   0.6717
## AGE              1.214e+01  2.265e+01   0.536   0.5921
## HOMEKIDS         2.291e+02  2.205e+02   1.039   0.2989
## I(YOJ^1.6)       1.451e+00  9.802e+00   0.148   0.8823
## PARENT1Yes       1.700e+02  6.172e+02   0.276   0.7830
## I(TRAVERSE^0.7)  1.111e+01  4.701e+01   0.236   0.8132
## CAR_USEPrivate   -4.413e+02  5.489e+02  -0.804   0.4214
## I(TIF^0.2)       -4.381e+02  7.321e+02  -0.598   0.5497
## RED_CARyes       -2.595e+02  5.217e+02  -0.497   0.6190
## CLM_FREQ        -1.520e+02  1.663e+02  -0.914   0.3610
## REVOKEDYes      -1.167e+03  5.430e+02  -2.149   0.0317 *
## MVR_PTS          1.135e+02  7.243e+01   1.566   0.1174
## I(CAR_AGE^0.5)   -3.511e+02  2.122e+02  -1.655   0.0982 .
## I(INCOME^0.4)    -2.820e+00  1.120e+01  -0.252   0.8012
## HOME_VAL         1.132e-03  2.066e-03   0.548   0.5837
## I(BLUEBOOK^0.5)  2.917e+01  7.245e+00   4.026 5.89e-05 ***
## OLDCLAIM         3.307e-02  2.366e-02   1.397   0.1624
## MSTATUSYes      -7.339e+02  5.226e+02  -1.404   0.1604
## SEXM             1.386e+03  6.789e+02   2.042   0.0413 *
## EDUCATIONBachelors  1.807e+01  6.843e+02   0.026   0.9789
## EDUCATIONHigh School -6.488e+02  5.425e+02  -1.196   0.2319
## EDUCATIONMasters   4.820e+02  1.130e+03   0.426   0.6699
## EDUCATIONPhD       1.746e+03  1.349e+03   1.294   0.1957
## JOBBlue Collar     7.549e+02  1.200e+03   0.629   0.5294
## JOBClerical        5.742e+02  1.255e+03   0.457   0.6474
## JOBDirector       -2.158e+03  1.814e+03  -1.189   0.2345
## JOBHome Maker      4.477e+02  1.391e+03   0.322   0.7477
## JOBLawyer          4.032e+02  1.074e+03   0.375   0.7075
## JOBManager        -7.432e+02  1.129e+03  -0.659   0.5103
## JOBProfessional    1.439e+03  1.179e+03   1.221   0.2223
```

```
## JOBStudent          4.161e+02  1.411e+03  0.295  0.7681
## CAR_TYPEPanel Truck -3.759e+02  9.639e+02 -0.390  0.6966
## CAR_TYPEPickup      -1.729e+02  6.290e+02 -0.275  0.7835
## CAR_TYPESports Car   1.111e+03  7.840e+02  1.417  0.1567
## CAR_TYPESUV          9.021e+02  6.886e+02  1.310  0.1903
## CAR_TYPEVan          2.934e+02  8.101e+02  0.362  0.7172
## URBANICITYHighly Urban/ Urban 9.856e+01  7.962e+02  0.124  0.9015
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 7821 on 1972 degrees of freedom
## (143 observations deleted due to missingness)
## Multiple R-squared:  0.03108,    Adjusted R-squared:  0.0129
## F-statistic:  1.71 on 37 and 1972 DF,  p-value: 0.00508
```

Residual analysis...



Logistic Regression Models

Model 1: Mean full model

Let build the first full model for logistiv regression...

```
##
## Call:
## glm(formula = TARGET_FLAG ~ ., family = "binomial", data = df_rm)
##
```

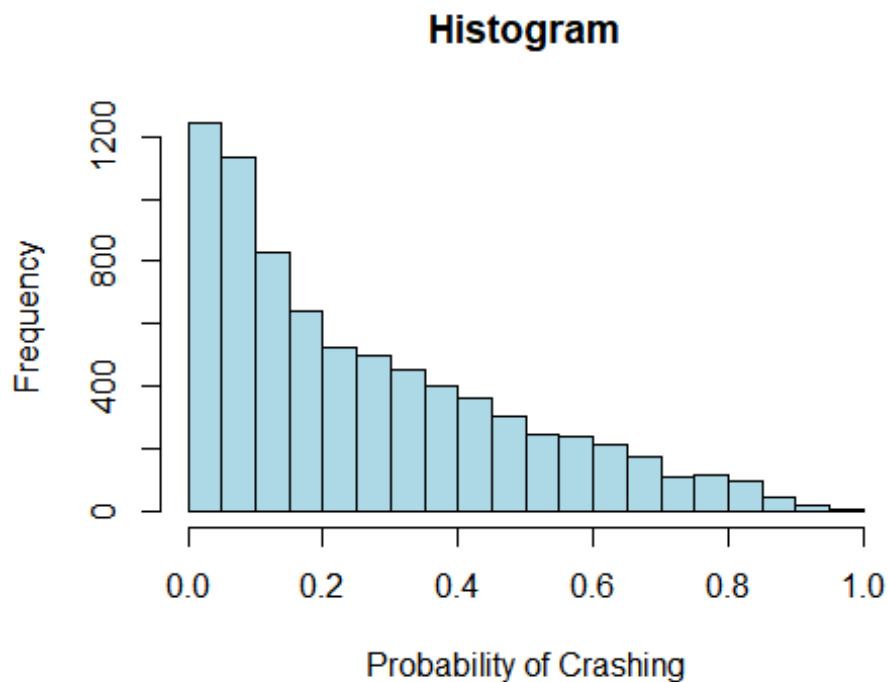
```

## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5654  -0.7168  -0.3972   0.6148   3.1535
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -2.994e+00  3.520e-01  -8.506  < 2e-16 ***
## KIDSDRIV       3.455e-01  6.381e-02   5.414  6.16e-08 ***
## AGE           3.193e-04  4.145e-03   0.077  0.938603
## HOMEKIDS       6.304e-02  3.822e-02   1.649  0.099118 .
## YOJ          -1.136e-02  8.872e-03  -1.280  0.200450
## PARENT1Yes     3.853e-01  1.131e-01   3.406  0.000659 ***
## TRAVTIME       1.492e-02  1.953e-03   7.641  2.16e-14 ***
## CAR_USEPrivate -7.534e-01  9.459e-02  -7.965  1.65e-15 ***
## TIF           -5.512e-02  7.590e-03  -7.261  3.83e-13 ***
## RED_CARYes     1.800e-03  8.927e-02   0.020  0.983911
## CLM_FREQ       2.025e-01  2.938e-02   6.892  5.51e-12 ***
## REVOKEDYes     8.804e-01  9.439e-02   9.327  < 2e-16 ***
## MVR_PTS        1.142e-01  1.405e-02   8.132  4.22e-16 ***
## CAR_AGE       -4.559e-03  7.797e-03  -0.585  0.558745
## INCOME        -3.107e-06  1.116e-06  -2.783  0.005386 **
## HOME_VAL      -1.290e-06  3.525e-07  -3.658  0.000254 ***
## BLUEBOOK      -2.147e-05  5.439e-06  -3.949  7.86e-05 ***
## OLDCLAIM      -1.288e-05  4.011e-06  -3.211  0.001324 **
## MSTATUSYes    -4.902e-01  8.656e-02  -5.663  1.49e-08 ***
## SEXM           8.549e-02  1.154e-01   0.741  0.458957
## EDUCATIONBachelors -3.626e-01  1.198e-01  -3.027  0.002470 **
## EDUCATIONHigh School  3.528e-02  9.786e-02   0.361  0.718470
## EDUCATIONMasters -2.129e-01  1.865e-01  -1.141  0.253692
## EDUCATIONPhD    -1.214e-01  2.231e-01  -0.544  0.586367
## JOBBBlue Collar  3.198e-01  1.922e-01   1.664  0.096189 .
## JOBClerical     4.362e-01  2.034e-01   2.144  0.031997 *
## JOBDoctor      -3.528e-01  2.716e-01  -1.299  0.193892
## JOBHome Maker   2.764e-01  2.176e-01   1.270  0.204072
## JOBLawyer       1.035e-01  1.749e-01   0.592  0.554008
## JOBManager     -6.151e-01  1.784e-01  -3.449  0.000563 ***
## JOBProfessional 1.439e-01  1.842e-01   0.781  0.434798
## JOBStudent      1.991e-01  2.226e-01   0.894  0.371160
## CAR_TYPEPanel Truck  6.109e-01  1.669e-01   3.661  0.000251 ***
## CAR_TYPEPickup   5.267e-01  1.041e-01   5.058  4.24e-07 ***
## CAR_TYPESports Car  9.979e-01  1.341e-01   7.440  1.01e-13 ***
## CAR_TYPESUV       7.535e-01  1.147e-01   6.570  5.03e-11 ***
## CAR_TYPEVan       5.872e-01  1.304e-01   4.503  6.71e-06 ***
## URBANICITYHighly Urban/ Urban 2.373e+00  1.164e-01  20.385  < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 8814.1  on 7650  degrees of freedom

```

```
## Residual deviance: 6831.7 on 7613 degrees of freedom
## (510 observations deleted due to missingness)
## AIC: 6907.7
##
## Number of Fisher Scoring iterations: 5
```

Confusion matrix and the curve...

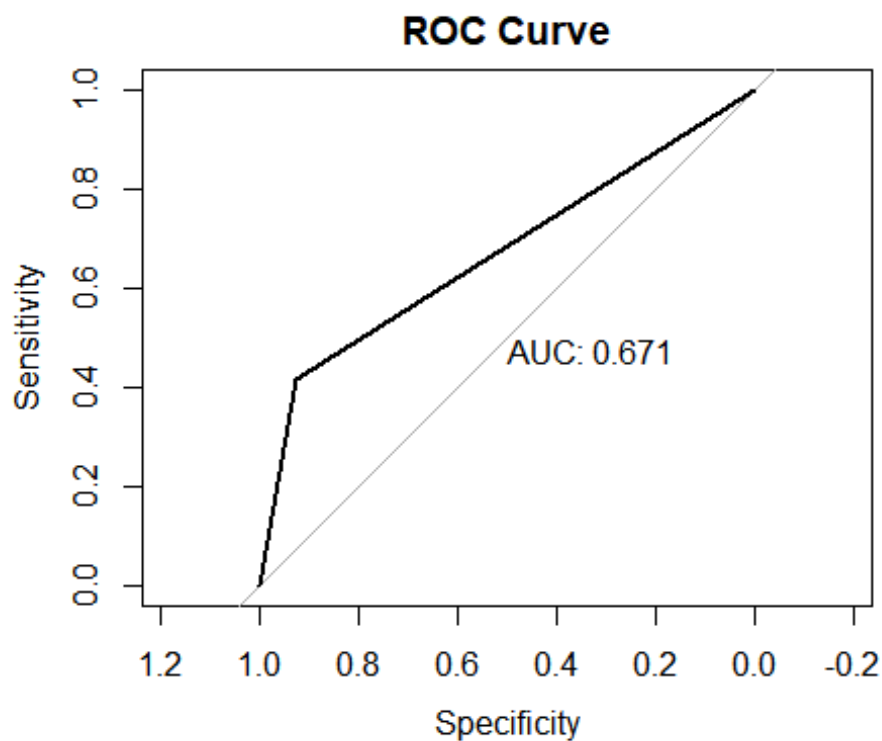


```
probabilities <- predict(rm_model1, df_rm, type = "response")
predicted1.classes <- ifelse(probabilities > 0.5, 1, 0)
df_rm$pred1.class <- predicted1.classes
table("Predictions" = df_rm$pred1.class, "Actual" = df_rm$TARGET_FLAG)
```

```
##           Actual
## Predictions  0    1
##           0 5216 1172
##           1  424  839
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction  0    1
##           0 5216 1172
##           1  424  839
##
##           Accuracy : 0.7914
##           95% CI : (0.7821, 0.8005)
```

```
##      No Information Rate : 0.7372
##      P-Value [Acc > NIR] : < 2.2e-16
##
##      Kappa : 0.3885
##
##      McNemar's Test P-Value : < 2.2e-16
##
##      Sensitivity : 0.9248
##      Specificity : 0.4172
##      Pos Pred Value : 0.8165
##      Neg Pred Value : 0.6643
##      Prevalence : 0.7372
##      Detection Rate : 0.6817
##      Detection Prevalence : 0.8349
##      Balanced Accuracy : 0.6710
##
##      'Positive' Class : 0
##
```

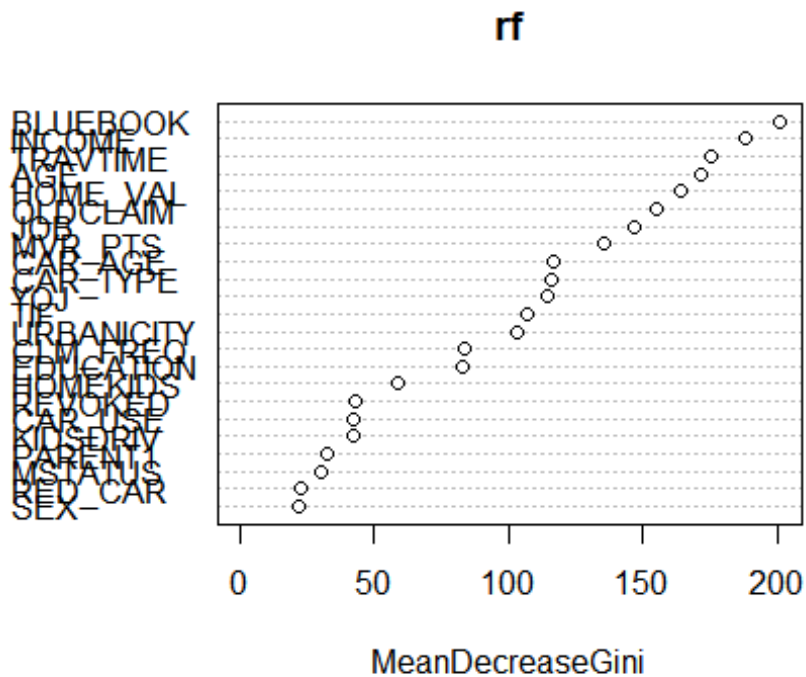


Model 2: Random forest

Logistic regression with Random forest...

```
##
## Call:
## randomForest(formula = factor(TARGET_FLAG) ~ ., data = rf_train,
## na.action = na.omit)
```

```
##           Type of random forest: classification
##           Number of trees: 500
## No. of variables tried at each split: 4
##
##           OOB estimate of  error rate: 21.16%
## Confusion matrix:
##           0   1 class.error
## 0 4233 284   0.06287359
## 1 1014 604   0.62669963
```



Confusion matrix...

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1062  262
##           1   61  131
##
##           Accuracy : 0.7869
##           95% CI : (0.7655, 0.8073)
##           No Information Rate : 0.7408
##           P-Value [Acc > NIR] : 1.63e-05
##
##           Kappa : 0.3346
##
##           Mcnemar's Test P-Value : < 2.2e-16
```

```
##
##          Sensitivity : 0.33333
##          Specificity : 0.94568
##          Pos Pred Value : 0.68229
##          Neg Pred Value : 0.80211
##          Prevalence : 0.25923
##          Detection Rate : 0.08641
##          Detection Prevalence : 0.12665
##          Balanced Accuracy : 0.63951
##
##          'Positive' Class : 1
##
```

Model 3: Stepwise

The stepwise model...

```
## Start:  AIC=5567.73
## TARGET_FLAG ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + RED_CAR + CLM_FREQ + REVOKED + MVR_PTS +
##      CAR_AGE + INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS +
##      SEX + EDUCATION + JOB + CAR_TYPE + URBANICITY
##
##          Df Deviance    AIC
## - RED_CAR      1   5491.8 5565.8
## - SEX          1   5491.8 5565.8
## - AGE          1   5491.9 5565.9
## - HOMEKIDS     1   5492.0 5566.0
## - YOJ          1   5492.4 5566.4
## - CAR_AGE      1   5492.5 5566.5
## <none>         1   5491.7 5567.7
## - OLDCLAIM     1   5499.8 5573.8
## - EDUCATION    4   5508.6 5576.6
## - BLUEBOOK     1   5503.9 5577.9
## - HOME_VAL     1   5504.0 5578.0
## - MSTATUS      1   5505.6 5579.6
## - PARENT1      1   5506.5 5580.5
## - INCOME       1   5508.4 5582.4
## - KIDSDRIV     1   5516.5 5590.5
## - CLM_FREQ     1   5522.5 5596.5
## - JOB          8   5541.5 5601.5
## - TIF          1   5534.4 5608.4
## - CAR_TYPE     5   5544.2 5610.2
## - TRAVTIME     1   5539.6 5613.6
## - CAR_USE      1   5547.7 5621.7
## - MVR_PTS      1   5553.4 5627.4
## - REVOKED      1   5564.9 5638.9
## - URBANICITY   1   5966.0 6040.0
##
## Step:  AIC=5565.75
```



```
## TARGET_FLAG ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
## CAR_USE + TIF + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE +
## INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + SEX +
## EDUCATION + JOB + CAR_TYPE + URBANICITY
##
```

	Df	Deviance	AIC
## - SEX	1	5491.8	5563.8
## - AGE	1	5491.9	5563.9
## - HOMEKIDS	1	5492.1	5564.1
## - YOJ	1	5492.4	5564.4
## - CAR_AGE	1	5492.5	5564.5
## <none>		5491.8	5565.8
## - OLDCLAIM	1	5499.8	5571.8
## - EDUCATION	4	5508.7	5574.7
## - BLUEBOOK	1	5503.9	5575.9
## - HOME_VAL	1	5504.0	5576.0
## - MSTATUS	1	5505.6	5577.6
## - PARENT1	1	5506.6	5578.6
## - INCOME	1	5508.4	5580.4
## - KIDSDRIV	1	5516.6	5588.6
## - CLM_FREQ	1	5522.5	5594.5
## - JOB	8	5541.6	5599.6
## - TIF	1	5534.4	5606.4
## - CAR_TYPE	5	5544.3	5608.3
## - TRAVTIME	1	5539.6	5611.6
## - CAR_USE	1	5547.8	5619.8
## - MVR_PTS	1	5553.4	5625.4
## - REVOKED	1	5565.0	5637.0
## - URBANICITY	1	5966.0	6038.0

```
##
```

```
## Step: AIC=5563.77
```

```
## TARGET_FLAG ~ KIDSDRIV + AGE + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
## CAR_USE + TIF + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE +
## INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + EDUCATION +
## JOB + CAR_TYPE + URBANICITY
##
```

	Df	Deviance	AIC
## - AGE	1	5491.9	5561.9
## - HOMEKIDS	1	5492.1	5562.1
## - YOJ	1	5492.4	5562.4
## - CAR_AGE	1	5492.5	5562.5
## <none>		5491.8	5563.8
## - OLDCLAIM	1	5499.8	5569.8
## - EDUCATION	4	5508.7	5572.7
## - HOME_VAL	1	5504.0	5574.0
## - MSTATUS	1	5505.6	5575.6
## - PARENT1	1	5506.6	5576.6
## - BLUEBOOK	1	5507.0	5577.0
## - INCOME	1	5508.5	5578.5
## - KIDSDRIV	1	5516.6	5586.6

```

## - CLM_FREQ      1    5522.5 5592.5
## - JOB           8    5541.6 5597.6
## - TIF           1    5534.4 5604.4
## - TRAVTIME      1    5539.7 5609.7
## - CAR_USE       1    5547.8 5617.8
## - CAR_TYPE      5    5557.8 5619.8
## - MVR_PTS       1    5553.4 5623.4
## - REVOKED       1    5565.0 5635.0
## - URBANICITY    1    5966.0 6036.0
##
## Step:  AIC=5561.91
## TARGET_FLAG ~ KIDSDRIV + HOMEKIDS + YOJ + PARENT1 + TRAVTIME +
##      CAR_USE + TIF + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE +
##      INCOME + HOME_VAL + BLUEBOOK + OLDCLAIM + MSTATUS + EDUCATION +
##      JOB + CAR_TYPE + URBANICITY
##
##           Df Deviance    AIC
## - HOMEKIDS      1    5492.5 5560.5
## - CAR_AGE       1    5492.7 5560.7
## - YOJ           1    5492.7 5560.7
## <none>          5491.9 5561.9
## - OLDCLAIM      1    5499.9 5567.9
## - EDUCATION     4    5508.8 5570.8
## - HOME_VAL      1    5504.4 5572.4
## - MSTATUS       1    5505.8 5573.8
## - PARENT1       1    5507.1 5575.1
## - BLUEBOOK      1    5507.6 5575.6
## - INCOME        1    5508.6 5576.6
## - KIDSDRIV      1    5516.8 5584.8
## - CLM_FREQ      1    5522.6 5590.6
## - JOB           8    5542.1 5596.1
## - TIF           1    5534.6 5602.6
## - TRAVTIME      1    5539.7 5607.7
## - CAR_USE       1    5547.8 5615.8
## - CAR_TYPE      5    5557.8 5617.8
## - MVR_PTS       1    5553.7 5621.7
## - REVOKED       1    5565.2 5633.2
## - URBANICITY    1    5966.6 6034.6
##
## Step:  AIC=5560.47
## TARGET_FLAG ~ KIDSDRIV + YOJ + PARENT1 + TRAVTIME + CAR_USE +
##      TIF + CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE + INCOME + HOME_VAL +
##      BLUEBOOK + OLDCLAIM + MSTATUS + EDUCATION + JOB + CAR_TYPE +
##      URBANICITY
##
##           Df Deviance    AIC
## - YOJ           1    5493.1 5559.1
## - CAR_AGE       1    5493.2 5559.2
## <none>          5492.5 5560.5
## - OLDCLAIM      1    5500.6 5566.6

```

```

## - EDUCATION      4    5509.5 5569.5
## - HOME_VAL       1    5505.3 5571.3
## - MSTATUS        1    5505.9 5571.9
## - BLUEBOOK       1    5508.3 5574.3
## - INCOME         1    5509.0 5575.0
## - PARENT1        1    5517.0 5583.0
## - CLM_FREQ       1    5523.2 5589.2
## - KIDSDRIV       1    5525.5 5591.5
## - JOB            8    5543.0 5595.0
## - TIF            1    5535.0 5601.0
## - TRAVTIME       1    5540.0 5606.0
## - CAR_USE        1    5548.6 5614.6
## - CAR_TYPE       5    5558.5 5616.5
## - MVR_PTS        1    5554.7 5620.7
## - REVOKED        1    5566.2 5632.2
## - URBANICITY     1    5967.2 6033.2
##
## Step:  AIC=5559.05
## TARGET_FLAG ~ KIDSDRIV + PARENT1 + TRAVTIME + CAR_USE + TIF +
##      CLM_FREQ + REVOKED + MVR_PTS + CAR_AGE + INCOME + HOME_VAL +
##      BLUEBOOK + OLDCLAIM + MSTATUS + EDUCATION + JOB + CAR_TYPE +
##      URBANICITY
##
##           Df Deviance    AIC
## - CAR_AGE      1    5493.8 5557.8
## <none>          5493.1 5559.1
## - OLDCLAIM     1    5501.2 5565.2
## - EDUCATION    4    5510.0 5568.0
## - HOME_VAL     1    5506.0 5570.0
## - MSTATUS      1    5507.3 5571.3
## - BLUEBOOK     1    5509.1 5573.1
## - INCOME       1    5510.1 5574.1
## - PARENT1      1    5517.3 5581.3
## - CLM_FREQ     1    5523.8 5587.8
## - KIDSDRIV     1    5525.9 5589.9
## - JOB          8    5543.5 5593.5
## - TIF          1    5536.0 5600.0
## - TRAVTIME     1    5540.5 5604.5
## - CAR_USE      1    5549.5 5613.5
## - CAR_TYPE     5    5559.2 5615.2
## - MVR_PTS      1    5555.8 5619.8
## - REVOKED      1    5566.8 5630.8
## - URBANICITY   1    5967.4 6031.4
##
## Step:  AIC=5557.78
## TARGET_FLAG ~ KIDSDRIV + PARENT1 + TRAVTIME + CAR_USE + TIF +
##      CLM_FREQ + REVOKED + MVR_PTS + INCOME + HOME_VAL + BLUEBOOK +
##      OLDCLAIM + MSTATUS + EDUCATION + JOB + CAR_TYPE + URBANICITY
##
##           Df Deviance    AIC

```

```

## <none>          5493.8 5557.8
## - OLDCLAIM      1  5501.9 5563.9
## - HOME_VAL      1  5506.6 5568.6
## - MSTATUS       1  5508.1 5570.1
## - EDUCATION     4  5515.5 5571.5
## - BLUEBOOK      1  5509.8 5571.8
## - INCOME        1  5511.1 5573.1
## - PARENT1       1  5518.1 5580.1
## - CLM_FREQ      1  5524.3 5586.3
## - KIDSDRIV      1  5526.6 5588.6
## - JOB           8  5544.2 5592.2
## - TIF           1  5536.9 5598.9
## - TRAVTIME      1  5541.2 5603.2
## - CAR_USE       1  5550.2 5612.2
## - CAR_TYPE      5  5560.0 5614.0
## - MVR_PTS       1  5556.6 5618.6
## - REVOKED       1  5567.6 5629.6
## - URBANICITY    1  5968.5 6030.5

##
## Call:
## glm(formula = TARGET_FLAG ~ KIDSDRIV + PARENT1 + TRAVTIME + CAR_USE +
##     TIF + CLM_FREQ + REVOKED + MVR_PTS + INCOME + HOME_VAL +
##     BLUEBOOK + OLDCLAIM + MSTATUS + EDUCATION + JOB + CAR_TYPE +
##     URBANICITY, family = "binomial", data = no_na_df)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5484  -0.7189  -0.3947   0.6354   3.1125
##
## Coefficients:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -2.772e+00  3.182e-01  -8.711  < 2e-16 ***
## KIDSDRIV       3.638e-01  6.317e-02   5.759 8.44e-09 ***
## PARENT1Yes     5.343e-01  1.083e-01   4.933 8.09e-07 ***
## TRAVTIME       1.496e-02  2.178e-03   6.870 6.41e-12 ***
## CAR_USEPrivate -7.829e-01  1.051e-01  -7.447 9.56e-14 ***
## TIF            -5.505e-02  8.538e-03  -6.448 1.13e-10 ***
## CLM_FREQ       1.817e-01  3.267e-02   5.564 2.64e-08 ***
## REVOKEDYes     9.110e-01  1.052e-01   8.664  < 2e-16 ***
## MVR_PTS        1.242e-01  1.577e-02   7.876 3.38e-15 ***
## INCOME         -5.254e-06  1.278e-06  -4.111 3.93e-05 ***
## HOME_VAL       -1.429e-06  4.007e-07  -3.565 0.000364 ***
## BLUEBOOK       -2.186e-05  5.513e-06  -3.966 7.31e-05 ***
## OLDCLAIM       -1.268e-05  4.477e-06  -2.832 0.004632 **
## MSTATUSYes     -3.512e-01  9.235e-02  -3.803 0.000143 ***
## EDUCATIONBachelors -3.810e-01  1.252e-01  -3.042 0.002352 **
## EDUCATIONHigh School  6.275e-02  1.078e-01   0.582 0.560533
## EDUCATIONMasters -2.379e-01  1.852e-01  -1.285 0.198896
## EDUCATIONPhD    -1.422e-01  2.332e-01  -0.610 0.541953

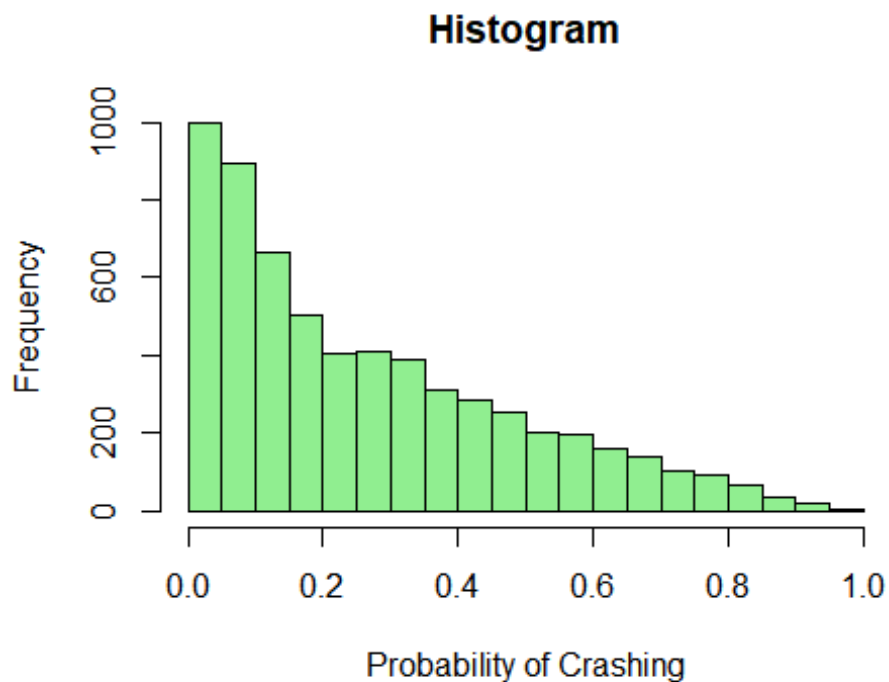
```

```

## JOBBlue Collar          1.491e-01  2.148e-01   0.694 0.487641
## JOBClerical             2.737e-01  2.260e-01   1.211 0.225827
## JOBDoctor              -4.619e-01  3.114e-01  -1.483 0.137996
## JOBHome Maker          2.027e-01  2.360e-01   0.859 0.390287
## JOBLawyer              9.464e-02  1.936e-01   0.489 0.624947
## JOBManager             -6.922e-01  1.995e-01  -3.470 0.000520 ***
## JOBProfessional        4.181e-02  2.045e-01   0.204 0.838022
## JOBStudent             3.952e-02  2.447e-01   0.161 0.871715
## CAR_TYPEPanel Truck    5.791e-01  1.766e-01   3.280 0.001038 **
## CAR_TYPEPickup         4.670e-01  1.158e-01   4.031 5.55e-05 ***
## CAR_TYPESports Car     8.509e-01  1.236e-01   6.885 5.78e-12 ***
## CAR_TYPESUV            6.545e-01  9.816e-02   6.668 2.60e-11 ***
## CAR_TYPEVan            5.255e-01  1.409e-01   3.731 0.000191 ***
## URBANICITYHighly Urban/ Urban 2.326e+00  1.273e-01  18.269 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 7078.9  on 6134  degrees of freedom
## Residual deviance: 5493.8  on 6103  degrees of freedom
## AIC: 5557.8
##
## Number of Fisher Scoring iterations: 5

```

Confusion matrix and the curve...

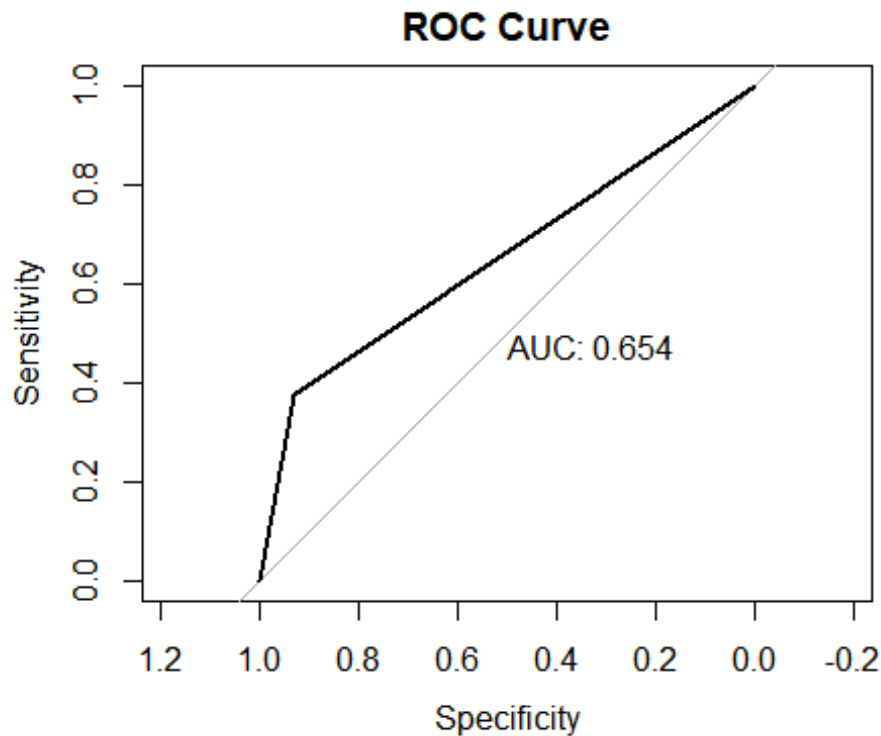


```

##           Actual
## Predictions    0    1
##           0 1137  258
##           1   82  155

## Confusion Matrix and Statistics
##
##           Reference
## Prediction    0    1
##           0 1137  258
##           1   82  155
##
##           Accuracy : 0.7917
##           95% CI : (0.7711, 0.8111)
##       No Information Rate : 0.7469
##       P-Value [Acc > NIR] : 1.272e-05
##
##           Kappa : 0.3585
##
## Mcnemar's Test P-Value : < 2.2e-16
##
##           Sensitivity : 0.37530
##           Specificity : 0.93273
##           Pos Pred Value : 0.65401
##           Neg Pred Value : 0.81505
##           Prevalence : 0.25306
##           Detection Rate : 0.09498
##       Detection Prevalence : 0.14522
##       Balanced Accuracy : 0.65402
##
##           'Positive' Class : 1
##

```



Select Models

We will use the Stepwise regression models for both questions, as the logistic stepwise model was the most accurate, had the highest sensitivity, and was the least complex. The stepwise linear model also had a higher adjusted R^2 value, and a smaller p-value. Thus, these should be our best bet in imputing the target values in the evaluation dataset.

```
## TARGET_FLAG TARGET_AMT KIDSDRIV AGE HOMEKIDS YOJ PARENT1 TRAVTIME
CAR_USE
## 1          NA          NA          0 48          0 11          No          26
Private
## 2          NA          NA          1 40          1 11          Yes          21
Private
## 3          NA          NA          0 44          2 12          Yes          30
Commercial
## 4          NA          NA          0 35          2 NA          Yes          74
Private
## 5          NA          NA          0 59          0 12          No          45
Private
## 6          NA          NA          0 46          0 14          No          7
Commercial
## TIF RED_CAR CLM_FREQ REVOKED MVR_PTS CAR_AGE INCOME HOME_VAL BLUEBOOK
## 1  1    yes      0      No      2     10  52881      0    21970
## 2  6    no       1      No      2      1  50815      0    18930
## 3 10    no       0      No      0     10  43486      0     5900
## 4  6    no       0     Yes      0      4  21204      0     9230
```

## 5	1	yes	2	No	4	1	87460	0	15420
## 6	1	no	1	No	2	12	NA	207519	25660
##	OLDCLAIM	MSTATUS	SEX	EDUCATION		JOB	CAR_TYPE		
URBAN	CITY								
## 1	0	No	M	Bachelors		Manager	Van	Highly Urban/	
Urban									
## 2	3295	No	M	High School		Manager	Minivan	Highly Urban/	
Urban									
## 3	0	No	F	High School		Blue Collar	SUV	Highly Rural/	
Rural									
## 4	0	No	M	High School		Clerical	Pickup	Highly Rural/	
Rural									
## 5	44857	No	M	High School		Manager	Minivan	Highly Urban/	
Urban									
## 6	2119	Yes	M	Bachelors		Professional	Panel Truck	Highly Urban/	
Urban									

Does the person crash the car?

The following shows if a person crash his or her car...

##	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
16															
##	0	0	0	0	0	0	0	0	0	0	0	1	1	0	0
1															
##	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31
32															
##	1	0	1	0	0	0	0	0	0	0	0	0	0	0	0
0															
##	33	34	35	36	37	38	39	40	41	42	43	44	45	46	47
48															
##	0	0	0	0	0	0	0	1	0	1	0	1	0	0	0
0															
##	49	50	51	52	53	54	55	56	57	58	59	60	61	62	63
64															
##	0	1	0	0	1	0	0	0	0	0	0	1	0	0	0
0															
##	65	66	67	68	69	70	71	72	73	74	75	76	77	78	79
80															
##	0	0	1	1	0	0	0	0	1	0	1	0	0	0	0
0															
##	81	82	83	84	85	86	87	88	89	90	91	92	93	94	95
96															
##	1	0	0	0	1	1	0	0	0	1	0	0	0	0	0
0															
##	97	98	99	100	101	102	103	104	105	106	107	108	109	110	111
112															
##	0	0	0	0	0	1	1	1	0	0	0	0	0	0	1
0															
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128															
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0															
##	129	130	131	132	133	134	135	136	137	138	139	140	141	142	143
144															
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0															
##	145	146	147	148	149	150	151	152	153	154	155	156	157	158	159
160															
##	0	1	0	0	0	0	1	0	1	1	0	0	0	0	1
1															
##	161	162	163	164	165	166	167	168	169	170	171	172	173	174	175
176															
##	0	0	0	0	1	0	0	0	0	0	0	1	0	0	0
0															
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192															
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1															
##	193	194	195	196	197	198	199	200	201	202	203	204	205	206	207
208															
##	0	0	0	1	0	0	0	0	0	0	0	0	0	0	1
0															
##	209	210	211	212	213	214	215	216	217	218	219	220	221	222	223
224															
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0															
##	225	226	227	228	229	230	231	232	233	234	235	236	237	238	239
240															
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1															
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256															
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0															
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272															
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0															
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288															
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0															
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304															
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0															
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320															
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0															
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336															
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0															
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352															
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0															
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368															
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0															
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384															
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0															
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400															
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0															
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416															
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0															
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432															
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0															
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448															
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0															
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464															
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0															
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480															
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0															
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496															
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1															
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512															
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0															
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528															
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0															
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544															
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0															
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560															
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0															
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576															
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0															
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592															
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0															
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608															
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0															
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624															
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0															
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640															
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0															
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656															
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0															
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672															
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1															
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688															
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0															
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0															
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736															
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0															
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752															
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0															
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768															
##	1	1	0	0	0	0	0	0	0	1	0	0	1	1	0
0															
##	769	770	771	772	773	774	775	776	777	778	779	780	781	782	783
784															
##	0	0	0	1	0	1	0	0	0	0	0	0	0	1	0
0															
##	785	786	787	788	789	790	791	792	793	794	795	796	797	798	799
800															
##	0	0	0	0	0	0	0	0	0	0	0	0	0	1	1
0															
##	801	802	803	804	805	806	807	808	809	810	811	812	813	814	815
816															
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0															
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832															
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0															
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848															
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0															
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864															
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0															
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880															
##	0	0	1	0	0	1	0	1	0	1	0	0	0	0	0
0															
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896															
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0															
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912															
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0															
##	913	914	915	916	917	918	919	920	921	922	923	924	925	926	927

928															
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0															
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944															
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0															
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960															
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0															
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976															
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0															
##	977	978	979	980	981	982	983	984	985	986	987	988	989	990	991
992															
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0															
##	993	994	995	996	997	998	999	1000	1001	1002	1003	1004	1005	1006	1007
1008															
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0															
##	1009	1010	1011	1012	1013	1014	1015	1016	1017	1018	1019	1020	1021	1022	1023
1024															
##	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
0															
##	1025	1026	1027	1028	1029	1030	1031	1032	1033	1034	1035	1036	1037	1038	1039
1040															
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0															
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1056															
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0															
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1072															
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0															
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1088															
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0															
##	1089	1090	1091	1092	1093	1094	1095	1096	1097	1098	1099	1100	1101	1102	1103
1104															
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0															
##	1105	1106	1107	1108	1109	1110	1111	1112	1113	1114	1115	1116	1117	1118	1119
1120															
##	0	0	0	0	0	0	1	0	0	0	0	0	0	1	1

```
0
## 1121 1122 1123 1124 1125 1126 1127 1128 1129 1130 1131 1132 1133 1134 1135
1136
## 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0
0
## 1137 1138 1139 1140 1141 1142 1143 1144 1145 1146 1147 1148 1149 1150 1151
1152
## 0 0 0 0 0 0 0 0 1 0 0 1 0 1 0
1
## 1153 1154 1155 1156 1157 1158 1159 1160 1161 1162 1163 1164 1165 1166 1167
1168
## 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0
0
## 1169 1170 1171 1172 1173 1174 1175 1176 1177 1178 1179 1180 1181 1182 1183
1184
## 0 0 0 1 1 1 0 0 0 0 0 1 0 1 0
0
## 1185 1186 1187 1188 1189 1190 1191 1192 1193 1194 1195 1196 1197 1198 1199
1200
## 1 0 0 0 0 0 0 0 0 1 0 0 0 0 0
0
## 1201 1202 1203 1204 1205 1206 1207 1208 1209 1210 1211 1212 1213 1214 1215
1216
## 0 0 0 0 0 0 1 0 0 0 0 0 0 0 0
0
## 1217 1218 1219 1220 1221 1222 1223 1224 1225 1226 1227 1228 1229 1230 1231
1232
## 1 0 0 0 0 0 1 0 1 0 0 0 0 1 0
0
## 1233 1234 1235 1236 1237 1238 1239 1240 1241 1242 1243 1244 1245 1246 1247
1248
## 0 1 0 0 0 1 0 0 0 0 0 0 0 1 0
0
## 1249 1250 1251 1252 1253 1254 1255 1256 1257 1258 1259 1260 1261 1262 1263
1264
## 0 0 0 1 0 0 0 0 1 0 0 0 0 0 0
1
## 1265 1266 1267 1268 1269 1270 1271 1272 1273 1274 1275 1276 1277 1278 1279
1280
## 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0
## 1281 1282 1283 1284 1285 1286 1287 1288 1289 1290 1291 1292 1293 1294 1295
1296
## 1 0 0 0 0 0 1 0 0 0 1 0 0 0 0
0
## 1297 1298 1299 1300 1301 1302 1303 1304 1305 1306 1307 1308 1309 1310 1311
1312
## 0 0 1 0 0 0 0 0 0 0 0 0 0 1 1
0
## 1313 1314 1315 1316 1317 1318 1319 1320 1321 1322 1323 1324 1325 1326 1327
```

```
1328
## 1 0 0 0 0 0 0 1 0 0 1 0 0 0 0
0
## 1329 1330 1331 1332 1333 1334 1335 1336 1337 1338 1339 1340 1341 1342 1343
1344
## 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0
0
## 1345 1346 1347 1348 1349 1350 1351 1352 1353 1354 1355 1356 1357 1358 1359
1360
## 1 0 0 0 0 0 0 1 0 0 0 0 0 0 0
0
## 1361 1362 1363 1364 1365 1366 1367 1368 1369 1370 1371 1372 1373 1374 1375
1376
## 0 0 1 0 0 0 1 1 0 1 0 0 0 0 0
0
## 1377 1378 1379 1380 1381 1382 1383 1384 1385 1386 1387 1388 1389 1390 1391
1392
## 0 0 0 0 1 1 0 0 0 0 0 0 0 0 1
0
## 1393 1394 1395 1396 1397 1398 1399 1400 1401 1402 1403 1404 1405 1406 1407
1408
## 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0
0
## 1409 1410 1411 1412 1413 1414 1415 1416 1417 1418 1419 1420 1421 1422 1423
1424
## 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0
0
## 1425 1426 1427 1428 1429 1430 1431 1432 1433 1434 1435 1436 1437 1438 1439
1440
## 0 1 0 0 1 0 0 0 0 0 0 0 0 1 0
0
## 1441 1442 1443 1444 1445 1446 1447 1448 1449 1450 1451 1452 1453 1454 1455
1456
## 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0
0
## 1457 1458 1459 1460 1461 1462 1463 1464 1465 1466 1467 1468 1469 1470 1471
1472
## 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0
## 1473 1474 1475 1476 1477 1478 1479 1480 1481 1482 1483 1484 1485 1486 1487
1488
## 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0
0
## 1489 1490 1491 1492 1493 1494 1495 1496 1497 1498 1499 1500 1501 1502 1503
1504
## 1 0 0 1 0 0 1 0 0 0 0 0 0 0 0
1
## 1505 1506 1507 1508 1509 1510 1511 1512 1513 1514 1515 1516 1517 1518 1519
1520
## 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0
```

```
0
## 1521 1522 1523 1524 1525 1526 1527 1528 1529 1530 1531 1532 1533 1534 1535
1536
## 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0
0
## 1537 1538 1539 1540 1541 1542 1543 1544 1545 1546 1547 1548 1549 1550 1551
1552
## 0 0 1 1 0 0 0 0 1 0 0 0 0 0 0
0
## 1553 1554 1555 1556 1557 1558 1559 1560 1561 1562 1563 1564 1565 1566 1567
1568
## 0 1 0 0 0 0 0 0 0 0 0 0 1 0 0
0
## 1569 1570 1571 1572 1573 1574 1575 1576 1577 1578 1579 1580 1581 1582 1583
1584
## 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0
0
## 1585 1586 1587 1588 1589 1590 1591 1592 1593 1594 1595 1596 1597 1598 1599
1600
## 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0
0
## 1601 1602 1603 1604 1605 1606 1607 1608 1609 1610 1611 1612 1613 1614 1615
1616
## 1 0 0 1 0 1 0 0 0 1 0 0 0 1 0
1
## 1617 1618 1619 1620 1621 1622 1623 1624 1625 1626 1627 1628 1629 1630 1631
1632
## 0 1 0 0 1 0 1 0 0 0 0 0 0 1 0
1
## 1633 1634 1635 1636 1637 1638 1639 1640 1641 1642 1643 1644 1645 1646 1647
1648
## 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0
0
## 1649 1650 1651 1652 1653 1654 1655 1656 1657 1658 1659 1660 1661 1662 1663
1664
## 0 1 0 0 0 0 1 0 0 0 0 0 0 1 1
1
## 1665 1666 1667 1668 1669 1670 1671 1672 1673 1674 1675 1676 1677 1678 1679
1680
## 1 0 0 0 1 0 1 0 1 0 0 0 0 0 0
0
## 1681 1682 1683 1684 1685 1686 1687 1688 1689 1690 1691 1692 1693 1694 1695
1696
## 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0
1
## 1697 1698 1699 1700 1701 1702 1703 1704 1705 1706 1707 1708 1709 1710 1711
1712
## 0 1 1 0 0 0 0 0 0 0 1 0 1 0 0
0
## 1713 1714 1715 1716 1717 1718 1719 1720 1721 1722 1723 1724 1725 1726 1727
```



```

1728
## 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0
0
## 1729 1730 1731 1732 1733 1734 1735 1736 1737 1738 1739 1740 1741 1742 1743
1744
## 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0
0
## 1745 1746 1747 1748 1749 1750 1751 1752 1753 1754 1755 1756 1757 1758 1759
1760
## 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0
0
## 1761 1762 1763 1764 1765 1766 1767 1768 1769 1770 1771 1772 1773 1774 1775
1776
## 1 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0
0
## 1777 1778 1779 1780 1781 1782 1783 1784 1785 1786 1787 1788 1789 1790 1791
1792
## 1 0 1 0 0 0 0 0 0 0 0 0 0 1 0 0
0
## 1793 1794 1795 1796 1797 1798 1799 1800 1801 1802 1803 1804 1805 1806 1807
1808
## 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1
1
## 1809 1810 1811 1812 1813 1814 1815 1816 1817 1818 1819 1820 1821 1822 1823
1824
## 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0
0
## 1825 1826 1827 1828 1829 1830 1831 1832 1833 1834 1835 1836 1837 1838 1839
1840
## 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0
1
## 1841 1842 1843 1844 1845 1846 1847 1848 1849 1850 1851 1852 1853 1854 1855
1856
## 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
0
## 1857 1858 1859 1860 1861 1862 1863 1864 1865 1866 1867 1868 1869 1870 1871
1872
## 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1
0
## 1873 1874 1875 1876 1877 1878 1879 1880 1881 1882 1883 1884 1885 1886 1887
1888
## 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0
0
## 1889 1890 1891 1892 1893 1894 1895 1896 1897 1898 1899 1900 1901 1902 1903
1904
## 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0
0
## 1905 1906 1907 1908 1909 1910 1911 1912 1913 1914 1915 1916 1917 1918 1919
1920
## 0 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0

```

0															
##	1921	1922	1923	1924	1925	1926	1927	1928	1929	1930	1931	1932	1933	1934	1935
1936															
##	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
0															
##	1937	1938	1939	1940	1941	1942	1943	1944	1945	1946	1947	1948	1949	1950	1951
1952															
##	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0
0															
##	1953	1954	1955	1956	1957	1958	1959	1960	1961	1962	1963	1964	1965	1966	1967
1968															
##	1	0	0	0	0	0	0	0	1	0	0	0	0	0	0
0															
##	1969	1970	1971	1972	1973	1974	1975	1976	1977	1978	1979	1980	1981	1982	1983
1984															
##	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
0															
##	1985	1986	1987	1988	1989	1990	1991	1992	1993	1994	1995	1996	1997	1998	1999
2000															
##	0	0	0	0	0	0	0	1	1	1	0	0	0	1	0
0															
##	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015
2016															
##	1	0	1	0	1	0	1	0	0	0	0	0	1	0	0
0															
##	2017	2018	2019	2020	2021	2022	2023	2024	2025	2026	2027	2028	2029	2030	2031
2032															
##	0	1	1	0	0	0	0	0	0	0	0	0	0	1	0
0															
##	2033	2034	2035	2036	2037	2038	2039	2040	2041	2042	2043	2044	2045	2046	2047
2048															
##	0	0	1	0	0	0	1	0	0	0	0	0	0	0	0
1															
##	2049	2050	2051	2052	2053	2054	2055	2056	2057	2058	2059	2060	2061	2062	2063
2064															
##	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
0															
##	2065	2066	2067	2068	2069	2070	2071	2072	2073	2074	2075	2076	2077	2078	2079
2080															
##	0	0	0	1	0	0	0	0	1	0	0	0	1	0	0
1															
##	2081	2082	2083	2084	2085	2086	2087	2088	2089	2090	2091	2092	2093	2094	2095
2096															
##	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0
1															
##	2097	2098	2099	2100	2101	2102	2103	2104	2105	2106	2107	2108	2109	2110	2111
2112															
##	0	0	1	1	1	0	1	0	0	0	1	0	0	0	1
0															
##	2113	2114	2115	2116	2117	2118	2119	2120	2121	2122	2123	2124	2125	2126	2127

2128														
##	0	0	0	0	0	0	1	0	0	0	1	0	0	0
0														
##	2129	2130	2131	2132	2133	2134	2135	2136	2137	2138	2139	2140	2141	
##	0	0	0	0	0	0	0	0	0	0	0	0	0	

What's the pay off?

The following shows the pay off for the person who has a crash car...

##	1	2	3	4	5	6	7	8
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	9	10	11	12	13	14	15	16
##	0.000	0.000	0.000	5475.670	NA	0.000	0.000	5696.563
##	17	18	19	20	21	22	23	24
##	NA	0.000	4005.702	0.000	0.000	0.000	0.000	0.000
##	25	26	27	28	29	30	31	32
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	33	34	35	36	37	38	39	40
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	5773.914
##	41	42	43	44	45	46	47	48
##	0.000	5222.274	0.000	6895.830	0.000	0.000	0.000	0.000
##	49	50	51	52	53	54	55	56
##	0.000	4122.707	0.000	0.000	6354.033	0.000	0.000	0.000
##	57	58	59	60	61	62	63	64
##	0.000	0.000	0.000	3956.597	0.000	0.000	0.000	0.000
##	65	66	67	68	69	70	71	72
##	0.000	0.000	5079.031	7313.395	0.000	0.000	0.000	0.000
##	73	74	75	76	77	78	79	80
##	5291.526	0.000	5617.353	0.000	0.000	0.000	0.000	0.000
##	81	82	83	84	85	86	87	88
##	4258.968	0.000	0.000	0.000	5389.727	6707.825	0.000	0.000
##	89	90	91	92	93	94	95	96
##	0.000	6004.846	0.000	0.000	0.000	0.000	0.000	0.000
##	97	98	99	100	101	102	103	104
##	0.000	0.000	0.000	0.000	0.000	6769.367	6191.289	6218.239
##	105	106	107	108	109	110	111	112
##	0.000	0.000	0.000	0.000	0.000	0.000	5465.364	0.000
##	113	114	115	116	117	118	119	120
##	0.000	0.000	5368.212	0.000	0.000	5420.656	3319.370	0.000
##	121	122	123	124	125	126	127	128
##	0.000	5301.919	4497.096	0.000	0.000	0.000	0.000	0.000
##	129	130	131	132	133	134	135	136
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	137	138	139	140	141	142	143	144
##	6802.093	6829.302	0.000	0.000	0.000	5537.473	0.000	0.000
##	145	146	147	148	149	150	151	152
##	0.000	4494.962	0.000	0.000	0.000	0.000	6005.710	0.000
##	153	154	155	156	157	158	159	160
##	6369.258	3709.136	0.000	0.000	0.000	0.000	5348.406	5846.070

##	161	162	163	164	165	166	167	168
##	0.000	0.000	0.000	0.000	7559.370	0.000	0.000	0.000
##	169	170	171	172	173	174	175	176
##	0.000	0.000	0.000	7417.136	0.000	0.000	0.000	0.000
##	177	178	179	180	181	182	183	184
##	8473.842	5261.994	7898.267	4553.929	6200.101	0.000	0.000	0.000
##	185	186	187	188	189	190	191	192
##	0.000	0.000	0.000	0.000	0.000	0.000	4548.520	4814.371
##	193	194	195	196	197	198	199	200
##	0.000	0.000	0.000	4972.555	0.000	0.000	0.000	0.000
##	201	202	203	204	205	206	207	208
##	0.000	0.000	0.000	0.000	0.000	0.000	5968.634	0.000
##	209	210	211	212	213	214	215	216
##	0.000	0.000	0.000	0.000	5545.745	0.000	0.000	0.000
##	217	218	219	220	221	222	223	224
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	225	226	227	228	229	230	231	232
##	0.000	0.000	5783.903	0.000	0.000	0.000	0.000	0.000
##	233	234	235	236	237	238	239	240
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	5629.963
##	241	242	243	244	245	246	247	248
##	0.000	0.000	6355.952	0.000	0.000	0.000	0.000	0.000
##	249	250	251	252	253	254	255	256
##	0.000	5206.449	0.000	5120.150	0.000	0.000	0.000	0.000
##	257	258	259	260	261	262	263	264
##	0.000	0.000	4650.007	0.000	0.000	0.000	0.000	0.000
##	265	266	267	268	269	270	271	272
##	0.000	0.000	0.000	0.000	6977.377	0.000	5488.436	0.000
##	273	274	275	276	277	278	279	280
##	0.000	0.000	0.000	0.000	6451.459	0.000	0.000	0.000
##	281	282	283	284	285	286	287	288
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	289	290	291	292	293	294	295	296
##	4036.865	5176.608	0.000	0.000	0.000	0.000	0.000	0.000
##	297	298	299	300	301	302	303	304
##	0.000	6578.373	0.000	0.000	0.000	0.000	0.000	0.000
##	305	306	307	308	309	310	311	312
##	0.000	0.000	0.000	0.000	0.000	0.000	7210.489	0.000
##	313	314	315	316	317	318	319	320
##	0.000	5813.319	0.000	0.000	0.000	0.000	5642.683	0.000
##	321	322	323	324	325	326	327	328
##	0.000	5232.695	0.000	0.000	5213.844	0.000	5652.385	0.000
##	329	330	331	332	333	334	335	336
##	0.000	0.000	0.000	0.000	5111.822	0.000	0.000	0.000
##	337	338	339	340	341	342	343	344
##	0.000	5345.916	0.000	0.000	4418.071	7482.650	0.000	5172.728
##	345	346	347	348	349	350	351	352
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	353	354	355	356	357	358	359	360
##	6432.744	5976.170	0.000	4121.160	0.000	0.000	0.000	0.000

##	361	362	363	364	365	366	367	368
##	5285.394	0.000	0.000	5054.093	0.000	0.000	0.000	0.000
##	369	370	371	372	373	374	375	376
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	5335.288
##	377	378	379	380	381	382	383	384
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	385	386	387	388	389	390	391	392
##	0.000	0.000	0.000	0.000	0.000	4248.370	0.000	0.000
##	393	394	395	396	397	398	399	400
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	401	402	403	404	405	406	407	408
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	409	410	411	412	413	414	415	416
##	0.000	0.000	0.000	4733.234	0.000	0.000	4750.170	0.000
##	417	418	419	420	421	422	423	424
##	0.000	0.000	0.000	0.000	4957.193	6953.478	6649.642	0.000
##	425	426	427	428	429	430	431	432
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	433	434	435	436	437	438	439	440
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	441	442	443	444	445	446	447	448
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	449	450	451	452	453	454	455	456
##	5891.145	5936.302	0.000	0.000	0.000	0.000	0.000	6248.901
##	457	458	459	460	461	462	463	464
##	0.000	7460.280	0.000	0.000	0.000	0.000	0.000	0.000
##	465	466	467	468	469	470	471	472
##	0.000	0.000	6808.783	4239.256	0.000	0.000	0.000	5350.144
##	473	474	475	476	477	478	479	480
##	0.000	0.000	0.000	0.000	7031.717	4821.085	0.000	0.000
##	481	482	483	484	485	486	487	488
##	0.000	0.000	0.000	0.000	5054.732	5156.531	0.000	0.000
##	489	490	491	492	493	494	495	496
##	0.000	6490.070	5404.977	0.000	0.000	0.000	0.000	4636.621
##	497	498	499	500	501	502	503	504
##	0.000	0.000	0.000	0.000	0.000	0.000	4606.546	0.000
##	505	506	507	508	509	510	511	512
##	6904.102	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	513	514	515	516	517	518	519	520
##	0.000	0.000	0.000	0.000	7244.168	0.000	0.000	0.000
##	521	522	523	524	525	526	527	528
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	529	530	531	532	533	534	535	536
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	537	538	539	540	541	542	543	544
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	545	546	547	548	549	550	551	552
##	0.000	0.000	0.000	7315.510	0.000	0.000	0.000	0.000
##	553	554	555	556	557	558	559	560

##	561	562	563	564	565	566	567	568
##	0.000	0.000	0.000	0.000	4547.021	0.000	6108.173	0.000
##	569	570	571	572	573	574	575	576
##	0.000	4685.575	0.000	0.000	0.000	0.000	0.000	0.000
##	577	578	579	580	581	582	583	584
##	0.000	0.000	0.000	0.000	0.000	5819.177	0.000	4281.428
##	585	586	587	588	589	590	591	592
##	0.000	0.000	0.000	0.000	5695.009	0.000	0.000	0.000
##	593	594	595	596	597	598	599	600
##	0.000	0.000	4882.506	4492.539	NA	0.000	0.000	3608.387
##	601	602	603	604	605	606	607	608
##	5516.032	0.000	0.000	0.000	0.000	0.000	7217.174	0.000
##	609	610	611	612	613	614	615	616
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	617	618	619	620	621	622	623	624
##	0.000	0.000	0.000	7257.339	0.000	0.000	0.000	0.000
##	625	626	627	628	629	630	631	632
##	0.000	8393.521	6160.348	0.000	0.000	5325.100	0.000	0.000
##	633	634	635	636	637	638	639	640
##	0.000	0.000	0.000	0.000	0.000	6169.289	0.000	0.000
##	641	642	643	644	645	646	647	648
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	649	650	651	652	653	654	655	656
##	0.000	0.000	0.000	0.000	5215.720	0.000	0.000	0.000
##	657	658	659	660	661	662	663	664
##	0.000	0.000	0.000	0.000	0.000	3346.967	0.000	0.000
##	665	666	667	668	669	670	671	672
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	4174.040
##	673	674	675	676	677	678	679	680
##	5705.315	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	681	682	683	684	685	686	687	688
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	689	690	691	692	693	694	695	696
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	697	698	699	700	701	702	703	704
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	705	706	707	708	709	710	711	712
##	0.000	0.000	0.000	5893.608	0.000	0.000	0.000	0.000
##	713	714	715	716	717	718	719	720
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	721	722	723	724	725	726	727	728
##	5262.107	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	729	730	731	732	733	734	735	736
##	0.000	0.000	3915.067	0.000	0.000	0.000	0.000	0.000
##	737	738	739	740	741	742	743	744
##	0.000	0.000	0.000	0.000	4859.104	0.000	4632.958	0.000
##	745	746	747	748	749	750	751	752
##	0.000	0.000	6333.132	0.000	0.000	0.000	0.000	0.000
##	753	754	755	756	757	758	759	760
##	4624.901	NA	0.000	0.000	0.000	0.000	0.000	0.000

##	761	762	763	764	765	766	767	768
##	0.000	5573.808	0.000	0.000	4879.736	5319.826	0.000	0.000
##	769	770	771	772	773	774	775	776
##	0.000	0.000	0.000	6227.692	0.000	6251.188	0.000	0.000
##	777	778	779	780	781	782	783	784
##	0.000	0.000	0.000	0.000	0.000	6695.185	0.000	0.000
##	785	786	787	788	789	790	791	792
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	793	794	795	796	797	798	799	800
##	0.000	0.000	0.000	0.000	0.000	4641.350	6332.597	0.000
##	801	802	803	804	805	806	807	808
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	809	810	811	812	813	814	815	816
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	817	818	819	820	821	822	823	824
##	0.000	6913.294	6594.283	0.000	4777.920	0.000	0.000	0.000
##	825	826	827	828	829	830	831	832
##	5693.148	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	833	834	835	836	837	838	839	840
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	841	842	843	844	845	846	847	848
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	849	850	851	852	853	854	855	856
##	6830.224	5524.305	6725.353	0.000	0.000	0.000	0.000	0.000
##	857	858	859	860	861	862	863	864
##	0.000	0.000	5213.194	0.000	0.000	5458.911	0.000	0.000
##	865	866	867	868	869	870	871	872
##	0.000	0.000	5947.022	0.000	0.000	5664.951	0.000	7598.274
##	873	874	875	876	877	878	879	880
##	0.000	5590.781	0.000	0.000	0.000	0.000	0.000	0.000
##	881	882	883	884	885	886	887	888
##	0.000	0.000	0.000	0.000	5295.700	0.000	6934.576	0.000
##	889	890	891	892	893	894	895	896
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	897	898	899	900	901	902	903	904
##	0.000	0.000	0.000	0.000	0.000	0.000	6346.439	0.000
##	905	906	907	908	909	910	911	912
##	0.000	0.000	NA	0.000	0.000	0.000	6837.345	0.000
##	913	914	915	916	917	918	919	920
##	0.000	0.000	0.000	0.000	7760.779	4775.519	0.000	0.000
##	921	922	923	924	925	926	927	928
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	929	930	931	932	933	934	935	936
##	5034.844	0.000	0.000	4856.103	0.000	0.000	0.000	0.000
##	937	938	939	940	941	942	943	944
##	0.000	0.000	0.000	0.000	5310.618	0.000	0.000	0.000
##	945	946	947	948	949	950	951	952
##	5539.785	0.000	0.000	0.000	0.000	4447.436	0.000	0.000
##	953	954	955	956	957	958	959	960
##	0.000	4843.225	0.000	0.000	0.000	0.000	0.000	0.000

##	961	962	963	964	965	966	967	968
##	0.000	0.000	0.000	0.000	0.000	6875.781	0.000	0.000
##	969	970	971	972	973	974	975	976
##	0.000	0.000	5309.937	0.000	0.000	0.000	0.000	0.000
##	977	978	979	980	981	982	983	984
##	0.000	0.000	0.000	0.000	0.000	0.000	5422.135	4999.322
##	985	986	987	988	989	990	991	992
##	5537.809	5727.931	0.000	0.000	0.000	6546.481	0.000	0.000
##	993	994	995	996	997	998	999	1000
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1001	1002	1003	1004	1005	1006	1007	1008
##	0.000	7289.280	6292.301	0.000	0.000	0.000	0.000	0.000
##	1009	1010	1011	1012	1013	1014	1015	1016
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1017	1018	1019	1020	1021	1022	1023	1024
##	0.000	0.000	0.000	0.000	0.000	0.000	7913.074	0.000
##	1025	1026	1027	1028	1029	1030	1031	1032
##	6506.584	4640.733	0.000	0.000	0.000	0.000	0.000	0.000
##	1033	1034	1035	1036	1037	1038	1039	1040
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1041	1042	1043	1044	1045	1046	1047	1048
##	0.000	0.000	3426.357	0.000	4214.611	0.000	0.000	0.000
##	1049	1050	1051	1052	1053	1054	1055	1056
##	0.000	0.000	5227.218	0.000	4948.486	0.000	0.000	0.000
##	1057	1058	1059	1060	1061	1062	1063	1064
##	0.000	0.000	4746.100	5385.210	0.000	4351.973	0.000	0.000
##	1065	1066	1067	1068	1069	1070	1071	1072
##	0.000	0.000	0.000	0.000	0.000	4176.099	0.000	0.000
##	1073	1074	1075	1076	1077	1078	1079	1080
##	0.000	6880.506	0.000	0.000	0.000	0.000	5607.576	0.000
##	1081	1082	1083	1084	1085	1086	1087	1088
##	6984.803	6709.498	0.000	0.000	0.000	6868.757	0.000	0.000
##	1089	1090	1091	1092	1093	1094	1095	1096
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1097	1098	1099	1100	1101	1102	1103	1104
##	0.000	0.000	0.000	4205.585	0.000	0.000	0.000	0.000
##	1105	1106	1107	1108	1109	1110	1111	1112
##	0.000	0.000	0.000	0.000	0.000	0.000	5182.965	0.000
##	1113	1114	1115	1116	1117	1118	1119	1120
##	0.000	0.000	0.000	0.000	0.000	6709.209	6223.129	0.000
##	1121	1122	1123	1124	1125	1126	1127	1128
##	0.000	0.000	0.000	0.000	0.000	0.000	8912.265	0.000
##	1129	1130	1131	1132	1133	1134	1135	1136
##	0.000	0.000	0.000	0.000	5712.423	0.000	0.000	0.000
##	1137	1138	1139	1140	1141	1142	1143	1144
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1145	1146	1147	1148	1149	1150	1151	1152
##	5463.160	0.000	0.000	7196.456	0.000	6822.587	0.000	5751.812
##	1153	1154	1155	1156	1157	1158	1159	1160
##	0.000	0.000	6275.183	0.000	0.000	0.000	0.000	0.000

##	1161	1162	1163	1164	1165	1166	1167	1168
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1169	1170	1171	1172	1173	1174	1175	1176
##	0.000	0.000	0.000	7041.006	6108.312	6815.829	0.000	0.000
##	1177	1178	1179	1180	1181	1182	1183	1184
##	0.000	0.000	0.000	5580.595	0.000	3747.483	0.000	0.000
##	1185	1186	1187	1188	1189	1190	1191	1192
##	3846.994	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1193	1194	1195	1196	1197	1198	1199	1200
##	0.000	5409.701	0.000	0.000	0.000	0.000	0.000	0.000
##	1201	1202	1203	1204	1205	1206	1207	1208
##	0.000	0.000	0.000	0.000	0.000	0.000	3536.531	0.000
##	1209	1210	1211	1212	1213	1214	1215	1216
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1217	1218	1219	1220	1221	1222	1223	1224
##	6260.202	0.000	0.000	0.000	0.000	0.000	6330.222	0.000
##	1225	1226	1227	1228	1229	1230	1231	1232
##	4989.314	0.000	0.000	0.000	0.000	6936.248	0.000	0.000
##	1233	1234	1235	1236	1237	1238	1239	1240
##	0.000	5751.840	0.000	0.000	0.000	6438.584	0.000	0.000
##	1241	1242	1243	1244	1245	1246	1247	1248
##	0.000	0.000	0.000	0.000	0.000	7192.531	0.000	0.000
##	1249	1250	1251	1252	1253	1254	1255	1256
##	0.000	0.000	0.000	4215.831	0.000	0.000	0.000	0.000
##	1257	1258	1259	1260	1261	1262	1263	1264
##	NA	0.000	0.000	0.000	0.000	0.000	0.000	NA
##	1265	1266	1267	1268	1269	1270	1271	1272
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1273	1274	1275	1276	1277	1278	1279	1280
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1281	1282	1283	1284	1285	1286	1287	1288
##	5428.112	0.000	0.000	0.000	0.000	0.000	4561.558	0.000
##	1289	1290	1291	1292	1293	1294	1295	1296
##	0.000	0.000	6678.837	0.000	0.000	0.000	0.000	0.000
##	1297	1298	1299	1300	1301	1302	1303	1304
##	0.000	0.000	6126.466	0.000	0.000	0.000	0.000	0.000
##	1305	1306	1307	1308	1309	1310	1311	1312
##	0.000	0.000	0.000	0.000	0.000	5076.144	4901.860	0.000
##	1313	1314	1315	1316	1317	1318	1319	1320
##	6085.861	0.000	0.000	0.000	0.000	0.000	0.000	NA
##	1321	1322	1323	1324	1325	1326	1327	1328
##	0.000	0.000	4236.719	0.000	0.000	0.000	0.000	0.000
##	1329	1330	1331	1332	1333	1334	1335	1336
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1337	1338	1339	1340	1341	1342	1343	1344
##	0.000	0.000	0.000	0.000	0.000	5490.831	0.000	0.000
##	1345	1346	1347	1348	1349	1350	1351	13

##	1361	1362	1363	1364	1365	1366	1367	1368
##	0.000	0.000	6768.419	0.000	0.000	0.000	6637.771	3595.377
##	1369	1370	1371	1372	1373	1374	1375	1376
##	0.000	6348.059	0.000	0.000	0.000	0.000	0.000	0.000
##	1377	1378	1379	1380	1381	1382	1383	1384
##	0.000	0.000	0.000	0.000	8113.193	6031.657	0.000	0.000
##	1385	1386	1387	1388	1389	1390	1391	1392
##	0.000	0.000	0.000	0.000	0.000	0.000	5366.899	0.000
##	1393	1394	1395	1396	1397	1398	1399	1400
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1401	1402	1403	1404	1405	1406	1407	1408
##	0.000	0.000	6504.564	4461.362	0.000	0.000	0.000	0.000
##	1409	1410	1411	1412	1413	1414	1415	1416
##	0.000	5960.671	0.000	0.000	0.000	0.000	0.000	0.000
##	1417	1418	1419	1420	1421	1422	1423	1424
##	0.000	0.000	0.000	0.000	0.000	6396.493	0.000	0.000
##	1425	1426	1427	1428	1429	1430	1431	1432
##	0.000	6749.695	0.000	0.000	3614.795	0.000	0.000	0.000
##	1433	1434	1435	1436	1437	1438	1439	1440
##	0.000	0.000	0.000	0.000	0.000	NA	0.000	0.000
##	1441	1442	1443	1444	1445	1446	1447	1448
##	0.000	5575.614	0.000	0.000	0.000	4881.547	0.000	0.000
##	1449	1450	1451	1452	1453	1454	1455	1456
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1457	1458	1459	1460	1461	1462	1463	1464
##	5645.008	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1465	1466	1467	1468	1469	1470	1471	1472
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1473	1474	1475	1476	1477	1478	1479	1480
##	0.000	0.000	0.000	0.000	4335.014	0.000	0.000	0.000
##	1481	1482	1483	1484	1485	1486	1487	1488
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1489	1490	1491	1492	1493	1494	1495	1496
##	5904.515	0.000	0.000	4381.385	0.000	0.000	6229.020	0.000
##	1497	1498	1499	1500	1501	1502	1503	1504
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	5267.192
##	1505	1506	1507	1508	1509	1510	1511	1512
##	0.000	0.000	5095.101	0.000	0.000	0.000	0.000	0.000
##	1513	1514	1515	1516	1517	1518	1519	1520
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1521	1522	1523	1524	1525	1526	1527	1528
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1529	1530	1531	1532	1533	1534	1535	1536
##	0.000	4973.152	0.000	0.000	0.000	0.000	0.000	0.000
##	1537	1538	1539	1540	1541	1542	1543	1544
##	0.000	0.000	5667.814	6760.523	0.000	0.000	0.000	0.000
##	1545	1546	1547	1548	1549	1550	1551	1552
##	5847.446	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1553	1554	1555	1556	1557	1558	1559	1560
##	0.000	5612.621	0.000	0.000	0.000	0.000	0.000	0.000

##	1561	1562	1563	1564	1565	1566	1567	1568
##	0.000	0.000	0.000	0.000	7166.470	0.000	0.000	0.000
##	1569	1570	1571	1572	1573	1574	1575	1576
##	0.000	0.000	5361.613	0.000	0.000	0.000	0.000	0.000
##	1577	1578	1579	1580	1581	1582	1583	1584
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1585	1586	1587	1588	1589	1590	1591	1592
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	5996.193
##	1593	1594	1595	1596	1597	1598	1599	1600
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1601	1602	1603	1604	1605	1606	1607	1608
##	5559.496	0.000	0.000	NA	0.000	7158.428	0.000	0.000
##	1609	1610	1611	1612	1613	1614	1615	1616
##	0.000	4873.533	0.000	0.000	0.000	6861.509	0.000	5559.262
##	1617	1618	1619	1620	1621	1622	1623	1624
##	0.000	6030.426	0.000	0.000	5233.554	0.000	6012.400	0.000
##	1625	1626	1627	1628	1629	1630	1631	1632
##	0.000	0.000	0.000	0.000	0.000	6496.079	0.000	5244.937
##	1633	1634	1635	1636	1637	1638	1639	1640
##	0.000	0.000	0.000	0.000	6024.966	0.000	0.000	0.000
##	1641	1642	1643	1644	1645	1646	1647	1648
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1649	1650	1651	1652	1653	1654	1655	1656
##	0.000	4095.290	0.000	0.000	0.000	0.000	6675.684	0.000
##	1657	1658	1659	1660	1661	1662	1663	1664
##	0.000	0.000	0.000	0.000	0.000	8356.651	4343.948	6755.248
##	1665	1666	1667	1668	1669	1670	1671	1672
##	6530.786	0.000	0.000	0.000	5371.681	0.000	5399.944	0.000
##	1673	1674	1675	1676	1677	1678	1679	1680
##	5815.950	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1681	1682	1683	1684	1685	1686	1687	1688
##	0.000	4690.875	5638.786	0.000	0.000	0.000	0.000	0.000
##	1689	1690	1691	1692	1693	1694	1695	1696
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	4865.409
##	1697	1698	1699	1700	1701	1702	1703	1704
##	0.000	5594.100	4691.226	0.000	0.000	0.000	0.000	0.000
##	1705	1706	1707	1708	1709	1710	1711	1712
##	0.000	0.000	NA	0.000	5597.763	0.000	0.000	0.000
##	1713	1714	1715	1716	1717	1718	1719	1720
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1721	1722	1723	1724	1725	1726	1727	1728
##	0.000	0.000	0.000	0.000	4925.682	0.000	0.000	0.000
##	1729	1730	1731	1732	1733	1734	1735	1736
##	6366.190	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1737	1738	1739	1740	1741	1742	1743	1744
##	0.000	0.000	0.000	0.000	5031.367	0.000	0.000	0.000
##	1745	1746	1747	1748	1749	1750	1751	1752
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1753	1754	1755	1756	1757	1758	1759	1760
##	0.000	4389.859	0.000	0.000	0.000	0.000	0.000	0.000

##	1761	1762	1763	1764	1765	1766	1767	1768
##	7176.761	0.000	0.000	0.000	0.000	4861.931	0.000	0.000
##	1769	1770	1771	1772	1773	1774	1775	1776
##	0.000	0.000	0.000	0.000	0.000	6824.111	0.000	0.000
##	1777	1778	1779	1780	1781	1782	1783	1784
##	4780.055	0.000	5018.836	0.000	0.000	0.000	0.000	0.000
##	1785	1786	1787	1788	1789	1790	1791	1792
##	0.000	0.000	0.000	0.000	4245.770	0.000	0.000	0.000
##	1793	1794	1795	1796	1797	1798	1799	1800
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1801	1802	1803	1804	1805	1806	1807	1808
##	0.000	0.000	0.000	0.000	0.000	0.000	5174.165	5898.036
##	1809	1810	1811	1812	1813	1814	1815	1816
##	0.000	0.000	0.000	0.000	5179.408	0.000	0.000	0.000
##	1817	1818	1819	1820	1821	1822	1823	1824
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1825	1826	1827	1828	1829	1830	1831	1832
##	0.000	4929.268	0.000	0.000	0.000	0.000	0.000	0.000
##	1833	1834	1835	1836	1837	1838	1839	1840
##	0.000	0.000	0.000	0.000	0.000	NA	0.000	5445.802
##	1841	1842	1843	1844	1845	1846	1847	1848
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1849	1850	1851	1852	1853	1854	1855	1856
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1857	1858	1859	1860	1861	1862	1863	1864
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1865	1866	1867	1868	1869	1870	1871	1872
##	0.000	0.000	0.000	0.000	0.000	4436.742	5887.182	0.000
##	1873	1874	1875	1876	1877	1878	1879	1880
##	0.000	0.000	0.000	0.000	0.000	6065.051	0.000	0.000
##	1881	1882	1883	1884	1885	1886	1887	1888
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1889	1890	1891	1892	1893	1894	1895	1896
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1897	1898	1899	1900	1901	1902	1903	1904
##	0.000	0.000	0.000	4204.661	0.000	0.000	0.000	0.000
##	1905	1906	1907	1908	1909	1910	1911	1912
##	0.000	0.000	0.000	0.000	0.000	6242.812	0.000	0.000
##	1913	1914	1915	1916	1917	1918	1919	1920
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1921	1922	1923	1924	1925	1926	1927	1928
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1929	1930	1931	1932	1933	1934	1935	1936
##	0.000	0.000	0.000	0.000	0.000	7316.304	0.000	0.000
##	1937	1938	1939	1940	1941	1942	1943	1944
##	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
##	1945	1946	1947	1948	1949	1950	1951	1952
##								

Appendix

```
library(corrplot)
library(tidyverse)
library(Hmisc)
library(PerformanceAnalytics)
library(mice)
library(gt)
library(caret)
library(bnstruct)
library(VIM)
library(corr)
library(kableExtra)
library(rpart)
library(gtsummary)
library(reshape)
library(pROC)
library(randomForest)

#Import data
df1_data <-
read.csv("https://raw.githubusercontent.com/aaitelmouden/DATA621/master/Homework4/data/insurance_training_data.csv")

df <- df1_data
glimpse(df)

# Get rid of INDEX, TARGET_FLAG, TARGET_AMT
df1 <- subset(df, select = -c(INDEX))

# Select money variables to clean before summary

df2 <- df1[, names(df1) %in% c("INCOME", "OLDCLAIM", "HOME_VAL", "BLUEBOOK")]

# Clean using string function

df2 <- apply(df2, 2, function(y) gsub("\\$", "", y))
df2 <- apply(df2, 2, function(y) gsub(",", "", y))
df2 <- apply(df2, 2, as.integer)
```

```

# Combine variables

df1 <- df1[, !names(df1) %in% c("INCOME", "OLDCLAIM", "HOME_VAL",
"BLUEBOOK")]
df1 <- cbind(df1, df2)

# Select characters variables to clean before summary

df2 <- df1[, names(df1) %in% c("MSTATUS", "SEX", "EDUCATION", "JOB",
"CAR_TYPE", "URBANICITY")]

# Clean using character function

df2 <- apply(df2, 2, function(y) gsub("z_", "", y))

df1 <- df1[, !names(df1) %in% c("MSTATUS", "SEX", "EDUCATION", "JOB",
"CAR_TYPE", "URBANICITY")]
df1 <- cbind(df1, df2)

# Factor

df1$SEX <- factor(df1$SEX)
df1$PARENT1 <- factor(df1$PARENT1)
df1$CAR_USE <- factor(df1$CAR_USE)
df1$RED_CAR <- factor(df1$RED_CAR)
df1$REVOKED <- factor(df1$REVOKED)
df1$MSTATUS <- factor(df1$MSTATUS)
df1$EDUCATION <- factor(df1$EDUCATION)
df1$CAR_TYPE <- factor(df1$CAR_TYPE)
df1$MSTATUS <- factor(df1$MSTATUS)
df1$URBANICITY <- factor(df1$URBANICITY)
df1$JOB <- factor(df1$JOB)

# Summary table

table1 <- tbl_summary(df1,
  statistic = list(all_continuous() ~ "{mean} ({sd}) {median} {min}
{max}"), missing = "no")
table1

# Boxplot

ggplot(melt(df1), aes(x=factor(variable), y=value)) + facet_wrap(~variable,
scale="free") + geom_boxplot()

# Histogram

ggplot(melt(df1), aes(x=value)) + facet_wrap(~variable, scale="free") +
geom_histogram(bins=50)

```

```

# Outliers plot

ggplot(stack(df1), aes(x = ind, y = values, fill=ind)) +
  geom_boxplot(outlier.colour = "red", outlier.alpha=.4) +
  coord_cartesian(ylim = c(0, 1000)) +
  theme_classic()+
  theme(axis.text.x=element_text(angle=45, hjust=1))

# Correlation between the numerical variables

df3 <- df1[,!names(df1) %in% c("PARENT1","RED_CAR",
"REVOKED","URBANICITY","INCOME","OLDCLAIM","HOME_VAL","BLUEBOOK","SEX","MSTAT
US","EDUCATION","JOB","CAR_USE","CAR_TYPE")]

a <- cor(df3, method="pearson", use="complete.obs")
a

corrplot(a)

# Imputing other NA variables by median before checking on minimum

df1[is.na(df1$HOME_VAL),]$HOME_VAL <-
median(df1[complete.cases(df1$HOME_VAL),]$HOME_VAL)
df1[is.na(df1$YOJ),]$YOJ <- median(df1[complete.cases(df1$YOJ),]$YOJ)
df1[is.na(df1$INCOME),]$INCOME <-
median(df1[complete.cases(df1$INCOME),]$INCOME)
df1[is.na(df1$AGE),]$AGE <- median(df1[complete.cases(df1$AGE),]$AGE)

# Looking at the minimum values in order to adjust for the BoxCox
Transformation

apply(df1,2,min)

# BoxCox Transformation

df4 <- df1 %>%
  filter(CAR_AGE >= 0) %>%
  mutate(CAR_AGE = CAR_AGE + 1, YOJ = YOJ + 1, INCOME = INCOME + 1) %>%
  select(TIF, BLUEBOOK, TRAVTIME, AGE, CAR_AGE, YOJ, INCOME)

apply(df4, 2, BoxCoxTrans)

# LM 1

```



```

df_lm <- df1 %>% select(-TARGET_FLAG) %>% filter(TARGET_AMT > 0)

# Linear regression

lm_model1 <- lm(TARGET_AMT~.,data = df_lm)

summary(lm_model1)

# Visualization

opar <- par(mfrow = c(2,2), oma = c(0, 0, 1.1, 0))
plot(lm_model1, las = 1)

# LM 2

lm_model2 <- step(lm_model1)
summary(lm_model2)

# LM 3

lm_model3 <- lm(TARGET_AMT ~ KIDSDRIV + AGE + HOMEKIDS + I(YOJ^1.6) + PARENT1
+ I(TRAFTIME^0.7) + CAR_USE + I(TIF^0.2) + RED_CAR + CLM_FREQ + REVOKED +
MVR_PTS + I(CAR_AGE^0.5) + I(INCOME^0.4) + HOME_VAL + I(BLUEBOOK^0.5) +
OLDCLAIM + MSTATUS + SEX + EDUCATION + JOB + CAR_TYPE + URBANICITY,
data=df_lm)
summary(lm_model3)

# Logistic Model

# Logistic M 1

df_rm <- df1 %>% select(-TARGET_AMT)

rm_model1 <- glm(TARGET_FLAG ~ ., family="binomial", df_rm)

summary(rm_model1)

# Logistic M 2

hist(rm_model1$fitted.values,main = " Histogram ",xlab = "Probability of
Crashing", col = 'light blue')

probabilities <- predict(rm_model1, df_rm, type = "response")
predicted1.classes <- ifelse(probabilities > 0.5, 1, 0)

```

```

df_rm$pred1.class <- predicted1.classes
table("Predictions" = df_rm$pred1.class, "Actual" = df_rm$TARGET_FLAG)

confusionMatrix(as.factor(predicted1.classes), as.factor(df_rm$TARGET_FLAG))

curve <- roc(response = df_rm$TARGET_FLAG,
  predictor = predicted1.classes,
  plot = TRUE,
  print.auc = TRUE,
  main = "ROC Curve")

df_rf <- subset(df_rm, select= -c(pred1.class))

trainIndex <- createDataPartition(df_rf$TARGET_FLAG, p = .8,
  list = FALSE,
  times = 1)

rf_train <- df_rf[ trainIndex,]
rf_test  <- df_rf[-trainIndex,]

rf <- randomForest(factor(TARGET_FLAG) ~ ., data = rf_train, na.action =
na.omit)
rf
varImpPlot(rf)

test_rf <- predict(rf, rf_test)
confusionMatrix(test_rf, factor(rf_test$TARGET_FLAG), positive = '1')

# Logistic M 3

no_na_df <- na.omit(rf_train)
rm_model1 <- glm(TARGET_FLAG ~ ., family="binomial", no_na_df)

model2 <- step(rm_model1)

summary(model2)

# Confusion matrix and the curve...

hist(model2$fitted.values,main = " Histogram ",xlab = "Probability of
Crashing", col = 'light green')

```

```

probabilities <- predict(model2, rf_test, type = "response")
predicted.classes <- ifelse(probabilities > 0.5, 1, 0)
rf_test$pred.class <- predicted.classes
table("Predictions" = rf_test$pred.class, "Actual" = rf_test$TARGET_FLAG)

confusionMatrix(as.factor(predicted.classes), as.factor(rf_test$TARGET_FLAG),
positive = '1')

curve <- roc(response = rf_test$TARGET_FLAG,
  predictor = predicted.classes,
  plot = TRUE,
  print.auc = TRUE,
  main = "ROC Curve")

# Select models

eval <-
read_csv('https://raw.githubusercontent.com/aaitelmouden/DATA621/master/Homework4/data/insurance-evaluation-data.csv')

# Get rid of INDEX, TARGET_FLAG, TARGET_AMT
eval <- subset(eval, select = -c(INDEX))

# Select money variables to clean before summary

eval2 <- eval[, names(eval) %in% c("INCOME", "OLDCLAIM", "HOME_VAL",
"BLUEBOOK")]

# Clean using string function

eval2 <- apply(eval2, 2, function(y) gsub("\\$", "", y))
eval2 <- apply(eval2, 2, function(y) gsub(",", "", y))
eval2 <- apply(eval2, 2, as.integer)

# Combine variables

eval1 <- eval[, !names(eval) %in% c("INCOME", "OLDCLAIM", "HOME_VAL",
"BLUEBOOK")]
eval1 <- cbind(eval1, eval2)

# Select characters variables to clean before summary

eval2 <- eval1[, names(eval1) %in% c("MSTATUS", "SEX", "EDUCATION", "JOB",
"CAR_TYPE", "URBANICITY")]

```

```

# Clean using character function

eval2 <- apply(eval2, 2, function(y) gsub("z_", "", y))

eval1 <- eval1[, !names(eval1) %in% c("MSTATUS", "SEX", "EDUCATION", "JOB",
"CAR_TYPE", "URBANICITY")]
eval1 <- cbind(eval1, eval2)
head(eval1)

### Does the person crash the car?

eval1[is.na(eval1$INCOME),]$INCOME <-
median(eval1[complete.cases(eval1$INCOME),]$INCOME)
eval1[is.na(eval1$HOME_VAL),]$HOME_VAL <- 0

probabilities <- predict(model2, eval1[, -1], type = "response")
preds <- ifelse(probabilities > 0.5, 1, 0)

preds[is.na(preds)] <- 0
preds

### What's the pay off?

amt <- ifelse(preds == 1, predict(lm_model2, eval1[, -1], type = "response"),
0)
amt

```

References

- [Regression Model Validation](#)
- [Binary Logistic Regression](#)
- [What are pseudo R-squareds?] (<https://stats.idre.ucla.edu/other/mult-pkg/faq/general/faq-what-are-pseudo-r-squareds/>)