Logistic_Regression_without_cross-validation.R

jas

Fri Sep 09 18:48:51 2016

```
drive <- "D:/R/Analytics vidhya/Hackathon/Loan Prediction"</pre>
setwd(drive)
train <- read.csv("TrainD.csv", header = TRUE, stringsAsFactors = FALSE)</pre>
test <- read.csv("TestD.csv", header = TRUE, stringsAsFactors = FALSE)
head(train)
##
      Loan ID ApplicantIncome CoapplicantIncome LoanAmount Loan Status Gender
## 1 LP001002
                           5849
                                                      4.959423
## 2 LP001003
                           4583
                                              1508
                                                      4.852030
                                                                           0
                                                                                   1
## 3 LP001005
                           3000
                                                  0
                                                      4.189655
                                                                           1
                                                                                   1
## 4 LP001006
                           2583
                                               2358
                                                      4.787492
                                                                           1
                                                                                   1
                                                                           1
                                                                                   1
## 5 LP001008
                           6000
                                                  0
                                                      4.948760
## 6 LP001011
                                              4196
                           5417
                                                      5.587249
                                                                           1
                                                                                   1
##
     Married Dependents0 Dependents1 Dependents2 Dependents3. Education
## 1
            0
                         1
                                                   0
                                      0
                                                                 0
## 2
            1
                         0
                                      1
                                                   0
                                                                 0
                                                                            1
## 3
            1
                         1
                                      0
                                                   0
                                                                            1
                                                                 0
## 4
            1
                         1
                                      0
                                                   0
                                                                 0
                                                                            0
## 5
            0
                         1
                                      0
                                                   0
                                                                 0
                                                                            1
## 6
                         0
                                      0
                                                   1
     Self_Employed Credit_History Property_AreaRural Property_AreaSemiurban
##
## 1
                                                       0
                                                                                0
                  0
                                   1
                                                                                0
## 2
                  0
                                   1
                                                       1
## 3
                  1
                                   1
                                                       0
                                                                                0
## 4
                  0
                                   1
                                                       0
                                                                                0
                  0
                                                       0
## 5
                                   1
                                                                                0
## 6
                  1
                                   1
                                                                                0
##
     Property_AreaUrban TotalIncome ApplicantIncome_Zero
## 1
                        1
                             8.674026
## 2
                        0
                             8.714568
                                                            0
## 3
                        1
                                                            0
                             8.006368
                                                            0
## 4
                        1
                             8.505323
                             8.699515
## 5
                        1
                                                            0
## 6
                        1
                             9.170872
     DebtRatio Mainapplicant DebtRatio TotalIncome CoappIncGApplinc
##
## 1
                   0.02436511
                                           0.02436511
                                                                        0
## 2
                   0.02792930
                                           0.02101461
## 3
                                                                        0
                   0.02200000
                                           0.02200000
## 4
                   0.04645761
                                           0.02428658
                                                                        0
## 5
                                                                        0
                   0.02350000
                                           0.02350000
## 6
                   0.04928927
                                           0.02777489
```

```
Loan Amount Term10.12.mnths Loan Amount Term112.120.mnths
## 1
## 2
                                0
                                                                0
## 3
                                0
                                                                0
## 4
                                0
                                                                0
## 5
                                0
                                                                0
## 6
##
     Loan_Amount_Term1120.240.mnths Loan_Amount_Term1240.360.mnths
## 1
## 2
                                   0
                                                                    1
## 3
                                   0
                                                                    1
                                   0
                                                                    1
## 4
                                   0
                                                                    1
## 5
## 6
                                   0
     Loan Amount Term1360..Mnths Gender Married Married Dep0 Married Dep1
##
## 1
                                0
                                                0
                                                              0
## 2
                                0
                                                1
                                                              0
                                                                            1
## 3
                                0
                                                1
                                                              1
                                                                            0
## 4
                                0
                                                1
                                                              1
                                                                            0
## 5
                                0
                                                0
                                                              0
                                                                            0
## 6
                                0
                                                1
                                                              0
                                                                            0
##
     Married Dep2 Married Dep3
## 1
                0
                              0
## 2
                0
                              0
                              0
## 3
                0
## 4
                0
                              0
## 5
                0
                              0
## 6
                1
                              0
######USing all parameters for prediction#######
logistic <- glm(Loan_Status ~ ., data=train[,!colnames(train) %in%</pre>
c("Loan_ID")],
                 family='binomial')
summary(logistic)
##
## Call:
## glm(formula = Loan_Status ~ ., family = "binomial", data = train[,
       !colnames(train) %in% c("Loan_ID")])
##
## Deviance Residuals:
##
       Min
                 10
                       Median
                                     3Q
                                             Max
## -2.2762 -0.3805
                       0.5052
                                0.6793
                                          2.5280
##
## Coefficients: (5 not defined because of singularities)
##
                                      Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                                     3.181e+00 5.699e+00
                                                             0.558 0.57673
## ApplicantIncome
                                    -4.079e-06 3.956e-05
                                                           -0.103 0.91788
## CoapplicantIncome
                                    -7.477e-05 6.455e-05 -1.158
                                                                    0.24671
## LoanAmount
                                     8.610e-01 6.566e-01
                                                             1.311 0.18977
```

```
## Gender
                                  -2.394e-01 3.732e-01
                                                         -0.641
                                                                 0.52124
## Married
                                  -4.839e-01 1.411e+00
                                                         -0.343 0.73168
## Dependents0
                                  -9.349e-01
                                              1.251e+00
                                                         -0.747
                                                                 0.45486
## Dependents1
                                  -1.581e+00 1.310e+00
                                                         -1.206
                                                                 0.22764
## Dependents2
                                   3.801e-01
                                              1.637e+00
                                                          0.232
                                                                 0.81638
## Dependents3.
                                          NA
                                                             NA
                                                                      NA
                                                     NA
                                   3.277e-01 2.707e-01
                                                          1.210
                                                                 0.22611
## Education
## Self_Employed
                                   6.284e-02 3.229e-01
                                                          0.195
                                                                 0.84570
                                                                 < 2e-16 ***
## Credit_History
                                   3.956e+00 4.107e-01
                                                          9.631
## Property AreaRural
                                  -2.164e-01
                                              2.704e-01
                                                         -0.800
                                                                 0.42362
                                                                 0.00646 **
## Property AreaSemiurban
                                   7.574e-01 2.781e-01
                                                          2.723
## Property AreaUrban
                                          NA
                                                     NA
                                                             NA
                                                                      NA
                                  -1.047e+00
                                              8.768e-01
                                                         -1.194
                                                                 0.23258
## TotalIncome
## ApplicantIncome Zero
                                          NA
                                                     NA
                                                             NA
                                                                      NΔ
## DebtRatio Mainapplicant
                                  -1.157e+00 2.491e+00
                                                         -0.465
                                                                 0.64218
## DebtRatio_TotalIncome
                                  -6.939e+01 3.173e+01
                                                         -2.187 0.02873 *
## CoappIncGApplinc
                                   3.132e-01 4.122e-01
                                                          0.760
                                                                 0.44727
## Loan Amount Term10.12.mnths
                                   1.327e+01 5.354e+02
                                                          0.025
                                                                 0.98022
## Loan Amount Term112.120.mnths
                                   7.493e-01
                                              9.485e-01
                                                          0.790
                                                                 0.42954
## Loan_Amount_Term1120.240.mnths
                                   1.522e+00 7.754e-01
                                                          1.963
                                                                 0.04961 *
## Loan Amount Term1240.360.mnths
                                   1.320e+00 6.429e-01
                                                          2.053
                                                                 0.04008 *
## Loan Amount Term1360..Mnths
                                          NA
                                                     NΑ
                                                             NA
                                                                      NA
## Gender_Married
                                   2.250e-01 6.487e-01
                                                          0.347
                                                                 0.72865
## Married Dep0
                                   8.338e-01
                                              1.346e+00
                                                          0.620
                                                                 0.53556
## Married Dep1
                                   1.223e+00 1.413e+00
                                                          0.865
                                                                 0.38688
## Married Dep2
                                  -2.710e-01
                                              1.727e+00
                                                         -0.157
                                                                 0.87535
                                          NA
## Married Dep3
                                                     NA
                                                             NA
                                                                      NΑ
## ---
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 762.89
                              on 613
                                      degrees of freedom
## Residual deviance: 542.42
                              on 588
                                      degrees of freedom
## AIC: 594.42
##
## Number of Fisher Scoring iterations: 12
pred = predict(logistic, newdata=train,type="response")
## Warning in predict.lm(object, newdata, se.fit, scale = 1, type =
## ifelse(type == : prediction from a rank-deficient fit may be misleading
pred <- as.integer(ifelse(pred>"0.499",1,0))
str(pred)
   int [1:614] 1 1 1 1 1 1 1 0 1 1 ...
library(caret)
## Warning: package 'caret' was built under R version 3.2.5
```

```
## Loading required package: lattice
## Loading required package: ggplot2
## Warning: package 'ggplot2' was built under R version 3.2.5
confusionMatrix(data=pred, train$Loan_Status)
## Confusion Matrix and Statistics
##
             Reference
##
## Prediction
               0
                    1
              93 12
##
            0
##
            1 99 410
##
##
                  Accuracy : 0.8192
                    95% CI: (0.7865, 0.8489)
##
##
       No Information Rate: 0.6873
##
       P-Value [Acc > NIR] : 9.197e-14
##
##
                     Kappa: 0.5202
   Mcnemar's Test P-Value: 3.275e-16
##
##
##
               Sensitivity: 0.4844
               Specificity: 0.9716
##
##
            Pos Pred Value: 0.8857
##
            Neg Pred Value: 0.8055
                Prevalence: 0.3127
##
##
            Detection Rate: 0.1515
##
      Detection Prevalence: 0.1710
##
         Balanced Accuracy: 0.7280
##
          'Positive' Class: 0
##
##
######USing relevant selected parameters for prediction#######
logistic1 <- glm(Loan_Status ~</pre>
Credit History+Property AreaSemiurban+DebtRatio TotalIncome+
                Loan Amount Term1120.240.mnths+Loan Amount Term1240.360.mnths
, data=train[,!colnames(train) %in% c("Loan ID")],
                family='binomial')
summary(logistic1)
##
## Call:
## glm(formula = Loan Status ~ Credit History + Property AreaSemiurban +
       DebtRatio_TotalIncome + Loan_Amount_Term1120.240.mnths +
##
       Loan Amount Term1240.360.mnths, family = "binomial", data = train[,
##
##
       !colnames(train) %in% c("Loan_ID")])
##
## Deviance Residuals:
```

```
Median
       Min
                 10
                                   30
                                           Max
                      0.5397
                               0.7364
## -2.2505 -0.3848
                                        2.3693
##
## Coefficients:
##
                                  Estimate Std. Error z value Pr(>|z|)
                                               0.6584 -4.476 7.61e-06 ***
## (Intercept)
                                   -2.9470
## Credit History
                                    3.7912
                                               0.3935
                                                        9.635 < 2e-16 ***
## Property_AreaSemiurban
                                    0.8270
                                               0.2345
                                                        3.527
                                                               0.00042 ***
## DebtRatio_TotalIncome
                                  -24.9974
                                              11.5929 -2.156 0.03106 *
## Loan Amount Term1120.240.mnths
                                    0.9774
                                               0.5931
                                                        1.648
                                                               0.09934 .
                                                        1.922 0.05456 .
## Loan_Amount_Term1240.360.mnths
                                    0.8742
                                               0.4547
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##
       Null deviance: 762.89
                              on 613
                                      degrees of freedom
## Residual deviance: 565.83 on 608 degrees of freedom
## AIC: 577.83
##
## Number of Fisher Scoring iterations: 5
pred1 = predict(logistic1, newdata=train,type="response")
pred1 <- as.integer(ifelse(pred>"0.499",1,0))
str(pred1)
   int [1:614] 1 1 1 1 1 1 1 0 1 1 ...
confusionMatrix(data=pred1, train$Loan Status)
## Confusion Matrix and Statistics
##
##
             Reference
                0
## Prediction
                    1
##
            0
              93 12
##
            1
              99 410
##
##
                  Accuracy : 0.8192
##
                    95% CI: (0.7865, 0.8489)
##
       No Information Rate: 0.6873
       P-Value [Acc > NIR] : 9.197e-14
##
##
##
                     Kappa : 0.5202
##
   Mcnemar's Test P-Value : 3.275e-16
##
##
               Sensitivity: 0.4844
##
               Specificity: 0.9716
##
            Pos Pred Value: 0.8857
##
            Neg Pred Value: 0.8055
                Prevalence: 0.3127
##
##
            Detection Rate: 0.1515
```

```
##
      Detection Prevalence: 0.1710
##
         Balanced Accuracy: 0.7280
##
          'Positive' Class: 0
##
##
####The accuracy doesnt improve######
##Accuracy at 0.8192##
##LB Accuracy is 0.76##
#Predict Output
predicted= predict(logistic,test,type = "response")
## Warning in predict.lm(object, newdata, se.fit, scale = 1, type =
## ifelse(type == : prediction from a rank-deficient fit may be misleading
head(predicted)
##
                                                              6
## 0.8577128 0.7662559 0.8066765 0.8921870 0.7043595 0.7848411
predicted <- as.integer(ifelse(predicted>"0.499",1,0))
head(predicted)
## [1] 1 1 1 1 1 1
library(dplyr)
## Warning: package 'dplyr' was built under R version 3.2.5
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
test1 <- test %>% select (Loan_ID)
comb <- data.frame(test1,predicted)</pre>
write.csv(comb, "12345.csv", row.names = FALSE)
```