data_challenge

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Read data and packages

```
# Upload packages for data clean and analysis
require(plyr) #data clean
## Loading required package: plyr
require(dplyr) #data clean
## Loading required package: dplyr
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:plyr':
##
##
       arrange, count, desc, failwith, id, mutate, rename, summarise,
##
       summarize
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
require(tidyr) #data clean
## Loading required package: tidyr
require(ggplot2) # visualzation
## Loading required package: ggplot2
require(XML) # web scriping
## Loading required package: XML
```

```
require(testthat) #test model
## Loading required package: testthat
require(kernlab) # kernel and SVM
## Loading required package: kernlab
##
## Attaching package: 'kernlab'
## The following object is masked from 'package:ggplot2':
##
##
       alpha
require(reshape2)
## Loading required package: reshape2
require(caret) # Cross - Validation k fold
## Loading required package: caret
## Loading required package: lattice
require(maps) # heatmap: map_data
## Loading required package: maps
##
## # maps v3.1: updated 'world': all lakes moved to separate new #
## # 'lakes' database. Type '?world' or 'news(package="maps")'. #
##
## Attaching package: 'maps'
## The following object is masked from 'package:plyr':
##
##
       ozone
require(ggmap) # for heatmap
```

Read file

Loading required package: ggmap

```
# Read Data and function files
LC <- read.csv(file.choose(), header = T) #Please read LC_biz_all.csv
source("clean_f2.R") # functions for cleaning the data
source("plot_f2.R") # functions for plot(heatmap)
source("analysis_f2.R") # functions for plot and data analysis</pre>
```

Data clean

Data Summary

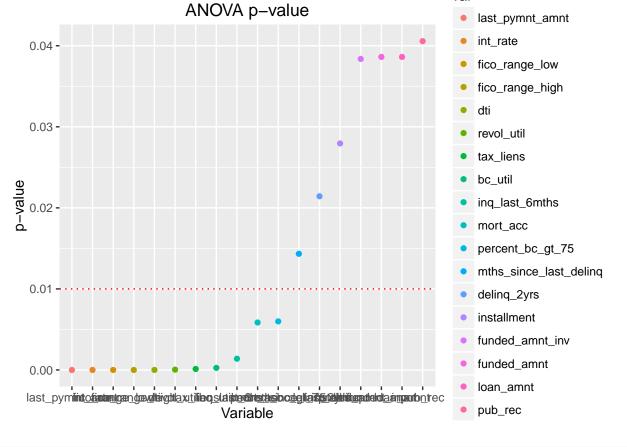
```
n.factor_all(lc)
     term emp_length home_ownership verification_status loan_status purpose
##
## 1
        2
                                   3
##
    addr_state
## 1
hasNA_all(lc)
##
        mths_since_last_delinq
                                     mths_since_last_record
##
##
                    revol_util mths_since_last_major_derog
##
                                                        3674
##
                       bc_util
                                       mths_since_recent_bc
##
##
      mths_since_recent_bc_dlq
                                      mths_since_recent_inq
##
                           3932
##
              num_tl_120dpd_2m
                                           percent_bc_gt_75
##
                            275
```

ANOVA

```
lc1 <- lc[, !(colnames(lc) %in% c(names(n.factor_all(lc))))]
lc1$loan_status <- lc$loan_status
lc1 <- lc1 %>% subset(!loan_status == "Current")
anova.p <- data.frame(var = names(anova(lc1)), p.value = anova(lc1))
anova.p_plot <- anova.p %>% subset(p.value < 0.05) %>%
    mutate(var = reorder(x = var, X = p.value, min)) %>%
    ggplot(aes(x = var, y = p.value, color = var)) +
```

```
geom_point() +
geom_hline(yintercept = 0.01, colour = "red", linetype = 3) +
    labs(x = "Variable", y = "p-value", title = "ANOVA p-value")
anova.p_plot
```

vai



```
sig.names <- as.character((anova.p %>% subset(p.value < 0.01))$var)
var.names <- c(sig.names, names(n.factor_all(lc)))</pre>
```

subset the data

```
lc2 <- lc[,c(var.names)]
lc2 <- lc2 %>% select(-purpose)
lc.categorical <- lc2[,names(n.factor_all(lc2))]
lc.categorical <- lc.categorical %>% select(-loan_status)
lc.numeric <- lc2[,!names(lc2) %in% names(n.factor_all(lc2))]
# Transform the categorical column to multiple numeric columns
categorical.list <- apply(lc.categorical, 2, function(x) model.matrix(~ x + 0))
lc3 <- as.data.frame(cbind(categorical.list, lc.numeric))
lc3$loan_status <- lc$loan_status
# Group the loan_status into 3 groups
lc3 <- lc3 %>% mutate(y = !(loan_status == "Current"))
lc3$y[lc3$loan_status == "Fully_Paid"] <- 2
lc3 <- lc3 %>% select(-loan_status)
```

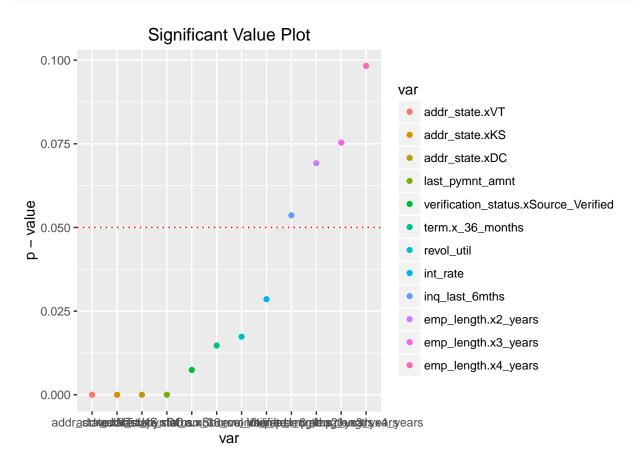
Logistic Regression

```
lc4 <- lc3 %>% subset(y != 0)
lc4$y[lc4$y == 2] <- 0
log.f <- logistic(lc4)</pre>
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
pred <- predict(log.f, lc4, type = "response")</pre>
## Warning in predict.lm(object, newdata, se.fit, scale = 1, type =
## ifelse(type == : prediction from a rank-deficient fit may be misleading
pred1 <- as.numeric(pred >= 0.5)
# in-sample prediction accuracy
1 - sum(pred1 == lc4$y, na.rm = T)/length(pred1)
## [1] 0.06681191
# out-of-sample prediction accracy
cv_error <- cv_k(lc4)</pre>
## Warning: glm.fit: algorithm did not converge
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## Warning in predict.lm(object, newdata, se.fit, scale = 1, type =
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```

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cv_error
```

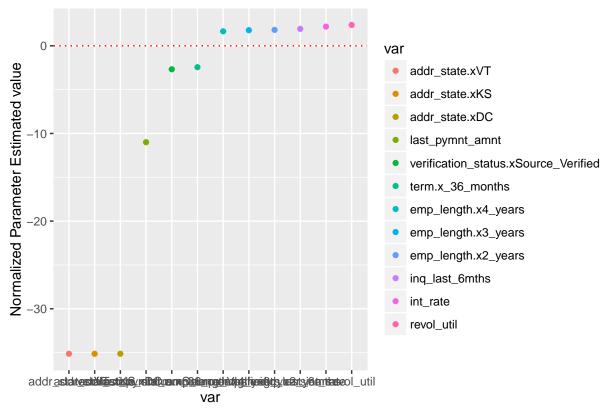
[1] 0.0879505

```
#Plot the significant value
significant_plot <- coeff_plot1(log.f, sig = T, 0.1)
print(significant_plot)</pre>
```



```
#Plot the coefficient value
coefficient_plot <- coeff_plot1(log.f, sig = F, 0.1)
print(coefficient_plot)</pre>
```

Normalized Parameter Estimated Value Plot



Kernel SVM

```
# Prepare for kernel SVM
lc4 <- lc4[-which(is.na(lc4$percent_bc_gt_75) | is.na(lc4$revol_util) | is.na(lc4$bc_util)),]</pre>
1.kern <- svm(lc4, "vanilladot")</pre>
    Setting default kernel parameters
## Warning in .local(x, ...): Variable(s) `' constant. Cannot scale data.
1.kern
## Support Vector Machine object of class "ksvm"
##
## SV type: C-svc (classification)
    parameter : cost C = 1
##
##
## Linear (vanilla) kernel function.
##
## Number of Support Vectors : 193
##
## Objective Function Value : -207.8558
## Training error: 0.051546
```

```
# New data
newdata <- as.matrix(lc4%>% select(-y))
pred.kern <- predict(l.kern, newdata)</pre>
# in-sample prediction accuracy
1 - sum(pred.kern == lc4$y, na.rm = T)/length(lc4$y)
## [1] 0.05154639
# cross validation error
cv.kernel.error <- cv_ksvm(lc4, "vanilladot", k = 10)</pre>
   Setting default kernel parameters
## Warning in .local(x, ...): Variable(s) `' constant. Cannot scale data.
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## Warning in .local(x, ...): Variable(s) `' constant. Cannot scale data.
   Setting default kernel parameters
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```

cv.kernel.error

[1] 0.05745098