E-M classification

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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) # data clean
## Loading required package: reshape2
require(datasets) # State data
require(caret) # Cross - Validation k fold
## Loading required package: caret
## Loading required package: lattice
require(RCurl) # load the website link
## Loading required package: RCurl
## Loading required package: bitops
## Attaching package: 'RCurl'
## The following object is masked from 'package:tidyr':
##
##
       complete
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

## Loading required package: ggmap

require(gridExtra)

## Loading required package: gridExtra

##
## Attaching package: 'gridExtra'

## The following object is masked from 'package:dplyr':
##
## combine
```

Read file

```
# 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>
```

Prepare for E-M algorithm

```
# Data for 36 months term
lc <- perToN(LC)  # transfer percentage to numeric
lc <- replaceBlank_all(lc) # replace the blank
# subset to 36 months
lc36 <- lc %>% subset(term == "_36_months") %>% droplevels()
# Transform earliest_cr_line, issue_d, last_pymnt_d from str to numeric
lc36 <- lc36 %>% dateToNum(., "earliest_cr_line") %>%
    dateToNum(., "issue_d") %>% dateToNum(., "last_pymnt_d")
lc36 <- lc36 %>% mutate(y1 = last_pymnt_d - issue_d, y2 = !(loan_status == "Fully_Paid")) %>% select(-c
# Select the important variables with ANOVA
lc1 <- lc36[, !(colnames(lc36) %in% c(names(n.factor_all(lc36))))]
lc1$loan_status <- lc36$loan_status
anova.p <- data.frame(var = names(anova(lc1)), p.value = anova(lc1))
sig.names <- as.character((anova.p %>% subset(p.value < 0.01))$var)
var.names <- c(sig.names, names(n.factor_all(lc36)))</pre>
```

```
1c36 <- 1c36[,c(var.names)] %>% select(-emp_title,-last_pymnt_d,-issue_d,
                                         -next_pymnt_d,-last_pymnt_amnt, -last_credit_pull_d, -zip_code)
# Set the loan_status "Charged_Off" with a large y value(100)
lc36$y1[lc36$loan_status == "Charged_Off"] = 61
lc36 <- lc36 %>% select(-loan_status)
lc.categorical <- lc36[,names(n.factor_all(lc36))]</pre>
lc.numeric <- lc36[,!names(lc36) %in% names(n.factor_all(lc36))]</pre>
# Transform the categorical column to multiple numeric columns
categorical.list <- apply(lc.categorical, 2, function(x) model.matrix(~ x + 0))</pre>
lc36 <- as.data.frame(cbind(categorical.list, lc.numeric))</pre>
# delete all the na
hasNA_all(1c36)
## revol_util
                      у1
lc36 <- lc36[!is.na(lc36$revol_util),]</pre>
lc36 <- lc36[!is.na(lc36$y1),] # One data missing y1
```

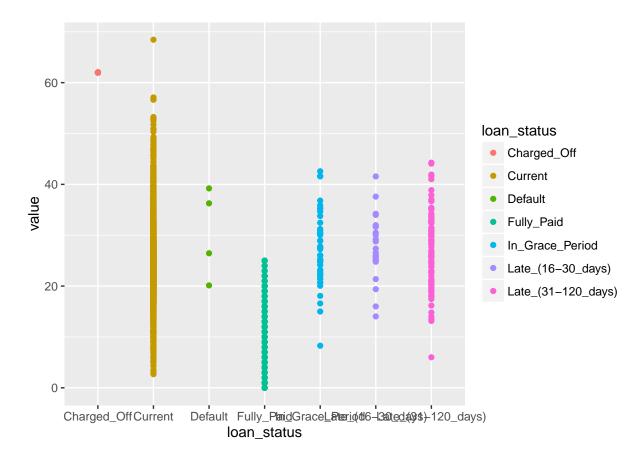
Use the EM algorithm to classify current and potential delinquency events(linear regression)

```
lc_em1 \leftarrow lc36
for (i in 1:20){
 k <- M_step(lc_em1)
 lc_em1 <- E_step(k,lc_em1)</pre>
}
## Warning in predict.lm(k, newdata): prediction from a rank-deficient fit may
## be misleading
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```

```
## Warning in predict.lm(k, newdata): prediction from a rank-deficient fit may
## be misleading
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## be misleading
## Warning in predict.lm(k, newdata): prediction from a rank-deficient fit may
## be misleading
# performance
true <- lc %>% subset(id %in% lc_em1$id) %>% select(loan_status)
compare.data <- data.frame(y1 = lc_em1$y1, y2 = lc_em1$y2, true = true) %>%
  mutate(y = y1 + y2)
plotdata <- compare.data %>% select(-y1,-y2) %>%
  melt()
## Using loan_status as id variables
totallm_plot <- plotdata %>% ggplot() +
  geom_point(aes(x = loan_status, y = value, color = loan_status))
print(totallm_plot)
```

Warning in predict.lm(k, newdata): prediction from a rank-deficient fit may

be misleading

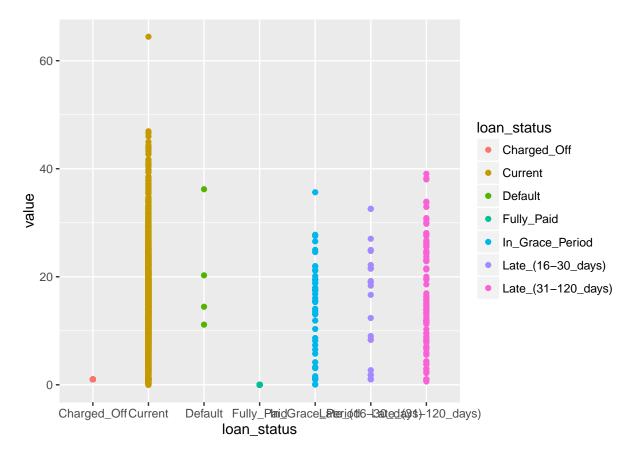


```
ggsave(filename = "totallm.png", plot = totallm_plot, path = ".",
    width = 10, height = 6, dpi = 400)

plotdata <- compare.data %>% select(-y1,-y) %>%
    melt()
```

Using loan_status as id variables

```
censorlm_plot <- plotdata %>% ggplot() +
  geom_point(aes(x = loan_status, y = value, color = loan_status))
print(censorlm_plot)
```



```
ggsave(filename = "censorlm.png", plot = censorlm_plot, path = ".",
    width = 10, height = 6, dpi = 400)
```

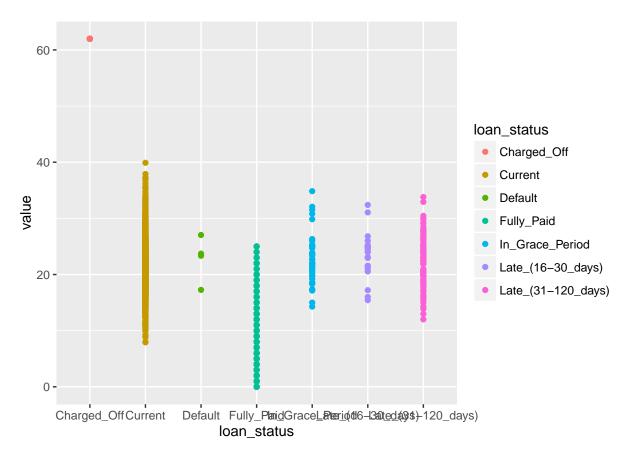
Use the EM algorithm to classify current and potential delinquency events(kernel SVM)

```
lc_em2 <- lc36
for (i in 1:20){
    k <- M_step_kern(lc_em2, kernel = "rbfdot")
    lc_em2 <- E_step_kern(k,lc_em2)
}

# performance
true <- lc %>% subset(id %in% lc_em2$id) %>% select(loan_status)
compare.data <- data.frame(y1 = lc_em2$y1, y2 = lc_em2$y2, true = true) %>%
    mutate(y = y1 + y2)
plotdata <- compare.data %>% select(-y1,-y2) %>%
    melt()
```

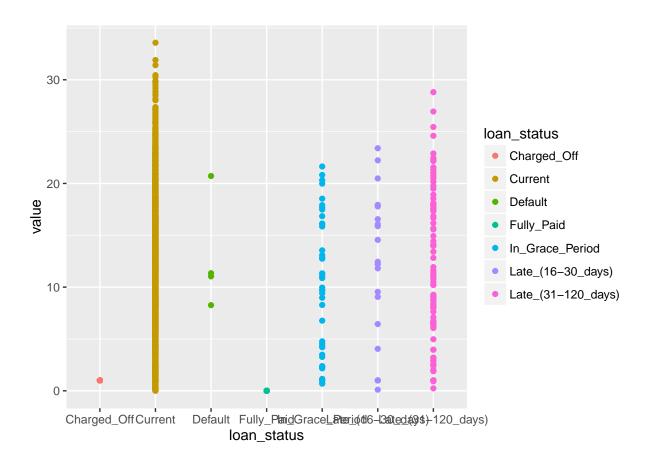
Using loan_status as id variables

```
totalsvm_plot <- plotdata %>% ggplot() +
   geom_point(aes(x = loan_status, y = value, color = loan_status))
print(totalsvm_plot)
```



Using loan_status as id variables

```
censorsvm_plot <- plotdata %>% ggplot() +
   geom_point(aes(x = loan_status, y = value, color = loan_status))
print(censorsvm_plot)
```



```
ggsave(filename = "censorsvm.png", plot = censorsvm_plot, path = ".",
    width = 10, height = 6, dpi = 400)
```