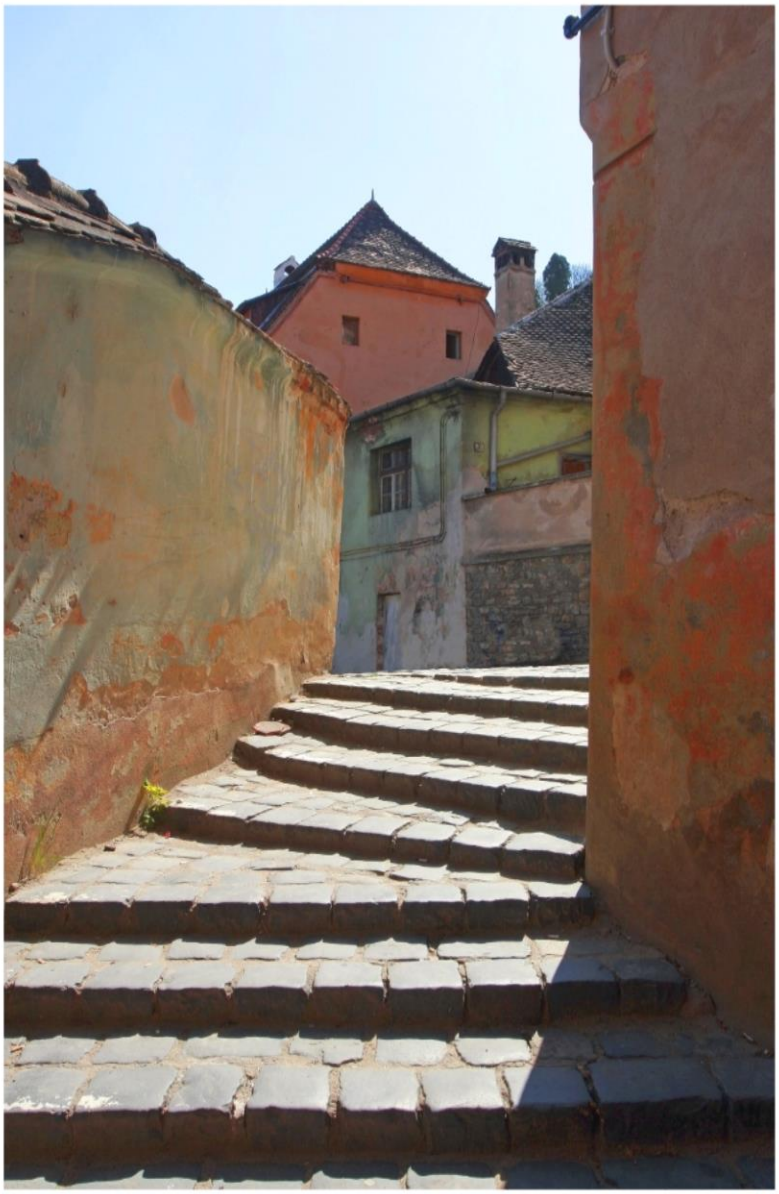

Detroit Blight Analysis

CAP STONE

Bala Subrahmanyam | No Restrictions | March 21, 2017

Contents

Get Started Right Away

This project is to analyze Detroit blight data to find appropriate model to predict future demolish judgments. The issue is due to various factors like crime, environment maintenance issues, neglecting properties due to people moving out of the town and poverty. We have different dimension data associate with a location like citizens reported issues and agency recorded incidents along with crime data.

Figure 1 : Data Dimensions

Data Representation and Visualization:

R programming google maps used to plot independent dataset as well as density overlapping to find correlation between datasets.

https://github.com/tubatibs/DsasVisualAnalytics/blob/master/DsasCapstone-DetroitBlightViolation.ipynb

Visualization to eliminate outliers, used CARTO and Microsoft excel power view to prepare data out of 311 calls and crime data sets as layers. Merged blight violation data set on those units to prepare data.

https://tubatibs.carto.com/builder/cf461ff8-c349-11e6-85e3-0ecd1babdde5/embed

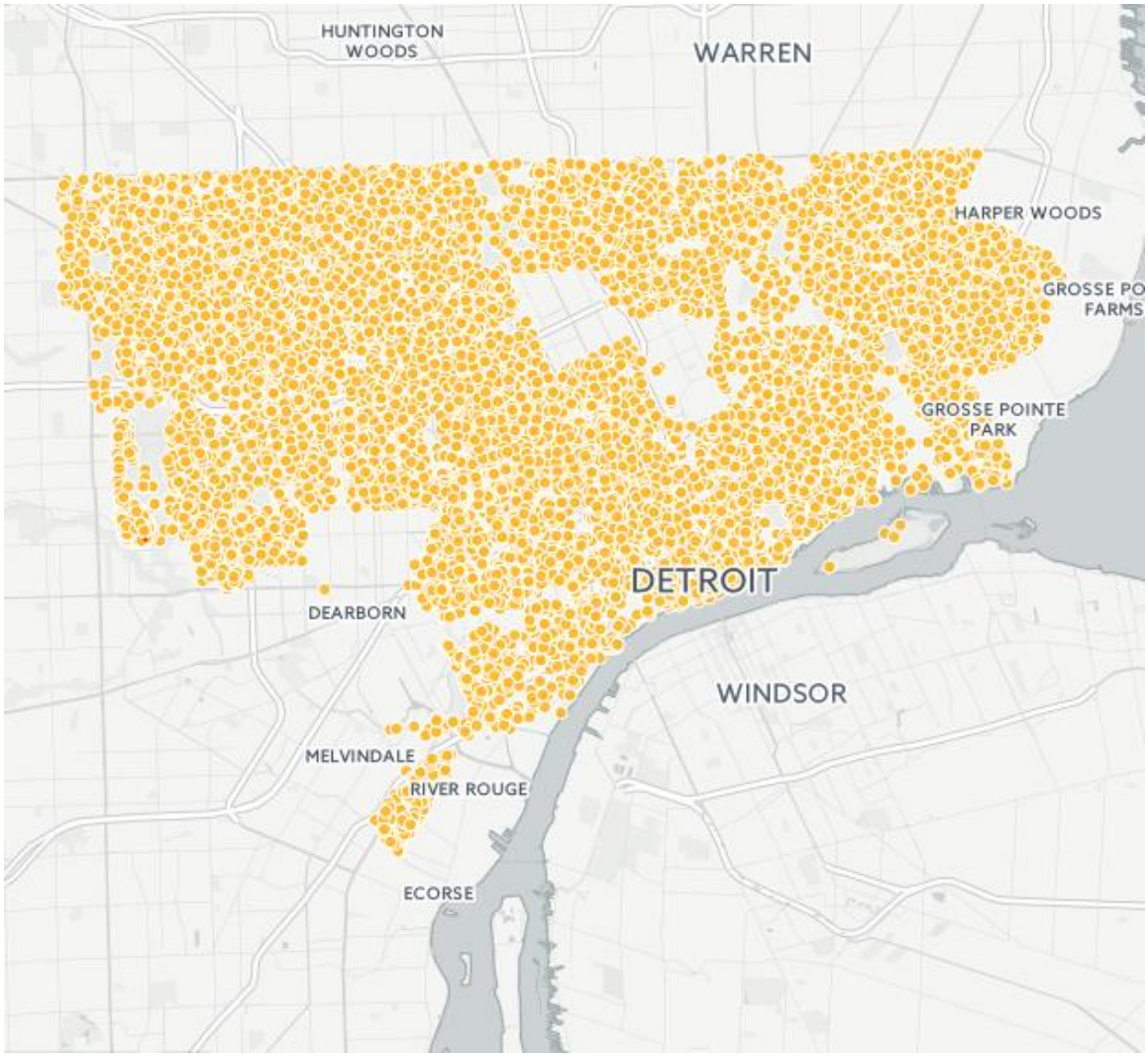

Figure 2: Data visualization and analysis

Understanding Data:

detroit-blight-violations.csv : Each record is a blight violation incident. Contains data associated to location observed and raised by different agencies.

detroit-demolition-permits.tsv: Each record represents a permit for a demolition. Contains fees and associated financial parties with dates.

detroit-311.csv: Each record represents a 311 call, typically a complaint. Contains citizen reported issues.

detroit-crime.csv: Each record represents a criminal incident. Contains location and details about criminal activity.

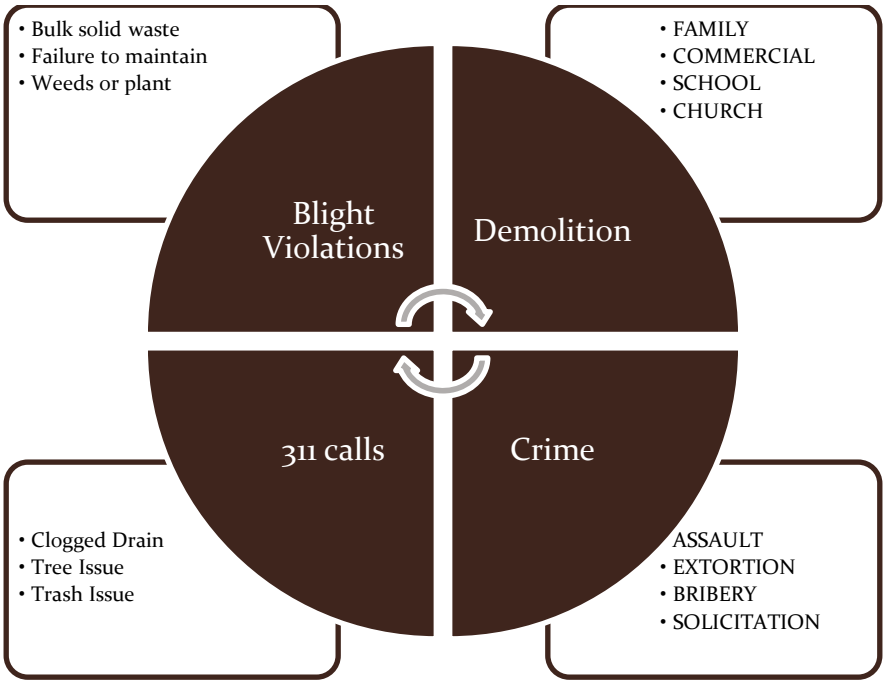

Figure 3: Data matrix with few items

Steps to capture data:

1. Filter out noisy data in all four files.
2. Extract latitude and longitude incase if it is part of address.
3. Crime data join with Blight violations aggregated crime count associated with location.
4. Join with 311 calls data with count of issue type and sum of ratings.
5. Validated consolidated data for any NaN or special characters.

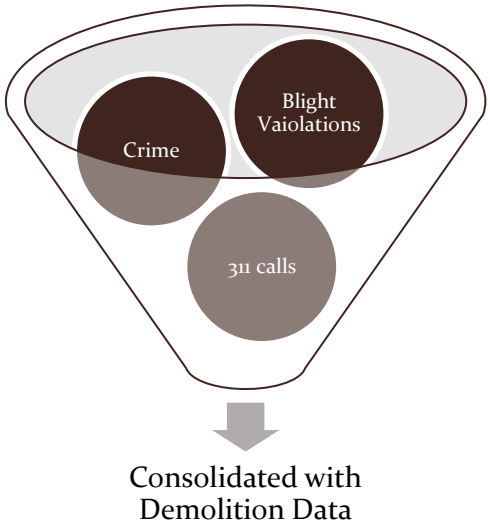

Figure 4: Data funnel representation

Training Dataset:

Mapped demolition permits records on above consolidated data set to derive demolished buildings. Other records will be treated as still in-use. Explored different models to validate predictions.

Evaluation of Advanced Predictive Data Models:

Data cleansing took long time to identify the important attributes with respect to available data. Using R programming to evaluate models. Explored predictive models as follows

- chi-squre analysis model :
  - dchisq gives the density, applied with demolished with crime data.
  - pchisq gives the distribution function, applied with demolished with crime data.
  - qchisq gives the quantile function, applied with demolished with crime data.
  - rchisq generates random deviates, applied with demolished with crime data.

- Survival analysis model:
  - Define a survival object as blighted building and fit a survival to a model.
    Surv(
    time = Blight violation as start event.
    time2 = Demolition time as end event.
    event = Demolition)

- Random Forest Model:
  - *Used below formula to run Random Forest Model*
    fol <- formula(m_Demolished ~ m_dbv_ngbr_8_ct + m_ddp_ngbr_8_ct + m_d311_ngbr_8_ct + s_CleanUpCost + s_JudgmentAmt + m_PaymentStatus + s_FineAmt + s_LateFee + m_AgencyName)

- XGBoost
    - Included only few features to evaluate XGBoost model "m_dbv_ngbr_8_ct","m_ddp_ngbr_8_ct","m_d311_ngbr_8_ct","m_ViolationCategory","s_LateFee","s_FineAmt","s_JudgmentAmt","s_CleanUpCost","m_PaymentStatus","m_AgencyName","m_Demolished"
      Configured number of rounds as 10, eval_metric as "mlogloss" or "merror" with objective as "multi:softprob". Generated importance matrix to find features.

Evaluated error/loss rate for each model. Observed models fit descending order as follows.

Chi squre > Survival > Random Forest Model > XGBoost

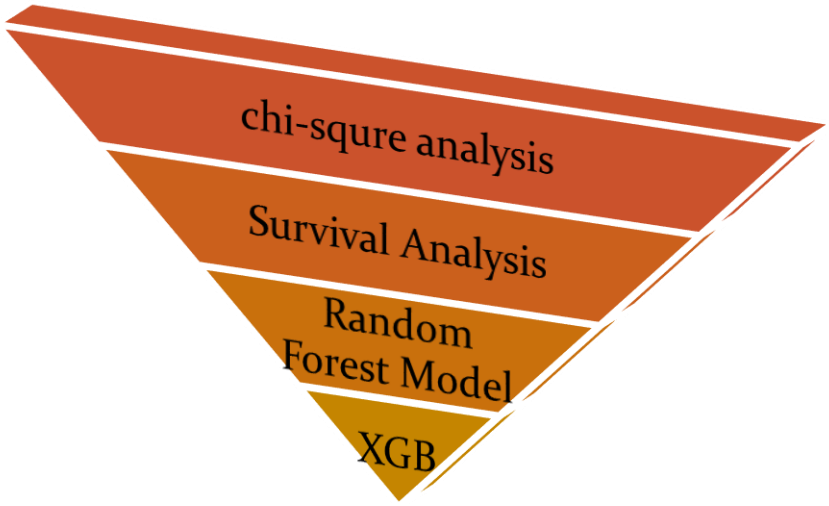

Figure 5: Inverted Pyramid depicts model fitness.

May need to repeat data and features evaluations for change in error rate. XGB model shows mlogloss values for 10 rounds as follows

[1] train-mlogloss:0.551635
[2] train-mlogloss:0.495870
[3] train-mlogloss:0.448311
[4] train-mlogloss:0.404216
[5] train-mlogloss:0.364788
[6] train-mlogloss:0.329524
[7] train-mlogloss:0.297907
[8] train-mlogloss:0.269515
[9] train-mlogloss:0.243987
[10] train-mlogloss:0.221007

With predictions

[1] 0.733373582 0.007193833 0.002647409 0.002647450 0.002647434 0.002647098

Please refer jupyter notebook for reproducible research.

https://github.com/tubatibs/DsasVisualAnalytics/blob/master/DsasCapstone-DetroitBlightViolation.ipynb

Optimizing features to improve accuracy:

Optimized above formula with important attributes identified by XGB model. But didn't observe much difference in statistics.

```
[1]     train-mlogloss:0.704789
[2]     train-mlogloss:0.516291
[3]     train-mlogloss:0.467244
[4]     train-mlogloss:0.423403
[5]     train-mlogloss:0.384140
[6]     train-mlogloss:0.346362
[7]     train-mlogloss:0.314879
[8]     train-mlogloss:0.286533
[9]     train-mlogloss:0.259020
[10]    train-mlogloss:0.234351
```

If I have enough time, I would like to

1. Evaluate how it works with derived sentiment values based on 311 call description and crime offence description.
2. Dig data deep into owner address outside Detroit and outside USA.
3. Would like to analyze housing, Job market and Schools.
4. Adding more dimensions data will help predict accurately.

Conclusion:

Looks like XGBoost model fits perfectly with multi-dimensional data provided here. Random Forest can come close and may need to explore with different features.

There was a clear indication that high crime rate and lack of maintenance contributes to blight, which costs billions of dollars to fix. Thanks for reading and I personally enjoyed analyzing blight data and learning R programming.

BIBLIOGRAPHY

https://www.theguardian.com/money/2014/sep/28/detroit-demolish-ruins-capitalists-abandoned-buildings-plan
https://en.wikipedia.org/wiki/Decimal_degrees

APPENDIX - Reproducible Research (Jupyter notebook)

https://github.com/tubatibs/DsasVisualAnalytics/blob/master/DsasCapstone-DetroitBlightViolation.ipynb

Please note that below is the sudo code. Please refer github jupyter notebook link above for latest version.

In [1]:

```
# Including libraries
library(dplyr)
library(tidyr)
library(stringr)
library(geohash)
```

In [2]:

```
# Download data from the links provided
library(downloader)
# Blight Violations
dbvurl <- "https://d18ky98rnyall9.cloudfront.net/_97bd1c1e5df9537bb13398c9898deed7_detroi
t-blight-violations.csv?Expires=1487808000&Signature=gHUOfvUDTW-h~HuT0YXBeapK~jrxVV~G~ItL
JCGvxfndaU-ZnP0OI1lurvBMMbxRy3JymGjsyrfMZvY8uXkywWwOeRT3JzfyXftxHVbDpw6rRsfOgmR0Bwu6HHIbc
SSANPjFG9p6FwpQh1YyJUKvMj8IQCoaanPuG10SRLWg7Bc_&Key-Pair-Id=APKAJLTNE6QMUY6HBC5A"
DBVFileName <- "detroit-blight-violations.csv"
if (!file.exists(DBVFileName))
    download(dbvurl, DBVFileName)
detBlightViol <- read.csv("detroit-blight-violations.csv", stringsAsFactors = FALSE,
                  strip.white = TRUE, na.strings ='' )

# Detroit Demolition Permits
ddpurl <- "https://d18ky98rnyall9.cloudfront.net/_dcebfb2135a2bf5a6392493bd61aba22_detroi
t-demolition-permits.tsv?Expires=1487808000&Signature=GIkiK8yRf70Fya8VNatb9t~lXh5VD4kX05G
ZMKlqb3l2lX-z9aXk4okJQao6dOfApCcdgM~-6L3KlBG1YKhFbCKQqagI2ALjFt-PTkJeCyfPFSQ5FqMcQlh7qUCl
pZCH7F~zJA9X-vutv6IIaS-tKt22sAGgwu9X6lCtw6raPBo_&Key-Pair-Id=APKAJLTNE6QMUY6HBC5A"
DDPFileName <- "detroit-demolition-permits.tsv"
if (!file.exists(DDPFileName))
    download(ddpurl, DDPFileName)
detDemolitionPermit <- read.delim("detroit-demolition-permits.tsv", header=TRUE, allowEsc
apes=FALSE, sep="\t", na.strings="", comment.char="")


# Detroit 311 calls
d311url <- "https://d18ky98rnyall9.cloudfront.net/_dcebfb2135a2bf5a6392493bd61aba22_detro
it-311.csv?Expires=1487808000&Signature=PTTniMA9xRitX6DycZSOORb45gCHgqeHDQABaMn54N6CswNJm
```

```
FIXEYolvrWNCXlp~K4gn9zaSUOm27eQJhmEg4n7FUTJ5ZrWnVWBOfFRxouPgcglrdqUHfx-HKqnTMByfTLcEPmEqh
ZLKg7d9SLYsx4Cc2vwxCFshMjhpEF7ZwA_&Key-Pair-Id=APKAJLTNE6QMUY6HBC5A"
D311FileName <- "detroit-311.csv"
if (!file.exists(D311FileName))
    download(d311url, D311FileName)
det311 <- read.csv("detroit-311.csv", stringsAsFactors = FALSE,
                   strip.white = TRUE, na.strings ='' )

# Detroit Crime
dcrurl <- "https://d18ky98rnyall9.cloudfront.net/_dcebfb2135a2bf5a6392493bd61aba22_detroi
t-crime.csv?Expires=1487808000&Signature=POU~pk3A00i-iFJpAT9ytnBpfygEdMrPcITocHFPPHbyHegk
i~dcECrUD1kWApMOYmymWt2Vrm5c5mWmKG1pwpIaMPLwFGkF5kUkMTgCLuask2b0LnKcDOI86WzRYmkBsN2VvSQxX
NK9y8CvCs2pUVzmwYbwGwstsOqbZei-Ohg_&Key-Pair-Id=APKAJLTNE6QMUY6HBC5A"
DCRFileName <- "detroit-crime.csv"
if (!file.exists(DCRFileName))
    download(dcrurl, DCRFileName)
detCrime <- read.csv("detroit-crime.csv", stringsAsFactors = FALSE,
                   strip.white = TRUE, na.strings ='' )
```

In [3]:

```
#dplyr mutate to extract (LAT,LNG) from ViolationAddress.
detBlightViol <- detBlightViol %>%
    #filter(TicketIssuedDT > "2005-01-01" & TicketIssuedDT <"2018-01-01") %>%
    mutate(latlng = gsub(".*\\n", "", ViolationAddress)) %>%
    transform(latlng = gsub('[()]','', latlng)) %>%
    filter(latlng != 'character(0)') %>%
    transform(latlng = gsub('[()]','', latlng)) %>%
    separate(latlng, c('LAT', 'LNG'), ', ')

#dplyr mutate to extract (LAT,LNG) from site_location
detDemolitionPermit <- detDemolitionPermit %>%
    mutate(latlng = gsub(".*\\n", "", site_location)) %>%
    transform(latlng = gsub('[()]','', latlng)) %>%
    filter(latlng != 'character(0)') %>%
    transform(latlng = gsub('[()]','', latlng)) %>%
    separate(latlng, c('LAT', 'LNG'), ', ')
```

In [4]:

```
#Visualize to glance through the data
library(ggplot2)
library(ggmap)
library(maps)
library(maptools)

# get Google map to plot the data into it
```

```
detroit_map <- get_map(location = "Detroit", zoom = 11,
                       maptype = "t", scale = 2)
#Crime incidents
ggmap(detroit_map) +
  geom_point(data = detCrime, aes(x = detCrime$LON, y = detCrime$LAT, fill = "red", alpha
 = 0.8), size = 2, shape = 17) +
  guides(fill=FALSE, alpha=FALSE, size=FALSE)
```

In [5]:

```
#311 Calls
ggmap(detroit_map) +
  geom_point(data = det311, aes(x = det311$lng, y = det311$lat, fill = "green", alpha = 0
.8), size = 2, shape = 25) +
  guides(fill=FALSE, alpha=FALSE, size=FALSE)
```

In [6]:

```
#Blight Violation
blight_viols <- detBlightViol %>%
  transform(LNG = as.numeric(LNG),
            LAT = as.numeric(LAT))

ggmap(detroit_map) +
  geom_point(data = blight_viols, aes(x = blight_viols$LNG, y = blight_viols$LAT, alpha =
 0.1), size = 1, shape = 1) +
  guides(fill=FALSE, alpha=FALSE, size=FALSE)
```

In [7]:

```
#Demolition permits
demo_permits <- detDemolitionPermit %>%
  transform(LNG = as.numeric(LNG),
            LAT = as.numeric(LAT))
ggmap(detroit_map) +
  geom_point(data = demo_permits, aes(x = LNG, y = LAT, fill = "yellow", alpha = 0.8), si
ze = 2, shape = 13) +
  guides(fill=FALSE, alpha=FALSE, size=FALSE)
```

In [8]:

```
#Plot Blight violations with Demolition permits
ggmap(detroit_map) + #geom_point(data=call311_dat, aes(x=lng, y=lat), color="dark green",
 alpha=.1, size=1.1) +
  geom_density2d(data=blight_viols, aes(x=blight_viols$LNG, y=blight_viols$LAT), size = 0
.3) +
  stat_density2d(data=demo_permits, aes(x=demo_permits$LNG, y=demo_permits$LAT, fill = ..
level.., alpha = ..level..), size = 0.01, bins = 16, geom = "polygon") +
  scale_fill_gradient(low = "green", high = "red") +
```

```
  scale_alpha(range = c(0, 0.3), guide = FALSE) + labs(title="Blight Violation vs Demolit
ion")
```

In [9]:

```
#dplyr filter to create gh_8 and uid colums
detBlightViol <- detBlightViol %>%
    mutate(gh_8 = gh_encode(as.numeric(LAT), as.numeric(LNG), 8),
           uid = paste0('blightviol_',row_number()))

detDemolitionPermit <- detDemolitionPermit %>%
    mutate(gh_8 = gh_encode(as.numeric(LAT), as.numeric(LNG), 8),
           uid = paste0('demolper',row_number()))

det311 <- det311 %>%
    mutate(gh_8 = gh_encode(as.numeric(lat), as.numeric(lng), 8),
           uid = paste0('det311',row_number()))

detCrime <- detCrime %>%
    mutate(gh_8 = gh_encode(as.numeric(LAT), as.numeric(LON), 8),
           uid = paste0('detcrim',row_number()))
```

In [10]:

```
#Summarize records based on gh_8
dbv_gh_8_grp <- detBlightViol %>%
    group_by(gh_8) %>%
    summarize(dbv_ngbr_8_ct = n()) %>%
    arrange(gh_8,desc(dbv_ngbr_8_ct))

#mutate(PERMIT_ISSUED=min(as.Date(detDemolitionPermit$PERMIT_ISSUED,format='%m/%d/%Y'), n
a.rm=TRUE)) %>%
ddp_gh_8_grp <- detDemolitionPermit %>%
    group_by(gh_8) %>%
    summarize(ddp_ngbr_8_ct = n()) %>%
    arrange(gh_8,desc(ddp_ngbr_8_ct))

#Summarize dates for survival analysis
ddp_gh_8_grp_d <- detDemolitionPermit %>%
    group_by(gh_8) %>%
    summarize(M_PERMIT_ISSUED=mean(as.numeric(PERMIT_ISSUED,format='%m/%d/%Y'), na.rm=TRU
E))

d311_gh_8_grp <- det311 %>%
    group_by(gh_8) %>%
    summarize(d311_ngbr_8_ct = n()) %>%
    arrange(gh_8,desc(d311_ngbr_8_ct))
```

```
dcr_gh_8_grp <- detCrime %>%
    group_by(gh_8) %>%
    summarize(dcr_ngbr_8_ct = n()) %>%
    arrange(gh_8,desc(dcr_ngbr_8_ct))
```

In [12]:

```
#join counts and demolished dates based on gh_8
detBlightViolV <- left_join(detBlightViol, dbv_gh_8_grp, by = c("gh_8" = "gh_8"))
detBlightViolVD <- left_join(detBlightViolV, ddp_gh_8_grp, by = c("gh_8" = "gh_8"))
detBlightViolVDD <- left_join(detBlightViolVD, ddp_gh_8_grp_d, by = c("gh_8" = "gh_8"))
detBlightViolVD3 <- left_join(detBlightViolVDD, d311_gh_8_grp, by = c("gh_8" = "gh_8"))
detBlightViolVD3C <- left_join(detBlightViolVD3, dcr_gh_8_grp, by = c("gh_8" = "gh_8"))
```

In [13]:

```
#Keeping only the colums required for analysis
keep <- c("LAT","LNG","gh_8", "dbv_ngbr_8_ct", "ddp_ngbr_8_ct", "d311_ngbr_8_ct", "dcr_ng
br_8_ct",  "ViolationCategory", "CleanUpCost", "JudgmentAmt", "PaymentStatus", "FineAmt",
 "AdminFee", "LateFee", "AgencyName", "TicketIssuedDT", "M_PERMIT_ISSUED")
detBlightViolVD3C <- detBlightViolVD3C[keep]

# Assing values
detBlightViolVD3C$PaymentStatus[detBlightViolVD3C$PaymentStatus=="PAID IN FULL"] <- 1
detBlightViolVD3C$PaymentStatus[detBlightViolVD3C$PaymentStatus=="PARTIAL PAYMENT MADE"]
<- 2
detBlightViolVD3C$PaymentStatus[detBlightViolVD3C$PaymentStatus=="NO PAYMENT APPLIED"] <-
 3
detBlightViolVD3C$PaymentStatus[detBlightViolVD3C$PaymentStatus=="NO PAYMENT ON RECORD"]
<- 4

detBlightViolVD3C$PaymentStatus <- as.numeric(detBlightViolVD3C$PaymentStatus)

detBlightViolVD3C$AgencyName[detBlightViolVD3C$AgencyName=="Department of Public Works"]
<- 1
detBlightViolVD3C$AgencyName[detBlightViolVD3C$AgencyName=="Building and Safety Engineeri
ng Department"] <- 2
detBlightViolVD3C$AgencyName[detBlightViolVD3C$AgencyName=="Health Department"] <- 3
detBlightViolVD3C$AgencyName[detBlightViolVD3C$AgencyName=="Detroit Police Department"] <
- 4

#Data Cleanup
detBlightViolVD3C$AgencyName[is.na(detBlightViolVD3C$AgencyName)] <- as.numeric(0)
detBlightViolVD3C$AgencyName <- as.numeric(detBlightViolVD3C$AgencyName)
detBlightViolVD3C$AgencyName[is.na(detBlightViolVD3C$AgencyName)] <- as.numeric(0)
```

```
Warning message in eval(expr, envir, enclos):
"NAs introduced by coercion"
```

In [14]:

```
# Data preparation to avoid non-numerics from the dataset
ifelse ((!is.na(detBlightViolVD3C$CleanUpCost) || detBlightViolVD3C$CleanUpCost != ""),
    detBlightViolVD3C$CleanUpCost <- as.numeric(sub('\\$','',(as.character(detBlightViolV
D3C$CleanUpCost)))),
    detBlightViolVD3C$CleanUpCost <- as.numeric(0))

ifelse ((!is.na(detBlightViolVD3C$JudgmentAmt) || detBlightViolVD3C$JudgmentAmt != ""),
    detBlightViolVD3C$JudgmentAmt <- as.numeric(sub('\\$','',(as.character(detBlightViolV
D3C$JudgmentAmt)))),
    detBlightViolVD3C$JudgmentAmt <- as.numeric(0))
detBlightViolVD3C$JudgmentAmt[is.na(detBlightViolVD3C$JudgmentAmt)] <- as.numeric(0)

ifelse ((!is.na(detBlightViolVD3C$FineAmt) || detBlightViolVD3C$FineAmt != ""),
    detBlightViolVD3C$FineAmt <- as.numeric(sub('\\$','',(as.character(detBlightViolVD3C$
FineAmt)))),
    detBlightViolVD3C$FineAmt <- as.numeric(0))
detBlightViolVD3C$FineAmt[is.na(detBlightViolVD3C$FineAmt)] <- as.numeric(0)

ifelse ((!is.na(detBlightViolVD3C$AdminFee) || detBlightViolVD3C$AdminFee != ""),
    detBlightViolVD3C$AdminFee <- as.numeric(sub('\\$','',(as.character(detBlightViolVD3C
$AdminFee)))),
    detBlightViolVD3C$AdminFee <- as.numeric(0))

ifelse ((!is.na(detBlightViolVD3C$LateFee) || detBlightViolVD3C$LateFee != ""),
    detBlightViolVD3C$LateFee <- as.numeric(sub('\\$','',(as.character(detBlightViolVD3C$
LateFee)))),
    detBlightViolVD3C$LateFee <- as.numeric(0))

detBlightViolVD3C$ddp_ngbr_8_ct[is.na(detBlightViolVD3C$ddp_ngbr_8_ct)] <- as.numeric(0)
detBlightViolVD3C$d311_ngbr_8_ct[is.na(detBlightViolVD3C$d311_ngbr_8_ct)] <- as.numeric(0
)
detBlightViolVD3C$dcr_ngbr_8_ct[is.na(detBlightViolVD3C$dcr_ngbr_8_ct)] <- as.numeric(0)

detBlightViolVD3C$Demolished[detBlightViolVD3C$ddp_ngbr_8_ct >0 ] <- 1
detBlightViolVD3C$Demolished[detBlightViolVD3C$ddp_ngbr_8_ct <=0 ] <- 0

ifelse ((!is.na(detBlightViolVD3C$TicketIssuedDT) || detBlightViolVD3C$TicketIssuedDT !=
""),
        detBlightViolVD3C$TicketIssuedDT <- as.numeric(as.Date(detBlightViolVD3C$TicketIs
suedDT,"%m/%d/%Y %H:%M:%S")),
        detBlightViolVD3C$TicketIssuedDT <- as.numeric(0))
```

```
ifelse ((!is.na(detBlightViolVD3C$M_PERMIT_ISSUED) || detBlightViolVD3C$M_PERMIT_ISSUED !
= ""),
        detBlightViolVD3C$M_PERMIT_ISSUED <- as.numeric(detBlightViolVD3C$M_PERMIT_ISSUED
),
        detBlightViolVD3C$M_PERMIT_ISSUED <- as.numeric(0))

nrow(detBlightViolVD3C)
summary(detBlightViolVD3C)
```

```
0
1680
1500
20
150
[1] NA
[1] NA
307804
```

```
     LAT                LNG                gh_8
 Length:307804      Length:307804      Length:307804
 Class :character   Class :character   Class :character
 Mode  :character   Mode  :character   Mode  :character

dbv_ngbr_8_ct   ddp_ngbr_8_ct    d311_ngbr_8_ct    dcr_ngbr_8_ct
 Min.   :    1   Min.   :  0.00   Min.   : 0.0000   Min.   : 0.0000
 1st Qu.:    4   1st Qu.:  0.00   1st Qu.: 0.0000   1st Qu.: 0.0000
 Median :    7   Median :  0.00   Median : 0.0000   Median : 0.0000
 Mean   : 1562   Mean   : 38.37   Mean   : 0.0419   Mean   : 0.8085
 3rd Qu.:   14   3rd Qu.:  0.00   3rd Qu.: 0.0000   3rd Qu.: 0.0000
 Max.   :21114   Max.   :548.00   Max.   :65.0000   Max.   :59.0000

 ViolationCategory   CleanUpCost         JudgmentAmt      PaymentStatus
 Min.   :0.000000   Min.   :    0.000   Min.   :    0.0   Min.   :1.000
 1st Qu.:0.000000   1st Qu.:    0.000   1st Qu.:  140.0   1st Qu.:3.000
 Median :0.000000   Median :    0.000   Median :  305.0   Median :3.000
 Mean   :0.006553   Mean   :    0.515   Mean   :  422.5   Mean   :2.744
 3rd Qu.:0.000000   3rd Qu.:    0.000   3rd Qu.:  305.0   3rd Qu.:3.000
 Max.   :1.000000   Max.   :13123.800   Max.   :11030.0   Max.   :4.000

    FineAmt           AdminFee     LateFee          AgencyName
 Min.   :    0.0   Min.   :20   Min.   :   0.0   Min.   :0.000
 1st Qu.:  100.0   1st Qu.:20   1st Qu.:  10.0   1st Qu.:1.000
 Median :  250.0   Median :20   Median :  25.0   Median :2.000
 Mean   :  357.9   Mean   :20   Mean   :  35.8   Mean   :1.746
 3rd Qu.:  250.0   3rd Qu.:20   3rd Qu.:  25.0   3rd Qu.:2.000
 Max.   :10000.0   Max.   :20   Max.   :1000.0   Max.   :4.000
```

```
 TicketIssuedDT   M_PERMIT_ISSUED    Demolished
 Min.   :-11407   Min.   :  2.0    Min.   :0.0000
 1st Qu.: 13738   1st Qu.:375.6    1st Qu.:0.0000
 Median : 14264   Median :381.5    Median :0.0000
 Mean   : 14508   Mean   :385.1    Mean   :0.1619
 3rd Qu.: 15106   3rd Qu.:409.0    3rd Qu.:0.0000
 Max.   : 16650   Max.   :782.0    Max.   :1.0000
 NA's   :38854    NA's   :257958
```

In [15]:

```
#Summarize based on gh_8
detBlightViolVD3C <- detBlightViolVD3C %>%
      group_by(gh_8) %>%
      summarize(m_dbv_ngbr_8_ct=mean(dbv_ngbr_8_ct, na.rm=TRUE),
                m_ddp_ngbr_8_ct=mean(ddp_ngbr_8_ct, na.rm=TRUE),
                m_d311_ngbr_8_ct=mean(d311_ngbr_8_ct, na.rm=TRUE),
                m_ViolationCategory=max(ViolationCategory, na.rm=TRUE),
                s_CleanUpCost=sum(CleanUpCost, na.rm=TRUE),
                s_JudgmentAmt=sum(JudgmentAmt, na.rm=TRUE),
                m_PaymentStatus=mean(PaymentStatus, na.rm=TRUE),
                s_FineAmt=sum(FineAmt, na.rm=TRUE),
                s_LateFee=sum(LateFee, na.rm=TRUE),
                m_AgencyName=max(AgencyName, na.rm=TRUE),
                m_Demolished=max(Demolished, na.rm=TRUE),
                m_TicketIssuedDT= min(TicketIssuedDT,na.rm=TRUE ),
                M_M_PERMIT_ISSUED=min(M_PERMIT_ISSUED,na.rm=TRUE))
```

In [16]:

```
# Using SQL queries on dataset to understand data.
library(sqldf)
sqldf("select count(*) from 'detBlightViolVD3C' where m_Demolished=1")
```

```
Loading required package: gsubfn
Loading required package: proto
Loading required package: RSQLite
Loading required package: tcltk
```

count(*)
3142

In [17]:

```
# Training and test data
indexes = sample(nrow(detBlightViolVD3C), size=0.1*nrow(detBlightViolVD3C))
```

```
train = detBlightViolVD3C[indexes,]
test = detBlightViolVD3C[-indexes,]
```

In [18]:

```
# Formula
fol <- formula(m_Demolished ~ m_dbv_ngbr_8_ct + m_ddp_ngbr_8_ct + m_d311_ngbr_8_ct  +
                s_CleanUpCost + s_JudgmentAmt + m_PaymentStatus +
                s_FineAmt + s_LateFee + m_AgencyName )
```

In [19]:

```
#library(rpart)
#library(caret)

#rpModel <- rpart(fol, method="class", data=train, na.action = na.omit)
#summary(rpModel)
#rpPred <- predict(rpModel, newdata = test, na.action = na.omit)
#NOT WORKING
```

In [20]:

```
#Random forest model analysis
library(party)
library(randomForest)

rfModel <- randomForest(fol, data=train, na.action = na.omit)
print(rfModel)
rfPred <- predict(rfModel, newdata = test, na.action = na.omit)
rfTbl <- table(rfPred, test$m_Demolished)
```

```
Warning message in randomForest.default(m, y, ...):
"The response has five or fewer unique values.  Are you sure you want to do regression?"
Call:
 randomForest(formula = fol, data = train, na.action = na.omit)
               Type of random forest: regression
                     Number of trees: 500
No. of variables tried at each split: 3

          Mean of squared residuals: 5.697523e-05
                    % Var explained: 99.87
```

In [30]:

```
head(rfTbl)
```

```
rfPred                    0   1
  -1.21277987652491e-16 414   0
  -1.17739151761498e-16   1   0
```

```
  -1.17530984944381e-16   1   0
  -1.17267306976032e-16   1   0
  -1.17253429188224e-16   1   0
  -1.17100773522338e-16   1   0
```

In [21]:

```
#chi-squre analysis
library(MASS)
print(chisq.test(test$m_Demolished, test$m_ddp_ngbr_8_ct))
```

```
Warning message in chisq.test(test$m_Demolished, test$m_ddp_ngbr_8_ct):
"Chi-squared approximation may be incorrect"
         Pearson's Chi-squared test

data:  test$m_Demolished and test$m_ddp_ngbr_8_ct
X-squared = 61177, df = 13, p-value < 2.2e-16
```

In [22]:

```
library("survival")
survfit(Surv(test$m_TicketIssuedDT,test$m_Demolished == 0)~1)
```

```
Call: survfit(formula = Surv(test$m_TicketIssuedDT, test$m_Demolished ==
    0) ~ 1)

   4490 observations deleted due to missingness
      n  events  median 0.95LCL 0.95UCL
  56687   53996   13906   13895   13913
```

In [23]:

```
library("xgboost")

xKeep <- c( "m_dbv_ngbr_8_ct","m_ddp_ngbr_8_ct","m_d311_ngbr_8_ct","m_ViolationCategory",
"s_LateFee","s_FineAmt","s_JudgmentAmt","s_CleanUpCost",
           "m_PaymentStatus","m_AgencyName","m_Demolished")
xtrain <- train[xKeep]
xtest <- test[xKeep]

trainMatrix <- as.matrix(xtrain)
testMatrix <- as.matrix(xtest)

xgb <- xgboost(data = data.matrix(xtrain[,-1]),
 label =xtrain$m_Demolished ,
 eta = 0.1,
 max_depth = 15,
 nround=10,
```

```
 subsample = 0.5,
 colsample_bytree = 0.5,
 seed = 1,
 eval_metric = "mlogloss",
 objective = "multi:softprob",
 num_class = 100,
 nthread = 2
)
```

```
[1]	train-mlogloss:0.551635
[2]	train-mlogloss:0.495870
[3]	train-mlogloss:0.448311
[4]	train-mlogloss:0.404216
[5]	train-mlogloss:0.364788
[6]	train-mlogloss:0.329524
[7]	train-mlogloss:0.297907
[8]	train-mlogloss:0.269515
[9]	train-mlogloss:0.243987
[10]	train-mlogloss:0.221007
```

In [24]:

```
dim(xtest)
pred <- predict(xgb, t(xtest[1,7]))

print(head(pred))
```

1. 61177
2. 11

```
[1] 0.733373582 0.007193833 0.002647409 0.002647450 0.002647434 0.002647098
```

In [25]:

```
model <- xgb.dump(xgb, with.stats = T)
model[1:10]
```

```
Warning message:
"'with.stats' is deprecated.
Use 'with_stats' instead.
See help("Deprecated") and help("xgboost-deprecated").”
```

1. 'booster[0]'
2. '0:[f9<0.5] yes=1,no=2,missing=1,gain=7242.32,cover=66.2706'
3. '1:leaf=4.92212,cover=63.2016'
4. '2:leaf=-0.0380929,cover=3.069'
5. 'booster[1]'
6. '0:[f9<0.5] yes=1,no=2,missing=1,gain=5721.53,cover=68.0526'
7. '1:leaf=-0.0497389,cover=64.9242'
8. '2:leaf=3.78888,cover=3.1284'
9. 'booster[2]'

10. '0:leaf=-0.0497558,cover=66.4092'

In [26]:

```
# Get the feature real names
names <- dimnames(trainMatrix)[[2]]

# Compute feature importance matrix
importance_matrix <- xgb.importance(names, model = xgb)

# Nice graph
xgb.plot.importance(importance_matrix[1:10,])
```