

Use the given link below:

<https://archive.ics.uci.edu/ml/machine-learning-databases/communities>

Note: As the dataset provided at the above link was incomplete i.e. headers were missing, So I have used Crimes data of Atlanta till 2017

```
library(readr)
library(data.table)

COBRA.YTD2017 <- read_csv("G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-
YTD2017.csv")

View(COBRA.YTD2017)
str(COBRA.YTD2017)
summary(COBRA.YTD2017)
sum(is.na(COBRA.YTD2017))

# finding the missing values
library(VIM)
missingvalue_plot <- aggr(COBRA.YTD2017, col=c('navyblue','red'), numbers=TRUE, sortVars=TRUE,
labels=names(COBRA.YTD2017), cex.axis=.7, gap=3, ylab=c("Histogram of missing data", "Pattern"))

COBRA_YTD <- COBRA.YTD2017[complete.cases(COBRA.YTD2017), ]

missingvalue_plot <- aggr(COBRA_YTD, col=c('green','yellow'), numbers=TRUE, sortVars=TRUE,
labels=names(COBRA_YTD), cex.axis=.7, gap=3, ylab=c("Histogram of missing data", "Pattern"))
```

```

> library(readr)
> library(data.table)
data.table 1.11.8 Latest news: r-datatable.com
> COBRA.YTD2017 <- read_csv("G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv")
Parsed with column specification:
cols(
  .default = col_character(),
  MI_PRINX = col_double(),
  offense_id = col_double(),
  occur_time = col_time(format = ""),
  poss_time = col_time(format = ""),
  beat = col_double(),
  dispo_code = col_double(),
  MaxOfnum_victims = col_double(),
  loc_type = col_double(),
  x = col_double(),
  y = col_double()
)
See spec(...) for full column specifications.
Warning: 9 parsing failures.
   row      col expected actual      file
1 3239 dispo_code a double    COS 'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'
2 7945 dispo_code a double    ADM 'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'
3 8527 dispo_code a double    ADM 'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'
4 10145 dispo_code a double    ADM 'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'
5 11912 dispo_code a double    ADM 'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'
.....
see problems(...) for more details.

```

	MI_PRINX	offense_id	rpt_date	occur_date	occur_time	poss_date	poss_time	beat	apt_office_prefix	apt_office_num	location	MinOfucr	MinOfib
1	8924155	173650072	12/31/2017	12/30/2017	23:15:00	12/31/2017	00:30:00	510	NA	NA	43 JESSE HILL JR DR NE	0640	2305
2	8924156	173650102	12/31/2017	12/18/2017	13:00:00	12/30/2017	22:00:00	501	NA	NA	1169 ATLANTIC DR NW	0640	2305
3	8924157	173650144	12/31/2017	12/30/2017	22:01:00	12/31/2017	01:00:00	303	NA	NA	633 PRYOR ST SW	0640	2305
4	8924158	173650149	12/31/2017	12/30/2017	20:00:00	12/31/2017	01:06:00	507	NA	NA	333 NELSON ST SW	0640	2305
5	8924159	173650159	12/31/2017	12/31/2017	00:41:00	12/31/2017	00:48:00	409	NA	NA	2348 CASCADE RD SW	0640	2305
6	8924160	173650180	12/31/2017	12/30/2017	23:00:00	12/31/2017	01:26:00	612	NA	NA	1245 GLENWOOD AVE SE	0650	2304
7	8924161	173650236	12/31/2017	12/31/2017	01:55:00	12/31/2017	01:59:00	605	NA	13	351 CHEROKEE AVE SE	0311	1212
8	8924162	173650241	12/31/2017	12/31/2017	00:00:00	12/31/2017	02:00:00	603	NA	NA	461 PONCE DE LEON AVE NE	0640	2305
9	8924163	173650295	12/31/2017	12/30/2017	00:00:00	12/31/2017	03:02:00	605	NA	NA	437 MEMORIAL DR SE	0640	2305
10	8924164	173650389	12/31/2017	12/31/2017	00:00:00	12/31/2017	03:34:00	304	NA	8	1053 LINAM ST SE	0531	2202A
11	8924165	173650449	12/31/2017	12/31/2017	00:40:00	12/31/2017	04:10:00	303	NA	NA	683 PRYOR ST SW	0710	2404
12	8924166	173650562	12/31/2017	12/31/2017	00:00:00	12/31/2017	05:53:00	104	NA	NA	192 CHICAMAUGA AVE SW	0640	2305

```

> str(COBRA.YTD2017)
Classes 'spec_tbl_df', 'tbl_df', 'tbl' and 'data.frame':    26759 obs. of  23 variables:
 $ MI_PRINX      : num  8924155 8924156 8924157 8924158 8924159 ...
 $ offense_id    : num  1.74e+08 1.74e+08 1.74e+08 1.74e+08 1.74e+08 ...
 $ rpt_date      : chr   "12/31/2017" "12/31/2017" "12/31/2017" "12/31/2017" ...
 $ occur_date    : chr   "12/30/2017" "12/18/2017" "12/30/2017" "12/30/2017" ...
 $ occur_time    : 'hms' num  23:15:00 13:00:00 22:01:00 20:00:00 ...
 .. attr(*, "units")= chr "secs"
 $ poss_date     : chr   "12/31/2017" "12/30/2017" "12/31/2017" "12/31/2017" ...
 $ poss_time     : 'hms' num   00:30:00 22:00:00 01:00:00 01:06:00 ...
 .. attr(*, "units")= chr "secs"
 $ beat         : num   510 501 303 507 409 612 605 603 605 304 ...
 $ apt_office_prefix: chr   NA NA NA NA ...
 $ apt_office_num : chr   NA NA NA NA ...
 $ location      : chr   "43 JESSE HILL JR DR NE" "1169 ATLANTIC DR NW" "633 PRYOR ST SW" "333 NELSON ST SW" ...
 $ minofucr      : chr   "0640" "0640" "0640" "0640" ...
 $ minofibr_code : chr   "2305" "2305" "2305" "2305" ...
 $ dispo_code    : num   NA NA NA NA NA NA NA NA NA NA ...
 $ Maxofnum_victims: num   2 1 1 2 1 1 1 1 1 ...
 $ shift         : chr   "Morn" "Unk" "Morn" "Eve" ...
 $ Avg Day       : chr   "Sat" "Unk" "Sat" "Sat" ...
 $ loc_type      : num   13 13 18 18 18 18 26 18 13 26 ...
 $ UC2 Literal   : chr   "LARCENY-FROM VEHICLE" "LARCENY-FROM VEHICLE" "LARCENY-FROM VEHICLE" "LARCENY-FROM VEHICLE" ...
 $ neighborhood  : chr   "Downtown" "Home Park" "Mechanicsville" "Castleberry Hill" ...
 $ npu           : chr   "M" "E" "V" "M" ...
 $ x             : num  -84.4 -84.4 -84.4 -84.4 -84.5 ...
 $ y             : num   33.8 33.8 33.7 33.8 33.7 ...
- attr(*, "problems")=Classes 'tbl_df', 'tbl' and 'data.frame':    9 obs. of  5 variables:
 .. $ row      : int  3239 7945 8527 10145 11912 12629 13305 17684 20632
 .. $ col      : chr   "dispo_code" "dispo_code" "dispo_code" "dispo_code" ...
 .. $ expected: chr   "a double" "a double" "a double" "a double" ...
 .. $ actual  : chr   "COS" "ADM" "ADM" "ADM" ...
 .. $ file    : chr   "'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'" "'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'" "'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'" ...
TA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv" "'G:/DATA ANALYTICS/DATA/crime-in-atlanta-2017/COBRA-YTD2017.csv'" ...
- attr(*, "spec")=
 .. cols()
 ..   MI_PRINX = col_double(),
 ..   offense_id = col_double(),
 ..   rpt_date = col_character(),
 ..   occur_date = col_character(),
 ..   occur_time = col_time(format = ""),
 ..   poss_date = col_character(),
 ..   poss_time = col_time(format = ""),
 ..   beat = col_double(),
 ..   apt_office_prefix = col_character(),
 ..   apt_office_num = col_character(),
 ..   location = col_character(),
 ..   minofucr = col_character(),
 ..   minofibr_code = col_character(),
 ..   dispo_code = col_double(),
 ..   Maxofnum_victims = col_double(),
 ..   shift = col_character(),
 ..   `Avg Day` = col_character(),
 ..   loc_type = col_double(),
 ..   `UC2 Literal` = col_character(),
 ..   neighborhood = col_character(),
 ..   npu = col_character(),
 ..   x = col_double(),
 ..   y = col_double()

```

```

> summary(COBRA.YTD2017)
  MI_PRINX      offense_id      rpt_date      occur_date      occur_time      poss_date      poss_time      beat
Min.   :8838438 Min.   :1.608e+08 Length:26759 Length:26759 Length:26759 Length:26759 Length:26759 Min.   :101.0
1st Qu.:8904204 1st Qu.:1.711e+08 Class :character Class :character Class1:hms      Class1:hms      1st Qu.:208.0
Median :8910894 Median :1.720e+08 Mode :character Mode :character Class2:difftime Class2:difftime Median :312.0
Mean   :8910851 Mean   :6.523e+08                                     Mode :numeric      Mode :numeric      Mean :355.6
3rd Qu.:8917584 3rd Qu.:1.728e+08                                     Mode :numeric      Mode :numeric      3rd Qu.:505.0
Max.   :8924410 Max.   :1.735e+11

```

apt_office_prefix	apt_office_num	location	minofucr	minofibr_code	dispo_code	Maxofnum_victims	Shift
Length:26759	Length:26759	Length:26759	Length:26759	Length:26759	Min. :10.00	Min. : 0.00	Length:26759
Class :character	Class :character	Class :character	Class :character	Class :character	1st Qu.:10.00	1st Qu.: 1.00	Class :character
Mode :character	Mode :character	Mode :character	Mode :character	Mode :character	Median :10.00	Median : 1.00	Mode :character
					Mean :13.32	Mean : 1.16	
					3rd Qu.:10.00	3rd Qu.: 1.00	
					Max. :60.00	Max. :27.00	
					NA's :22968	NA's :75	

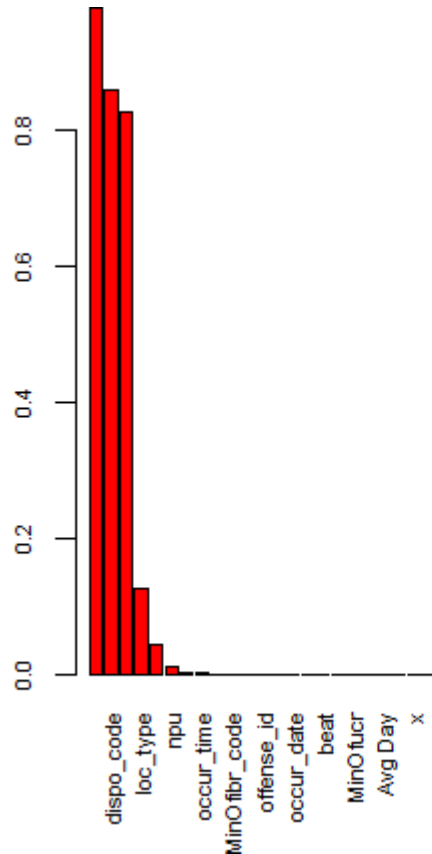
Avg Day	loc_type	UC2 Literal	neighborhood	npu	x	y
Length:26759	Min. : 1.00	Length:26759	Length:26759	Length:26759	Min. : -84.55	Min. : 0.00
Class :character	1st Qu.:13.00	Class :character	Class :character	Class :character	1st Qu.: -84.43	1st Qu.:33.73
Mode :character	Median :18.00	Mode :character	Mode :character	Mode :character	Median : -84.40	Median :33.76
	Mean :20.76				Mean : -83.69	Mean :33.47
	3rd Qu.:20.00				3rd Qu.: -84.37	3rd Qu.:33.79
	Max. :99.00				Max. : 0.00	Max. :33.88
	NA's :3344					

```

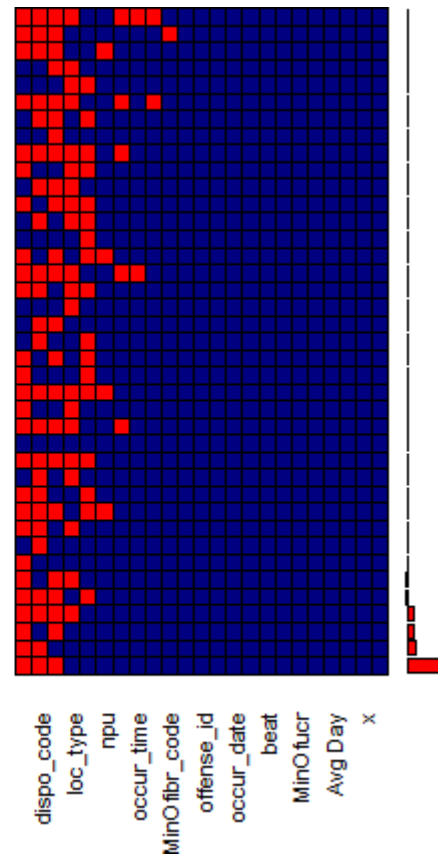
> sum(is.na(COBRA.YTD2017))
[1] 76194
>

```

Histogram of missing data



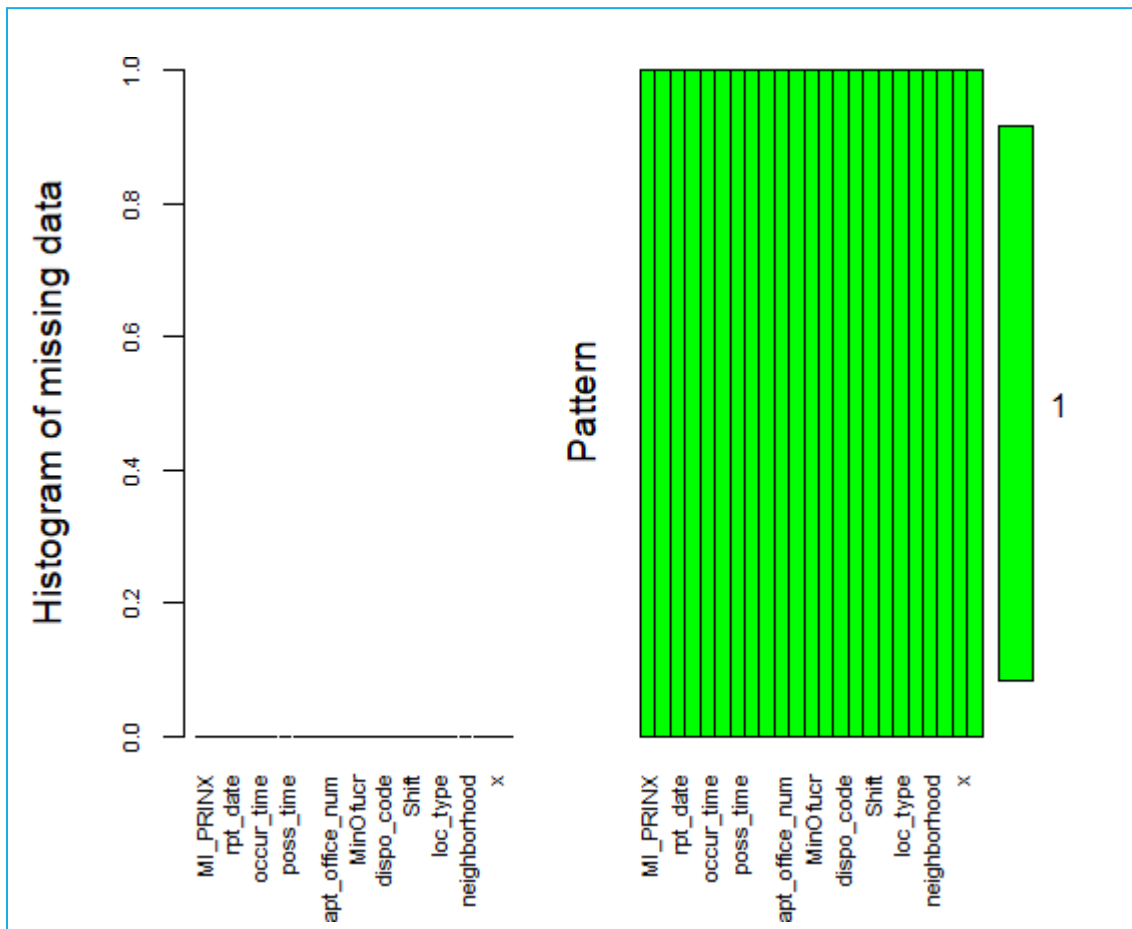
Pattern



```
> missingvalue_plot <- aggr(COBRA_YTD, col=c('black','yellow'), numbers=TRUE, sortVars=TRUE,
  sing data","Pattern"))
```

variables sorted by number of missings:

Variable	Count
MI_PRINX	0
offense_id	0
rpt_date	0
occur_date	0
occur_time	0
poss_date	0
poss_time	0
beat	0
apt_office_prefix	0
apt_office_num	0
location	0
MinOfucr	0
Minofibr_code	0
dispo_code	0
MaxOfnum_victims	0
shift	0
Avg Day	0
loc_type	0
uc2 Literal	0
neighborhood	0
npu	0
x	0
y	0



Perform the below operations:

- Find out top 5 attributes having highest correlation (select only Numeric features)

#a. Find out top 5 attributes having highest correlation (select only Numeric features).

```
fit<-lm(beat~MinOfucr+MaxOfnum_victims+loc_type+neighborhood+x+y,data =COBRA.YTD2017,
na.action = na.omit)
```

```
fit
```

```
summary(fit)
```

```
fit1<-lm(formula=MinOfucr~beat+MaxOfnum_victims+loc_type+neighborhood+x+y,data
=COBRA.YTD2017)
```

```
fit1
```

```
summary(fit1)
```

```
vif(fit)
```

```
vif(fit1)
```

```
vif(fit)>5
```

```
vif(fit1)>5
```

```
> fit<-lm(beat~Minofucr+Maxofnum_victims+loc_type+neighborhood+x+y,data =COBRA.YTD2017, na.action = na.omit)
> fit
```

Call:
lm(formula = beat ~ Minofucr + Maxofnum_victims + loc_type +
neighborhood + x + y, data = COBRA.YTD2017, na.action = na.omit)

Coefficients:

(Intercept)	Minofucr0220	Minofucr0311
2.995e+02	-1.754e+01	4.436e+03
Minofucr0312	Minofucr0313	Minofucr0314
4.435e+03	4.436e+03	4.437e+03
Minofucr0315	Minofucr0316	Minofucr0317
4.438e+03	4.435e+03	4.438e+03
Minofucr0321	Minofucr0322	Minofucr0323
4.436e+03	4.437e+03	4.436e+03
Minofucr0324	Minofucr0325	Minofucr0327
4.436e+03	4.436e+03	4.434e+03
Minofucr0331	Minofucr0332	Minofucr0333
4.449e+03	4.435e+03	4.435e+03
Minofucr0334	Minofucr0335	Minofucr0336
4.435e+03	4.435e+03	4.438e+03
Minofucr0337	Minofucr0341	Minofucr0342
4.436e+03	4.432e+03	4.439e+03
Minofucr0343	Minofucr0344	Minofucr0345
4.440e+03	4.435e+03	4.429e+03
Minofucr0347	Minofucr0410	Minofucr0420
4.435e+03	4.434e+03	4.438e+03

```
> summary(fit)
```

Call:
lm(formula = beat ~ Minofucr + Maxofnum_victims + loc_type +
neighborhood + x + y, data = COBRA.YTD2017, na.action = na.omit)

Residuals:

Min	1Q	Median	3Q	Max
-502.97	-0.79	0.05	1.11	400.21

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.995e+02	3.177e+00	94.280	< 2e-16 ***
Minofucr0220	-1.754e+01	9.905e+00	-1.770	0.07667 .
Minofucr0311	4.436e+03	2.674e+03	1.659	0.09715 .
Minofucr0312	4.435e+03	2.674e+03	1.659	0.09717 .
Minofucr0313	4.436e+03	2.674e+03	1.659	0.09714 .
Minofucr0314	4.437e+03	2.674e+03	1.660	0.09702 .
Minofucr0315	4.438e+03	2.674e+03	1.660	0.09698 .
Minofucr0316	4.435e+03	2.674e+03	1.659	0.09720 .
Minofucr0317	4.438e+03	2.674e+03	1.660	0.09700 .
Minofucr0321	4.436e+03	2.674e+03	1.659	0.09709 .
Minofucr0322	4.437e+03	2.674e+03	1.659	0.09707 .
Minofucr0323	4.436e+03	2.674e+03	1.659	0.09717 .
Minofucr0324	4.436e+03	2.674e+03	1.659	0.09712 .
Minofucr0325	4.436e+03	2.674e+03	1.659	0.09715 .
Minofucr0327	4.434e+03	2.674e+03	1.658	0.09730 .
Minofucr0331	4.449e+03	2.674e+03	1.664	0.09615 .
Minofucr0332	4.435e+03	2.674e+03	1.659	0.09715 .
Minofucr0333	4.435e+03	2.674e+03	1.659	0.09719 .
Minofucr0334	4.435e+03	2.674e+03	1.659	0.09722 .
Minofucr0335	4.435e+03	2.674e+03	1.658	0.09724 .
Minofucr0336	4.438e+03	2.674e+03	1.660	0.09693 .
Minofucr0337	4.436e+03	2.674e+03	1.659	0.09714 .
Minofucr0341	4.432e+03	2.674e+03	1.657	0.09744 .
Minofucr0342	4.439e+03	2.674e+03	1.660	0.09688 .
Minofucr0343	4.440e+03	2.674e+03	1.661	0.09679 .
Minofucr0344	4.435e+03	2.674e+03	1.659	0.09722 .
Minofucr0345	4.429e+03	2.674e+03	1.656	0.09768 .
Minofucr0347	4.435e+03	2.674e+03	1.659	0.09718 .
Minofucr0410	4.434e+03	2.674e+03	1.658	0.09731 .
Minofucr0420	4.438e+03	2.674e+03	1.660	0.09699 .
Minofucr0430	4.436e+03	2.674e+03	1.659	0.09711 .
Minofucr0440	4.440e+03	2.674e+03	1.660	0.09683 .
Minofucr0511	4.435e+03	2.674e+03	1.659	0.09720 .
Minofucr0512	4.435e+03	2.674e+03	1.659	0.09717 .
Minofucr0521	4.435e+03	2.674e+03	1.659	0.09717 .
Minofucr0522	4.438e+03	2.674e+03	1.660	0.09700 .
Minofucr0531	4.434e+03	2.674e+03	1.658	0.09727 .
Minofucr0532	4.440e+03	2.674e+03	1.660	0.09684 .
Minofucr0610	4.436e+03	2.674e+03	1.659	0.09713 .
Minofucr0620	4.428e+03	2.674e+03	1.656	0.09775 .
Minofucr0630	4.436e+03	2.674e+03	1.659	0.09711 .
Minofucr0640	4.436e+03	2.674e+03	1.659	0.09714 .
Minofucr0650	4.436e+03	2.674e+03	1.659	0.09712 .
Minofucr0660	4.436e+03	2.674e+03	1.659	0.09709 .
Minofucr0670	4.435e+03	2.674e+03	1.659	0.09721 .
Minofucr0680	4.443e+03	2.674e+03	1.662	0.09659 .


```

Console ~/
neighborhoodGrant Park      3.034e+02  2.792e+00 108.654 < 2e-16 ***
neighborhoodGreen Acres Valley 1.034e+02  1.183e+01  8.744 < 2e-16 ***
neighborhoodGreen Forest Acres 1.038e+02  9.532e+00 10.891 < 2e-16 ***
neighborhoodGreenbriar      1.017e+02  3.669e+00 27.723 < 2e-16 ***
neighborhoodGreenbriar Village 1.054e+02  9.953e+00 10.589 < 2e-16 ***
neighborhoodGrove Park      -1.867e+02  3.108e+00 -60.074 < 2e-16 ***
neighborhoodHammond Park     1.245e+00  3.379e+00  0.368 0.71261
neighborhoodHanover West    -8.289e+01  1.192e+01 -6.956 3.59e-12 ***
neighborhoodHarland Terrace  6.973e+01  3.438e+00 20.282 < 2e-16 ***
neighborhoodHarris Chiles   -1.898e+02  3.977e+00 -47.723 < 2e-16 ***
neighborhoodHarvel Homes Community -1.957e+02  1.832e+01 -10.681 < 2e-16 ***
neighborhoodHeritage Valley  1.056e+02  7.222e+00 14.622 < 2e-16 ***
neighborhoodHigh Point      1.494e+00  6.541e+00  0.228 0.81933
neighborhoodHills Park      -8.706e+01  4.349e+00 -20.020 < 2e-16 ***
neighborhoodHome Park       2.077e+02  2.970e+00 69.910 < 2e-16 ***
neighborhoodHorseshoe Community 1.050e+02  1.826e+01  5.750 9.07e-09 ***
neighborhoodHunter Hills    -1.900e+02  3.301e+00 -57.576 < 2e-16 ***
neighborhoodHuntington      9.959e+01  1.531e+01  6.503 8.02e-11 ***
neighborhoodInman Park      3.063e+02  3.092e+00 99.063 < 2e-16 ***
neighborhoodIvan Hill       1.064e+02  8.624e+00 12.336 < 2e-16 ***
neighborhoodJoyland         2.876e+00  5.018e+00  0.573 0.56647
neighborhoodJust Us         -1.945e+02  2.549e+01 -7.628 2.48e-14 ***
neighborhoodKings Forest    1.038e+02  5.123e+00 20.258 < 2e-16 ***
neighborhoodKingswood       -7.802e+01  1.846e+01 -4.227 2.38e-05 ***
neighborhoodKirkwood        3.136e+02  3.882e+00 80.783 < 2e-16 ***
neighborhoodKnight Park/Howell Station -1.871e+02  5.181e+00 -36.112 < 2e-16 ***
neighborhoodLake Claire     3.132e+02  4.936e+00 63.462 < 2e-16 ***
neighborhoodLake Estates    1.017e+02  2.577e+01  3.945 7.99e-05 ***
neighborhoodLakewood        3.189e+00  4.945e+00  0.645 0.51903
neighborhoodLakewood Heights 1.256e+00  2.813e+00  0.447 0.65524
neighborhoodLaurens Valley  1.028e+02  1.819e+01  5.651 1.62e-08 ***
neighborhoodLeila Valley    9.150e-01  4.786e+00  0.191 0.84838
neighborhoodLenox           -7.186e+01  4.580e+00 -15.690 < 2e-16 ***
neighborhoodLincoln Homes   -1.777e+02  6.240e+00 -28.485 < 2e-16 ***
neighborhoodLindbergh/Morosgo -7.476e+01  3.924e+00 -19.052 < 2e-16 ***
neighborhoodLindridge/Martin Manor -7.303e+01  4.154e+00 -17.579 < 2e-16 ***
neighborhoodLoring Heights  -8.428e+01  3.594e+00 -23.452 < 2e-16 ***
neighborhoodMagnum Manor    1.026e+02  9.038e+00 11.351 < 2e-16 ***
neighborhoodMargaret Mitchell -8.366e+01  1.108e+01 -7.549 4.57e-14 ***
neighborhoodMarietta Street Artery 2.108e+02  3.381e+00 62.332 < 2e-16 ***
neighborhoodMays            1.069e+02  4.777e+00 22.377 < 2e-16 ***
neighborhoodMeadowbrook Forest 1.037e+02  8.465e+00 12.245 < 2e-16 ***
neighborhoodMechanicsville  3.271e+00  2.492e+00  1.312 0.18938
neighborhoodMellwood        -5.193e+01  1.834e+01 -2.832 0.00463 **
neighborhoodMemorial Park   -8.163e+01  1.830e+01 -4.461 8.20e-06 ***
neighborhoodMidtown         2.107e+02  2.809e+00 75.008 < 2e-16 ***
neighborhoodMidwest Cascade  1.085e+02  5.999e+00 18.085 < 2e-16 ***
neighborhoodMonroe Heights  -1.788e+02  5.409e+00 -33.066 < 2e-16 ***
neighborhoodMorningside/Lenox Park -7.618e+01  3.660e+00 -20.816 < 2e-16 ***
[ reached getoption("max.print") -- omitted 90 rows ]
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 25.39 on 22091 degrees of freedom
(4378 observations deleted due to missingness)
Multiple R-squared:  0.9774,    Adjusted R-squared:  0.9771
F-statistic: 3306 on 289 and 22091 DF,  p-value: < 2.2e-16

```

```
> fit1<-lm(formula=MinOfucr~beat+MaxOfnum_victims+loc_type+neighborhood+x+y,data =COBRA.YTD2017)
> fit1
```

Call:
lm(formula = MinOfucr ~ beat + MaxOfnum_victims + loc_type +
neighborhood + x + y, data = COBRA.YTD2017)

Coefficients:

(Intercept)	beat	MaxOfnum_victims
2.218e+02	7.126e-03	-1.229e+01
loc_type	neighborhoodAdams Park	neighborhoodAdamsville
-4.508e-02	-9.749e+00	-9.269e+00
neighborhoodAlmond Park	neighborhoodAmal Heights	neighborhoodAnsley Park
-1.903e+01	-1.164e-01	-1.197e+00
neighborhoodArden/Habersham	neighborhoodArdmore	neighborhoodArgonne Forest
-3.377e+00	2.239e+01	6.209e+01
neighborhoodArlington Estates	neighborhoodAshley Courts	neighborhoodAshview Heights
-3.531e+01	-8.837e+00	-3.203e+01
neighborhoodAtkins Park	neighborhoodAtlanta Industrial Park	neighborhoodAtlanta University Center
3.540e+00	4.923e+01	6.809e+01
neighborhoodAtlantic Station	neighborhoodAudobon Forest	neighborhoodAudobon Forest West
1.603e+01	8.961e+00	5.861e+01
neighborhoodBaker Hills	neighborhoodBakers Ferry	neighborhoodBankhead
-2.243e+01	-7.371e+00	-7.140e+01
neighborhoodBankhead/Bolton	neighborhoodBeecher Hills	neighborhoodBen Hill
3.446e+01	4.918e+00	-3.544e+00
neighborhoodBen Hill Acres	neighborhoodBen Hill Forest	neighborhoodBen Hill Pines
-1.669e+01	1.236e+02	-4.936e+00

```
> summary(fit1)
```

Call:
lm(formula = MinOfucr ~ beat + MaxOfnum_victims + loc_type +
neighborhood + x + y, data = COBRA.YTD2017)

Residuals:

Min	1Q	Median	3Q	Max
-333.92	-35.53	20.95	61.95	429.07

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.218e+02	1.461e+01	15.184	< 2e-16 ***
beat	7.126e-03	2.641e-02	0.270	0.787314
MaxOfnum_victims	-1.229e+01	1.003e+00	-12.255	< 2e-16 ***
loc_type	-4.508e-02	4.229e-02	-1.066	0.286378
neighborhoodAdams Park	-9.749e+00	1.633e+01	-0.597	0.550649
neighborhoodAdamsville	-9.269e+00	1.117e+01	-0.830	0.406568
neighborhoodAlmond Park	-1.903e+01	1.808e+01	-1.052	0.292635
neighborhoodAmal Heights	-1.164e-01	2.291e+01	-0.005	0.995947
neighborhoodAnsley Park	-1.197e+00	1.866e+01	-0.064	0.948858
neighborhoodArden/Habersham	-3.377e+00	4.704e+01	-0.072	0.942767
neighborhoodArdmore	2.239e+01	2.280e+01	0.982	0.326053
neighborhoodArgonne Forest	6.209e+01	3.833e+01	1.620	0.105302
neighborhoodArlington Estates	-3.531e+01	2.595e+01	-1.361	0.173611
neighborhoodAshley Courts	-8.837e+00	1.854e+01	-0.477	0.633634
neighborhoodAshview Heights	-3.203e+01	1.351e+01	-2.371	0.017734 *
neighborhoodAtkins Park	3.540e+00	5.926e+01	0.060	0.952370
neighborhoodAtlanta Industrial Park	4.923e+01	2.278e+01	2.161	0.030671 *
neighborhoodAtlanta University Center	6.809e+01	1.396e+01	0.049	0.961095
neighborhoodAtlantic Station	1.603e+01	1.448e+01	1.107	0.268284
neighborhoodAudobon Forest	8.961e+00	2.658e+01	0.337	0.735972
neighborhoodAudobon Forest West	5.861e+01	3.888e+01	1.507	0.131752
neighborhoodBaker Hills	-2.243e+01	2.084e+01	-1.076	0.281915
neighborhoodBakers Ferry	-7.371e+00	4.557e+01	-0.162	0.871524
neighborhoodBankhead	-7.140e+01	1.342e+01	-5.322	1.04e-07 ***
neighborhoodBankhead/Bolton	3.446e+01	2.491e+01	1.383	0.166544
neighborhoodBeecher Hills	4.918e+00	2.909e+01	0.169	0.865739
neighborhoodBen Hill	-3.544e+00	2.230e+01	-0.159	0.873721
neighborhoodBen Hill Acres	-1.669e+01	2.347e+01	-0.711	0.476929
neighborhoodBen Hill Forest	1.236e+02	7.181e+01	1.722	0.085147 .
neighborhoodBen Hill Pines	-4.936e+00	4.275e+01	-0.115	0.908080
neighborhoodBen Hill Terrace	1.217e+00	2.194e+01	0.055	0.955768
neighborhoodBentzen Park	-2.507e+01	1.990e+01	-1.260	0.207583
neighborhoodBerkeley Park	2.098e+01	1.298e+01	1.616	0.106162
neighborhoodBetmar Lavilla	2.739e+01	1.687e+01	1.623	0.104539
neighborhoodBlair Villa/Poole Creek	2.762e+01	1.720e+01	1.606	0.108317
neighborhoodBlantown	3.626e+01	1.273e+01	2.848	0.004406 **
neighborhoodBolton	1.308e+01	1.527e+01	0.857	0.391730
neighborhoodBolton Hills	2.209e+00	4.210e+01	0.052	0.958159
neighborhoodBoulder Park	-2.241e+01	4.180e+01	-0.536	0.591895
neighborhoodBoulevard Heights	4.861e+00	1.947e+01	0.250	0.802878
neighborhoodBrandon	5.670e+01	3.161e+01	1.794	0.072852 .
neighborhoodBrentwood	-7.443e+01	4.012e+01	-1.855	0.063549 .
neighborhoodBriar Glen	4.873e+01	4.238e+01	1.150	0.250262
neighborhoodBrookhaven	4.735e+00	3.674e+01	0.129	0.897448
neighborhoodBrookview Heights	9.546e+00	2.162e+01	0.441	0.658876
neighborhoodBrookwood	-8.286e+00	2.136e+01	-0.388	0.698026

neighborhoodMorningside/Lenox Park	1.639e+01	1.445e+01	1.135	0.256477	
neighborhoodMozley Park	-2.716e+01	1.429e+01	-1.900	0.057434	.
neighborhoodMt. Gilead Woods	-4.631e+01	3.944e+01	-1.174	0.240313	
neighborhoodMt. Paran Parkway	2.164e+01	1.013e+02	0.213	0.830942	
neighborhoodMt. Paran/Northside	-1.300e+01	2.964e+01	-0.439	0.660913	
neighborhoodNiskey Cove	3.107e+01	7.154e+01	0.434	0.664004	
neighborhoodNiskey Lake	6.724e+01	5.117e+01	1.314	0.188860	
neighborhoodNorth Buckhead	1.867e+01	1.819e+01	1.026	0.304853	
neighborhoodNorwood Manor	-1.205e+01	1.809e+01	-0.666	0.505368	
neighborhoodOakcliff	3.736e+01	3.895e+01	0.959	0.337453	
neighborhoodOakland	4.533e+00	2.489e+01	0.182	0.855482	
neighborhoodOakland City	-2.498e+01	1.108e+01	-2.255	0.024165	*
neighborhoodOld Fairburn Village	4.873e+01	1.004e+02	0.485	0.627339	
neighborhoodOld Fourth ward	-2.104e+00	1.343e+01	-0.157	0.875535	
neighborhoodOld Gordon	-1.341e+01	2.382e+01	-0.563	0.573585	
neighborhoodOrchard Knob	2.727e+00	1.970e+01	0.138	0.889909	
neighborhoodOrmewood Park	-2.503e+01	1.470e+01	-1.702	0.088719	.
neighborhoodPaces	3.021e+01	2.446e+01	1.235	0.216919	
neighborhoodPeachtree Battle Alliance	5.466e+01	3.078e+01	1.776	0.075811	.
neighborhoodPeachtree Heights East	2.207e+01	3.130e+01	0.705	0.480826	
neighborhoodPeachtree Heights West	1.580e+01	1.856e+01	0.852	0.394469	
neighborhoodPeachtree Hills	2.217e+01	1.988e+01	1.115	0.264844	
neighborhoodPeachtree Park	-1.739e+00	2.034e+01	-0.085	0.931870	
neighborhoodPenelope Neighbors	-9.972e+01	3.041e+01	-3.279	0.001043	***
neighborhoodPeopletown	-2.815e+01	1.172e+01	-2.403	0.016289	*
neighborhoodPerkerson	1.361e+01	1.191e+01	1.143	0.253221	
neighborhoodPeyton Forest	2.500e+01	3.278e+01	0.763	0.445761	
neighborhoodPiedmont Heights	3.562e-01	1.482e+01	0.024	0.980827	
neighborhoodPine Hills	-2.234e+00	1.860e+01	-0.120	0.904364	
neighborhoodPittsburgh	-2.804e+00	1.025e+01	-0.274	0.784408	
neighborhoodPleasant Hill	3.624e+01	4.720e+01	0.768	0.442645	
neighborhoodPolar Rock	4.473e+00	2.027e+01	0.221	0.825378	
neighborhoodPomona Park	1.397e+01	7.127e+01	0.196	0.844614	
neighborhoodPoncey-Highland	2.103e+01	1.548e+01	1.358	0.174405	
neighborhoodPrinceton Lakes	5.334e+01	1.607e+01	3.320	0.000902	***
neighborhoodRandall Mill	2.865e+01	2.346e+01	1.221	0.221968	
neighborhoodRebel Valley Forest	-7.690e+00	1.941e+01	-0.396	0.692056	
neighborhoodReynoldstown	-1.769e+01	1.593e+01	-1.111	0.266663	
neighborhoodRidgecrest Forest	-1.680e+01	3.149e+01	-0.533	0.593810	
neighborhoodRidgedale Park	-5.404e+00	2.641e+01	-0.205	0.837852	
neighborhoodRidgewood Heights	1.853e+01	3.132e+01	0.591	0.554223	
neighborhoodRiverside	1.021e+01	1.471e+01	0.694	0.487558	
neighborhoodRockdale	2.647e+01	1.817e+01	1.457	0.145122	
neighborhoodRosedale Heights	-2.206e+01	1.861e+01	-1.185	0.235855	
neighborhoodRue Royal	-1.624e+02	7.170e+01	-2.265	0.023504	*
neighborhoodSandlewood Estates	1.123e+01	3.141e+01	0.357	0.720790	
neighborhoodScotts Crossing	-2.423e+01	1.787e+01	-1.356	0.175150	
neighborhoodSherwood Forest	-3.203e+01	5.168e+01	-0.620	0.535337	
neighborhoodSouth Atlanta	-2.876e+01	1.279e+01	-2.247	0.024619	*

[reached getOption("max.print") -- omitted 42 rows]

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 99.8 on 22139 degrees of freedom
(4378 observations deleted due to missingness)

Multiple R-squared: 0.147, Adjusted R-squared: 0.1377

F-statistic: 15.83 on 241 and 22139 DF, p-value: < 2.2e-16

```

> vif(fit)
              GVIF   Df GVIF^(1/(2*Df))
MinOfucr      NaN   49             NaN
MaxOfnum_victims NaN   1             NaN
loc_type      NaN   1             NaN
neighborhood   NaN  236             NaN
x              NaN   1             NaN
y              NaN   1             NaN
> vif(fit1)
              GVIF   Df GVIF^(1/(2*Df))
beat          4.411556e+01   1   6.641955
MaxOfnum_victims 1.017245e+00   1   1.008586
loc_type        1.025699e+00   1   1.012768
neighborhood    3.705006e+03  236   1.017562
x              1.883397e+05   1  433.981262
y              1.884420e+05   1  434.099027
> vif(fit)>5
              GVIF   Df GVIF^(1/(2*Df))
MinOfucr       NA  TRUE             NA
MaxOfnum_victims NA FALSE             NA
loc_type        NA FALSE             NA
neighborhood    NA  TRUE             NA
x               NA FALSE             NA
y               NA FALSE             NA
> vif(fit1)>5
              GVIF   Df GVIF^(1/(2*Df))
beat           TRUE FALSE             TRUE
MaxOfnum_victims FALSE FALSE             FALSE
loc_type        FALSE FALSE             FALSE
neighborhood    TRUE  TRUE             FALSE
x               TRUE FALSE             TRUE
y               TRUE FALSE             TRUE
> |

```

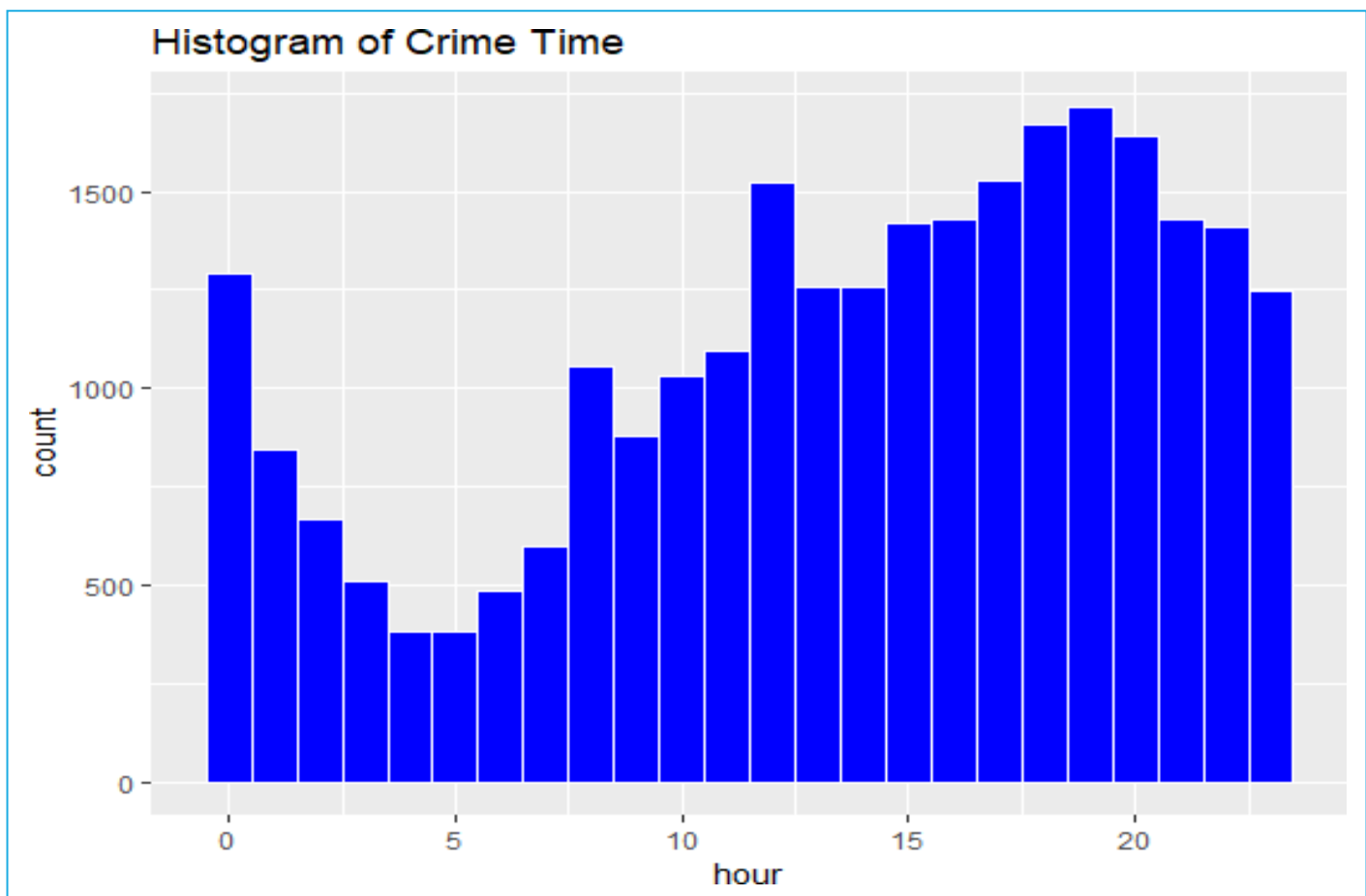
- P-values are very important because, we can consider linear model to be statistically significant only.
- When both these p-values are less than the pre-determined statistical determined level, which is ideally 0.05.
- This is visually interpreted by the significance stars at the end of the row.
- The more stars beside the variable's p-value, the more significant the variable.
- When there is a p-value, there is a null and alternative hypothesis associated with it.
- Null and Alternate Hypothesis
- In Linear regression, the Null Hypothesis is that the coefficients associated with the variables is equal to zero.
- The alternate hypothesis is that the coefficients are not equal to zero.
- There exists a relationship between the independent variable in question and the dependent variable.

b) Find out top 3 reasons for having more crime in a city.

```
#b. Find out top 3 reasons for having more crime in a city.
```

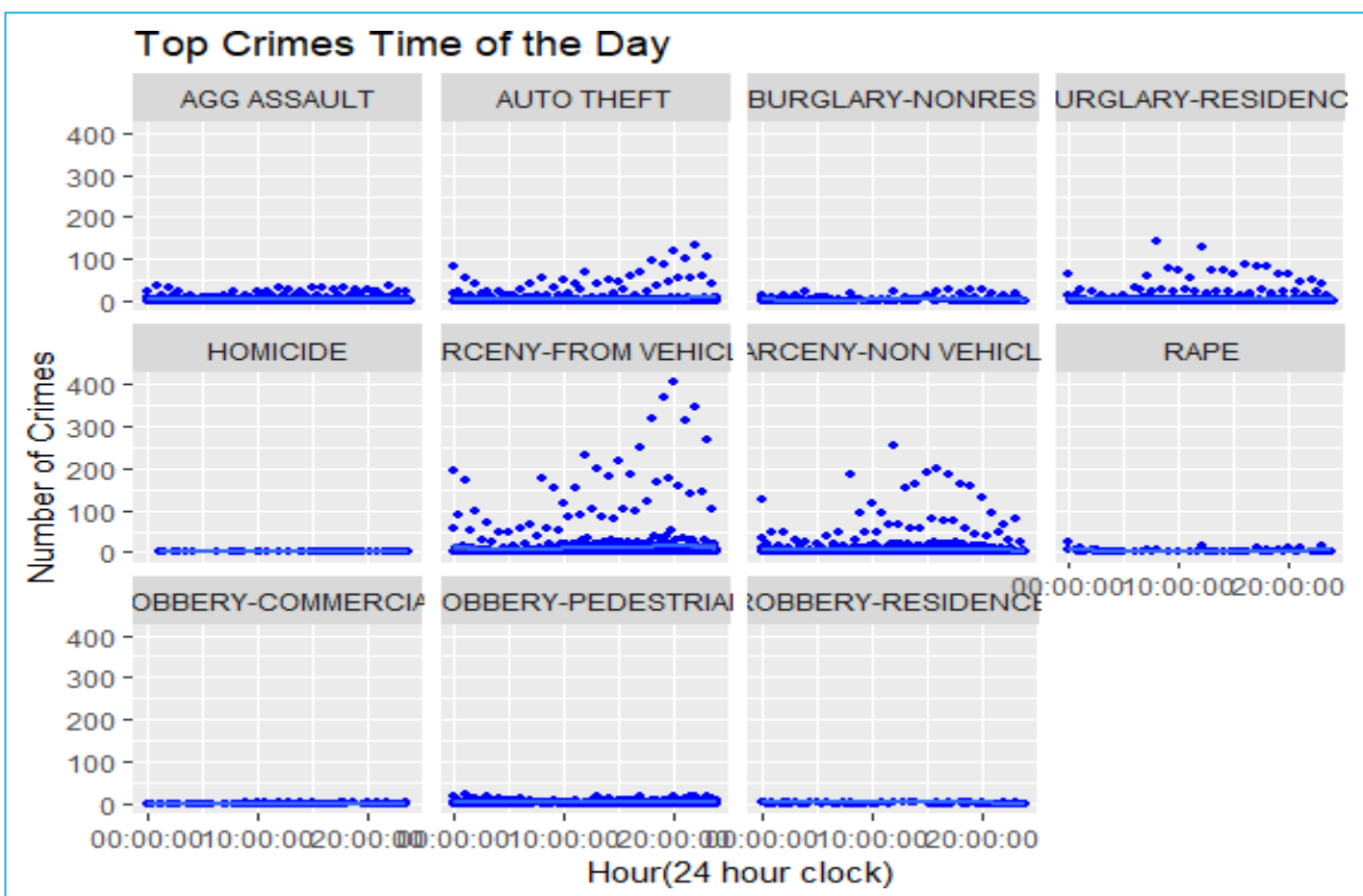
```
library(ggplot2)
COBRA.YTD2017$hour <- sub(":.:", "", COBRA.YTD2017$occur_time)
COBRA.YTD2017$hour <- as.numeric(COBRA.YTD2017$hour)
ggplot(aes(x=hour), data=COBRA.YTD2017)+geom_histogram(bins=24, color='white', fill='blue')
+
  ggtitle('Histogram of Crime Time')
```

```
UC2<-table(COBRA.YTD2017$`UC2 Literal`)
hist(UC2)
```



```
library(dplyr)
COBRA <- COBRA.YTD2017 %>% group_by(`UC2 Literal`, occur_time) %>%
  summarise(total = n())

ggplot(aes(x=occur_time, y=total), data=COBRA) + geom_point(colour="blue", size=1) +
  geom_smooth(method="loess") + xlab('Hour(24 hour clock)') +
  ylab('Number of Crimes') + ggtitle('Top Crimes Time of the Day') + facet_wrap(~`UC2 Literal`)
```



#Downtown and midtown are the most common locations where crimes take place, followed by Old Fourth Ward and West End. larceny theft are the top crimes in Atlanta followed by aggravated assault

```
library(knitr)
library(kableExtra)
```

```
kable(count(COBRA.YTD2017, COBRA.YTD2017$`UC2 Literal`, sort=TRUE), "html",
col.names=c("Crime Type", "Frequency")) %>%
kable_styling(bootstrap_options="striped", full_width=FALSE)
```

Crime Type	Frequency
LARCENY-FROM VEHICLE	9840
LARCENY-NON VEHICLE	6589
AUTO THEFT	3197
BURGLARY-RESIDENCE	2635
AGG ASSAULT	2024
ROBBERY-PEDESTRIAN	1126
BURGLARY-NONRES	758
RAPE	226
ROBBERY-COMMERCIAL	157
ROBBERY-RESIDENCE	132
HOMICIDE	75

c) Which all attributes have high correlation with crime rate?

#c. Which all attributes have correlation with crime rate?

```
library(ggplot2)
```

```
library(corrplot)
```

```
pairs(COBRA.YTD2017)
```

```
rank1<-sample(COBRA.YTD2017[1:100,22:23], 20, replace=T)
```

```
rank2<-sample(COBRA.YTD2017[1:100,22:23], 20, replace=T)
```

```
cbind(rank1,rank2)
```

```
plot(rank1, rank2)
```

```
cor(rank1,rank2, method="spearman")
```

```
cor(rank1,rank2, method="pearson")
```

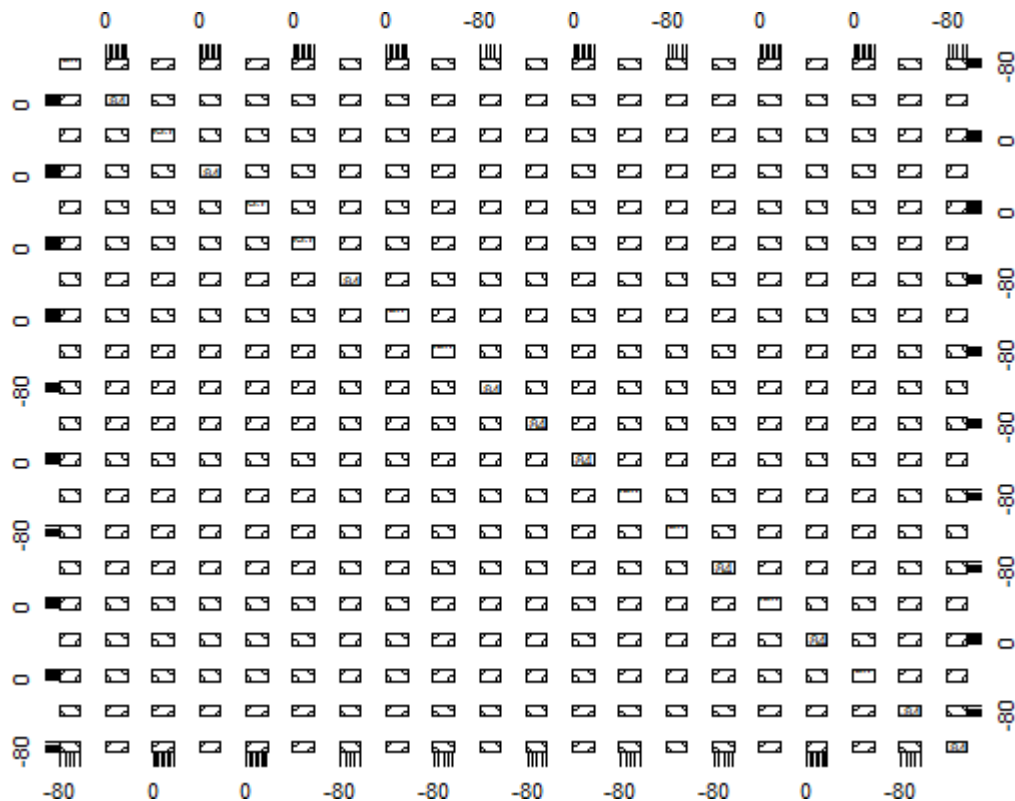
```
> #c. which all attributes have correlation with crime rate?
> library(ggplot2)
> library(corrplot)
> rank1<-sample(COBRA.YTD2017[1:100,22:23], 20, replace=T)
> rank2<-sample(COBRA.YTD2017[1:100,22:23], 20, replace=T)
> cbind(rank1,rank2)
```

[illegible]

```

      y      y      x      y      x      y      x      x
1  33.75582 33.75582 -84.38013 33.75582 -84.38013 33.75582 -84.38013 -84.38013
2  33.78674 33.78674 -84.39745 33.78674 -84.39745 33.78674 -84.39745 -84.39745
3  33.73760 33.73760 -84.39486 33.73760 -84.39486 33.73760 -84.39486 -84.39486
4  33.75156 33.75156 -84.39887 33.75156 -84.39887 33.75156 -84.39887 -84.39887
5  33.72146 33.72146 -84.46522 33.72146 -84.46522 33.72146 -84.46522 -84.46522
6  33.74006 33.74006 -84.34660 33.74006 -84.34660 33.74006 -84.34660 -84.34660
7  33.74505 33.74505 -84.37373 33.74505 -84.37373 33.74505 -84.37373 -84.37373
8  33.77303 33.77303 -84.37190 33.77303 -84.37190 33.77303 -84.37190 -84.37190
9  33.74639 33.74639 -84.37285 33.74639 -84.37285 33.74639 -84.37285 -84.37285
10 33.72579 33.72579 -84.38625 33.72579 -84.38625 33.72579 -84.38625 -84.38625
11 33.73616 33.73616 -84.39495 33.73616 -84.39495 33.73616 -84.39495 -84.39495
12 33.74974 33.74974 -84.43192 33.74974 -84.43192 33.74974 -84.43192 -84.43192
13 33.68243 33.68243 -84.40766 33.68243 -84.40766 33.68243 -84.40766 -84.40766
14 33.71691 33.71691 -84.39276 33.71691 -84.39276 33.71691 -84.39276 -84.39276
15 33.73048 33.73048 -84.38742 33.73048 -84.38742 33.73048 -84.38742 -84.38742
16 33.70087 33.70087 -84.46574 33.70087 -84.46574 33.70087 -84.46574 -84.46574
17 33.75831 33.75831 -84.43107 33.75831 -84.43107 33.75831 -84.43107 -84.43107
18 33.79671 33.79671 -84.47527 33.79671 -84.47527 33.79671 -84.47527 -84.47527
19 33.70678 33.70678 -84.38031 33.70678 -84.38031 33.70678 -84.38031 -84.38031
20 33.84746 33.84746 -84.36983 33.84746 -84.36983 33.84746 -84.36983 -84.36983
21 33.75757 33.75757 -84.33868 33.75757 -84.33868 33.75757 -84.33868 -84.33868
22 33.77067 33.77067 -84.38023 33.77067 -84.38023 33.77067 -84.38023 -84.38023
23 33.70369 33.70369 -84.40795 33.70369 -84.40795 33.70369 -84.40795 -84.40795
24 33.75437 33.75437 -84.43089 33.75437 -84.43089 33.75437 -84.43089 -84.43089
25 33.74647 33.74647 -84.42057 33.74647 -84.42057 33.74647 -84.42057 -84.42057
[ reached 'max' / getOption("max.print") -- omitted 75 rows ]
>

```



```
> cor(rank1,rank2, method="pearson")
```