

# BA Pundits

*Ushnik*

*December 12, 2016*

## NYC Yellow Taxi Traffic Data Analysis

Using NYC Taxi data available on the NYC Taxi and Limousine Commission website, we intend to analyze the geography of pickup and drop-offs made by the cabs during peak hours of the day. The data set includes 11.2 million trips made during the month of June 2016 and drills down each trip to particulars like pickup and drop-off dates, time, longitude, latitude, trip distance, fare amount, etc. We aim to identify the area that is the busiest during the peak hours and target the average commute time during weekdays and weekends. This would give us a sense of how much time it takes for a person to get from Point A to Point B with and without traffic (assuming weekends have little or no traffic as compared to weekdays).

```
####Install packages: "ggplot2" and "lubridate"
options(warn=-1)
library(ggplot2)
library(hexbin)
library(readr)
library(dplyr)
```

```
##
## 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
```

```
library(extrafont)
```

```
## Registering fonts with R
```

```
library(scales)
```

```
##
## Attaching package: 'scales'
```

```
## The following objects are masked from 'package:readr':
##
##   col_factor, col_numeric
```

```
library(grid)
library(RColorBrewer)
library(digest)
library(stringr)
library(methods)
library(lubridate)
```

```
##
## Attaching package: 'lubridate'
```

```
## The following object is masked from 'package:base':
##
##   date
```

```
options(repr.plot.mimetypes = 'image/png', repr.plot.width=4, repr.plot.height=3,
repr.plot.res=300)
```

```
df <- read.csv("C:/Users/Ushnik/Downloads/yellow_tripdata_2016-06.csv", header=T)
str(df)
```

```
## 'data.frame':   11135470 obs. of  19 variables:
##  $ VendorID           : int  2 2 2 2 2 2 2 1 1 1 ...
##  $ tpep_pickup_datetime : Factor w/ 2395442 levels "2016-06-01 00:00:00",...: 706487 706487 70
6487 706487 706487 706487 706488 706488 706488 ...
##  $ tpep_dropoff_datetime: Factor w/ 2399652 levels "1996-06-20 16:23:24",...: 707233 708555 70
7235 708614 707847 707606 708257 707452 707389 707881 ...
##  $ passenger_count     : int  2 1 1 1 1 1 5 1 1 2 ...
##  $ trip_distance       : num  0.79 5.22 1.26 7.39 3.1 2.17 6.02 1.4 1.2 1.9 ...
##  $ pickup_longitude    : num  -74 -74 -74 -74 -74 ...
##  $ pickup_latitude     : num  40.8 40.7 40.8 40.8 40.7 ...
##  $ RatecodeID          : int  1 1 1 1 1 1 1 1 1 1 ...
##  $ store_and_fwd_flag   : Factor w/ 2 levels "N","Y": 1 1 1 1 1 1 1 1 1 1 ...
##  $ dropoff_longitude   : num  -74 -74 -74 -73.9 -74 ...
##  $ dropoff_latitude    : num  40.8 40.7 40.7 40.9 40.8 ...
##  $ payment_type        : int  2 1 1 1 1 1 2 1 2 1 ...
##  $ fare_amount         : num  6 22 6.5 26 13.5 10.5 21.5 8.5 8 12 ...
##  $ extra               : num  0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
##  $ mta_tax             : num  0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 0.5 ...
##  $ tip_amount          : num  0 4 1.56 1 2.96 2.36 0 1.95 0 3.33 ...
##  $ tolls_amount        : num  0 0 0 0 0 0 0 0 0 0 ...
##  $ improvement_surcharge: num  0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 ...
##  $ total_amount        : num  7.3 27.3 9.36 28.3 17.76 ...
```

```
head(df)
```

```
## VendorID tpep_pickup_datetime tpep_dropoff_datetime passenger_count
## 1 2 2016-06-09 21:06:36 2016-06-09 21:13:08 2
## 2 2 2016-06-09 21:06:36 2016-06-09 21:35:11 1
## 3 2 2016-06-09 21:06:36 2016-06-09 21:13:10 1
## 4 2 2016-06-09 21:06:36 2016-06-09 21:36:10 1
## 5 2 2016-06-09 21:06:36 2016-06-09 21:23:23 1
## 6 2 2016-06-09 21:06:36 2016-06-09 21:19:21 1
## trip_distance pickup_longitude pickup_latitude RatecodeID
## 1 0.79 -73.98336 40.76094 1
## 2 5.22 -73.98172 40.73667 1
## 3 1.26 -73.99432 40.75107 1
## 4 7.39 -73.98236 40.77389 1
## 5 3.10 -73.98711 40.73317 1
## 6 2.17 -73.99520 40.73949 1
## store_and_fwd_flag dropoff_longitude dropoff_latitude payment_type
## 1 N -73.97746 40.75398 2
## 2 N -73.98164 40.67024 1
## 3 N -74.00423 40.74217 1
## 4 N -73.92947 40.85154 1
## 5 N -73.98591 40.76645 1
## 6 N -73.99320 40.76264 1
## fare_amount extra_mta_tax tip_amount tolls_amount improvement_surcharge
## 1 6.0 0.5 0.5 0.00 0 0.3
## 2 22.0 0.5 0.5 4.00 0 0.3
## 3 6.5 0.5 0.5 1.56 0 0.3
## 4 26.0 0.5 0.5 1.00 0 0.3
## 5 13.5 0.5 0.5 2.96 0 0.3
## 6 10.5 0.5 0.5 2.36 0 0.3
## total_amount
## 1 7.30
## 2 27.30
## 3 9.36
## 4 28.30
## 5 17.76
## 6 14.16
```

The data set includes 11135470 obs. of 19 variables

## Data Filtering

We filter the data set to analyze peak hours of the morning. Assuming 8 AM to 9 AM to be the peak travelling hour for people going to work, we rewrite the data set with the filter. We then convert the pick up and drop-off dates to the day of the week. Finally we include the relevant data and create the data set we work with. The resulting data has a few latitude and longitude outliers, and we therefore establish the limits of our plot within NYC coordinates: Minimum Latitude = 40.5774, Maximum Latitude = 40.9176, Minimum Longitude = -74.15, Maximum Longitude = -73.7004.

```
df<- with(df,df[ hour (df$tpep_pickup_datetime)>=8 & hour(df$tpep_pickup_datetime)< 9 & hour
(df$tpep_dropoff_datetime)>=8 & hour(df$tpep_dropoff_datetime)< 9 , ] )
head(df,10)
```

##	VendorID	tpep_pickup_datetime	tpep_dropoff_datetime	passenger_count	
## 146436	1	2016-06-01 08:00:00	2016-06-01 08:00:37	1	
## 146437	2	2016-06-01 08:00:00	2016-06-01 08:08:08	2	
## 146438	2	2016-06-01 08:00:00	2016-06-01 08:11:48	6	
## 146439	2	2016-06-01 08:00:00	2016-06-01 08:14:35	1	
## 146440	2	2016-06-01 08:00:00	2016-06-01 08:05:22	1	
## 146441	2	2016-06-01 08:00:00	2016-06-01 08:06:23	1	
## 146442	1	2016-06-01 08:00:01	2016-06-01 08:14:07	1	
## 146443	1	2016-06-01 08:00:01	2016-06-01 08:08:48	1	
## 146444	2	2016-06-01 08:00:01	2016-06-01 08:15:15	1	
## 146445	2	2016-06-01 08:00:01	2016-06-01 08:12:37	2	
##	trip_distance	pickup_longitude	pickup_latitude	RatecodeID	
## 146436	1.30	-73.98035	40.74580	1	
## 146437	1.39	-73.97506	40.79034	1	
## 146438	0.72	-73.97356	40.76101	1	
## 146439	2.31	-74.00020	40.74219	1	
## 146440	0.66	-73.98213	40.77029	1	
## 146441	0.72	-73.97213	40.75356	1	
## 146442	2.00	-73.95631	40.76780	1	
## 146443	1.20	-73.98000	40.74558	1	
## 146444	2.15	-73.96409	40.77639	1	
## 146445	0.84	-73.98239	40.76849	1	
##	store_and_fwd_flag	dropoff_longitude	dropoff_latitude	payment_type	
## 146436	N	-73.98153	40.74644	1	
## 146437	N	-73.97030	40.78390	1	
## 146438	N	-73.98186	40.76588	2	
## 146439	N	-73.97338	40.75508	1	
## 146440	N	-73.98729	40.77891	2	
## 146441	N	-73.98418	40.75870	1	
## 146442	N	-73.96792	40.78740	1	
## 146443	N	-73.99100	40.73148	1	
## 146444	N	-73.96641	40.79074	2	
## 146445	N	-73.99028	40.77182	1	
##	fare_amount	extra	mta_tax	tip_amount	tolls_amount
## 146436	2.5	0	0.5	0.65	0
## 146437	7.5	0	0.5	1.66	0
## 146438	8.5	0	0.5	0.00	0
## 146439	11.0	0	0.5	1.00	0
## 146440	5.5	0	0.5	0.00	0
## 146441	6.0	0	0.5	0.50	0
## 146442	11.0	0	0.5	2.35	0
## 146443	7.5	0	0.5	2.05	0
## 146444	11.5	0	0.5	0.00	0
## 146445	9.0	0	0.5	1.96	0
##	improvement_surcharge	total_amount			
## 146436	0.3	3.95			
## 146437	0.3	9.96			
## 146438	0.3	9.30			
## 146439	0.3	12.80			
## 146440	0.3	6.30			
## 146441	0.3	7.30			
## 146442	0.3	14.15			
## 146443	0.3	10.35			

## 146444	0.3	12.30
## 146445	0.3	11.76

```
summary(df)
```

```

##      VendorID      tpep_pickup_datetime
## Min.      :1.000    2016-06-21 08:13:12:    20
## 1st Qu.:1.000    2016-06-01 08:36:11:    17
## Median :2.000    2016-06-03 08:10:57:    17
## Mean    :1.531    2016-06-03 08:13:35:    17
## 3rd Qu.:2.000    2016-06-03 08:17:01:    17
## Max.    :2.000    2016-06-03 08:17:28:    17
##      (Other)      :390581
##      tpep_dropoff_datetime passenger_count trip_distance
## 2016-06-01 08:55:44:    18    Min.      :0.000    Min.      : 0.000
## 2016-06-01 08:50:04:    17    1st Qu.:1.000    1st Qu.: 0.900
## 2016-06-03 08:43:45:    17    Median :1.000    Median : 1.400
## 2016-06-03 08:44:06:    17    Mean    :1.597    Mean    : 2.078
## 2016-06-20 08:44:08:    17    3rd Qu.:1.000    3rd Qu.: 2.300
## 2016-06-02 08:44:20:    16    Max.    :9.000    Max.    :32.000
## (Other)      :390584
## pickup_longitude pickup_latitude RatecodeID    store_and_fwd_flag
## Min.      :-79.10    Min.      : 0.00    Min.      : 1.00    N:388728
## 1st Qu.: -73.99    1st Qu.:40.74    1st Qu.: 1.00    Y: 1958
## Median : -73.98    Median :40.76    Median : 1.00
## Mean     :-73.08    Mean     :40.26    Mean     : 1.02
## 3rd Qu.: -73.96    3rd Qu.:40.77    3rd Qu.: 1.00
## Max.      : 0.00    Max.      :43.43    Max.      :99.00
##
## dropoff_longitude dropoff_latitude payment_type    fare_amount
## Min.      :-118.19    Min.      : 0.00    Min.      :1.000    Min.      : -60.00
## 1st Qu.: -73.99    1st Qu.:40.74    1st Qu.:1.000    1st Qu.: 6.00
## Median : -73.98    Median :40.76    Median :1.000    Median : 8.50
## Mean     :-73.16    Mean     :40.31    Mean     :1.295    Mean     :10.44
## 3rd Qu.: -73.97    3rd Qu.:40.77    3rd Qu.:2.000    3rd Qu.:12.00
## Max.      : 0.00    Max.      :45.63    Max.      :4.000    Max.      :450.00
##
##      extra      mta_tax      tip_amount      tolls_amount
## Min.      :-0.500000    Min.      :-0.5000    Min.      : -8.000    Min.      : -5.5400
## 1st Qu.: 0.000000    1st Qu.: 0.5000    1st Qu.: 0.000    1st Qu.: 0.0000
## Median : 0.000000    Median : 0.5000    Median : 1.280    Median : 0.0000
## Mean     : 0.000304    Mean     : 0.4983    Mean     : 1.521    Mean     : 0.1489
## 3rd Qu.: 0.000000    3rd Qu.: 0.5000    3rd Qu.: 2.060    3rd Qu.: 0.0000
## Max.      : 4.500000    Max.      : 0.5000    Max.      :175.000    Max.      :55.5500
##
## improvement_surcharge total_amount
## Min.      :-0.3000    Min.      : -62.67
## 1st Qu.: 0.3000    1st Qu.: 7.80
## Median : 0.3000    Median :10.55
## Mean     : 0.2997    Mean     :12.91
## 3rd Qu.: 0.3000    3rd Qu.:14.75
## Max.      : 0.3000    Max.      :490.80
##

```

```
View(df)
```

```
df$day<- strptime(df[,2],"%Y-%m-%d")
df$day<-weekdays(df$day)
df1 <- subset(df,df$weekday == "Saturday" | df$weekday == "Sunday")
df2 <- subset(df,df$weekday != "Saturday" & df$weekday != "Sunday")
View(df)

df<-data.frame(df[c(2,3,5:7,10,11,19,20)])
head(df,10)
```

```
##          tpep_pickup_datetime tpep_dropoff_datetime trip_distance
## 146436 2016-06-01 08:00:00 2016-06-01 08:00:37 1.30
## 146437 2016-06-01 08:00:00 2016-06-01 08:08:08 1.39
## 146438 2016-06-01 08:00:00 2016-06-01 08:11:48 0.72
## 146439 2016-06-01 08:00:00 2016-06-01 08:14:35 2.31
## 146440 2016-06-01 08:00:00 2016-06-01 08:05:22 0.66
## 146441 2016-06-01 08:00:00 2016-06-01 08:06:23 0.72
## 146442 2016-06-01 08:00:01 2016-06-01 08:14:07 2.00
## 146443 2016-06-01 08:00:01 2016-06-01 08:08:48 1.20
## 146444 2016-06-01 08:00:01 2016-06-01 08:15:15 2.15
## 146445 2016-06-01 08:00:01 2016-06-01 08:12:37 0.84
##          pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude
## 146436 -73.98035 40.74580 -73.98153 40.74644
## 146437 -73.97506 40.79034 -73.97030 40.78390
## 146438 -73.97356 40.76101 -73.98186 40.76588
## 146439 -74.00020 40.74219 -73.97338 40.75508
## 146440 -73.98213 40.77029 -73.98729 40.77891
## 146441 -73.97213 40.75356 -73.98418 40.75870
## 146442 -73.95631 40.76780 -73.96792 40.78740
## 146443 -73.98000 40.74558 -73.99100 40.73148
## 146444 -73.96409 40.77639 -73.96641 40.79074
## 146445 -73.98239 40.76849 -73.99028 40.77182
##          total_amount      day
## 146436 3.95 Wednesday
## 146437 9.96 Wednesday
## 146438 9.30 Wednesday
## 146439 12.80 Wednesday
## 146440 6.30 Wednesday
## 146441 7.30 Wednesday
## 146442 14.15 Wednesday
## 146443 10.35 Wednesday
## 146444 12.30 Wednesday
## 146445 11.76 Wednesday
```

```
df<- df %>% filter(df$trip_distance < 5)
View(df)

sprintf("# of Rows in Dataframe: %s", nrow(df))
```

```
## [1] "# of Rows in Dataframe: 362161"
```

```
sprintf("Dataframe Size: %s", format(object.size(df), units = "MB"))
```

```
## [1] "Dataframe Size: 387.9 Mb"
```

```
min_lat <- 40.5774
max_lat <- 40.9176
min_long <- -74.15
max_long <- -73.7004
```

The filtered data set now consists of 9.4 million entries.

## Establishing Plot themes and parameters

We create a theme for our plots with a black background and establish plot margins, title sizes and font sizes, and legend particulars

```
theme_map_dark <- function(palate_color = "Greys") {

  palate <- brewer.pal(palate_color, n=12)
  color.background = "black"
  color.grid.minor = "black"
  color.grid.major = "black"
  color.axis.text = palate[1]
  color.axis.title = palate[1]
  color.title = palate[1]

  font.title <- "Source Sans Pro"
  font.axis <- "Open Sans Condensed Bold"

  theme_bw(base_size=5) +
    theme(panel.background=element_rect(fill=color.background, color=color.background)) +
    theme(plot.background=element_rect(fill=color.background, color=color.background)) +
    theme(panel.border=element_rect(color=color.background)) +
    theme(panel.grid.major=element_blank()) +
    theme(panel.grid.minor=element_blank()) +
    theme(axis.ticks=element_blank()) +
    theme(plot.title=element_text(colour=color.title,family=font.title, size=10, face = 'bold'))
  +
    theme(axis.text.x=element_blank()) +
    theme(axis.text.y=element_blank()) +
    theme(axis.title.y=element_blank()) +
    theme(axis.title.x=element_blank()) +
    theme(plot.margin = unit(c(0.0, 0.5, 1, 0.75), "mm")) +
    theme(strip.background = element_rect(fill=color.background, color=color.background),strip.t
ext=element_text(size=7, colour=color.axis.title,family=font.title))

}
```

## Data Analysis and Plot



We analyze the data on the basis of pickup, drop-offs and whether it is a weekday or a weekend and plot 4 graphs which identify places in Manhattan that are the busiest.

```
#####WEEKDAY#####
```

```
dfweekday<- df %>% filter(df$day!="Saturday" & df$day!= "Sunday")
```

```
#Pickup
```

```
plot1 <- ggplot(data=dfweekday , aes(x=dfweekday$pickup_longitude, y=dfweekday$pickup_latitude,  
z=dfweekday$tpep_pickup_datetime)) +  
  geom_point(size=0.06, color="#777777") +  
  scale_x_continuous(limits=c(-74.0224, -73.8521)) +  
  scale_y_continuous(limits=c(40.6959, 40.8348)) +  
  theme_map_dark() +  
  labs(title = "NYC Taxi: Pickup Location during weekdays in June 2016 between 8 AM to 9 AM") +  
  coord_equal()  
print(plot1)
```



The plot above shows a high concentration of Taxi Pickups in Midtown East, Sutton Place, and areas between 6th and 8th Avenues and W42nd and W58th Streets.

*#Drop-Off*

```

plot2 <- ggplot(data=dfweekday , aes(x=dfweekday$dropoff_longitude,
y=dfweekday$dropoff_latitude, z=dfweekday$tpep_dropoff_datetime )) +
  geom_point(size=0.06, color="#777777") +
  stat_summary_hex(fun = dfweekday$tpep_dropoff_datetime, bins= 100, alpha=0.5) +
  scale_x_continuous(limits=c(-74.0224, -73.8521)) +
  scale_y_continuous(limits=c(40.6959, 40.8348)) +
  theme_map_dark() +
  scale_fill_gradient(low="#CCCCCC", high="#8E44AD", trans="log", breaks=c("00:30")) +
  labs(title = "NYC Taxi: Drop-off Location during weekdays in June 2016 between 8 AM to 9 AM")
+
  coord_equal()
print(plot2)

```

**NYC Taxi: Drop-off Location during weekdays in June 2016 between 8 AM to 9 AM**



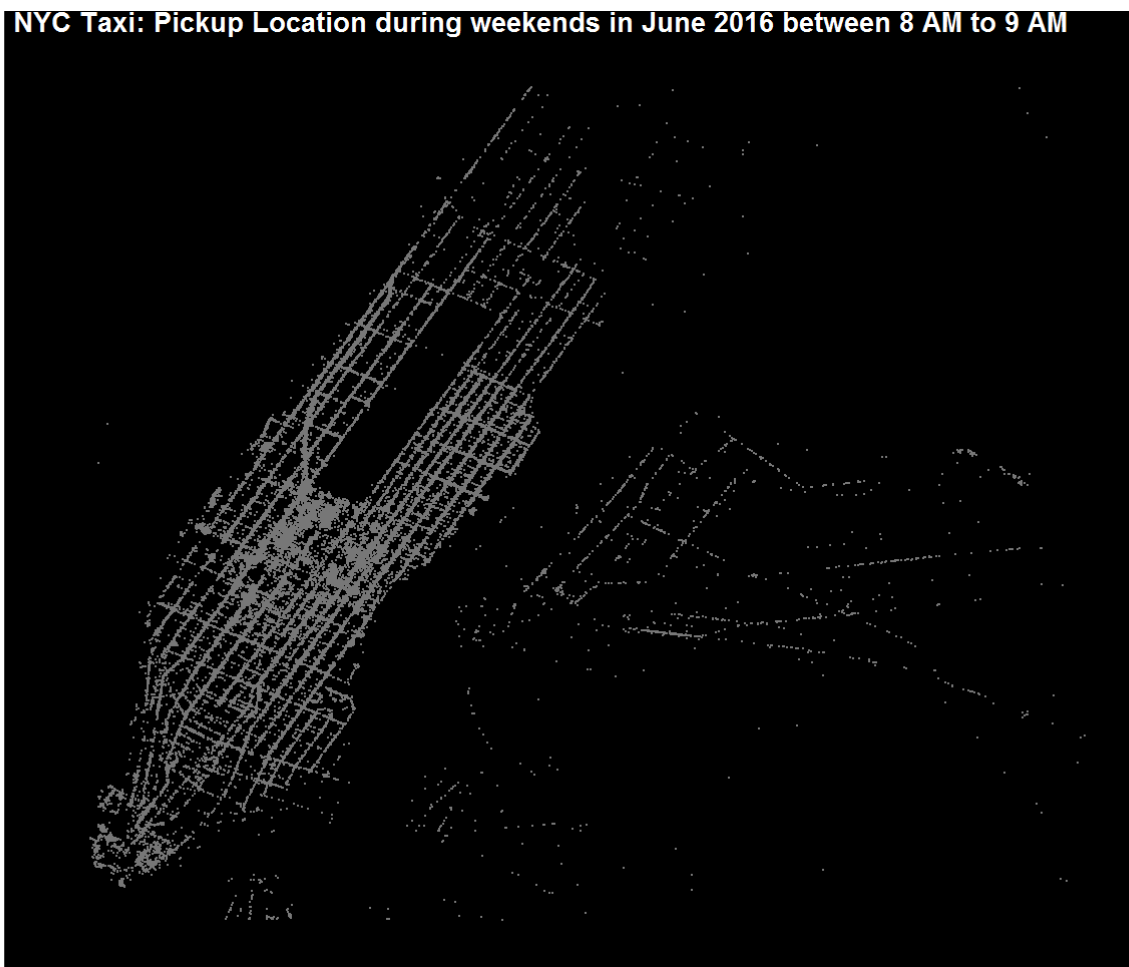
The plot above shows a high concentration of Taxi Drop-offs in Central and Eastern Midtown, some of which include the Garment District, Murray Hill, the Times Square Area, Columbus Circle, etc.

```
#####WEEKEND#####
```

```
dfweekend<- df %>% filter(df$day=="Saturday" | df$day== "Sunday")
```

```
#Pickup
```

```
plot3 <- ggplot(data=dfweekend , aes(x=dfweekend$pickup_longitude, y=dfweekend$pickup_latitude,
z=dfweekend$tpep_pickup_datetime)) +
  geom_point(size=0.06, color="#777777") +
  stat_summary_hex(fun = dfweekend$tpep_pickup_datetime, bins= 100, alpha=0.5) +
  scale_x_continuous(limits=c(-74.0224, -73.8521)) +
  scale_y_continuous(limits=c(40.6959, 40.8348)) +
  theme_map_dark() +
  scale_fill_gradient(low="#FFFFFF", high="#E74C3C", trans="log", breaks=c("00:30")) +
  labs(title = "NYC Taxi: Pickup Location during weekends in June 2016 between 8 AM to 9 AM") +
  coord_equal()
print(plot3)
```



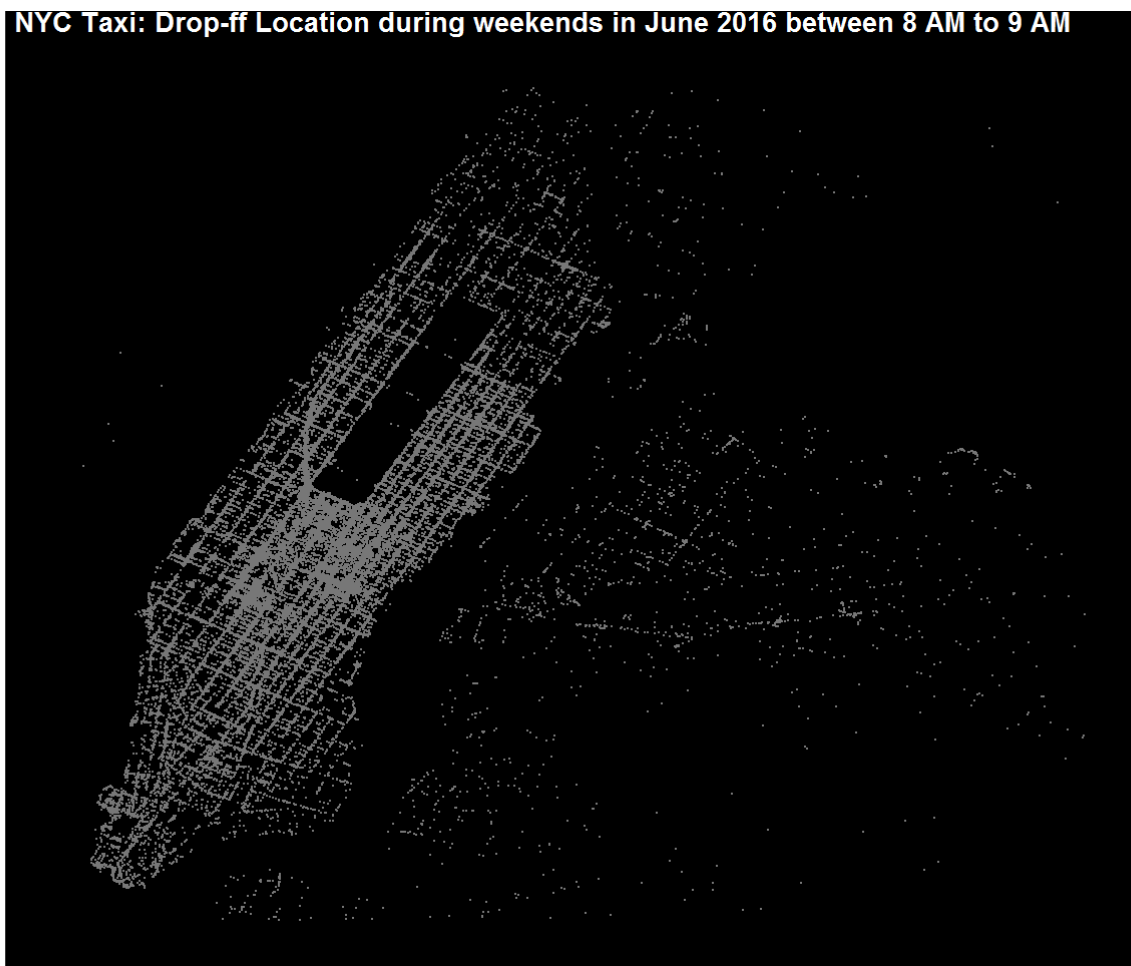
The weekend pickups show a relatively sparse distribution. Some of the areas with comparatively high pickups include the Times Square Area (a lot of hotels) and Upper Midtown on 5th Avenue (shopping).

### #Drop-Off

```

plot4 <- ggplot(data=dfweekend , aes(x=dfweekend$dropoff_longitude,
y=dfweekend$dropoff_latitude, z=dfweekend$tpep_dropoff_datetime )) +
  geom_point(size=0.06, color="#777777") +
  stat_summary_hex(fun = dfweekend$tpep_dropoff_datetime, bins= 100, alpha=0.5) +
  scale_x_continuous(limits=c(-74.0224, -73.8521)) +
  scale_y_continuous(limits=c(40.6959, 40.8348)) +
  theme_map_dark() +
  scale_fill_gradient(low="#FFFFFF", high="#E74C3C", trans="log", breaks=c("00:30")) +
  labs(title = "NYC Taxi: Drop-ff Location during weekends in June 2016 between 8 AM to 9 AM") +
  coord_equal()
print(plot4)

```



Drop-offs during the weekends are concentrated mostly around the tourist attractions, some of which include Columbus Circle, Times Square and the Rockefeller Center. People also travel to the St. Patricks Cathedral area on 50th street and Madison Avenue.

## Calculation of time in spent in traffic

We consider 3 neighborhoods for this analysis - people who travel from Midtown East to the Garment District and from Midtown East to the Times Square area on weekdays and weekends respectively.

We define these areas according to the following coordinates:

#### Midtown East = Min Long: -73.9808, Max Long: -73.9591 & Min Lat: 40.7480, Max Lat: 40.7643 ####  
 Garment District = Min Long: -73.9963, Max Long: -73.9841 & Min Lat: 40.7478, Max Lat: 40.7583 #### Times  
 Square = Min Long: -73.9916, Max Long: -73.9808 & Min Lat: 40.7542, Max Lat: 40.7613

We calculate the average time differences between these coordinates on weekdays and weekends. Assuming that there is little traffic during the weekends, the average difference would give us the extra time a person spends sitting in traffic during the weekdays.

*##Creating a new column for pickup and drop-off times during weekdays and weekends*

```
Hours1 <- format(as.POSIXct(strptime(dfweekday$tpep_pickup_datetime,"%Y-%m-%d %H:%M:%S",tz=""))
,format = "%H:%M")
head(Hours1)
```

```
## [1] "08:00" "08:00" "08:00" "08:00" "08:00" "08:00"
```

```
dfweekday$pickuptime<-Hours1
```

```
Hours2 <- format(as.POSIXct(strptime(dfweekday$tpep_dropoff_datetime,"%Y-%m-%d
%H:%M:%S",tz="")) ,format = "%H:%M")
head(Hours2)
```

```
## [1] "08:00" "08:08" "08:11" "08:14" "08:05" "08:06"
```

```
dfweekday$dropofftime<-Hours2
```

```
Hours3 <- format(as.POSIXct(strptime(dfweekend$tpep_pickup_datetime,"%Y-%m-%d %H:%M:%S",tz=""))
,format = "%H:%M")
head(Hours3)
```

```
## [1] "08:07" "08:07" "08:07" "08:07" "08:07" "08:07"
```

```
dfweekend$pickuptime<-Hours3
```

```
Hours4 <- format(as.POSIXct(strptime(dfweekend$tpep_dropoff_datetime,"%Y-%m-%d
%H:%M:%S",tz="")) ,format = "%H:%M")
head(Hours4)
```

```
## [1] "08:24" "08:16" "08:26" "08:20" "08:09" "08:11"
```

```

dfweekend$dropofftime<-Hours4

## Converting time to Hour and Minute format to be read by R and creating a column for time differences

###Weekdays

pickupWeekday<-as.POSIXct(dfweekday$pickuptime,format="%H:%M")
dropoffWeekday<-as.POSIXct(dfweekday$dropofftime,format="%H:%M")

dfweekday$diffinmin <- difftime(dropoffWeekday,pickupWeekday,tz,units = "mins")

###Weekends

pickupWeekend<-as.POSIXct(dfweekend$pickuptime,format="%H:%M")
dropoffWeekend<-as.POSIXct(dfweekend$dropofftime,format="%H:%M")

dfweekend$diffinmin <- difftime(dropoffWeekend,pickupWeekend,tz,units = "mins")

```

## Midtown East to Times Square

```

#For Weekdays (Monday to Friday)

M_T_Weekday<- subset(dfweekday,subset= dfweekday$pickup_longitude >=-73.9808 & dfweekday$pickup_
longitude <=-73.9591
      & dfweekday$pickup_latitude >= 40.7480 & dfweekday$pickup_latitude <= 40.7643
      & dfweekday$dropoff_longitude >= -73.9916 & dfweekday$dropoff_longitude <= -73.980
8
      & dfweekday$dropoff_latitude >= 40.7542 & dfweekday$dropoff_latitude <= 40.7613
)

head(M_T_Weekday)

```

```
##      tpep_pickup_datetime tpep_dropoff_datetime trip_distance
## 6      2016-06-01 08:00:00      2016-06-01 08:06:23          0.72
## 23     2016-06-01 08:00:05      2016-06-01 08:12:49          1.15
## 40     2016-06-01 08:00:09      2016-06-01 08:09:36          1.20
## 169    2016-06-01 08:03:42      2016-06-01 08:08:12          0.00
## 208    2016-06-01 08:00:32      2016-06-01 08:07:23          0.63
## 222    2016-06-01 08:00:34      2016-06-01 08:15:01          1.66
##      pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude
## 6      -73.97213      40.75356      -73.98418      40.75870
## 23     -73.97250      40.76207      -73.98604      40.75847
## 40     -73.96573      40.75453      -73.98104      40.75611
## 169    -73.97315      40.75353      -73.98140      40.75775
## 208    -73.97465      40.75735      -73.98412      40.75770
## 222    -73.96870      40.76147      -73.99004      40.75647
##      total_amount      day pickuptime dropofftime diffinmin
## 6          7.30 Wednesday      08:00      08:06      6 mins
## 23         10.30 Wednesday      08:00      08:12     12 mins
## 40          9.95 Wednesday      08:00      08:09      9 mins
## 169         6.89 Wednesday      08:03      08:08      5 mins
## 208         7.82 Wednesday      08:00      08:07      7 mins
## 222        11.30 Wednesday      08:00      08:15     15 mins
```

```
mean(M_T_Weekday$trip_distance)
```

```
## [1] 0.9566856
```

```
#For Weekends (Saturday and Sunday)
```

```
M_T_Weekend<- subset(dfweekend,subset= dfweekend$pickup_longitude >=-73.9808 & dfweekend$pickup_
longitude <=-73.9591
      & dfweekend$pickup_latitude >=40.7480 & dfweekend$pickup_latitude <=40.7643
      & dfweekend$dropoff_longitude >=-73.9916 & dfweekend$dropoff_longitude <=-73.9808
      & dfweekend$dropoff_latitude >=40.7542 & dfweekend$dropoff_latitude <=40.7613
)
head(M_T_Weekend)
```

```
##      tpep_pickup_datetime tpep_dropoff_datetime trip_distance
## 213  2016-06-04 08:17:08   2016-06-04 08:27:16           1.47
## 240  2016-06-04 08:17:20   2016-06-04 08:22:07           0.90
## 363  2016-06-04 08:24:32   2016-06-04 08:27:16           0.53
## 476  2016-06-04 08:19:02   2016-06-04 08:25:18           1.10
## 501  2016-06-04 08:19:11   2016-06-04 08:25:56           1.61
## 707  2016-06-04 08:00:26   2016-06-04 08:04:28           0.90
##      pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude
## 213      -73.97655      40.74805      -73.98595      40.75878
## 240      -73.97714      40.75862      -73.98866      40.75683
## 363      -73.97950      40.76386      -73.98590      40.75756
## 476      -73.97482      40.76241      -73.98942      40.75680
## 501      -73.97022      40.75955      -73.99081      40.75541
## 707      -73.96986      40.76403      -73.98162      40.75768
##      total_amount      day pickuptime dropofftime diffinmin
## 213          8.80 Saturday      08:17      08:27    10 mins
## 240          6.30 Saturday      08:17      08:22     5 mins
## 363          5.76 Saturday      08:24      08:27     3 mins
## 476          7.30 Saturday      08:19      08:25     6 mins
## 501          9.00 Saturday      08:19      08:25     6 mins
## 707          7.55 Saturday      08:00      08:04     4 mins
```

```
mean(M_T_Weekend$trip_distance)
```

```
## [1] 1.225527
```

```
##Calculating extra time taken during weekdays
```

```
mean(M_T_Weekend$diffinmin)
```

```
## Time difference of 7.85755 mins
```

```
mean(M_T_Weekday$diffinmin)
```

```
## Time difference of 11.07932 mins
```

```
Time_Difference1 <- mean(M_T_Weekday$diffinmin) - mean(M_T_Weekend$diffinmin)
Time_Difference1
```

```
## Time difference of 3.22177 mins
```

It takes an average of 3.22 mins more to get to Times Square from Midtown East on a weekday between 8AM and 9AM.

## Midtown to Garment District



```
#For Weekdays (Monday to Friday)
```

```
M_G_Weekday<- subset(dfweekday,subset= dfweekday$pickup_longitude >=-73.9808 & dfweekday$pickup_longitude <=-73.9591
& dfweekday$pickup_latitude >=40.7480 & dfweekday$pickup_latitude <=40.7643
& dfweekday$dropoff_longitude >=-73.9963 & dfweekday$dropoff_longitude <=-73.9841
& dfweekday$dropoff_latitude >=40.7478 & dfweekday$dropoff_latitude <=40.7583
)
head(M_G_Weekday)
```

```
##      tpep_pickup_datetime tpep_dropoff_datetime trip_distance
## 208 2016-06-01 08:00:32 2016-06-01 08:07:23 0.63
## 222 2016-06-01 08:00:34 2016-06-01 08:15:01 1.66
## 361 2016-06-01 08:03:54 2016-06-01 08:20:11 0.80
## 533 2016-06-01 08:04:01 2016-06-01 08:08:39 0.82
## 543 2016-06-01 08:04:03 2016-06-01 08:09:48 0.91
## 615 2016-06-01 08:01:31 2016-06-01 08:07:41 0.80
##      pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude
## 208 -73.97465 40.75735 -73.98412 40.75770
## 222 -73.96870 40.76147 -73.99004 40.75647
## 361 -73.97387 40.74913 -73.98521 40.75288
## 533 -73.97863 40.75222 -73.98804 40.75214
## 543 -73.97594 40.76010 -73.98429 40.74857
## 615 -73.97821 40.75216 -73.98733 40.75114
##      total_amount      day pickuptime dropofftime diffinmin
## 208 7.82 Wednesday 08:00 08:07 7 mins
## 222 11.30 Wednesday 08:00 08:15 15 mins
## 361 14.12 Wednesday 08:03 08:20 17 mins
## 533 6.96 Wednesday 08:04 08:08 4 mins
## 543 8.16 Wednesday 08:04 08:09 5 mins
## 615 6.80 Wednesday 08:01 08:07 6 mins
```

```
mean(M_G_Weekday$trip_distance)
```

```
## [1] 1.207938
```

```
#For Weekends (Saturday and Sunday)
```

```
M_G_Weekend <- subset(dfweekend,subset= dfweekend$pickup_longitude >= -73.9808 & dfweekend$pickup_longitude <= -73.9591
& dfweekend$pickup_latitude >= 40.7480 & dfweekend$pickup_latitude <= 40.7643
& dfweekend$dropoff_longitude >= -73.9963 & dfweekend$dropoff_longitude <= -73.9841
& dfweekend$dropoff_latitude >= 40.7478 & dfweekend$dropoff_latitude <= 40.7583
)
head(M_G_Weekend)
```

```
##      tpep_pickup_datetime tpep_dropoff_datetime trip_distance
## 70      2016-06-04 08:07:48      2016-06-04 08:12:49          1.08
## 149     2016-06-04 08:16:38      2016-06-04 08:24:53          1.43
## 240     2016-06-04 08:17:20      2016-06-04 08:22:07          0.90
## 280     2016-06-04 08:17:43      2016-06-04 08:22:50          1.01
## 363     2016-06-04 08:24:32      2016-06-04 08:27:16          0.53
## 381     2016-06-04 08:18:24      2016-06-04 08:26:45          1.70
##      pickup_longitude pickup_latitude dropoff_longitude dropoff_latitude
## 70      -73.97872      40.74991      -73.99385      40.74988
## 149     -73.97296      40.74910      -73.99301      40.74968
## 240     -73.97714      40.75862      -73.98866      40.75683
## 280     -73.97754      40.75455      -73.99055      40.75113
## 363     -73.97950      40.76386      -73.98590      40.75756
## 381     -73.96922      40.76064      -73.99085      40.75092
##      total_amount      day pickuptime dropofftime diffinmin
## 70      6.80 Saturday      08:07      08:12      5 mins
## 149     8.30 Saturday      08:16      08:24      8 mins
## 240     6.30 Saturday      08:17      08:22      5 mins
## 280     7.56 Saturday      08:17      08:22      5 mins
## 363     5.76 Saturday      08:24      08:27      3 mins
## 381    10.56 Saturday      08:18      08:26      8 mins
```

```
View(M_G_Weekend)
mean(M_G_Weekend$trip_distance)
```

```
## [1] 1.518019
```

```
mean(M_G_Weekend$diffinmin)
```

```
## Time difference of 9.068339 mins
```

```
mean(M_G_Weekday$diffinmin)
```

```
## Time difference of 12.88285 mins
```

```
Time_Difference2 <- mean(M_G_Weekday$diffinmin)-mean(M_G_Weekend$diffinmin)
Time_Difference2
```

```
## Time difference of 3.814514 mins
```

It takes an average of 3.82 mins longer on a weekday between 8AM and 9AM, to reach the Garment District from Midtown East.