

Screenshots showing files and user IDs are not present for all 4 members of the team

aws Services Resource Groups

Amazon S3 > data-science-deepak

Overview Properties Permissions Management

Q Type a prefix and press Enter to search. Press ESC to clear.

Upload Create folder More

US West (Oregon)

Viewing 1 to 3

Name	Last modified	Size	Storage class
Parking_Violations_Issued_-_Fiscal_Year_2015.csv	Jul 12, 2018 11:06:57 AM GMT+0530	2.7 GB	Standard
Parking_Violations_Issued_-_Fiscal_Year_2016.csv	Jul 12, 2018 11:03:50 AM GMT+0530	2.0 GB	Standard
Parking_Violations_Issued_-_Fiscal_Year_2017.csv	Jul 12, 2018 11:01:44 AM GMT+0530	1.9 GB	Standard

Viewing 1 to 3

aws Services Resource Groups

Amazon S3 > upgradkumar2018

Overview Properties Permissions Management

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US West (Oregon)

Viewing 1 to 3

Name	Last modified	Size	Storage class
Parking_Violations_Issued_-_Fiscal_Year_2015.csv	Jul 10, 2018 9:56:43 PM GMT+0530	2.7 GB	Standard
Parking_Violations_Issued_-_Fiscal_Year_2016.csv	Jul 10, 2018 9:59:01 PM GMT+0530	2.0 GB	Standard
Parking_Violations_Issued_-_Fiscal_Year_2017.csv	Jul 10, 2018 10:01:07 PM GMT+0530	1.9 GB	Standard

Viewing 1 to 3

AnalysisNYC-PaS3 ManRStudioRStudioRStudioRStudioRStudioRStudioGgplotRStudio

Secure | https://s3.console.aws.amazon.com/s3/buckets/datascience-dharma/Assignment/?region=us-west-2&tab=overview

Tata Sky online rechState Bank of IndiaHoliday Packages | T akamaicdn.webex.coGmailGoogleGoogle Cloud ConscPD200: Developing

awsServicesResource Groupsdharmanandanareddy.pothula

Amazon S3 > datascience-dharma / Assignment

Overview

Type a prefix and press Enter to search. Press ESC to clear.

UploadCreate folderMore

<input type="checkbox"/>	Name	Last modified		
<input type="checkbox"/>	Parking_Violations_Issued_-_Fiscal_Year_2015.csv	Jul 12, 2018 7:52:03 AM GMT+0530	2.7 GB	Standard
<input type="checkbox"/>	Parking_Violations_Issued_-_Fiscal_Year_2016.csv	Jul 12, 2018 7:54:02 AM GMT+0530	2.0 GB	Standard
<input type="checkbox"/>	Parking_Violations_Issued_-_Fiscal_Year_2017.csv	Jul 12, 2018 7:55:09 AM GMT+0530	1.9 GB	Standard

IAM User:
dharmanandanareddy.pothula@iitb.net

Account:
3273-8586-4877

My Account

My Organization

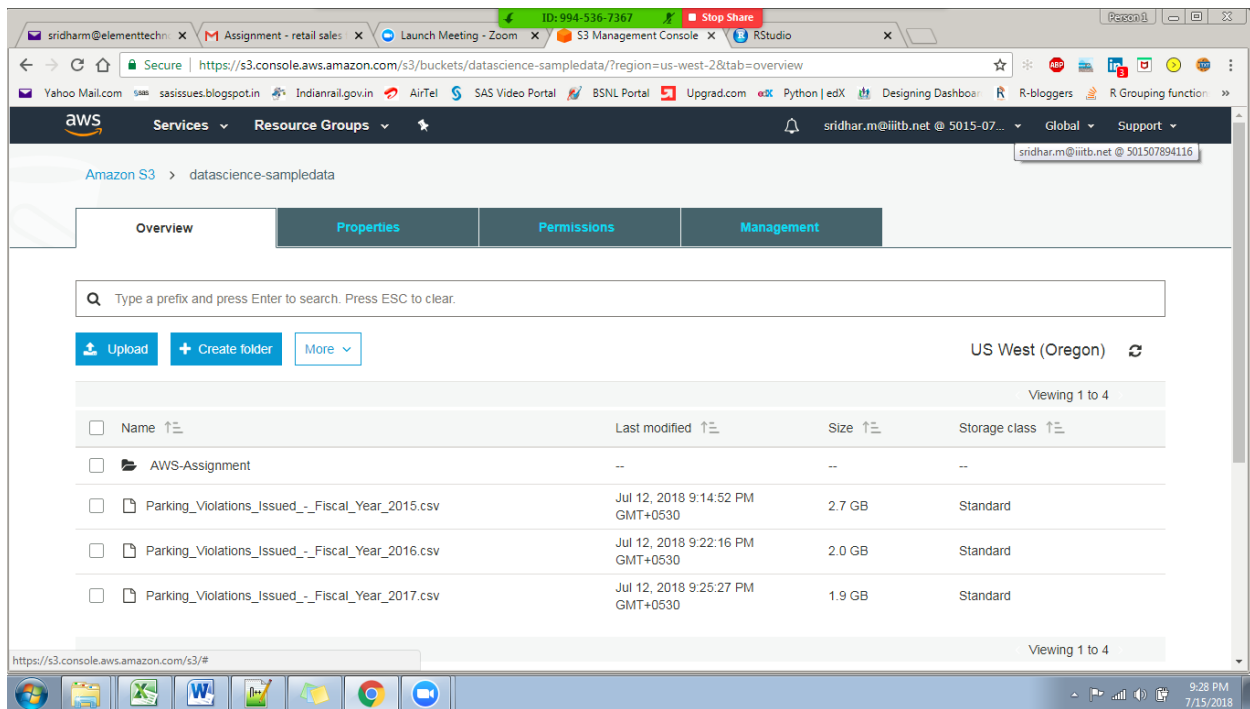
My Billing Dashboard

My Security Credentials

Switch Role

Sign Out

Type here to search



Examine the data.

1. Find total number of tickets for each year.

2015: 10951257 (51 columns)

2016: 10626899 (51 Columns)

2017: 10803028

2. Find out how many unique states the cars which got parking tickets came from.

2015: 68 (99 excluded)

2016: 68 (99 excluded)

2017: 67

3. Some parking tickets don't have addresses on them, which is cause for concern. Find out how many such tickets there are.

2015: 5463

2016: 8274

2017: 4009

Aggregation tasks

1. How often does each violation code occur? (frequency of violation codes - find the top 5)

2015

#ViolationCode	count
----------------	-------

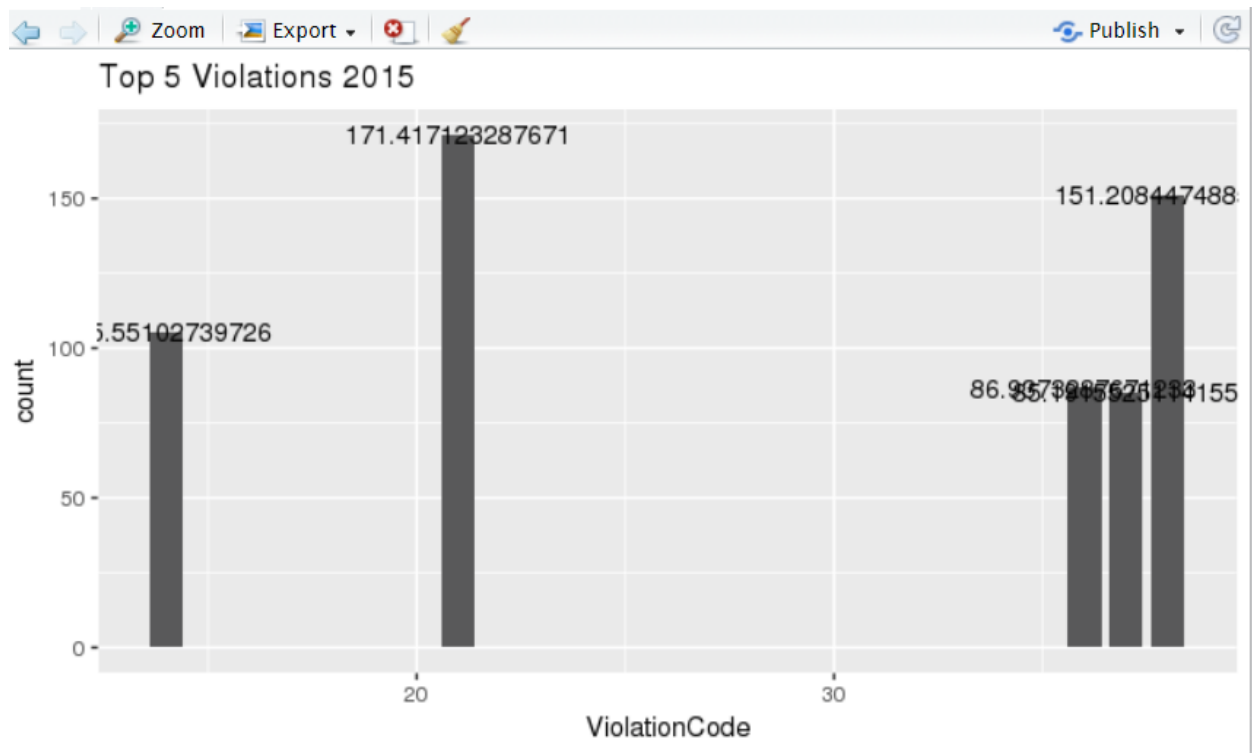
#1	21 171.41712
----	--------------

#2	38 151.20845
----	--------------

#3	14 105.55103
----	--------------

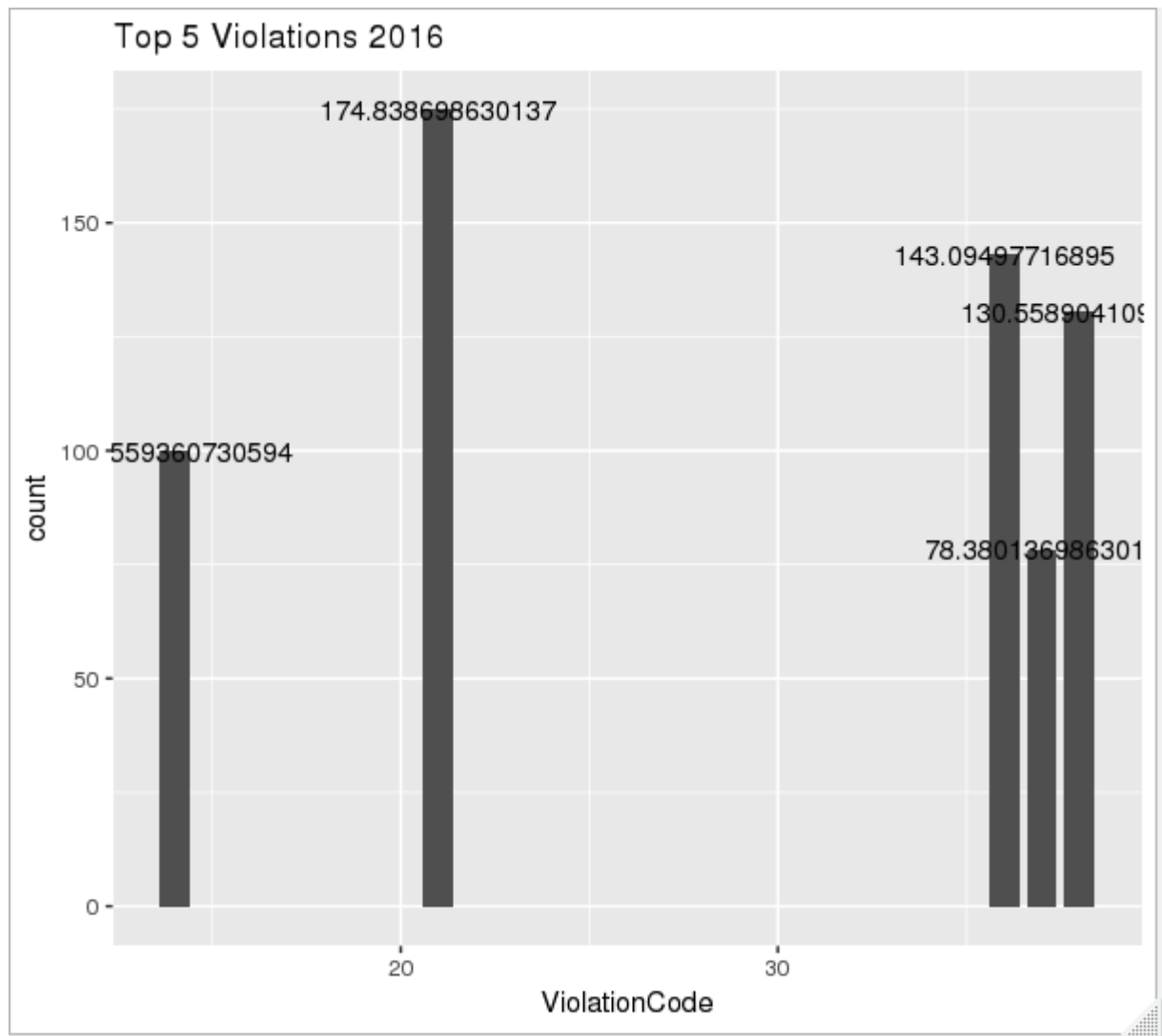
#4	36 86.93733
----	-------------

#5	37 85.19155
----	-------------



2016

```
#ViolationCode    count
#                21 174.83870
#                38 130.55890
#                14  99.95594
#                36 143.09498
#                37  78.38014
```



2017

ViolationCode count

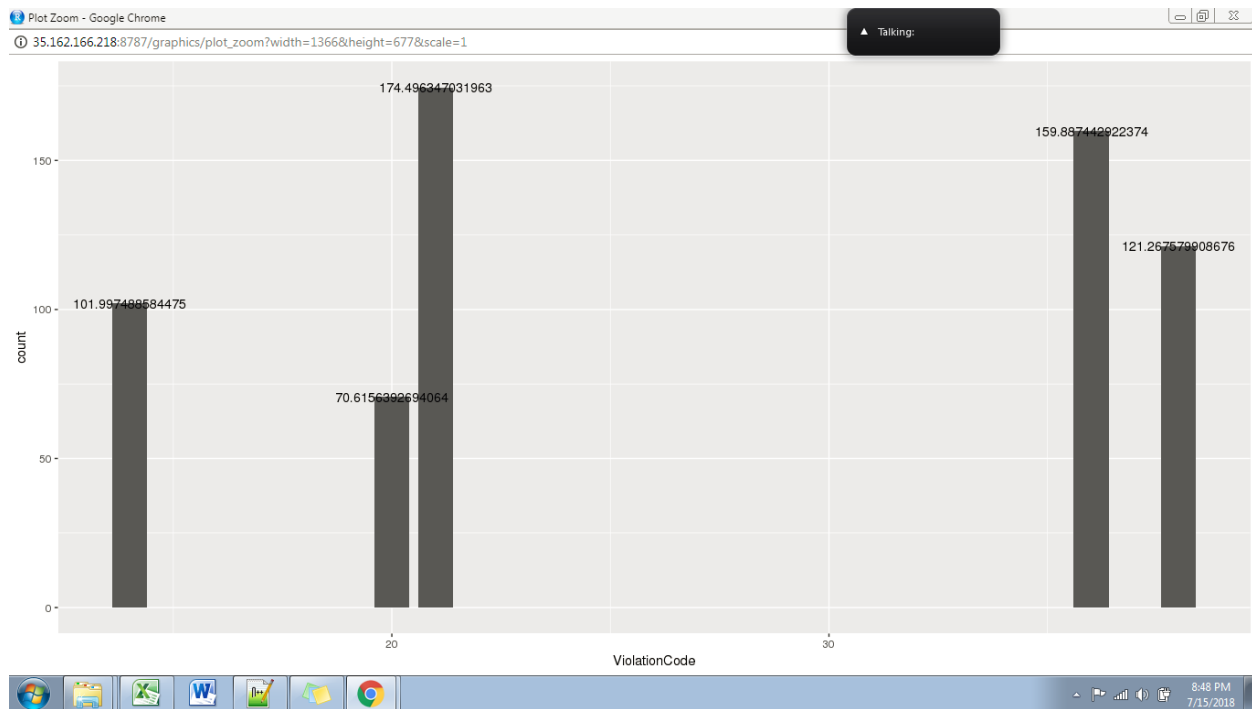
#1 21 174.49635

#2 36 159.88744

#3 38 121.26758

#4 14 101.99749

#5 20 70.61564

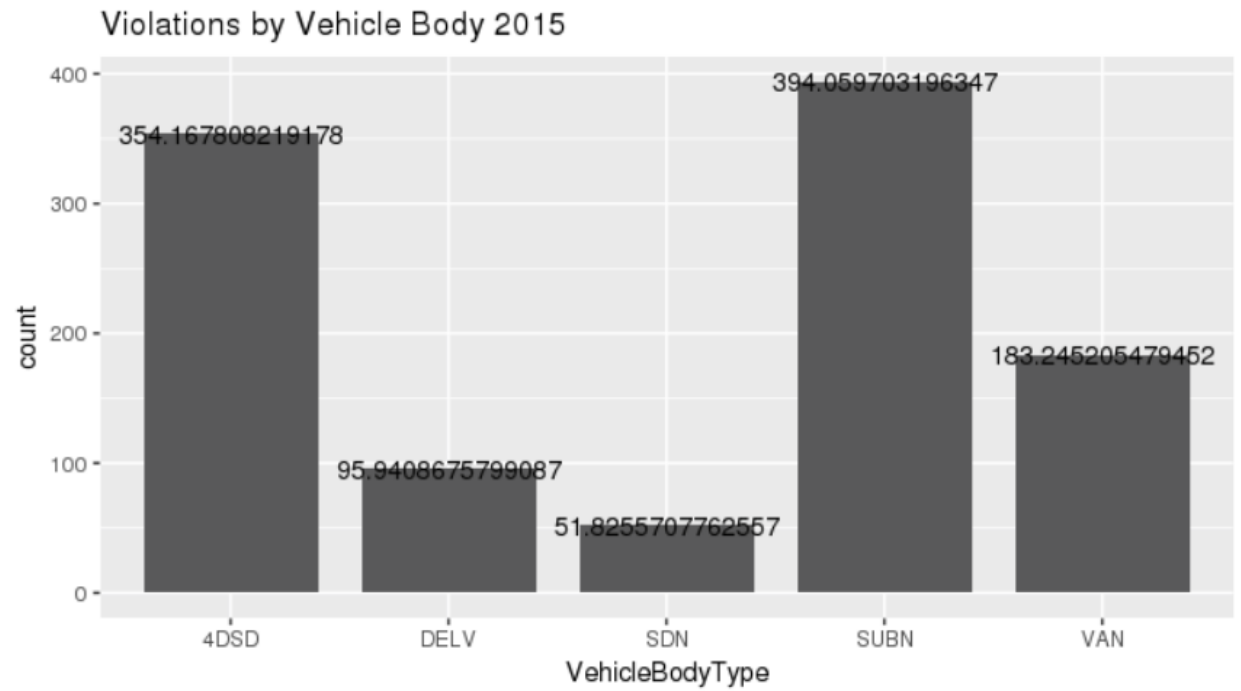


2. How often does each vehicle body type get a parking ticket? How about the vehicle make? (find the top 5 for both)

2015

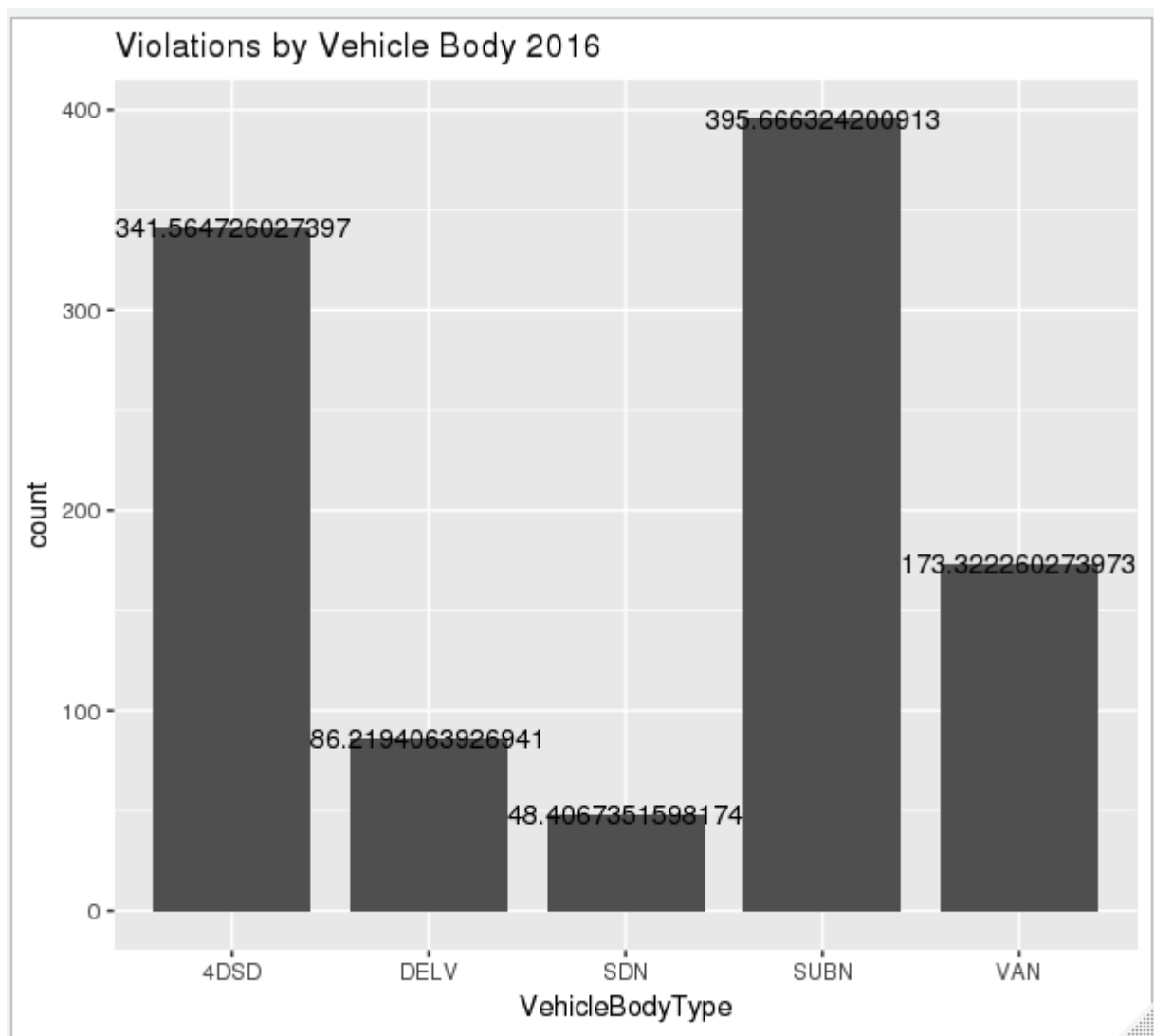
VehicleBodyType count

- 1 SUBN 394.05970
- 2 4DSD 354.16781
- 3 VAN 183.24521
- 4 DELV 95.94087
- 5 SDN 51.82557



2016

#VehicleBodyType	count
#1 SUBN	395.66632
#2 4DSD	341.56473
#3 VAN	173.32226
#4 DELV	86.21941
#5 SDN	48.40674



2017

VehicleBodyType count

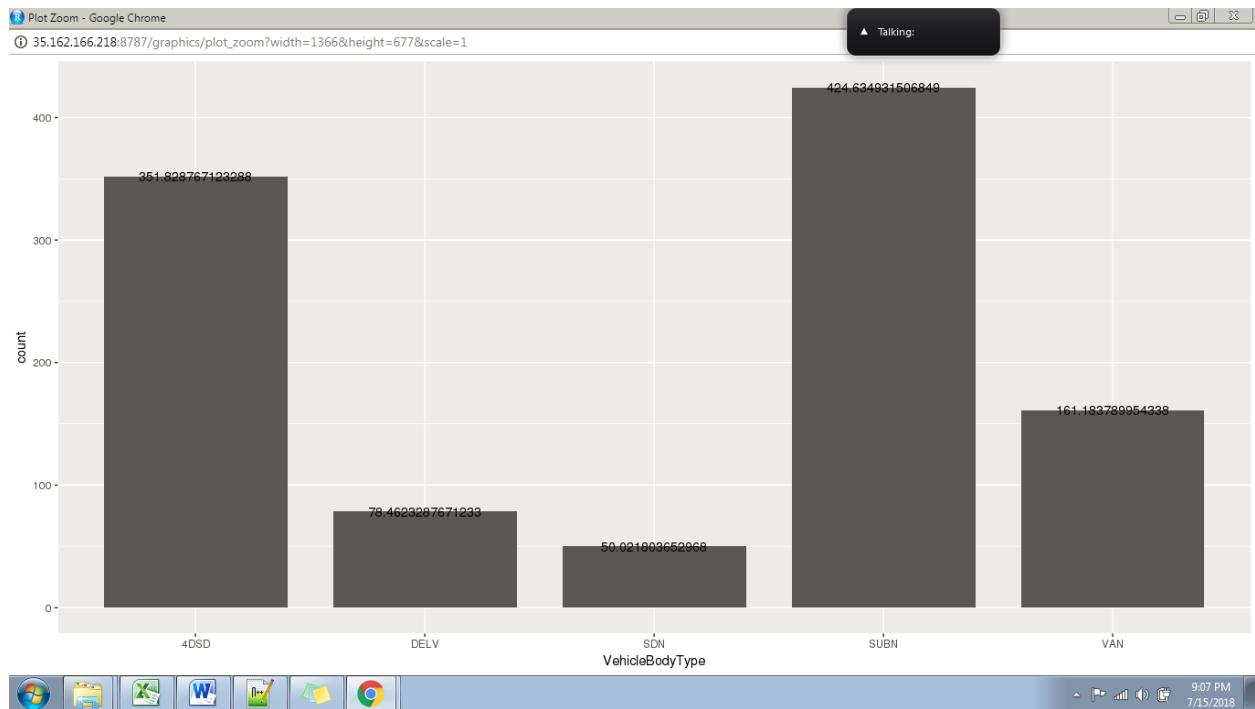
#1 SUBN 424.63493

#2 4DSD 351.82877

#3 VAN 161.18379

#4 DELV 78.46233

#5 SDN 50.02180



Vehicle Make Violations

2015 (Ford)

VehicleMake count

- 1 FORD 161.7926
- 2 TOYOT 128.2561
- 3 HONDA 116.2156
- 4 NISSA 95.6129
- 5 CHEVR 95.4782

2016

#VehicleMake	count
#1 FORD	151.22991
#2 TOYOT	131.82534
#3 HONDA	115.76187
#4 NISSA	95.30057
#5 CHEVR	86.71952

2017

VehicleMake count

#1 FORD 146.22808

#2 TOYOT 138.29349

#3 HONDA 123.20068

#4 NISSA 104.86187

#5 CHEVR 81.58162

3. A precinct is a police station that has a certain zone of the city under its command. Find the (5 highest) frequencies of:

1. Violating Precincts (this is the precinct of the zone where the violation occurred)
2. Issuing Precincts (this is the precinct that issued the ticket)

2015

#ViolationPrecinct count

#1 0 1633006

#2 19 559716

#3 18 400887

#4 14 384596

#5 1 307808

#IssuerPrecinct count

#1 0 1834343

#2 19 544946

#3 18 391501

#4 14 369725

#5 1 298594

2016

#ViolationPrecinct count

#1 0 1868655

#2 19 554465

#3 18 331704

#4 14 324467

#5 1 303850

#IssuerPrecinct count

#1 0 1834343

#2 19 544946

#3 18 391501

#4 14 369725

#5 1 298594

2017

ViolationPrecinct count

#1 0 2072400

#2 19 535671

#3 14 352450

#4 1 331810

#5 18 306920

IssuerPrecinct count

#1 0 2388479

#2 19 521513

#3 14 344977

#4 1 321170

#5 18 296553

[1] 3206545

#ViolationCode count

#1 36 1400614

#2 7 516389

#3 21 326255

[1] 3206545

[1] 2914991

ViolationCode count

#1 36 1400614

#2 7 516388

#3 14 154780

4. Find the violation code frequency across 3 precincts which have issued the most number of tickets - do these precinct zones have an exceptionally high frequency of certain violation codes? Are these codes common across precincts?

2015

#ViolationPrecinct count

#1 0 1619126

#2 25 9

#3 18 3

#4 104 3

#5 7 2

#6 106 2

#7 14 2

#8 26 1

#9 44 1

#10	122	1
#11	47	1
#12	1	1
#13	13	1
#14	48	1
#15	41	1
#16	33	1
#17	10	1
#18	77	1
#19	102	1
#20	50	1
#21	113	1
#22	121	1
#23	75	1
#24	71	1
#25	123	1
#26	901	1
#27	66	1

2016

#Violation	Precinct	count
#1	0	1619126
#2	25	9
#3	18	3
#4	104	3
#5	7	2
#6	106	2
#7	14	2
#8	26	1
#9	44	1
#10	122	1
#11	47	1
#12	1	1

#13	13	1
#14	48	1
#15	41	1
#16	33	1
#17	10	1
#18	77	1
#19	102	1
#20	50	1
#21	113	1
#22	121	1
#23	75	1
#24	71	1
#25	123	1
#26	901	1
#27	66	1

2017

[1] 2062664

ViolationPrecinct count

#1	0	2062644
#2	13	5
#3	7	3
#4	1	2
#5	78	2
#6	14	2

#7	34	1
#8	6	1
#9	17	1
#10	84	1
#11	70	1
#12	10	1

5. You'd want to find out the properties of parking violations across different times of the day:

- The Violation Time field is specified in a strange format. Find a way to make this into a time attribute that you can use to divide into groups.
- Find a way to deal with missing values, if any.
- Divide 24 hours into 6 equal discrete bins of time. The intervals you choose are at your discretion. For each of these groups, find the 3 most commonly occurring violations
- Now, try another direction. For the 3 most commonly occurring violation codes, find the most common times of day (in terms of the bins from the previous part)

2015

#bin_number ViolationCode count

#1 1 21 63574

#2 1 40 36490

#3 1 78 34842

#1 2 14 134458

#2 2 21 106858

#3 2 40 91344

#bin_number ViolationCode count

#1 3 21 1192163

#2 3 38 449070

#3 3 36 360365

#bin_number ViolationCode count

#1 4 38 432287

#2 4 37 324905

#3 4 36 220663

#1 5 38 241327

#2 5 37 175802

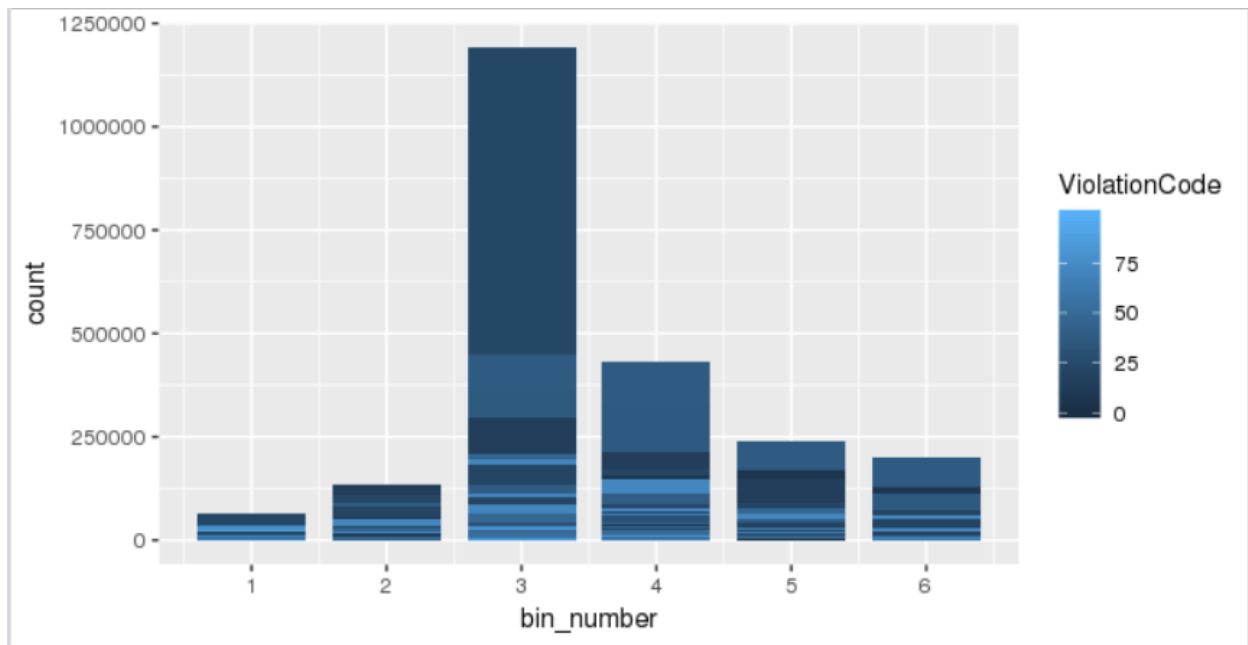
#3 5 7 168888

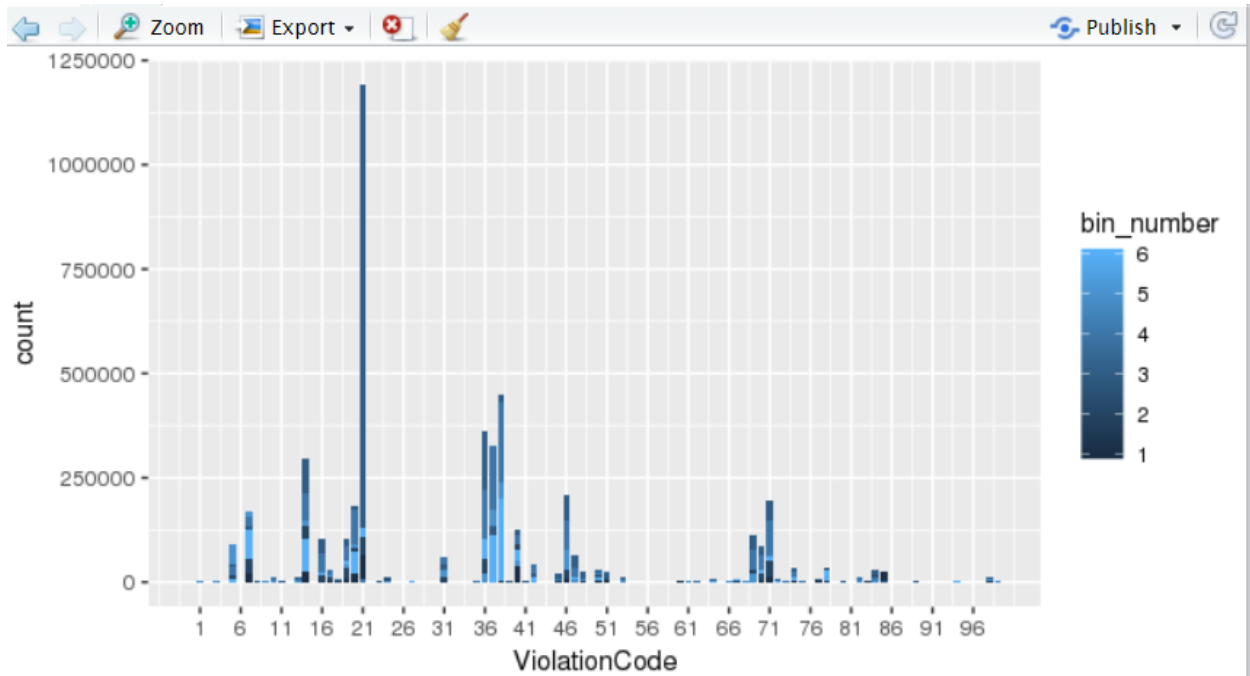
#bin_number ViolationCode count

#1 6 38 198472

#2 6 21 130164

#3 6 7 124456





2016

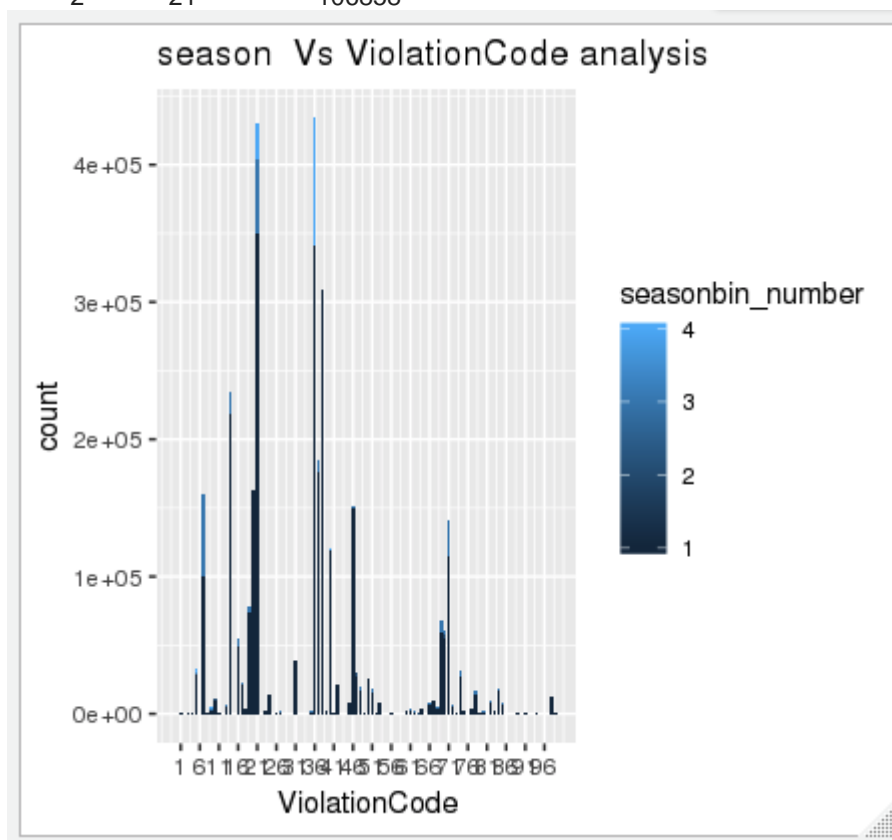
bin_number	ViolationCode	count
1	21	63574
1	40	36490
1	78	34842
1	14	26545
1	85	24865
2	14	134458
2	21	106858
2	40	91344
2	20	81103
2	36	56550
3	21	1192163
3	38	449070
3	36	360365
3	14	297711
3	46	210978
4	38	432287
4	37	324905
4	36	220663
4	14	211833
4	20	171943
5	38	241327
5	37	175802
5	7	168888
5	14	148538

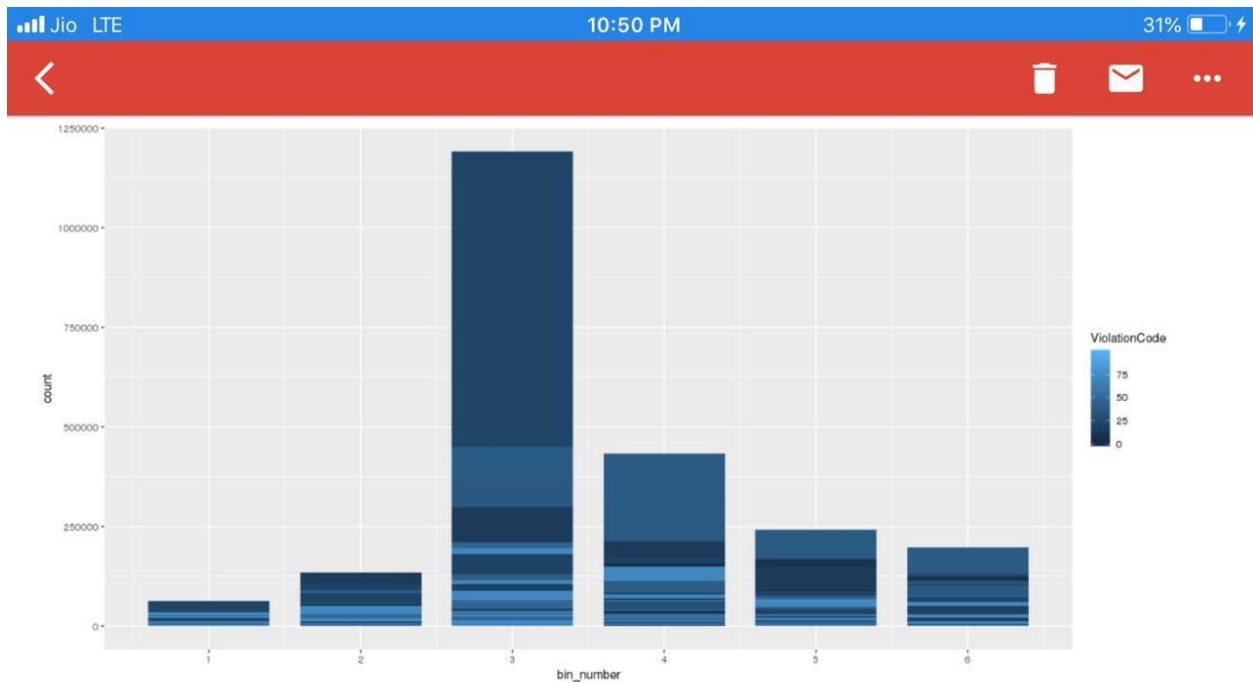
5	5	89709
6	38	198472
6	21	130164
6	7	124456
6	37	112638
6	14	105542

bin_number	ViolationCode	count
3	36	360365
4	36	220663
6	36	102881

bin_number	ViolationCode	count
3	38	449070
4	38	432287
5	38	241327

bin_number	ViolationCode	count
3	21	1192163
6	21	130164
2	21	106858





2017

```
# bin_number ViolationCode count
```

```
#1      3      21 1182689
```

```
#2      3      36 751422
```

```
#3      4      36 376961
```

```
#4      4      38 356354
```

```
#5      3      38 346518
```

```
#6      3      14 274288
```

```
# [1] 572 3
```

```
#bin_number ViolationCode count
```

```
#1      1      21 73160
```

#2	1	40 45960
----	---	----------

#3	1	14 29311
----	---	----------

# bin_number	ViolationCode	count
--------------	---------------	-------

#1	2	14 141276
----	---	-----------

#2	2	21 119469
----	---	-----------

#3	2	40 112186
----	---	-----------

#bin_number	ViolationCode	count
-------------	---------------	-------

#1	3	21 1182689
----	---	------------

#2	3	36 751422
----	---	-----------

#3	3	38 346518
----	---	-----------

#bin_number	ViolationCode	count
-------------	---------------	-------

#1	4	36 376961
----	---	-----------

#2	4	38 356354
----	---	-----------

#3	4	37 265869
----	---	-----------

#bin_number	ViolationCode	count
-------------	---------------	-------

#1	5	38 203232
----	---	-----------

#2	5	37 145784
----	---	-----------

#3	5	14 144749
----	---	-----------

#bin_number	ViolationCode	count
-------------	---------------	-------

#1	6	36 211434
----	---	-----------

#2	6	38 153537
----	---	-----------

#3	6	21 144082
----	---	-----------

# bin_number	ViolationCode	count
--------------	---------------	-------

#1	4	38 356354
----	---	-----------

#2	3	38 346518
----	---	-----------

#3 5 38 203232

#bin_number ViolationCode count

#1 3 21 1182689

#2 6 21 144082

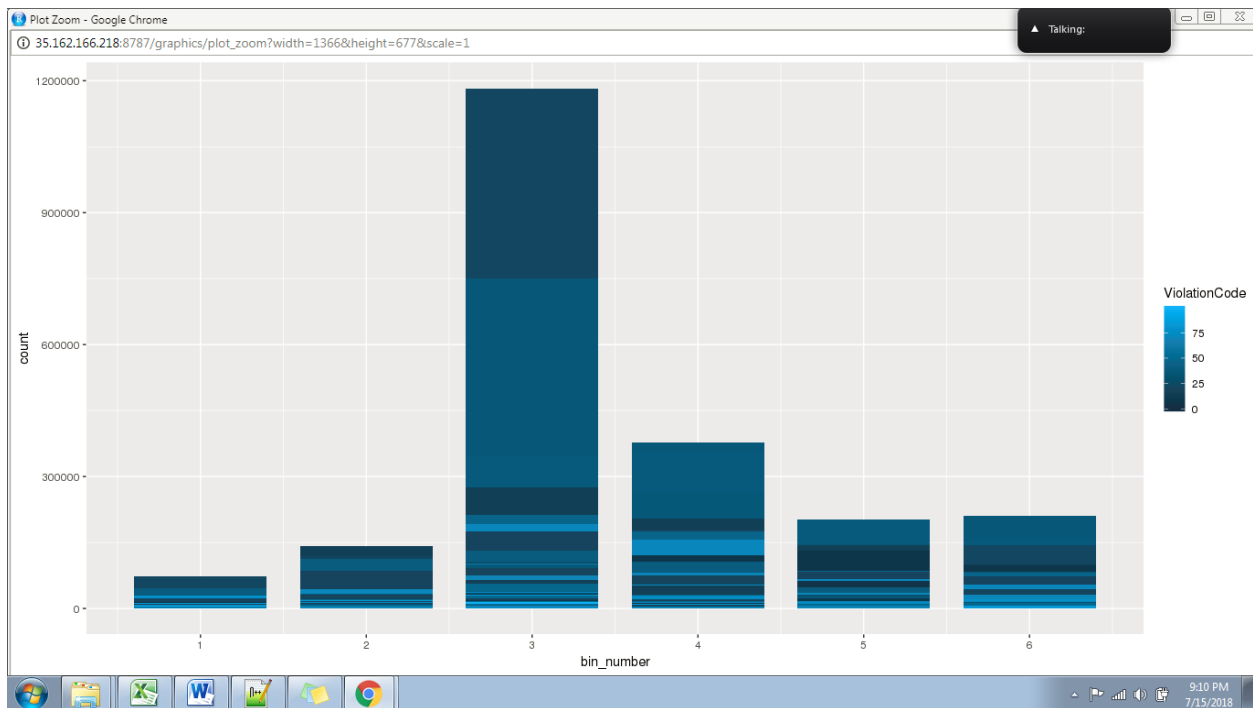
#3 2 21 119469

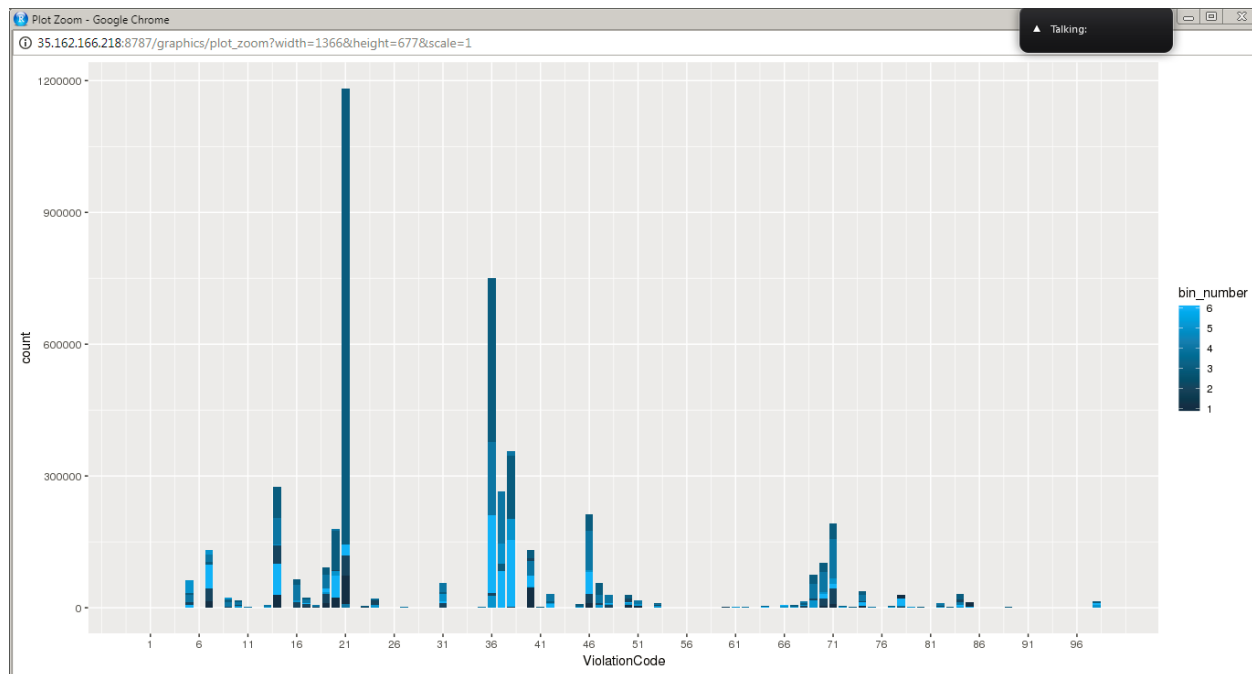
#bin_number ViolationCode count

#1 3 36 751422

#2 4 36 376961

#3 6 36 211434





6. Let's try and find some seasonality in this data

- First, divide the year into some number of seasons, and find frequencies of tickets for each season.
- Then, find the 3 most common violations for each of these season

2015

#seasonbin_number ViolationCode count

#1 1 38 336762

#2 1 21 281600

#3 1 14 220029

#seasonbin_number ViolationCode count

#1 2 21 471580

#2 2 38 346719

#3 2 14 262595

#seasonbin_number ViolationCode count

#1 3 21 397871

#2 3 38 348466

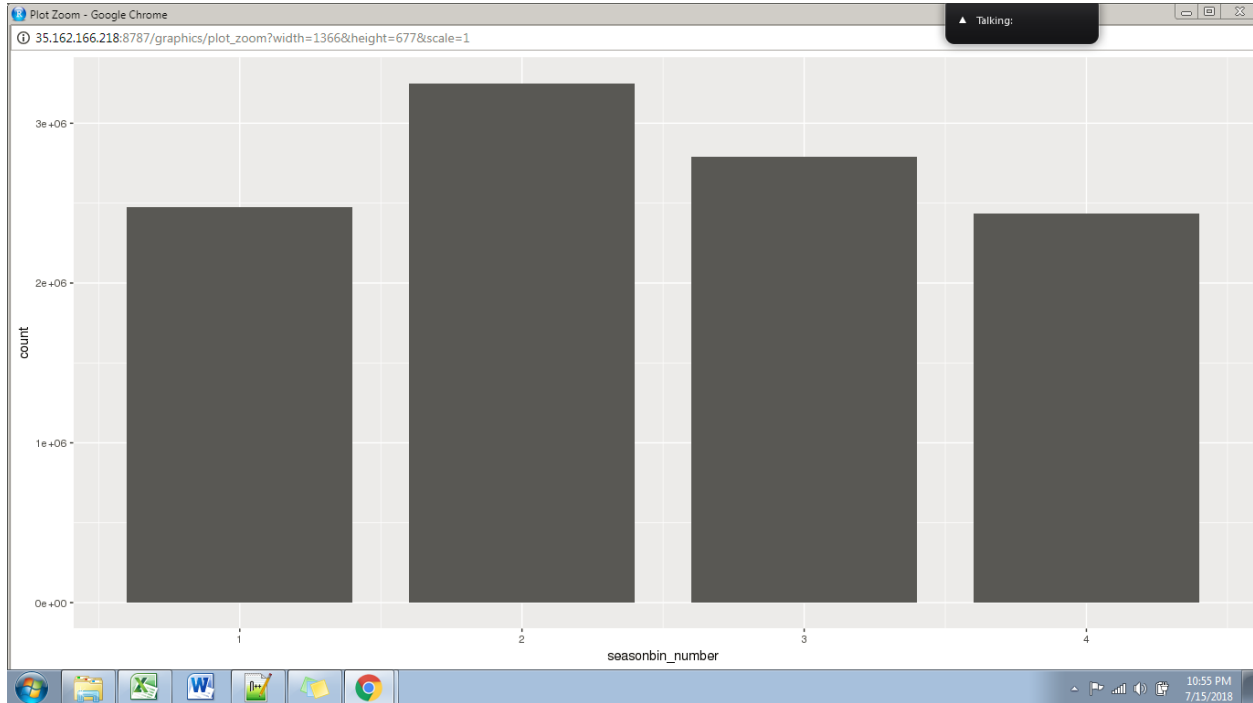
#3 3 14 234606

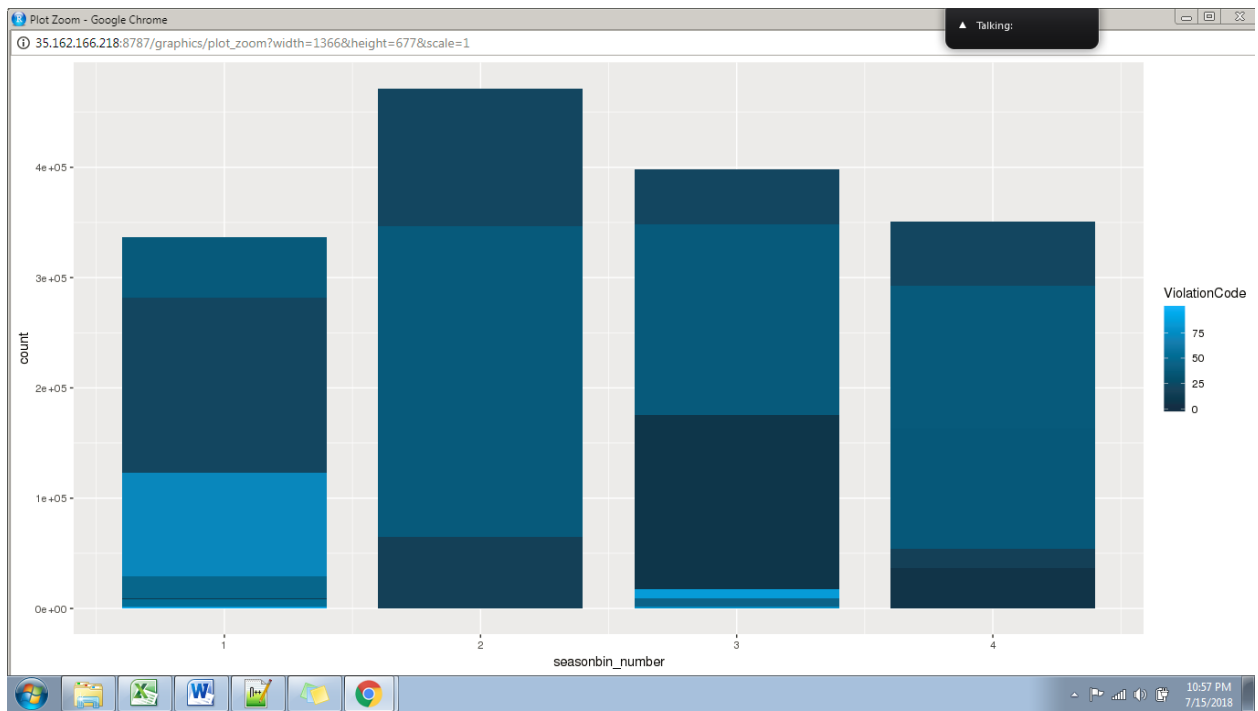
#seasonbin_number ViolationCode count

#1 4 21 350563

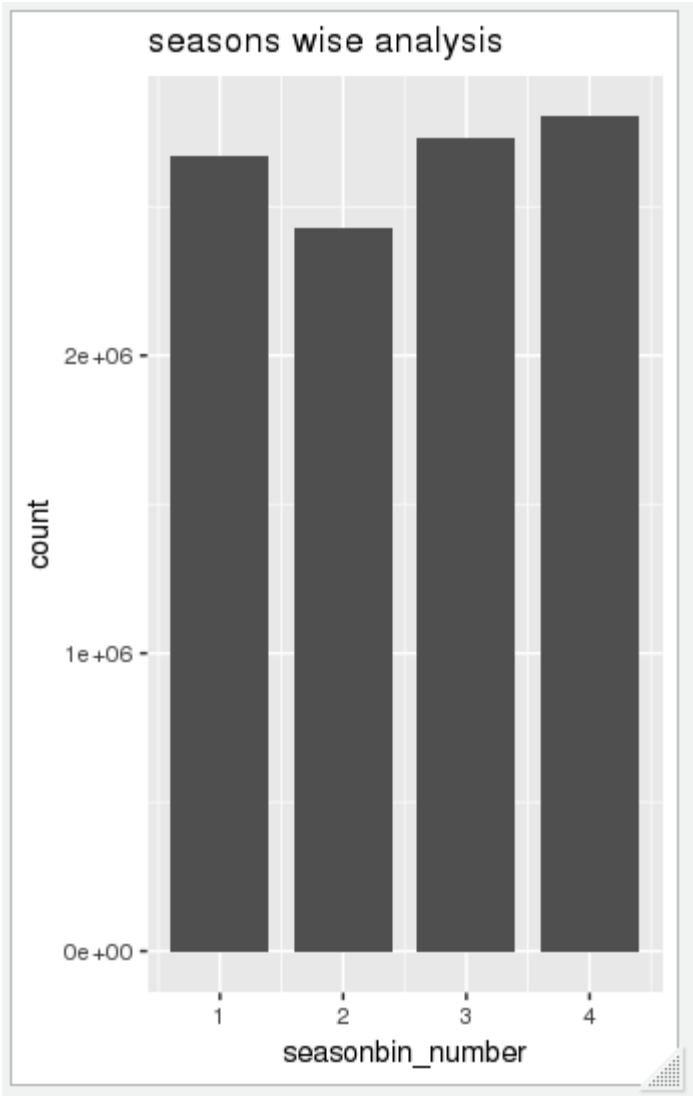
#2 4 38 292639

#3 4 14 207397

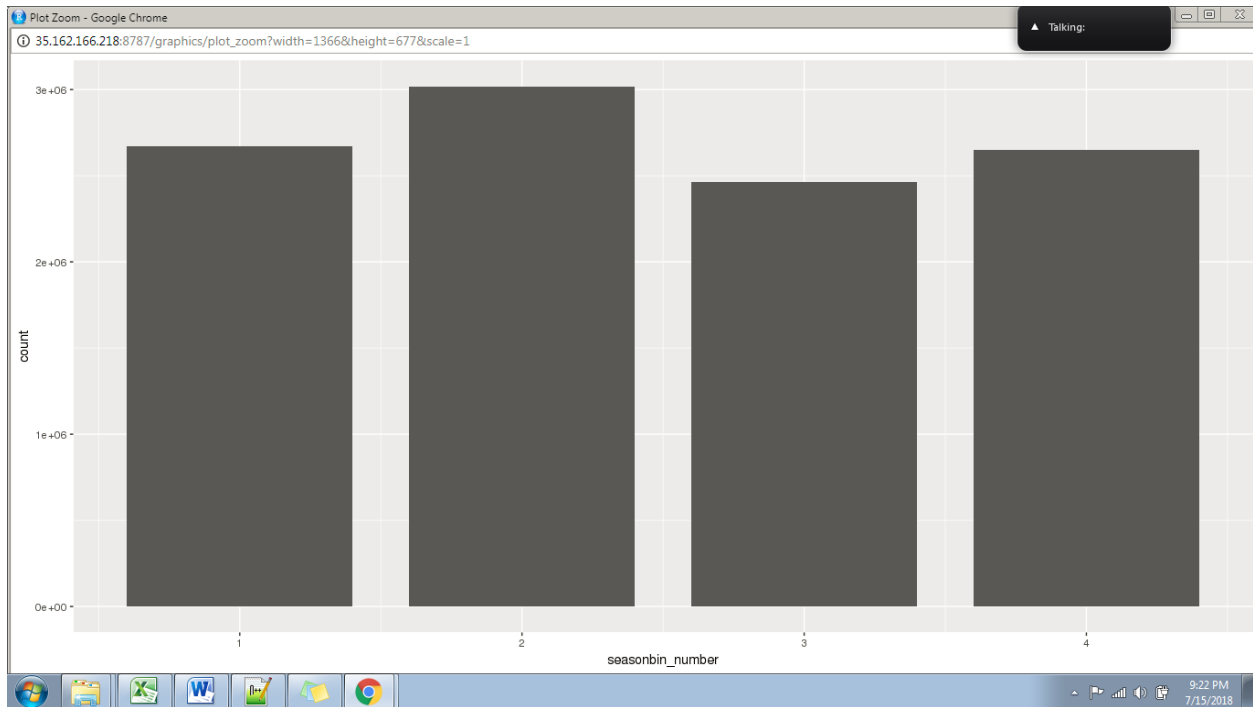




2016



2017



7 The fines collected from all the parking violation constitute a revenue source for the NYC police department. Let's take an example of estimating that for the 3 most commonly occurring codes.

- Find total occurrences of the 3 most common violation codes
- Then, search the internet for NYC parking violation code fines. You will find a website (on the nyc.gov URL) that lists these fines. They're divided into two categories, one for the highest-density locations of the city, the other for the rest of the city. For simplicity, take an average of the two.
- Using this information, find the total amount collected for all of the fines. State the code which has the highest total collection.
- What can you intuitively infer from these findings?

2015

#ViolationCode count fine

#1 21 1501614 75080700

#2 38 1324586 66229300

#3 14 924627 106332105

2016

	#ViolationCode	count	fine
--	----------------	-------	------

#1	21	1531587	76579350
----	----	---------	----------

#2	36	1253512	62675600
----	----	---------	----------

#3	38	1143696	131525040
----	----	---------	-----------

2017

	#ViolationCode	count	fine
#1	21	1528588	76429400
#2	36	1400614	70030700
#3	38	1062304	122164960