

Problem Set I

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1. **PS1:** Due Sat Oct 5 at 5:00PM Central. Worth 50 points.

We use (✱) to indicate a problem that we think might be time consuming.

Steps to submit (5 points on PS1)

1. "This submission is my work alone and complies with the 30538 integrity policy." Add your initials to indicate your agreement: ****
2. "I have uploaded the names of anyone I worked with on the problem set [here](#)" **** (1 point)
3. Late coins used this pset: **\1** Late coins left after submission: **\3**
4. Knit your `ps1.qmd` to HTML
5. Convert your HTML to PDF by printing to PDF from a browser.
6. Submit resulting `ps1.pdf` to Gradescope (4 points)
7. Tag your submission in Gradescope

```
# set up
import pandas as pd
import altair as alt

import warnings
warnings.filterwarnings('ignore')
```

Read in one percent sample (15 Points)

- 1.

```
##I looked at a stackflow post to find how to do it!
import time

start_time= time.time()

data_set_1 = pd.read_csv("/Users/samarnegahdar/Documents/school/Fall quarter 2024/Python

end_time= time.time()
time_spent= end_time - start_time
print("the time spent running this csv file is", time_spent)
## The Assert statement to see if the length of dataset is 287458 rows
```

```
assert len(data_set_1)== 287458
print("length of dataset is:",len(data_set_1))
```

the time spent running this csv file is 1.0854239463806152
length of dataset is: 287458

2.

```
import os
##I want to see how many Bytes are there in the csv file:
File_size_in_bytes = os.path.getsize("/Users/samarnegahdar/Documents/school/Fall quarter
File_size_in_GB = File_size_in_bytes / (1024 ** 3)
print("CSV file is:", File_size_in_GB, "GB")
##If we wanna predict how large the dataset is we have to multiply the file size by 100 (
Full_size_file= File_size_in_GB * 100
##since each Terabyte is 1024 GBs, I want to convert it to a more readable number
```

CSV file is: 0.07791011407971382 GB

3. I took a look at the CSV file and realized they are ordered based on date and time (issue-date column). now I wanna check if these column was supposed to be the default ordering column.

```
import pandas as pd

##so now I have to find that one column based on which the dataset is ordered.

##I made a mistake! I had forgotten to remove the first
##column that is the index column so:
valid_columns = data_set_1.columns[data_set_1.columns.str.contains('') == False]
def find_ordering_column(dataframe):
    for column in valid_columns:
        if dataframe[column].is_monotonic_increasing or dataframe[column].is_monotonic_decreasing:
            print(f"The dataset is ordered by the column: {column}")
            return column
    print("No column appears to order the dataset.")
    return None ## I could not get the
                ##function right so I looked this up and it
                ## turns out adding "return" means exit the
                ## function once you have found one column
                ## that meets the if clause
    print("No single column appears to order the dataset")
    return None
```

No column is the base for ordering it but the first column which is the index?? should we consider that as the desired column based on which the dataset is ordered? if so, it will be like this:

```

import pandas as pd

def find_ordering_column(dataframe):
    for column in dataframe.columns:
        if dataframe[column].is_monotonic_increasing or dataframe[column].is_monotonic_decreasing:
            print(f"The dataset is ordered by the column: {column}")
            return column
    print("No column appears to order the dataset.")
    return None

ordering_column = find_ordering_column(data_set_1)

##the subsetted dataset
first_500 = data_set_1.head(500)

##the function is going to be the same as above!
def is_column_ordered(subset, column):
    if subset[column].is_monotonic_increasing or subset[column].is_monotonic_decreasing:
        return f"The column '{column}' is ordered."
    else:
        return f"The column '{column}' is not ordered."

##now I am going to test it on the subsetted dataset
desired_col = find_ordering_column(data_set_1)

if desired_col:
    result = is_column_ordered(first_500, desired_col)
    print(result)

```

The dataset is ordered by the column: Unnamed: 0
 The dataset is ordered by the column: Unnamed: 0
 The column 'Unnamed: 0' is ordered.

Cleaning the data and benchmarking (15 Points)

1.

```

# Define start and end dates for 2017
start_2017 = "2017-01-01 00:00:00"
end_2017 = "2017-12-31 23:59:59"

##I had not written this line and the whole file
#crashed and I was panicking so I asked GPT for
#help by putting in the error and added
#this line to it!
data_set_1["issue_date"] = pd.to_datetime(data_set_1["issue_date"])

```

```
# I want a new dataset that filters all 2017!
data_2017_aggregated = data_set_1[(data_set_1["issue_date"] >= start_2017) &
(data_set_1["issue_date"] <= end_2017)]

# Print the number of tickets issued in 2017
print("The number of tickets issued in 2017 is", len(data_2017_aggregated))
```

The number of tickets issued in 2017 is 22364

2.

```
##I first see how many times each violation has occurred
violation_occurance= data_set_1["violation_description"].value_counts()
##now I see the top 20
top_20_viol_table= violation_occurance.head(20).reset_index()

##I am going to create a dataset(or a table)
##containing the top 20 violations and their
##frequencies
top_20_viol_table.columns = ['violation_type', 'frequency']
print(top_20_viol_table)
```

	violation_type	frequency
0	EXPIRED PLATES OR TEMPORARY REGISTRATION	44811
1	STREET CLEANING	28712
2	RESIDENTIAL PERMIT PARKING	23683
3	EXP. METER NON-CENTRAL BUSINESS DISTRICT	20600
4	PARKING/STANDING PROHIBITED ANYTIME	19753
5	EXPIRED METER OR OVERSTAY	18756
6	REAR AND FRONT PLATE REQUIRED	15829
7	NO CITY STICKER VEHICLE UNDER/EQUAL TO 16,000 ...	14246
8	RUSH HOUR PARKING	11965
9	NO CITY STICKER OR IMPROPER DISPLAY	10773
10	EXPIRED METER CENTRAL BUSINESS DISTRICT	9736
11	NO STANDING/PARKING TIME RESTRICTED	8640
12	WITHIN 15' OF FIRE HYDRANT	6104
13	PARK OR STAND IN BUS/TAXI/CARRIAGE STAND	6004
14	TRUCK,RV,BUS, OR TAXI RESIDENTIAL STREET	4789
15	STREET CLEANING OR SPECIAL EVENT	3370
16	DOUBLE PARKING OR STANDING	2904
17	EXPIRED PLATE OR TEMPORARY REGISTRATION	2720
18	STOP SIGN OR TRAFFIC SIGNAL	2191
19	PARK OR BLOCK ALLEY	2050

```
##drawing the plot I'm guessing we should do it through altair?
import altair as alt
```

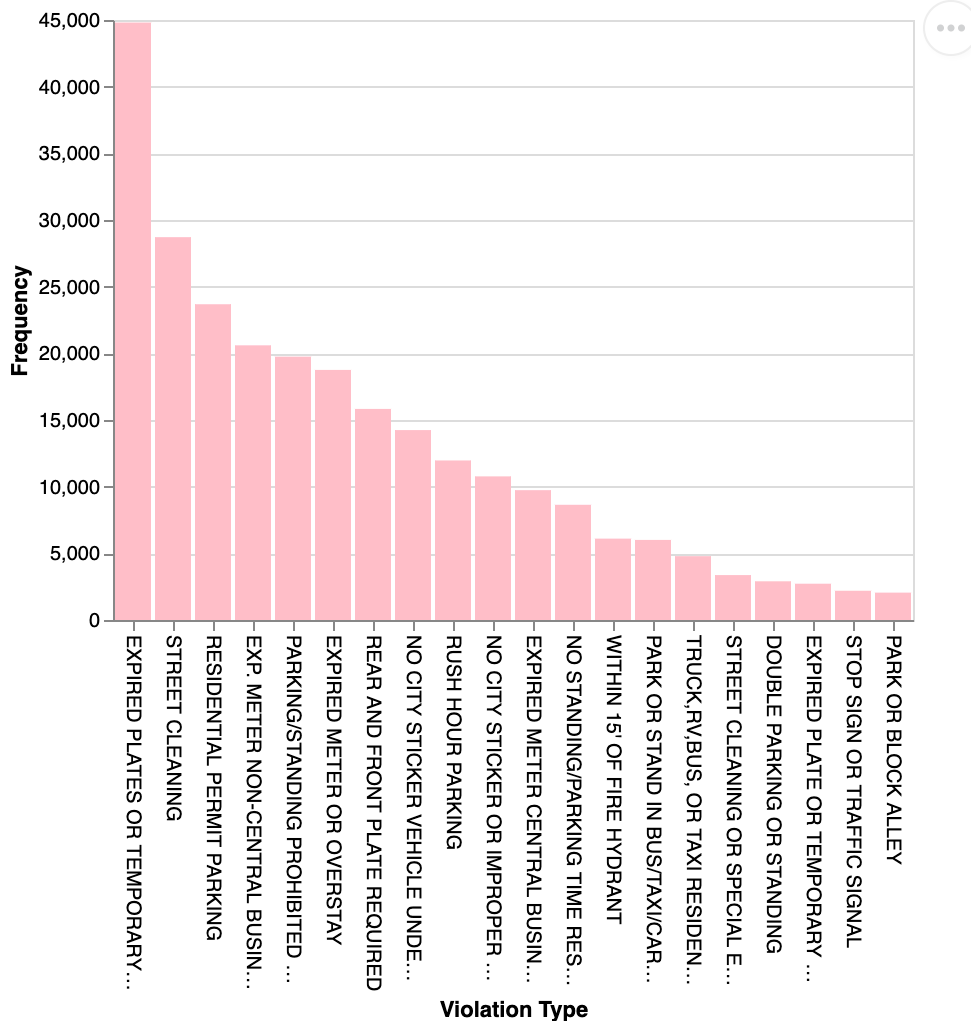
```
bar_chart = alt.Chart(top_20_viol_table).mark_bar(color= "pink").encode(
    x=alt.X('violation_type:N', sort='-y', title='Violation Type'),
```

```

y=alt.Y('frequency:Q', title='Frequency'),
tooltip=['violation_type', 'frequency']
).configure_axisX(
    labelAngle=90 ##rotating the labels on X axis
    ##so they're readable
)

# Display the chart
bar_chart

```



Visual Encoding (15 Points)

1.

```

##first of all I wanna see how many columns I have
columns= data_set_1.shape[1]
print(columns)
##I wanna see the name of columns:
col_names = data_set_1.columns
print(col_names)

```

```
column_data_types = data_set_1.dtypes
print(column_data_types)
```

24

```
Index(['Unnamed: 0', 'ticket_number', 'issue_date', 'violation_location',
      'license_plate_number', 'license_plate_state', 'license_plate_type',
      'zipcode', 'violation_code', 'violation_description', 'unit',
      'unit_description', 'vehicle_make', 'fine_level1_amount',
      'fine_level2_amount', 'current_amount_due', 'total_payments',
      'ticket_queue', 'ticket_queue_date', 'notice_level',
      'hearing_disposition', 'notice_number', 'officer', 'address'],
      dtype='object')
```

```
Unnamed: 0                int64
ticket_number             float64
issue_date                datetime64[ns]
violation_location        object
license_plate_number      object
license_plate_state       object
license_plate_type        object
zipcode                   object
violation_code            object
violation_description     object
unit                     float64
unit_description          object
vehicle_make              object
fine_level1_amount        int64
fine_level2_amount        int64
current_amount_due        float64
total_payments            float64
ticket_queue              object
ticket_queue_date         object
notice_level              object
hearing_disposition       object
notice_number             float64
officer                   object
address                   object
dtype: object
```

```
##Now I am going to create a new dataset:
##this table will have 23 rows since I am
##ignoring the first column since it is
##index column
Column_types= {
    "Variable Name": [
        'ticket_number', 'issue_date', 'violation_location',
        'license_plate_number', 'license_plate_state', 'license_plate_type',
        'zipcode', 'violation_code', 'violation_description', 'unit',
        'unit_description', 'vehicle_make', 'fine_level1_amount',
        'fine_level2_amount', 'current_amount_due', 'total_payments',
        'ticket_queue', 'ticket_queue_date', 'notice_level',
```

```

        'hearing_disposition', 'notice_number', 'officer', 'address'
    ],
    "Data Types": [
        'Ordinal', 'Temporal', 'Nominal', 'Nominal', 'Nominal', 'Nominal', 'Ordinal/Nominal'
    ]
}
column_types_table = pd.DataFrame(Column_types)

# Display the DataFrame
print(column_types_table)

```

	Variable Name	Data Types
0	ticket_number	Ordinal
1	issue_date	Temporal
2	violation_location	Nominal
3	license_plate_number	Nominal
4	license_plate_state	Nominal
5	license_plate_type	Nominal
6	zipcode	Ordinal/Nominal
7	violation_code	Nominal
8	violation_description	Nominal
9	unit	Quantitative
10	unit_description	Nominal
11	vehicle_make	Nominal
12	fine_level1_amount	Quantitative
13	fine_level2_amount	quantitative
14	current_amount_due	Quantitative
15	total_payments	Quantitative
16	ticket_queue	Nominal
17	ticket_queue_date	Temporal
18	notice_level	Nominal
19	hearing_disposition	Nominal
20	notice_number	Ordinal
21	officer	Nominal
22	address	Nominal

I personally think for the zip code we can have a kind of ranking especially in this dataset since we were talking about how lower income neighborhoods were affected so there can be a kind of order (higher income neighborhoods and lower income neighborhoods!) Or even like we can order neighborhoods from North to South... so many ways we can rank them based on Zip code!

- At first I did not get the question (I think it is bc the column called vehicle make is weird) but now that I know, I will group my data based on the type of the car and sum the total number of tickets paid for each of them:

```

## I first check what values exist in the 'ticket_queue' column
print(data_set_1['ticket_queue'].unique())

## So I created a dummy column for whether the

```

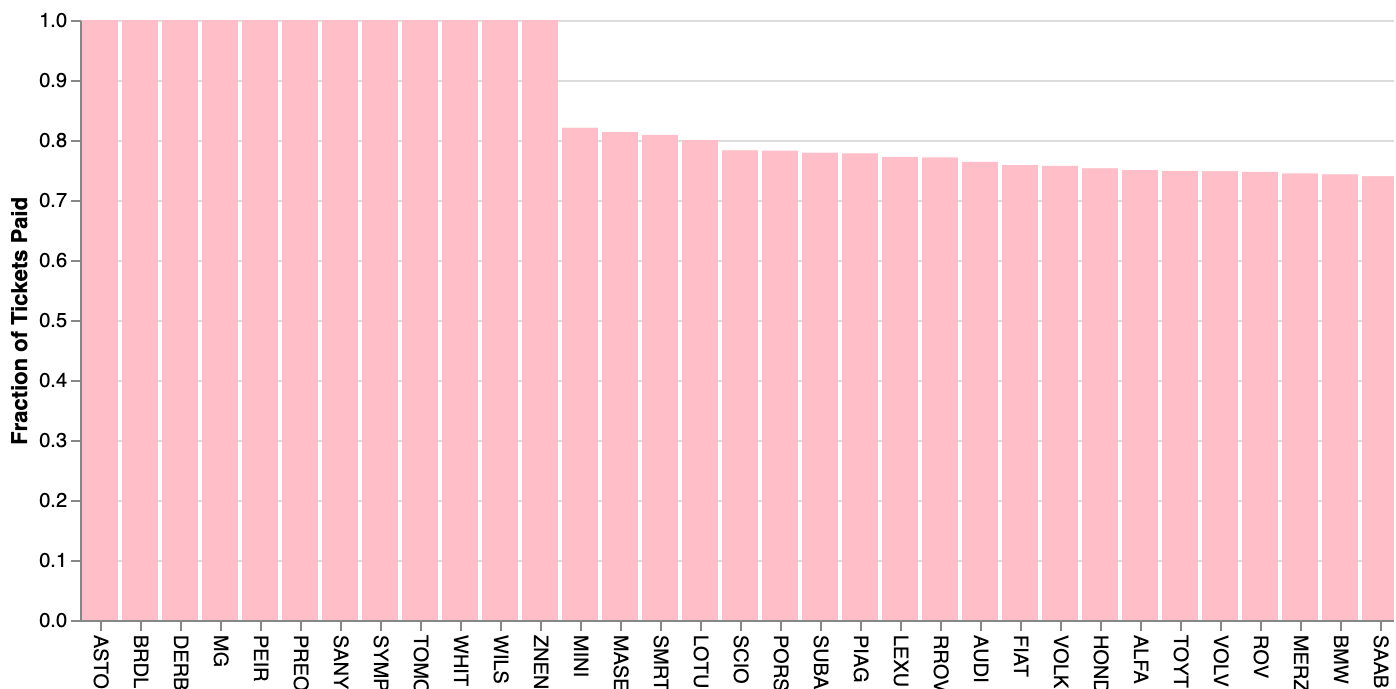
```
## ticket_queue is "Paid" or not, case insensitive
data_set_1['is_paid'] = data_set_1['ticket_queue'].str.lower() == 'paid'.lower()

grouped_data = data_set_1.groupby('vehicle_make').agg(
    T_tickets=('ticket_number', 'count'),
    tickets_P=('is_paid', 'sum')
).reset_index()

grouped_data['paid_fraction'] = grouped_data['tickets_P'] / grouped_data['T_tickets']
bar_chart = alt.Chart(grouped_data).mark_bar(color= "pink").encode(
    x=alt.X('vehicle_make:N', sort='-y', title='Vehicle Make'),
    y=alt.Y('paid_fraction:Q', title='Fraction of Tickets Paid'),
    tooltip=['vehicle_make', 'paid_fraction']
).properties(
    title='Fraction of Tickets Paid by Vehicle Make',
).configure_axisX(
    labelAngle=90 # Rotate x-axis labels for readability
)

# Display the chart
bar_chart
```

```
['Paid' 'Notice' 'Define' 'Dismissed' 'Bankruptcy' 'Court' 'Hearing Req']
```



So just like the article also talked about the burden it had on lower income citizens, certain social classes drive certain types of car for example the fraction of tickets paid by vehicle ASTO is q (I am guessing it is Aston Martin) which is a luxury high-end car so mostly super rich people own such brands and therefore they all have paid their tickets!

3.

```
##first I am just making sure "issue_date" column is
##time type
data_set_1["issue_date"] = pd.to_datetime(data_set_1["issue_date"])

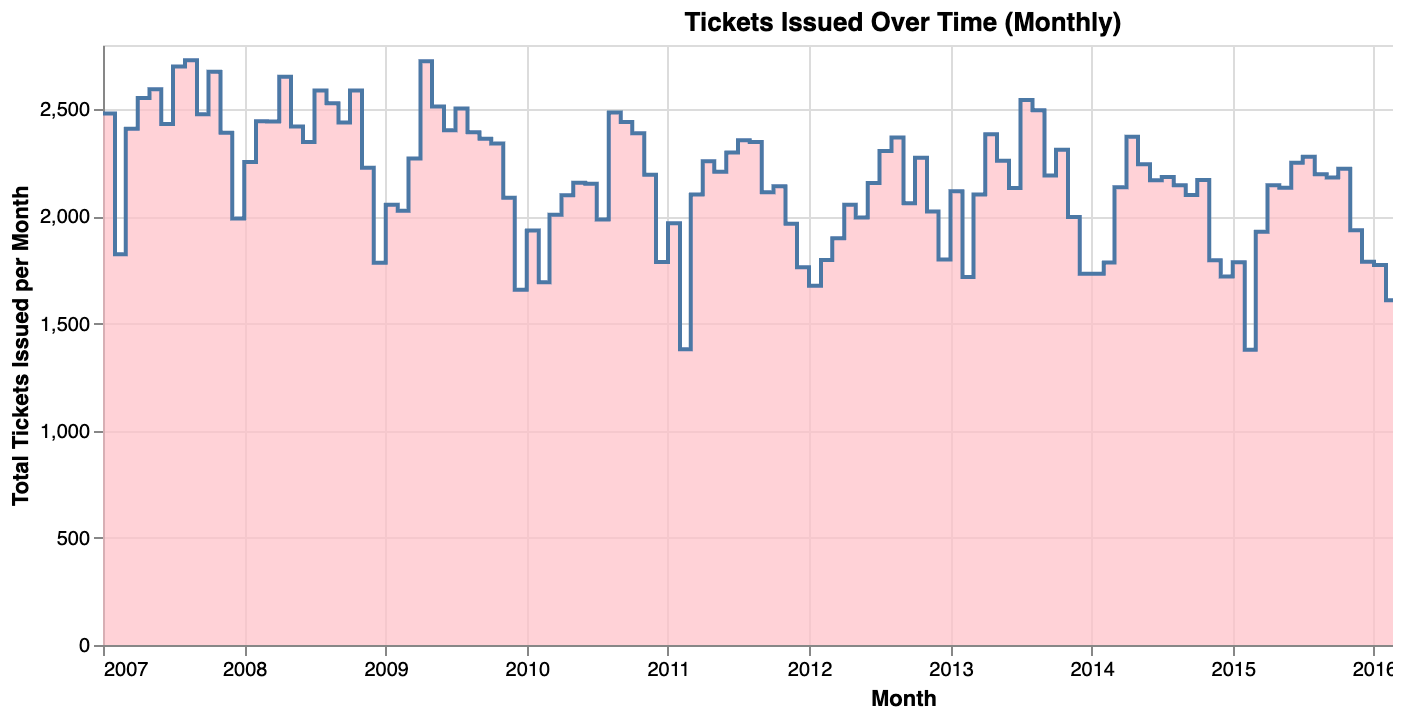
##now since the data is huge, I will group it by the
##total number of tickets issued every month!
##(I had to ask GPT cause I had no idea how to do it

Monthly_tickets = data_set_1.groupby(data_set_1['issue_date'].dt.to_period('M')).size().reset_index()
Monthly_tickets['issue_date'] = Monthly_tickets['issue_date'].dt.to_timestamp()

step_chart = alt.Chart(Monthly_tickets).mark_area(
    color="pink",
    interpolate='step-after',
    line=True
).encode(
    x=alt.X('issue_date:T', title='Month'),
    y=alt.Y('ticket_count:Q', title='Total Tickets Issued per Month'),
    tooltip=['issue_date', 'ticket_count']
).properties(
    title='Tickets Issued Over Time (Monthly)',
    width= 800,
)

step_chart

# Display the chart
step_chart
```



4.

```
##I am adding these two lines after searching
##for the error I had(Max rows)

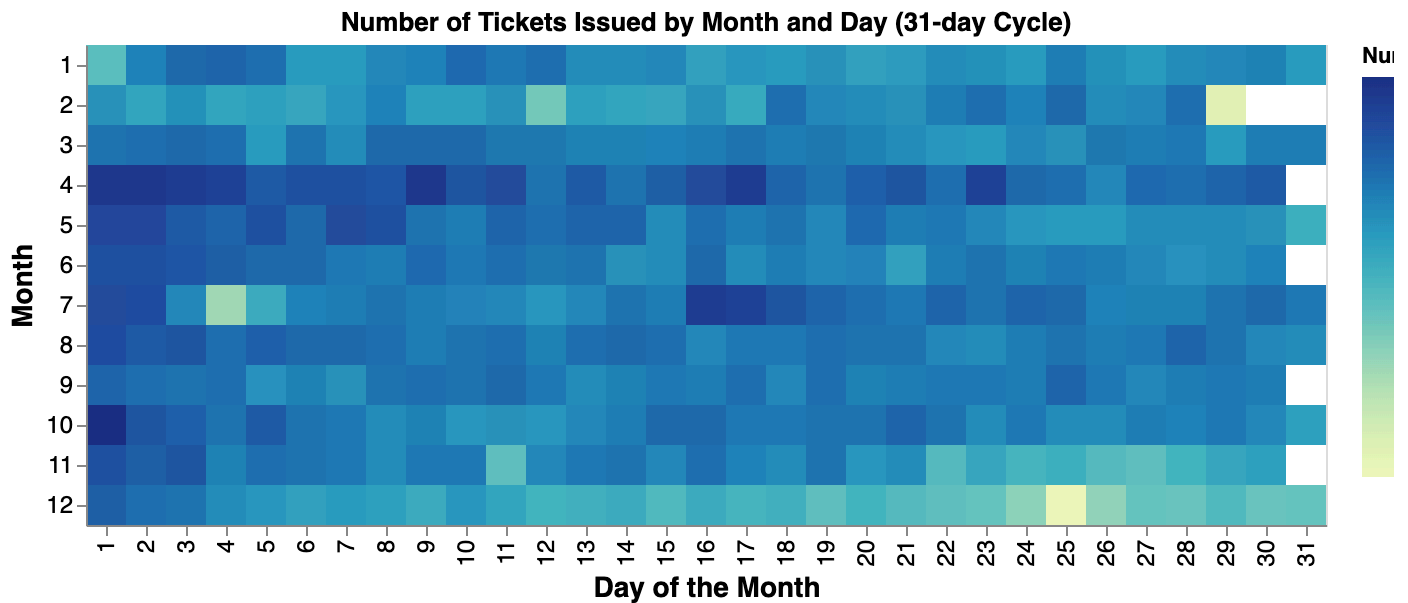
data_set_1['issue_date'] = pd.to_datetime(data_set_1['issue_date'])

data_set_1['day'] = data_set_1['issue_date'].dt.day
data_set_1['month'] = data_set_1['issue_date'].dt.month

tickets_by_day_month = data_set_1.groupby(['day', 'month']).size().reset_index(name='ticket_count')

heatmap = alt.Chart(tickets_by_day_month).mark_rect().encode(
    x=alt.X('day:0', scale=alt.Scale(domain=list(range(1, 32))), title='Day of the Month'),
    y=alt.Y('month:0', title='Month'),
    color=alt.Color('ticket_count:Q', title='Number of Tickets'),
    tooltip=['day', 'month', 'ticket_count']
).properties(
    title='Number of Tickets Issued by Month and Day (31-day Cycle)',
).configure_axis(
    labelFontSize=12,
    titleFontSize=14
)

heatmap
```



5.

```
data_set_1['issue_date'] = pd.to_datetime(data_set_1['issue_date'])

top_5_violations = data_set_1['violation_description'].value_counts().head(5).index

subset_data = data_set_1[data_set_1['violation_description'].isin(top_5_violations)]

subset_data['issue_month'] = subset_data['issue_date'].dt.to_period('M')

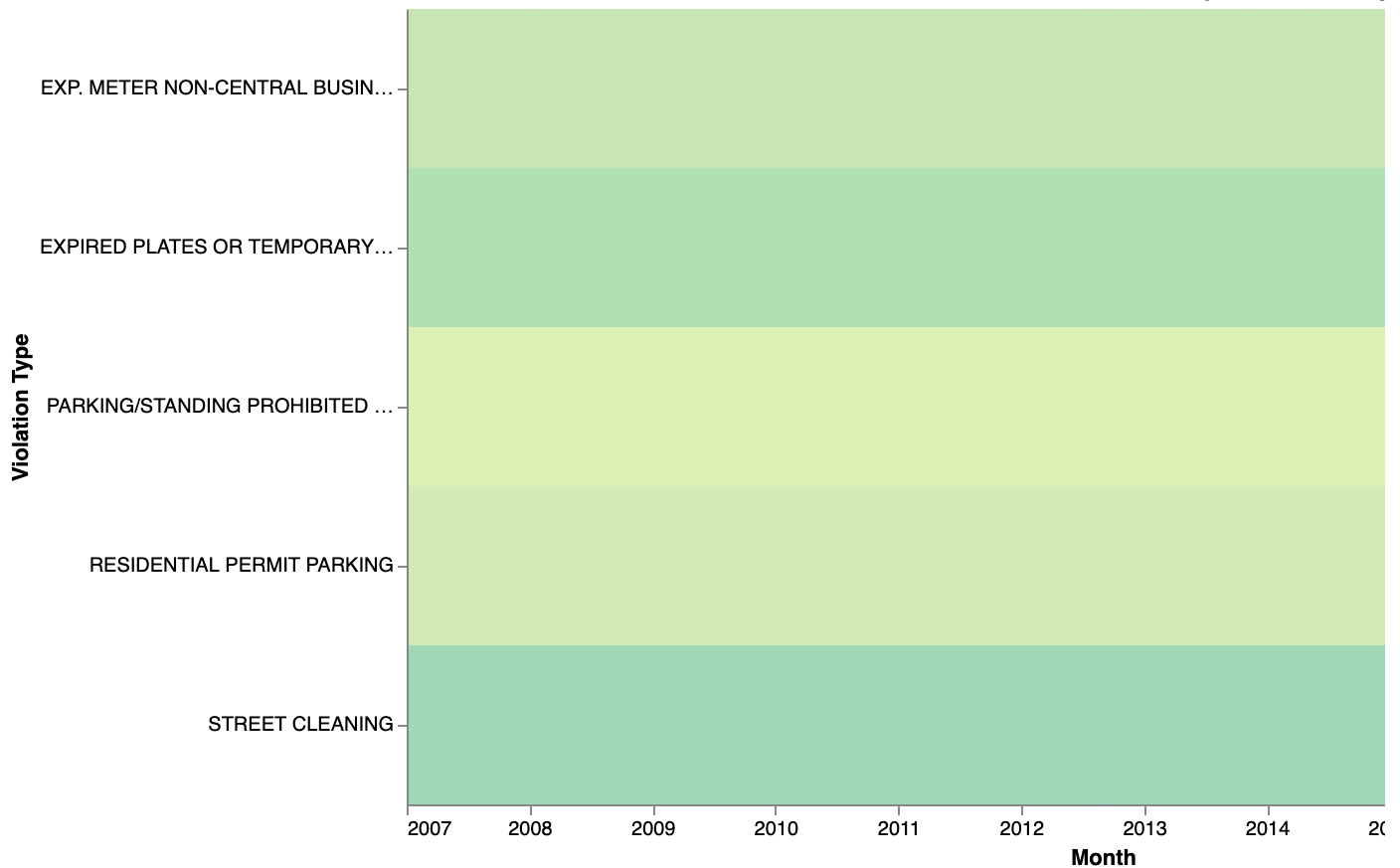
tickets_over_time = subset_data.groupby(['issue_month', 'violation_description']).size()

tickets_over_time['issue_month'] = tickets_over_time['issue_month'].dt.to_timestamp()

lasagna_plot = alt.Chart(tickets_over_time).mark_rect().encode(
    x=alt.X('issue_month:T', title='Month'),
    y=alt.Y('violation_description:N', title='Violation Type'),
    color=alt.Color('ticket_count:Q', title='Tickets Issued'),
).properties(
    title='Tickets Issued Over Time for the Top 5 Violation Types',
    width=700,
    height=400
)

lasagna_plot
```

Tickets Issued Over Time for the Top 5 Violation Ty



6. Bar plot: the first thing is that bar plots are easiest to understand and read, they give some general information but the problem is that it is limited, like we only mentioned 20 violations but imagine having more (it would have been either a huge plot or nonlegible). Heatmap: I personally really like heatmaps for actual climate related but in our case, it shows the intensity of tickets issued per day each month but first you cannot get all 11 years, second it does not give us specification of violations and other details. It can also be very overwhelming if we have such a large dataset! (although knowing that certain days have higher issuance rate to others it wouldn't really give us the cause like why did that happen? (violation type)) Lasagna map: I would say this is a combination of bar plots and heatmaps where you can see the number of tickets per month and the violation tied to those issuances, but it is not that delicate (it only shows very large bins of different number of tickets issued).
7. As I mentioned in the previous question since Lasagna plot is somehow the combination of barplot and heatmap I would say that is our best bet since it is giving info both on violation type and also the date.