Bike_Share_Analysis

July 30, 2019

1 2016 US Bike Share Activity Snapshot

1.1 Table of Contents

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- ## Introduction

Tip: Quoted sections like this will provide helpful instructions on how to navigate and use a Jupyter notebook.

Over the past decade, bicycle-sharing systems have been growing in number and popularity in cities across the world. Bicycle-sharing systems allow users to rent bicycles for short trips, typically 30 minutes or less. Thanks to the rise in information technologies, it is easy for a user of the system to access a dock within the system to unlock or return bicycles. These technologies also provide a wealth of data that can be used to explore how these bike-sharing systems are used.

In this project, you will perform an exploratory analysis on data provided by Motivate, a bike-share system provider for many major cities in the United States. You will compare the system usage between three large cities: New York City, Chicago, and Washington, DC. You will also see if there are any differences within each system for those users that are registered, regular users and those users that are short-term, casual users.

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2.1 Posing Questions

Before looking at the bike sharing data, you should start by asking questions you might want to understand about the bike share data. Consider, for example, if you were working for Motivate. What kinds of information would you want to know about in order to make smarter business

decisions? If you were a user of the bike-share service, what factors might influence how you would want to use the service?

Question 1: Write at least two questions related to bike sharing that you think could be answered by data.

Answer:

What is the price per 30 minutes of bicycle rent?

Will the user pay in advance or after he finishes his ride?

How far can the user drive his rented bike from the origin of where he rented it?

Can the user rent the bike for a whole day or a few hours?

What are the sizes of bikes the company offers and is it suitable for all ages?

Can the user take someone else at the back seat of the bike? or it's not allowed?

What maintenance services do we provide to the user in case his rented bike has a problem during his trip?

Can the user pay in both cash and credit?

Does the company offer discounts for regular users using the service?

Tip: If you double click on this cell, you will see the text change so that all of the formatting is removed. This allows you to edit this block of text. This block of text is written using Markdown, which is a way to format text using headers, links, italics, and many other options using a plain-text syntax. You will also use Markdown later in the Nanodegree program. Use **Shift** + **Enter** or **Shift** + **Return** to run the cell and show its rendered form.

Data Collection and Wrangling

Now it's time to collect and explore our data. In this project, we will focus on the record of individual trips taken in 2016 from our selected cities: New York City, Chicago, and Washington, DC. Each of these cities has a page where we can freely download the trip data.:

- New York City (Citi Bike): Link
- Chicago (Divvy): Link
- Washington, DC (Capital Bikeshare): Link

If you visit these pages, you will notice that each city has a different way of delivering its data. Chicago updates with new data twice a year, Washington DC is quarterly, and New York City is monthly. However, you do not need to download the data yourself. The data has already been collected for you in the /data/ folder of the project files. While the original data for 2016 is spread among multiple files for each city, the files in the /data/ folder collect all of the trip data for the year into one file per city. Some data wrangling of inconsistencies in timestamp format within each city has already been performed for you. In addition, a random 2% sample of the original data is taken to make the exploration more manageable.

Question 2: However, there is still a lot of data for us to investigate, so it's a good idea to start off by looking at one entry from each of the cities we're going to analyze. Run the first code cell below to load some packages and functions that you'll be using in your analysis. Then, complete the second code cell to print out the first trip recorded from each of the cities (the second line of each data file).

Tip: You can run a code cell like you formatted Markdown cells above by clicking on the cell and using the keyboard shortcut **Shift** + **Enter** or **Shift** + **Return**. Alternatively, a code cell can be executed using the **Play** button in the toolbar after selecting it. While

the cell is running, you will see an asterisk in the message to the left of the cell, i.e. In [*]:. The asterisk will change into a number to show that execution has completed, e.g. In [1]. If there is output, it will show up as Out [1]:, with an appropriate number to match the "In" number.

```
In [83]: ## import all necessary packages and functions.
         import csv # read and write csv files
         from datetime import datetime # operations to parse dates
         from pprint import pprint # use to print data structures like dictionaries in
                                   # a nicer way than the base print function.
In [84]: def print_first_point(filename):
             This function prints and returns the first data point (second row) from
             a csv file that includes a header row.
             # print city name for reference
             city = filename.split('-')[0].split('/')[-1]
             print('\nCity: {}'.format(city))
             with open(filename, 'r') as f_in:
                 ## TODO: Use the csv library to set up a DictReader object. ##
                 ## see https://docs.python.org/3/library/csv.html
                 trip_reader = csv.DictReader(f_in)
                 ## TODO: Use a function on the DictReader object to read the
                 ## first trip from the data file and store it in a variable.
                 ## see https://docs.python.orq/3/library/csv.html#reader-objects ##
                 first_trip = next(trip_reader)
                 ## TODO: Use the pprint library to print the first trip. ##
                 ## see https://docs.python.org/3/library/pprint.html
                 pprint(first_trip)
             # output city name and first trip for later testing
             return (city, first_trip)
         # list of files for each city
         data_files = ['./data/NYC-CitiBike-2016.csv',
                       './data/Chicago-Divvy-2016.csv',
                       './data/Washington-CapitalBikeshare-2016.csv',]
         # print the first trip from each file, store in dictionary
         example_trips = {}
         for data_file in data_files:
             city, first_trip = print_first_point(data_file)
             example_trips[city] = first_trip
```

City: NYC

```
OrderedDict([('tripduration', '839'),
             ('starttime', '1/1/2016 00:09:55'),
             ('stoptime', '1/1/2016 00:23:54'),
             ('start station id', '532'),
             ('start station name', 'S 5 Pl & S 4 St'),
             ('start station latitude', '40.710451'),
             ('start station longitude', '-73.960876'),
             ('end station id', '401'),
             ('end station name', 'Allen St & Rivington St'),
             ('end station latitude', '40.72019576'),
             ('end station longitude', '-73.98997825'),
             ('bikeid', '17109'),
             ('usertype', 'Customer'),
             ('birth year', ''),
             ('gender', '0')])
City: Chicago
OrderedDict([('trip_id', '9080545'),
             ('starttime', '3/31/2016 23:30'),
             ('stoptime', '3/31/2016 23:46'),
             ('bikeid', '2295'),
             ('tripduration', '926'),
             ('from_station_id', '156'),
             ('from_station_name', 'Clark St & Wellington Ave'),
             ('to_station_id', '166'),
             ('to_station_name', 'Ashland Ave & Wrightwood Ave'),
             ('usertype', 'Subscriber'),
             ('gender', 'Male'),
             ('birthyear', '1990')])
City: Washington
OrderedDict([('Duration (ms)', '427387'),
             ('Start date', '3/31/2016 22:57'),
             ('End date', '3/31/2016 23:04'),
             ('Start station number', '31602'),
             ('Start station', 'Park Rd & Holmead Pl NW'),
             ('End station number', '31207'),
             ('End station', 'Georgia Ave and Fairmont St NW'),
             ('Bike number', 'W20842'),
             ('Member Type', 'Registered')])
```

If everything has been filled out correctly, you should see below the printout of each city name (which has been parsed from the data file name) that the first trip has been parsed in the form of a dictionary. When you set up a DictReader object, the first row of the data file is normally interpreted as column names. Every other row in the data file will use those column names as keys, as a dictionary is generated for each row.

This will be useful since we can refer to quantities by an easily-understandable label instead

of just a numeric index. For example, if we have a trip stored in the variable row, then we would rather get the trip duration from row['duration'] instead of row[0].

Condensing the Trip Data

It should also be observable from the above printout that each city provides different information. Even where the information is the same, the column names and formats are sometimes different. To make things as simple as possible when we get to the actual exploration, we should trim and clean the data. Cleaning the data makes sure that the data formats across the cities are consistent, while trimming focuses only on the parts of the data we are most interested in to make the exploration easier to work with.

You will generate new data files with five values of interest for each trip: trip duration, starting month, starting hour, day of the week, and user type. Each of these may require additional wrangling depending on the city:

- **Duration**: This has been given to us in seconds (New York, Chicago) or milliseconds (Washington). A more natural unit of analysis will be if all the trip durations are given in terms of minutes.
- Month, Hour, Day of Week: Ridership volume is likely to change based on the season, time
 of day, and whether it is a weekday or weekend. Use the start time of the trip to obtain these
 values. The New York City data includes the seconds in their timestamps, while Washington
 and Chicago do not. The datetime package will be very useful here to make the needed
 conversions.
- User Type: It is possible that users who are subscribed to a bike-share system will have different patterns of use compared to users who only have temporary passes. Washington divides its users into two types: 'Registered' for users with annual, monthly, and other longer-term subscriptions, and 'Casual', for users with 24-hour, 3-day, and other short-term passes. The New York and Chicago data uses 'Subscriber' and 'Customer' for these groups, respectively. For consistency, you will convert the Washington labels to match the other two.

Question 3a: Complete the helper functions in the code cells below to address each of the cleaning tasks described above.

```
duration = int(datum['Duration (ms)'])/60000
             else:
                 #converting sec to minutes
                 duration= int(datum['tripduration'])/60
             return duration
         # Some tests to check that your code works. There should be no output if all of
         # the assertions pass. The `example_trips` dictionary was obtained from when
         # you printed the first trip from each of the original data files.
         tests = {'NYC': 13.9833,
                  'Chicago': 15.4333,
                  'Washington': 7.1231}
         for city in tests:
             assert abs(duration_in_mins(example_trips[city], city) - tests[city]) < .001
In [86]: def time_of_trip(datum, city):
             Takes as input a dictionary containing info about a single trip (datum) and
             its origin city (city) and returns the month, hour, and day of the week in
             which the trip was made.
             Remember that NYC includes seconds, while Washington and Chicago do not.
             HINT: You should use the datetime module to parse the original date
             strings into a format that is useful for extracting the desired information.
             see https://docs.python.org/3/library/datetime.html#strftime-and-strptime-behavior
             # YOUR CODE HERE
             if city == 'NYC':
                 #Matching time format with the data
                 time= datetime.strptime(datum['starttime'],'%m/%d/%Y %H:%M:%S')
                 month= int(time.strftime("%-m")) #storing month value
                 day_of_week= str(datetime.strptime(datum['starttime'],'%m/%d/%Y %H:%M:%S').strf
                 hour= int(time.strftime("%-H")) #storing hour value
             elif city=='Chicago':
                 time= datetime.strptime(datum['starttime'],'%m/%d/%Y %H:%M')
                 month= int(time.strftime("%-m"))
                 day_of_week= str(datetime.strptime(datum['starttime'],'%m/%d/%Y %H:%M').strftim
                 hour= int(time.strftime("%-H"))
             elif city== 'Washington':
                 time= datetime.strptime(datum['Start date'],'%m/%d/%Y %H:%M')
                 month= int(time.strftime("%-m"))
                 day_of_week= str(datetime.strptime(datum['Start date'],'%m/%d/%Y %H:%M').strfti
                 hour= int(time.strftime("%-H"))
```

```
return (month, hour, day_of_week)
         # Some tests to check that your code works. There should be no output if all of
         # the assertions pass. The `example_trips` dictionary was obtained from when
         # you printed the first trip from each of the original data files.
         tests = {'NYC': (1, 0, 'Friday'),
                  'Chicago': (3, 23, 'Thursday'),
                  'Washington': (3, 22, 'Thursday')}
         for city in tests:
             assert time_of_trip(example_trips[city], city) == tests[city]
In [87]: def type_of_user(datum, city):
             Takes as input a dictionary containing info about a single trip (datum) and
             its origin city (city) and returns the type of system user that made the
             trip.
             Remember that Washington has different category names compared to Chicago
             11 11 11
             # YOUR CODE HERE
             if city == 'Washington':
                 if datum['Member Type'] == 'Registered':
                     user_type= 'Subscriber'
                 else:
                     user_type= 'Customer'
                 user_type= datum['usertype']
             return user_type
         # Some tests to check that your code works. There should be no output if all of
         # the assertions pass. The `example_trips` dictionary was obtained from when
         # you printed the first trip from each of the original data files.
         tests = {'NYC': 'Customer',
                  'Chicago': 'Subscriber',
                  'Washington': 'Subscriber'}
         for city in tests:
             assert type_of_user(example_trips[city], city) == tests[city]
```

Question 3b: Now, use the helper functions you wrote above to create a condensed data file for each city consisting only of the data fields indicated above. In the <code>/examples/</code> folder, you will see an example datafile from the Bay Area Bike Share before and after conversion. Make sure that your output is formatted to be consistent with the example file.

```
In [88]: def condense_data(in_file, out_file, city):
```

```
and writes the condensed data to a specified output file. The city
             argument determines how the input file will be parsed.
             HINT: See the cell below to see how the arguments are structured!
             with open(out_file, 'w') as f_out, open(in_file, 'r') as f_in:
                 # set up csv DictWriter object - writer requires column names for the
                 # first row as the "fieldnames" argument
                 out_colnames = ['duration', 'month', 'hour', 'day_of_week', 'user_type']
                 trip_writer = csv.DictWriter(f_out, fieldnames = out_colnames)
                 trip_writer.writeheader()
                 ## TODO: set up csv DictReader object ##
                 trip_reader = csv.DictReader(f_in)
                 # collect data from and process each row
                 for row in trip_reader:
                     # set up a dictionary to hold the values for the cleaned and trimmed
                     # data point
                     new_point = {}
                     ## TODO: use the helper functions to get the cleaned data from
                     ## the original data dictionaries.
                                                                                      ##
                     ## Note that the keys for the new_point dictionary should match ##
                     ## the column names set in the DictWriter object above.
                                                                                      ##
                     duration= duration_in_mins(row, city)
                     month, hour, day_of_week= time_of_trip(row, city)
                     user_type= type_of_user(row, city)
                     new_point.update({'duration':duration, 'month':month, 'hour':hour,
                                       'day_of_week':day_of_week, 'user_type':user_type})
                     ## TODO: write the processed information to the output file.
                     ## see https://docs.python.org/3/library/csv.html#writer-objects ##
                     trip_writer.writerow(new_point)
In [89]: # Run this cell to check your work
         city_info = {'Washington': {'in_file': './data/Washington-CapitalBikeshare-2016.csv',
                                     'out_file': './data/Washington-2016-Summary.csv'},
                      'Chicago': {'in_file': './data/Chicago-Divvy-2016.csv',
                                  'out_file': './data/Chicago-2016-Summary.csv'},
                      'NYC': {'in_file': './data/NYC-CitiBike-2016.csv',
                              'out_file': './data/NYC-2016-Summary.csv'}}
         for city, filenames in city_info.items():
```

This function takes full data from the specified input file

```
condense_data(filenames['in_file'], filenames['out_file'], city)
             print_first_point(filenames['out_file'])
City: Washington
OrderedDict([('duration', '7.123116666666666'),
             ('month', '3'),
             ('hour', '22'),
             ('day_of_week', 'Thursday'),
             ('user_type', 'Subscriber')])
City: Chicago
OrderedDict([('duration', '15.4333333333333334'),
             ('month', '3'),
             ('hour', '23'),
             ('day_of_week', 'Thursday'),
             ('user_type', 'Subscriber')])
City: NYC
OrderedDict([('duration', '13.98333333333333'),
             ('month', '1'),
             ('hour', '0'),
             ('day_of_week', 'Friday'),
             ('user_type', 'Customer')])
```

Tip: If you save a jupyter Notebook, the output from running code blocks will also be saved. However, the state of your workspace will be reset once a new session is started. Make sure that you run all of the necessary code blocks from your previous session to reestablish variables and functions before picking up where you last left off.

Exploratory Data Analysis

Now that you have the data collected and wrangled, you're ready to start exploring the data. In this section you will write some code to compute descriptive statistics from the data. You will also be introduced to the matplotlib library to create some basic histograms of the data.

Statistics

First, let's compute some basic counts. The first cell below contains a function that uses the csv module to iterate through a provided data file, returning the number of trips made by subscribers and customers. The second cell runs this function on the example Bay Area data in the /examples/folder. Modify the cells to answer the question below.

Question 4a: Which city has the highest number of trips? Which city has the highest proportion of trips made by subscribers? Which city has the highest proportion of trips made by short-term customers?

Answer:

The highest number of trips in NYC

The highest proportion of trips made by sbuscribers in BayArea

The highest proportion of trips made by customers in Chicago

```
In [90]: def number_of_trips(filename):
             This function reads in a file with trip data and reports the proportion of
             trips made by subscribers, customers, the total trips number
             also, returning number of subscriber_trips and customer_trips
             with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f_in)
                 # initialize count variables
                 subscriber_trips = 0
                 customer_trips = 0
                 # tally up ride types
                 for row in reader:
                     if row['user_type'] == 'Subscriber':
                         subscriber_trips+= 1
                     else:
                         customer_trips+= 1
                 # compute total number of rides
                 total_trips = subscriber_trips + customer_trips
                 # return tallies as a tuple
                 subscriber_prop = subscriber_trips/total_trips*100
                 customer_prop = customer_trips/total_trips*100
                 return(subscriber_prop, customer_prop, total_trips)
In [91]: ## Modify this and the previous cell to answer Question 4a. Remember to run ##
         ## the function on the cleaned data files you created from Question 3.
                                                                                      ##
         city_info = {'Washington': {'out_file': './data/Washington-2016-Summary.csv'},
                      'Chicago': {'out_file': './data/Chicago-2016-Summary.csv'},
                      'NYC': {'out_file': './data/NYC-2016-Summary.csv'},
                      'BayArea': {'out_file':'./examples/BayArea-Y3-Summary.csv'}}
         for city, filenames in city_info.items():
             print(city+':')
             print(number_of_trips(filenames['out_file']))
Washington:
(78.0282242257938, 21.971775774206193, 66326)
Chicago:
(76.22520136973007, 23.774798630269924, 72131)
NYC:
```

```
(88.83590199351151, 11.164098006488485, 276798)
BayArea:
(89.95078583902206, 10.049214160977932, 6299)
```

Tip: In order to add additional cells to a notebook, you can use the "Insert Cell Above" and "Insert Cell Below" options from the menu bar above. There is also an icon in the toolbar for adding new cells, with additional icons for moving the cells up and down the document. By default, new cells are of the code type; you can also specify the cell type (e.g. Code or Markdown) of selected cells from the Cell menu or the dropdown in the toolbar.

Now, you will write your own code to continue investigating properties of the data.

Question 4b: Bike-share systems are designed for riders to take short trips. Most of the time, users are allowed to take trips of 30 minutes or less with no additional charges, with overage charges made for trips of longer than that duration. What is the average trip length for each city? What proportion of rides made in each city are longer than 30 minutes?

Answer:

Average duration of trips in Washington: 18.93 minutes Proportion of long duration trips in Washington: 10.83% Average duration of trips in Chicago: 16.56 minutes Proportion of long duration trips in Chicago: 8.33% Average duration of trips in NYC: 15.81 minutes Proportion of long duration trips in NYC: 7.30% Average duration of trips in BayArea: 14.03 minutes Proportion of long duration trips in BayArea: 3.52%

```
##
In [92]: ## Use this and additional cells to answer Question 4b.
                                                                                   ##
         ## HINT: The csv module reads in all of the data as strings, including
         ## numeric values. You will need a function to convert the strings
                                                                                   ##
         ## into an appropriate numeric type before you aggregate data.
                                                                                   ##
         ## TIP: For the Bay Area example, the average trip length is 14 minutes ##
         ## and 3.5% of trips are longer than 30 minutes.
                                                                                   ##
         def average_trip_duration(filename):
             This function reads in a file with trip data and reports the average duration of
             trips made by users and the proportion of rides longer than 30 minutes.
             11 11 11
             with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f_in)
                 # initialize count variables
                 total_duration=0.0
                 total_trips = 0
                 long_trips = 0
```

```
#counting trips, long_trips and total_duration
                 for row in reader:
                     if float(row['duration']) > 30:
                         long_trips +=1
                     total_trips +=1
                     total_duration +=float(row['duration'])
                 # compute average length and proportion of long duration trips
                 avg_length = total_duration/total_trips
                 prop_long= long_trips/total_trips*100
                 return(avg_length, prop_long)
         city_info = {'Washington': {'out_file': './data/Washington-2016-Summary.csv'},
                      'Chicago': {'out_file': './data/Chicago-2016-Summary.csv'},
                      'NYC': {'out_file': './data/NYC-2016-Summary.csv'},
                      'BayArea': {'out_file':'./examples/BayArea-Y3-Summary.csv'}}
         for city, filenames in city_info.items():
             print(city+':')
             print(average_trip_duration(filenames['out_file']))
Washington:
(18.93287355913721, 10.83888671109369)
Chicago:
(16.563629368787335, 8.332062497400562)
(15.81259299802294, 7.3024371563378345)
BayArea:
(14.038656929671422, 3.5243689474519764)
```

Question 4c: Dig deeper into the question of trip duration based on ridership. Choose one city. Within that city, which type of user takes longer rides on average: Subscribers or Customers? **Answer**: for **NYC** the average trip duration for Customers is 32.98 minutes and 13.68 minutes for Subscribers and it is clear that customers take longer trips than subscribers

```
In [93]: ## Use this and additional cells to answer Question 4c. If you have ##
    ## not done so yet, consider revising some of your previous code to ##
    ## make use of functions for reusability. ##
    ##
    ## TIP: For the Bay Area example data, you should find the average ##
    ## Subscriber trip duration to be 9.5 minutes and the average Customer ##
    ## trip duration to be 54.6 minutes. Do the other cities have this ##
    ## level of difference? ##
    def average_user_duration(filename):
```

```
This function reads in a file with trip data and reports the total
             average trip duration for Customers and subscribers.
             with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f_in)
                 # initialize count variables
                 duration_customer=0.0
                 duration_subscriber=0.0
                 customer_trips=0
                 subscriber_trips=0
                 # tally up ride types
                 for row in reader:
                     if row['user_type'] == 'Customer':
                         customer_trips+=1
                         duration_customer+=float(row['duration'])
                     elif row['user_type'] == 'Subscriber':
                         subscriber_trips+=1
                         duration_subscriber+=float(row['duration'])
                 # compute average length
                 customer_avg=duration_customer/customer_trips
                 subscriber_avg= duration_subscriber/subscriber_trips
                 return(customer_avg, subscriber_avg)
         city_info = {'NYC': {'out_file': './data/NYC-2016-Summary.csv'},
                      'BayArea': {'out_file':'./examples/BayArea-Y3-Summary.csv'}}
         for city, filenames in city_info.items():
             print(city+':')
             print(average_user_duration(filenames['out_file']))
NYC:
(32.982004306775025, 13.680790523907177)
BayArea:
(54.55121116377032, 9.512633839275217)
```

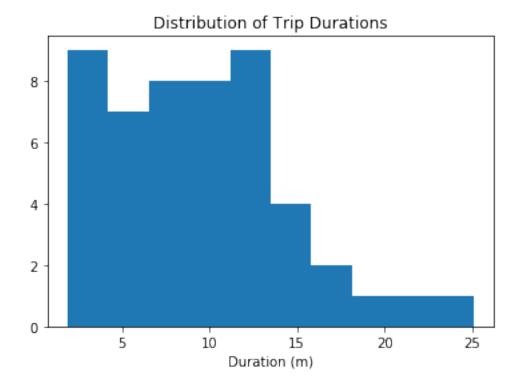
Visualizations

nnn

The last set of values that you computed should have pulled up an interesting result. While the mean trip time for Subscribers is well under 30 minutes, the mean trip time for Customers is actually *above* 30 minutes! It will be interesting for us to look at how the trip times are distributed.

In order to do this, a new library will be introduced here, matplotlib. Run the cell below to load the library and to generate an example plot.

```
In [94]: # load library
        import matplotlib.pyplot as plt
        # this is a 'magic word' that allows for plots to be displayed
        # inline with the notebook. If you want to know more, see:
        # http://ipython.readthedocs.io/en/stable/interactive/magics.html
        %matplotlib inline
        # example histogram, data taken from bay area sample
        data = [ 7.65, 8.92, 7.42, 5.50, 16.17, 4.20, 8.98, 9.62, 11.48, 14.33,
                19.02, 21.53, 3.90, 7.97, 2.62, 2.67, 3.08, 14.40, 12.90, 7.83,
                25.12, 8.30, 4.93, 12.43, 10.60, 6.17, 10.88, 4.78, 15.15,
                 9.43, 13.32, 11.72, 9.85, 5.22, 15.10, 3.95, 3.17, 8.78,
                 4.55, 12.68, 12.38, 9.78, 7.63, 6.45, 17.38, 11.90, 11.52, 8.63,]
        plt.hist(data)
        plt.title('Distribution of Trip Durations')
        plt.xlabel('Duration (m)')
        plt.show()
```

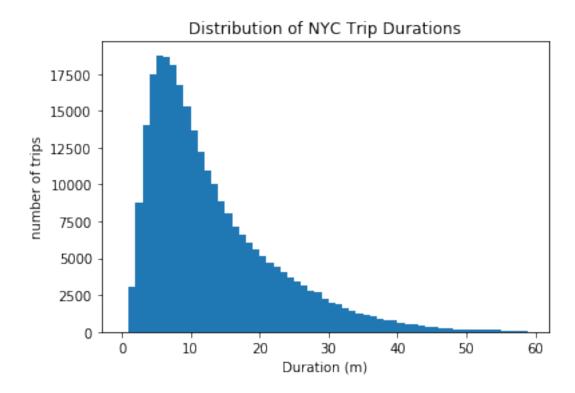


In the above cell, we collected fifty trip times in a list, and passed this list as the first argument to the <code>.hist()</code> function. This function performs the computations and creates plotting objects

for generating a histogram, but the plot is actually not rendered until the .show() function is executed. The .title() and .xlabel() functions provide some labeling for plot context.

You will now use these functions to create a histogram of the trip times for the city you selected in question 4c. Don't separate the Subscribers and Customers for now: just collect all of the trip times and plot them.

```
In [95]: ## Use this and additional cells to collect all of the trip times as a list ##
         ## and then use pyplot functions to generate a histogram of trip times.
         import matplotlib.pyplot as plt
         def build_duration_hist(filename, city):
             This function takes filename with trip data and a city name
             and builds a histogram with trips durations data.
             with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f_in)
                 # initialize count variables
                 durations=[]
                 # collect trips durations
                 for row in reader:
                     durations.append(float(row['duration']))
             #create histogram with duration data
             %matplotlib inline
             plt.hist(durations,range(0,60))
             plt.title('Distribution of '+city+' Trip Durations')
             plt.xlabel('Duration (m)')
             plt.ylabel('number of trips')
             plt.show()
             return 'The histogram above represnts the duration of '+city+' city trips data'
         city_info = {'NYC': {'out_file': './data/NYC-2016-Summary.csv'}}
         for city, filenames in city_info.items():
             print(build_duration_hist(filenames['out_file'], city))
```



The histogram above represnts the duration of NYC city trips data

If you followed the use of the .hist() and .show() functions exactly like in the example, you're probably looking at a plot that's completely unexpected. The plot consists of one extremely tall bar on the left, maybe a very short second bar, and a whole lot of empty space in the center and right. Take a look at the duration values on the x-axis. This suggests that there are some highly infrequent outliers in the data. Instead of reprocessing the data, you will use additional parameters with the .hist() function to limit the range of data that is plotted. Documentation for the function can be found [here].

Question 5: Use the parameters of the .hist() function to plot the distribution of trip times for the Subscribers in your selected city. Do the same thing for only the Customers. Add limits to the plots so that only trips of duration less than 75 minutes are plotted. As a bonus, set the plots up so that bars are in five-minute wide intervals. For each group, where is the peak of each distribution? How would you describe the shape of each distribution?

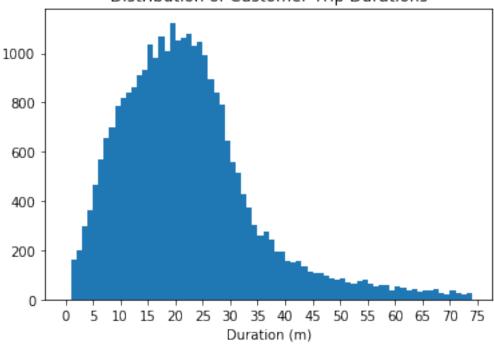
```
Answer: For NYC: Customers Histogram:
```

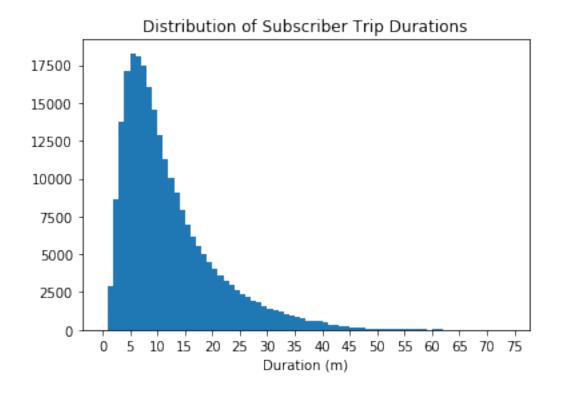
```
 The peak is 20 minutes for more than 1000 trips 
 The distribution is right skewed, range from 2 to 73 minutes and no outrliers 

Subscribers Histogram:
```

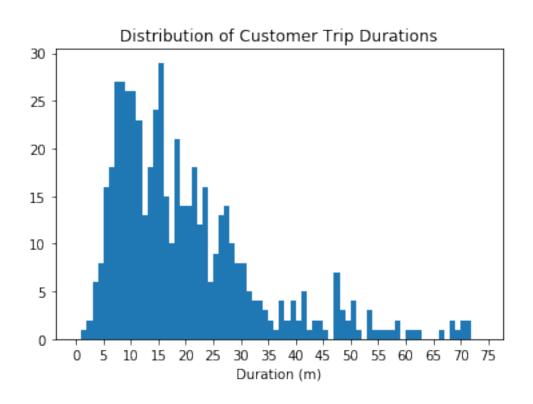
```
The peak is 5 minutes for more than 17500 trips 
       The distribution is right skewed, range from 2 to 63 minutes and no outliers
       </01>
    </10>
In [96]: ## Use this and additional cells to answer Question 5. ##
                                         ##
         def build_user_duration_hist(filename, city):
             This function reads in a file with trip data and a city name
             and builds two histograms with trips durations data for each
             subscribers and customers.
            with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f_in)
                 # initialize count variables
                 duration_customer=[]
                 duration_subscriber=[]
                 # tally up ride types
                 for row in reader:
                     if row['user_type'] == 'Customer':
                         duration_customer.append(float(row['duration']))
                     elif row['user_type'] == 'Subscriber':
                         duration_subscriber.append(float(row['duration']))
                 # build customers trip durations histogram
                 %matplotlib inline
                 plt.hist(duration_customer,range(0,75))
                 plt.title('Distribution of Customer Trip Durations')
                 plt.xlabel('Duration (m)')
                 plt.xticks(range(0,80,5))
                 plt.show()
                 #build subscribers trip druations histogram
                 plt.hist(duration_subscriber,range(0,75))
                 plt.title('Distribution of Subscriber Trip Durations')
                 plt.xlabel('Duration (m)')
                 plt.xticks(range(0,80,5))
                 plt.show()
                 return 'The histograms above represnts the duration of '+city+' city trips data
```

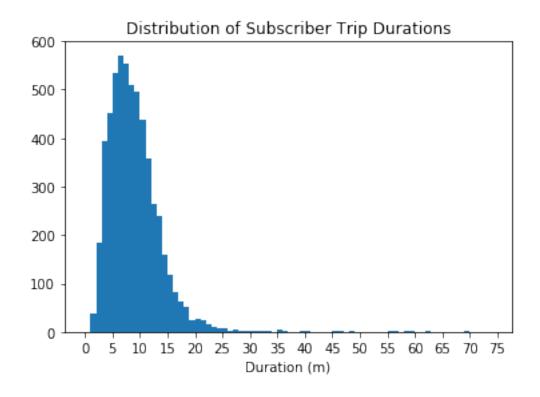
Distribution of Customer Trip Durations





The histograms above represnts the duration of NYC city trips data





The histograms above represnts the duration of BayArea city trips data

Performing Your Own Analysis

So far, you've performed an initial exploration into the data available. You have compared the relative volume of trips made between three U.S. cities and the ratio of trips made by Subscribers and Customers. For one of these cities, you have investigated differences between Subscribers and Customers in terms of how long a typical trip lasts. Now it is your turn to continue the exploration in a direction that you choose. Here are a few suggestions for questions to explore:

- How does ridership differ by month or season? Which month / season has the highest ridership? Does the ratio of Subscriber trips to Customer trips change depending on the month or season?
- Is the pattern of ridership different on the weekends versus weekdays? On what days are Subscribers most likely to use the system? What about Customers? Does the average duration of rides change depending on the day of the week?
- During what time of day is the system used the most? Is there a difference in usage patterns for Subscribers and Customers?

If any of the questions you posed in your answer to question 1 align with the bullet points above, this is a good opportunity to investigate one of them. As part of your investigation, you

will need to create a visualization. If you want to create something other than a histogram, then you might want to consult the Pyplot documentation. In particular, if you are plotting values across a categorical variable (e.g. city, user type), a bar chart will be useful. The documentation page for <code>.bar()</code> includes links at the bottom of the page with examples for you to build off of for your own use.

Question 6: Continue the investigation by exploring another question that could be answered by the data available. Document the question you want to explore below. Your investigation should involve at least two variables and should compare at least two groups. You should also use at least one visualization as part of your explorations.

- How does ridership differ by month or season? Which month / season has the highest ridership? Does the ratio of Subscriber trips to Customer trips change depending on the month or season?

Answer: For **NYC**: Months Histogram:

```
The lowest number of trips is in January of 10k trips and the number keeps increasing the second of the second
                                           to more than 30k trips in September and then goes down to 15k trips in December
                             The higest trips in September with more than 30,000 trips 
                             <B>Seasons Histogram: </B>
                             <01>
                             The lowest number of trips is in the winter of 40k trips and the number keeps increase.
                                           to more than 80k trips in the Summer and then goes down to 75k trips in the Autumn <
                             The higest trips in the Summer with more than 80,000 trips 
                             <B>Subscribers and Customers ratio (Months) Histogram: </B>
                             The ratio of Customers trips is 5% in Jan, Feb, Dec and 5%-10% in Mar, Jun, Oct, No.
                                           and increases to reach from 10% to 15% in Apr, May, Jul, Aug, Sep with maximum of 15%
                             The ratio of Subscribers trips is maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Maximum as 95% in Jan, Feb, Dec and 90%-95% in Ja
                                           and decreases to 85%-90% in Apr, May, Jul, Aug, Sep with minimum of 85% at Aug and Se
                             <B>Subscribers and Customers ratio (Seasons) Histogram: </B>
                             The ratio of Customers trips is 5:10% in Winter and Autumn and increases
                                           in the Summer and the Spring to reach from 10-15% with maximum of 15% in Summer 
                             The ratio of subscribers trips is 90:95% in Winter and Autumn and decreases
                                           in the Summer and the Spring to reach from 85-90% with maximum of 95% in Winter
                             </10>
In [97]: ## Use this and additional cells to continue to explore the dataset. ##
```

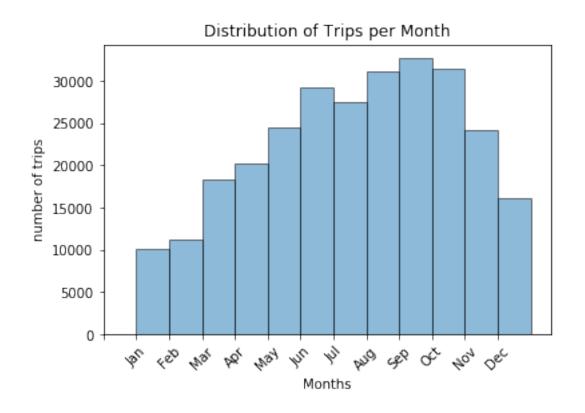
Once you have performed your exploration, document your findings

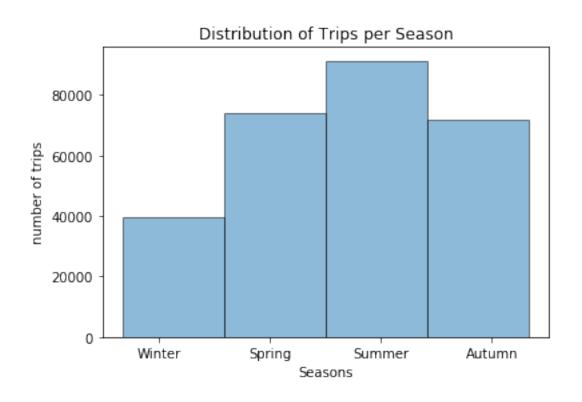
```
import csv
         import calendar
         import pandas as pd
         import numpy as np
         def trips_data(filename):
             HHHH
             This function reads in a file with trip data and reports trips data
             based on months for customers and subscribers for the use of builidng
             histograms...
             11 11 11
             with open(filename, 'r') as f_in:
                 # set up csv reader object
                 reader = csv.DictReader(f in)
                 # initialize count variables
                 # for counting each trip in each month
                 t_month=[]
                 c_month=[]
                 s_month=[]
                 # counting each trip in each month
                 for row in reader:
                     if row['user_type'] == 'Customer':
                         c_month.append(int(row['month']))
                     elif row['user_type'] == 'Subscriber':
                         s_month.append(int(row['month']))
                     t_month.append(int(row['month']))
             return (s_month, c_month, t_month)
In [98]: city_info = {'NYC': {'out_file': './data/NYC-2016-Summary.csv'}}
         # getting trips number in each month
         s_month, c_month, t_month = trips_data(city_info['NYC']['out_file'])
         # for storing each month trips numbers
         month_data=[]
         season_data=[]
         customer_data=[]
         subscriber_data=[]
         # for storing the ratio of user type to the total month trips
         customer_ratio=[]
         subscriber_ratio=[]
         customer_seas_ratio=[]
```

##

in the Markdown cell above.

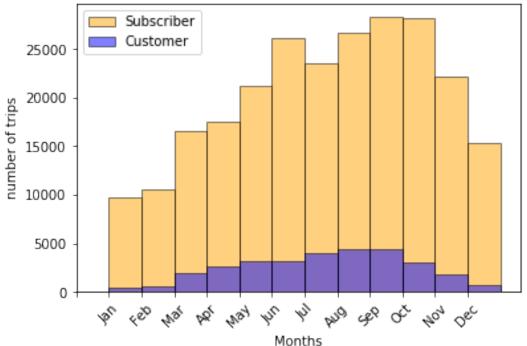
```
subscriber_seas_ratio=[]
%matplotlib inline
# histogram of months trips
n, bins, patches_month= plt.hist(t_month, range(1,14), alpha=0.5, histtype='bar', ec='bar', ec='
plt.xticks(range(13), calendar.month_abbr[0:15], rotation=45)
plt.title('Distribution of Trips per Month')
plt.ylabel('number of trips')
plt.xlabel('Months')
plt.show()
#store month trips data from the histogram
for bar in patches_month:
             month_data.append(bar.get_height())
# histogram for seasons trips
n, bins, patches_season= plt.hist(t_month, bins=4, alpha=0.5, histtype='bar', ec='black
plt.xticks([2,5,8,11], ['Winter', 'Spring', 'Summer', 'Autumn'], rotation=0)
plt.title('Distribution of Trips per Season')
plt.ylabel('number of trips')
plt.xlabel('Seasons')
plt.show()
#store season trips data from the histogram
for bar in patches_season:
             season_data.append(bar.get_height())
```



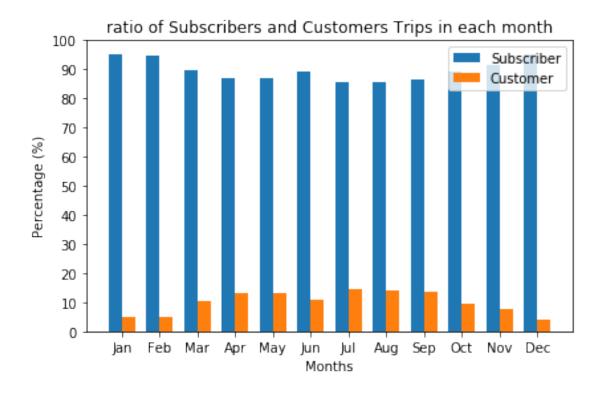


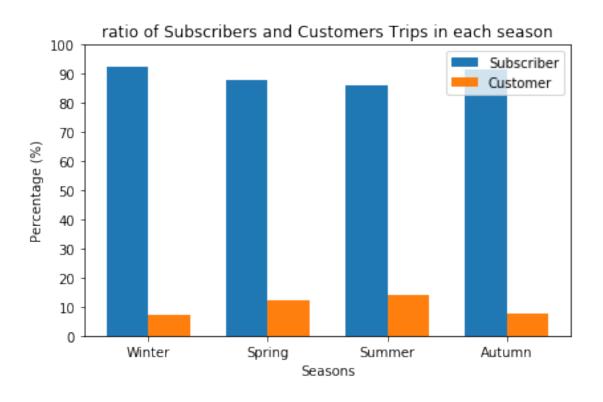
```
In [99]: # storing subscriber_data trips each month
         n, bins, patches_subscriber =plt.hist(s_month, range(1,14), alpha=0.5, histtype='bar',
         for bar in patches_subscriber:
             subscriber_data.append(bar.get_height())
         # storing customer_data trips each month
         n, bins, patches_customer=plt.hist(c_month, range(1,14), alpha=0.5, histtype='bar', ec=
         for bar in patches_customer:
             customer_data.append(bar.get_height())
         # customers and subscribers monthly trips comparison histogram
         plt.xticks( range(13), calendar.month_abbr[0:15], rotation=45 )
         plt.title('Distribution of Subscribers and Customers Trips in each month')
         plt.ylabel('number of trips')
         plt.xlabel('Months')
         plt.legend()
         plt.show()
         #calculating ratio of customers and subscribers trips in each month
         for i in range(len(month_data)):
             customer_ratio.append(customer_data[i]/month_data[i]*100)
             subscriber_ratio.append(subscriber_data[i]/month_data[i]*100)
```

Distribution of Subscribers and Customers Trips in each month



```
In [100]: #building histogram bar comparison between subscribers and customers monthly trips rat
          labels = calendar.month_abbr[1:]
          x = np.arange(12) # the label locations
          width = 0.35 # the width of the bars
          fig, ax = plt.subplots()
          rects2 = ax.bar(x - width/2, subscriber_ratio, width, label='Subscriber')
          rects1 = ax.bar(x + width/2, customer_ratio, width, label='Customer')
          ax.set_ylabel('Percentage (%)')
          ax.set_xlabel('Months')
          ax.set_title('ratio of Subscribers and Customers Trips in each month')
          ax.set_xticks(x)
          ax.set_yticks(range(0,101,10))
          ax.set_xticklabels(labels)
          ax.legend()
          fig.tight_layout()
          plt.show()
          s=0
          z=3
          #calculating ratio of customers and subscribers trips in each month
          for i in range(len(season_data)):
              customer_seas_ratio.append(sum(customer_data[s:z])/season_data[i]*100)
              subscriber_seas_ratio.append(sum(subscriber_data[s:z])/season_data[i]*100)
              s + = 3
              z + = 3
          #building histogram bar comparison between subscribers and customers seasons trips rat
          labels = ['Winter', 'Spring', 'Summer', 'Autumn']
          x = np.arange(4) # the label locations
          width = 0.35 # the width of the bars
          fig, ax = plt.subplots()
          rects2 = ax.bar(x - width/2, subscriber_seas_ratio, width, label='Subscriber')
          rects1 = ax.bar(x + width/2, customer_seas_ratio, width, label='Customer')
          ax.set_ylabel('Percentage (%)')
          ax.set_xlabel('Seasons')
          ax.set_title('ratio of Subscribers and Customers Trips in each season')
          ax.set_xticks(x)
          ax.set_yticks(range(0,101,10))
          ax.set_xticklabels(labels)
          ax.legend()
          fig.tight_layout()
          plt.show()
```





Conclusions

Congratulations on completing the project! This is only a sampling of the data analysis process: from generating questions, wrangling the data, and to exploring the data. Normally, at this point in the data analysis process, you might want to draw conclusions about the data by performing a statistical test or fitting the data to a model for making predictions. There are also a lot of potential analyses that could be performed on the data which are not possible with only the data provided. For example, detailed location data has not been investigated. Where are the most commonly used docks? What are the most common routes? As another example, weather has potential to have a large impact on daily ridership. How much is ridership impacted when there is rain or snow? Are subscribers or customers affected more by changes in weather?

Question 7: Putting the bike share data aside, think of a topic or field of interest where you would like to be able to apply the techniques of data science. What would you like to be able to learn from your chosen subject?

Answer: I am interested in E-commerce industry I want to learn more about customers The reasons they buy and the reasons they don't

What average/ratio of customers buy from social media channels and direct search
How much engagment and clicks or page visits do a user do before he purchase an item
Do guests buy in average more than registered users and do they spend more on products
how many products customers search for and they don't find on my website

Tip: If we want to share the results of our analysis with others, we aren't limited to giving them a copy of the jupyter Notebook (.ipynb) file. We can also export the Notebook output in a form that can be opened even for those without Python installed. From the **File** menu in the upper left, go to the **Download as** submenu. You can then choose a different format that can be viewed more generally, such as HTML (.html) or PDF (.pdf). You may need additional packages or software to perform these exports.

If you are working on this project via the Project Notebook page in the classroom, you can also submit this project directly from the workspace. **Before you do that**, you should save an HTML copy of the completed project to the workspace by running the code cell below. If it worked correctly, the output code should be a 0, and if you click on the jupyter icon in the upper left, you should see your .html document in the workspace directory. Alternatively, you can download the .html copy of your report following the steps in the previous paragraph, then *upload* the report to the directory (by clicking the jupyter icon).

Either way, once you've gotten the .html report in your workspace, you can complete your submission by clicking on the "Submit Project" button to the lower-right hand side of the workspace.