Workbook

Use this notebook to complete the exercises throughout the workshop.

Table of Contents

- Section 1 Getting Started with Pandas
- Section 2 Data Wrangling
- Section 3 Data Visualization

Section 1

Exercise 1.1

Create a DataFrame by reading in the 2019_Yellow_Taxi_Trip_Data.csv file. Examine the first 5 rows.

```
In []: import pandas as pd

df = pd.read_csv('../data/2019_Yellow_Taxi_Trip_Data.csv')
```

Exercise 1.2

Find the dimensions (number of rows and number of columns) in the data.

```
In []: import pandas as pd
    df = pd.read_csv('../data/2019_Yellow_Taxi_Trip_Data.csv')
    df.shape
Out[]: (10000, 18)
```

Exercise 1.3

Using the data in the 2019_Yellow_Taxi_Trip_Data.csv file, calculate summary statistics for the fare_amount , tip_amount , tolls_amount , and total_amount columns.

```
In []: import pandas as pd

df = pd.read_csv('../data/2019_Yellow_Taxi_Trip_Data.csv')

df[['fare_amount','tip_amount','totls_amount','total_amount']].describe()
```

	fare_amount	tip_amount	tolls_amount	total_amount
count	10000.000000	10000.000000	10000.000000	10000.000000
mean	15.106313	2.634494	0.623447	22.564659
std	13.954762	3.409800	6.437507	19.209255
min	-52.000000	0.000000	-6.120000	-65.920000
25%	7.000000	0.000000	0.000000	12.375000
50%	10.000000	2.000000	0.000000	16.300000
75%	16.000000	3.250000	0.000000	22.880000
max	176.000000	43.000000	612.000000	671.800000

Exercise 1.4

Out[]:

Isolate the fare_amount, tip_amount, tolls_amount, and total_amount for the longest trip by distance (trip_distance).

Section 2

Exercise 2.1

Read in the meteorite data from the Meteorite_Landings.csv file, rename the mass (g) column to mass, and drop all the latitude and longitude columns. Sort the result by mass in descending order.

```
In []: import pandas as pd

df = pd.read_csv('../data/Meteorite_Landings.csv')

df.rename(columns={'mass (g)': 'mass'}, inplace=True)

df.drop(['reclat','reclong'], axis=1, inplace=True)

df.sort_values(['mass'], ascending=False, inplace=True)

df.head()
```

Out[]:

GeoLc	year	fall	mass	recclass	nametype	id	name	
(- 19.	01/01/1920 12:00:00 AM	Found	60000000.0	Iron, IVB	Valid	11890	Hoba	16392
(76. -64.9	01/01/1818 12:00:00 AM	Found	58200000.0	Iron, IIIAB	Valid	5262	Cape York	5373
(-27. -60.	12/22/1575 12:00:00 AM	Found	50000000.0	lron, IAB-MG	Valid	5247	Campo del Cielo	5365
-111.0	01/01/1891 12:00:00 AM	Found	30000000.0	Iron, IAB-MG	Valid	5257	Canyon Diablo	5370
(47.0	01/01/1898 12:00:00 AM	Found	28000000.0	Iron, IIIE	Valid	2335	Armanty	3455

Exercise 2.2

Using the meteorite data from the Meteorite_Landings.csv file, update the year column to only contain the year, convert it to a numeric data type, and create a new column indicating whether the meteorite was observed falling before 1970. Set the index to the id column and extract all the rows with IDs between 10,036 and 10,040 (inclusive) with loc[].

Hint 1: Use year.str.slice() to grab a substring.

Hint 2: Make sure to sort the index before using <code>loc[]</code> to select the range.

Bonus: There's a data entry error in the year column. Can you find it? (Don't spend too much time on this.)

```
In []: import pandas as pd

df = pd.read_csv('../data/Meteorite_Landings.csv')

df['year']=df['year'].str.slice(start=6, stop=10)

df.dropna(subset = ['year'],inplace = True)

df['year']=df['year'].astype('int64')

df = df.assign(before1970=lambda x: x.year < 1970)

df.set_index('id')

df.sort_index()

between = df.loc[10036:10040]

between.head()</pre>
```

Out[]:

		name	id	nametype	recclass	mass (g)	fall	year	reclat	rec
	10036	Elephant Moraine 90022	8432	Valid	CK5	15.5	Found	1990	-76.28573	156.4!
	10037	Elephant Moraine 90023	8433	Valid	CK5	31.5	Found	1990	-76.27507	156.41
	10038	Elephant Moraine 90024	8434	Valid	Eucrite- br	22.8	Found	1990	-76.28843	156.47
	10039	Elephant Moraine 90025	8435	Valid	CK5	45.8	Found	1990	-76.28200	156.39
	10040	Elephant Moraine 90026	8436	Valid	CK5	61.5	Found	1990	-76.29226	156.45

Exercise 2.3

Using the meteorite data from the Meteorite_Landings.csv file, create a pivot table that shows both the number of meteorites and the 95th percentile of meteorite mass for those that were found versus observed falling per year from 2005 through 2009 (inclusive). Hint: Be sure to convert the year column to a number as we did in the previous exercise.

```
In []: import pandas as pd

df = pd.read_csv('../data/Meteorite_Landings.csv')

df['year'] = df['year'].str.slice(start=6, stop=10)
    df.dropna(subset=['year'], inplace=True)
    df['year'] = df['year'].astype('int64')

df_filtered = df.query('2005 <= year <= 2009')

count_df = df_filtered.groupby(['year', 'fall'])['id'].count().unstack()

p95_df = df_filtered.groupby(['year', 'fall'])['mass (g)'].apply(lambda x

final_df = pd.merge(count_df, p95_df, left_index=True, right_index=True,

final_df.head()</pre>
```

Out[]:	fall	Fell_count	Found_count	Fell_p95_mass	Found_p95_mass		
	year						
	2005	NaN	875.0	NaN	4500.00		
	2006	5.0	2451.0	25008.0	1600.50		
	2007	8.0	1181.0	89675.0	1126.90		
	2008	9.0	948.0	106000.0	106000.0 2274.80		
	2009	5.0	1492.0	8333.4	1397.25		

Exercise 2.4

Using the meteorite data from the Meteorite_Landings.csv file, compare summary statistics of the mass column for the meteorites that were found versus observed falling.

```
In [ ]: import pandas as pd
        df = pd.read_csv('../data/Meteorite_Landings.csv')
        print(df.groupby('fall')['mass (g)'].describe())
                                                                              75%
                count
                                               std min
                                                            25%
                                                                     50%
                               mean
       fall
       Fell
               1075.0 47070.715023 717067.125826 0.1
                                                         686.00
                                                                 2800.0
                                                                          10450.0
       Found 44510.0 12461.922983 571105.752311 0.0
                                                            6.94
                                                                    30.5
                                                                            178.0
                     max
       fall
       Fell
              23000000.0
       Found
              60000000.0
```

Exercise 2.5

Using the taxi trip data in the 2019_Yellow_Taxi_Trip_Data.csv file, resample the data to an hourly frequency based on the dropoff time. Calculate the total trip_distance, fare_amount, tolls_amount, and tip_amount, then find the 5 hours with the most tips.

```
In []: import pandas as pd

df = pd.read_csv('../data/2019_Yellow_Taxi_Trip_Data.csv')

df['tpep_dropoff_datetime'] = pd.to_datetime(df['tpep_dropoff_datetime'])

df.set_index('tpep_dropoff_datetime', inplace = True)

df = df.resample('H').agg({'trip_distance': 'sum', 'fare_amount': 'sum', print(df.nlargest(5, 'tip_amount'))

#df.head()
```

	trip_distance	fare_amount	tolls_amount	tip_amoun
t				
tpep_dropoff_datetime				
2019-10-23 16:00:00	10676.95	67797.76	699.04	12228.6
4				
2019-10-23 17:00:00	16052.83	70131.91	4044.04	12044.0
3				
2019-10-23 18:00:00	3104.56	11565.56	1454.67	1907.6
4				
2019-10-23 15:00:00	14.34	213.50	0.00	51.7
5				
2019-10-23 19:00:00	98.59	268.00	24.48	25.7
4				

Out[]:

trip_distance fare_amount tolls_amount tip_amount

tpep_dropoff_datetime								
2019-10-23 07:00:00	0.67	4.5	0.0	0.0				
2019-10-23 08:00:00	17.07	62.5	0.0	4.0				
2019-10-23 09:00:00	1.58	58.0	0.0	0.0				
2019-10-23 10:00:00	0.00	0.0	0.0	0.0				
2019-10-23 11:00:00	0.00	0.0	0.0	0.0				

Section 3

Exercise 3.1

Using the TSA traveler throughput data in the tsa_melted_holiday_travel.csv file, create box plots for traveler throughput for each year in the data. Hint: Pass kind='box' into the plot() method to generate box plots.

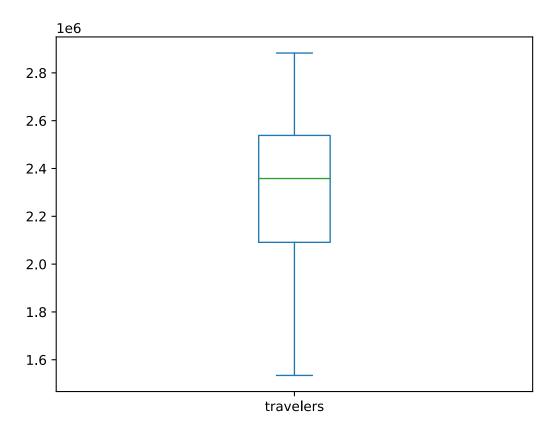
```
In []: import pandas as pd
import matplotlib_inline
from utils import mpl_svg_config

matplotlib_inline.backend_inline.set_matplotlib_formats(
    'svg',
    **mpl_svg_config('section-3')
)

df = pd.read_csv('../data/tsa_melted_holiday_travel.csv',parse_dates=True
box2019 = df.query('year == 2019').travelers.plot(kind='box')

#box2020 = df.query('year == 2020').travelers.plot(kind='box')

#box2021 = df.query('year == 2021').travelers.plot(kind='box')
```



Exercise 3.2

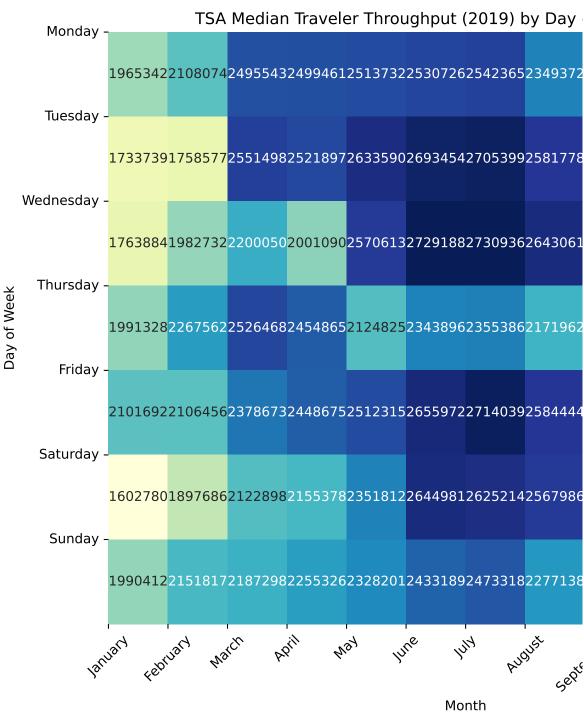
Using the TSA traveler throughput data in the tsa_melted_holiday_travel.csv file, create a heatmap that shows the 2019 TSA median traveler throughput by day of week and month.

```
In [ ]:
        import pandas as pd
        import matplotlib_inline
        from utils import mpl_svg_config
        import matplotlib.pyplot as plt
        import seaborn as sns
        matplotlib_inline.backend_inline.set_matplotlib_formats(
            'svg', # output images using SVG format
            **mpl_svg_config('section-3') # optional: configure metadata
        )
        # Load the TSA traveler throughput data
        df = pd.read_csv('../data/tsa_melted_holiday_travel.csv', parse_dates=Tru
        # Extract year, month, and day of week from the 'date' column
        df['Year'] = df.index.year
        df['Month'] = df.index.month
        df['DayOfWeek'] = df.index.dayofweek
        df_2019 = df[df['Year'] == 2019]
        pivot_table = pd.pivot_table(df_2019, values='travelers', index='DayOfWee
        month_order = ['January', 'February', 'March', 'April', 'May', 'June', 'J
        day_order = ['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Sat
```

```
plt.figure(figsize=(12, 8))
sns.heatmap(pivot_table, cmap='YlGnBu', annot=True, fmt=".0f", cbar_kws={
  plt.title('TSA Median Traveler Throughput (2019) by Day of Week and Month
  plt.xlabel('Month')
  plt.ylabel('Day of Week')

plt.xticks(ticks=range(12), labels=month_order, rotation=45)
  plt.yticks(ticks=range(7), labels=day_order, rotation=0)

plt.show()
```

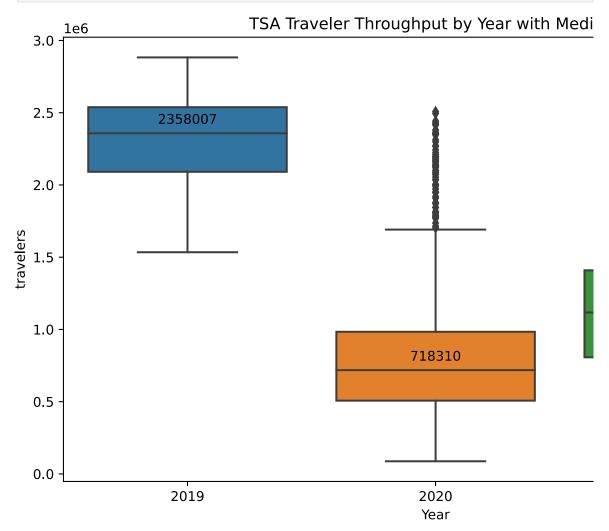


Exercise 3.3

Annotate the medians in the box plot from *Exercise 3.1*. Hint: The \times coordinates will be 1, 2, and 3 for 2019, 2020, and 2021, respectively. Alternatively, to avoid

hardcoding values, you can use the Axes get_xticklabels() method, in which case you should look at the documentation for the Text class.

```
In []:
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
        df = pd.read_csv('../data/tsa_melted_holiday_travel.csv')
        df['date'] = pd.to_datetime(df['date'])
        df['Year'] = df['date'].dt.year
        plt.figure(figsize=(10, 6))
        ax = sns.boxplot(x='Year', y='travelers', data=df)
        medians = df.groupby(['Year'])['travelers'].median()
        for x_tick, (year, median) in enumerate(medians.items()):
            ax.annotate(f'{median:.0f}', xy=(x_tick, median),
                        xytext=(0,5),
                        textcoords='offset points',
                        ha='center', va='bottom')
        plt.title('TSA Traveler Throughput by Year with Medians')
        plt.show()
```



In []: