

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

## Exercise

For these exercises we are using a [dataset \(https://www.kaggle.com/dgomonov/new-york-city-airbnb-open-data/kernels\)](https://www.kaggle.com/dgomonov/new-york-city-airbnb-open-data/kernels) provided by Airbnb for a Kaggle competition. It describes its offer for New York City in 2019, including types of appartments, price, location etc.

### 1. Create a dataframe

Create a dataframe of a few lines with objects and their poperties (e.g fruits, their weight and colour). Calculate the mean of your Dataframe.

```
In [2]: dict_of_list = {'fruit_name': ["apple", "pear", "watermelon"], 'weight': [100, 94, 95], 'colour': ['green', "yellow", "rosa"]}
fruits = pd.DataFrame(dict_of_list)
```

```
In [3]: fruits.describe()
# calculates common statistical values
# and makes it only for the columns that make sense
```

Out[3]:

	weight
count	3.000000
mean	96.333333
std	3.214550
min	94.000000
25%	94.500000
50%	95.000000
75%	97.500000
max	100.000000

```
In [4]: fruits.mean()
```

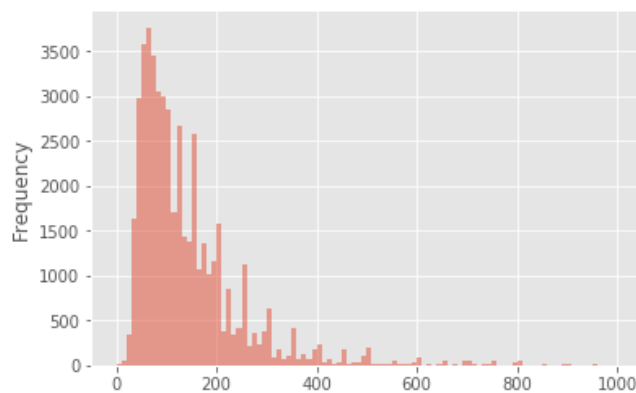
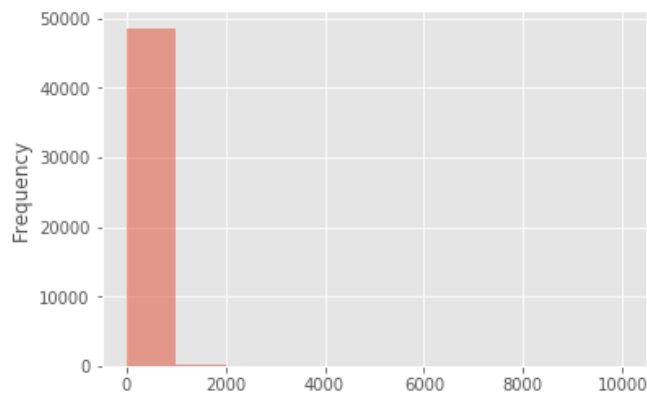
```
Out[4]: weight    96.333333
dtype: float64
```

### 2. Import

- Import the table called AB\_NYC\_2019.csv as a dataframe. It is located in the Datasets folder. Have a look at the beginning of the table (head).
- Create a histogram of prices

```
In [5]: mydata = pd.read_csv('Datasets/AB_NYC_2019.csv')
# mydata
```

```
In [6]: plt.style.use('ggplot')
mydata['price'].plot.hist(alpha = 0.5)
plt.show()
# to have nicer plot (more bars)
mydata['price'].plot.hist(alpha = 0.5, bins=range(0,1000,10))
plt.show()
```



### 3. Operations

Create a new column in the dataframe by multiplying the "price" and "availability\_365" columns to get an estimate of the maximum yearly income.

```
In [7]: mydata['max_yearly_income'] = mydata['price'] * mydata['availability_365']
```

```
/usr/local/lib/python3.5/dist-packages/pandas/core/computation/check.py:1
9: UserWarning: The installed version of numexpr 2.4.3 is not supported i
n pandas and will be not be used
The minimum supported version is 2.6.1
```

```
ver=ver, min_ver=_MIN_NUMEXPR_VERSION), UserWarning)
```

```
In [8]: # what can be done with numpy can be done
# np.log(mydata['price'])
```

```
In [9]: # mydata
```

### 3b. Subselection and plotting

Create a new Dataframe by first subselecting yearly incomes between 1 and 100'000. Then make a scatter plot of yearly income versus number of reviews

```
In [10]: #mydata_sub = mydata[ (mydata['max_yearly_income'] >= 1) and (mydata['max_yearly_income'] <= 100000) ]
#mydata_sub = mydata[ (mydata.max_yearly_income >= 1) and (mydata.max_yearly_income <= 100000) ]
mydata_sub = mydata[ (mydata['max_yearly_income'] >= 1) & (mydata['max_yearly_income'] <= 100000) ].copy()
# mydata[(mydata.max_yearly_income>=1)&(mydata.max_yearly_income <= 100000)].copy()
# mydata_sub
```

```
In [11]: mydata_sub.plot(x = 'number_of_reviews', y = 'max_yearly_income', kind = 'scatter')
max(mydata_sub['max_yearly_income'])
```

Out[11]: 99900



## 4. Combine

We provide below an additional table that contains the number of inhabitants of each of New York's boroughs ("neighbourhood\_group" in the table). Use merge to add this population information to each element in the original dataframe.

```
In [12]: borough_dt = pd.read_excel('Datasets/ny_boroughs.xlsx')
#borough_dt
```

```
In [13]: #mydata
```

```
In [14]: merged_dt = pd.merge(mydata, borough_dt, left_on='neighbourhood_group',
right_on='borough', how='left')
#merged_dt
```

## 5. Groups

- Using groupby calculate the average price for each type of room (room\_type) in each neighbourhood\_group.  
What is the average price for an entire home in Brooklyn ?
- Unstack the multi-level Dataframe into a regular Dataframe with `unstack()` and create a bar plot with the resulting table

```
In [15]: merged_dt.groupby(['neighbourhood_group', 'room_type']).price.mean()
```

```
Out[15]: neighbourhood_group room_type
Bronx      Entire home/apt      127.506596
           Private room         66.788344
           Shared room         59.800000
Brooklyn   Entire home/apt      178.327545
           Private room         76.500099
           Shared room         50.527845
Manhattan  Entire home/apt      249.239109
           Private room        116.776622
           Shared room         88.977083
Queens     Entire home/apt      147.050573
           Private room         71.762456
           Shared room         69.020202
Staten Island Entire home/apt  173.846591
           Private room         62.292553
           Shared room         57.444444
Name: price, dtype: float64
```

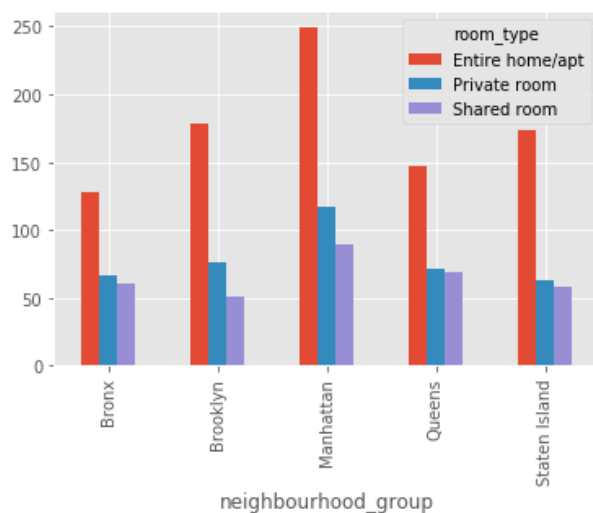
```
In [16]: merged_dt.groupby(['neighbourhood_group', 'room_type']).price.mean()['Brooklyn']['Entire home/apt']
```

```
Out[16]: 178.32754472225128
```

```
In [17]: merged_dt.groupby(['neighbourhood_group', 'room_type'])['price'].mean()['Brooklyn']['Entire home/apt']
```

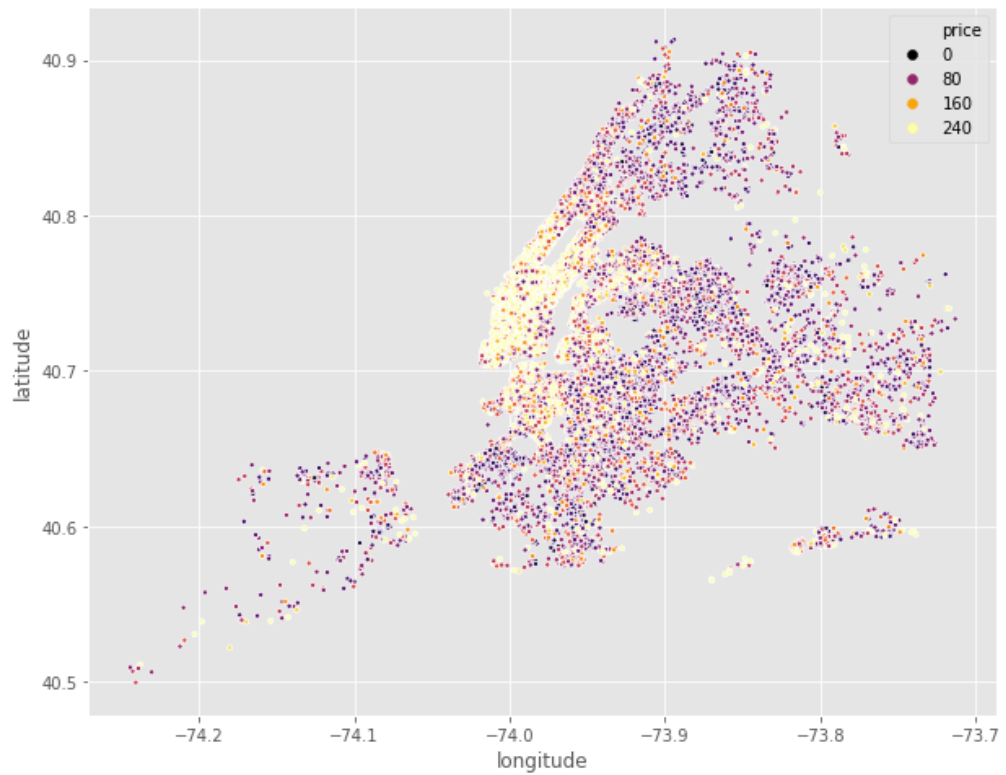
```
Out[17]: 178.32754472225128
```

```
In [18]: unstd_dt = merged_dt.groupby(['neighbourhood_group', 'room_type']).price.mean().unstack()
unstd_dt.plot(kind = 'bar');
```



## 6. Advanced plotting

```
In [19]: fig, ax = plt.subplots(figsize=(10,8))
g = sns.scatterplot(data = merged_dt, y = 'latitude', x = 'longitude', hue = 'price',
                    hue_norm=(0,200), s=10, palette='inferno')
```



Using Seaborn, create a scatter plot where x and y positions are longitude and latitude, the color reflects price and the shape of the marker the borough (neighbourhood\_group). Can you recognize parts of new york ? Does the map make sense ?