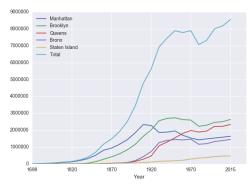
MfA: Python in the City



Katherine St. John City University of New York American Museum of Natural History

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Goal: Every table have at most one from each school, one from each discipline.

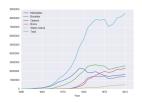
K. St. John (Hunter & AMNH)

Session 2

20 November 2019 1 / 51

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Outline

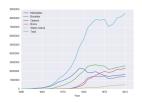


- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

3

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Recap: Workshop Overview





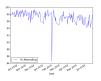


Three sessions:

- Flood Maps (arrays & images)
- 2 Noisiest Street (structured data, file I/O)
- 3 Mapping Collisions (using objects, mapping coordinates)

Recap: Workshop Overview







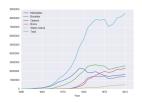
Three sessions:

- Flood Maps (arrays & images)
- 2 Noisiest Street (structured data, file I/O)
- 3 Mapping Collisions (using objects, mapping coordinates)

Each session:

- Design Challenge
 - Analyze a publicly available dataset
 - Introduce computing concepts & packages
 - Write a program to solve the problem
- Variations on the theme
- Design a Challenge

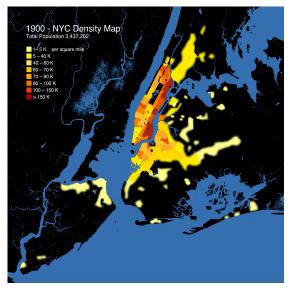
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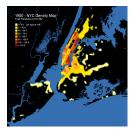
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(Myles Zhang, wiki)

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(Myles Zhang, wiki)

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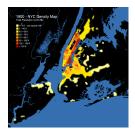
(Myles Zhang, wiki)

Working in Groups: graph NYC population (and growth) since 1900.



(Myles Zhang, wiki)

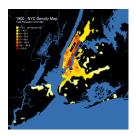
Working in Groups: graph NYC population (and growth) since 1900.



(Myles Zhang, wiki)

Working in Groups: graph NYC population (and growth) since 1900.

- Input: What data do you need?
- Process: How can you compute the growth?
- Output: How can you present your data?

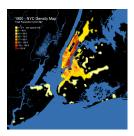


(Myles Zhang, wiki)

Input: What data do you need?

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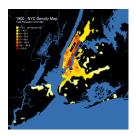
(Myles Zhang, wiki)

Input: What data do you need?

• Populations of the city in 1900 to today.

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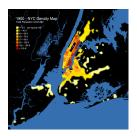
Image: A match a ma



(Myles Zhang, wiki)

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- Wikipedia page on NYC Historical Population.

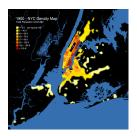


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CSV File: stjohn.github.io/service/mfa/f19.html

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Year, Manhattan, Brooklyn, Queens, Bronx, Staten Island, Total 1698,4937,2017,,,727,7681 1771,21863,3623,,,2847,28423 1790, 33131, 4549, 6159, 1781, 3827, 49447 1800,60515,5740,6642,1755,4563,79215 1810,96373,8303,7444,2267,5347,119734 1820, 123706, 11187, 8246, 2782, 6135, 152056 1830, 202589, 20535, 9049, 3023, 7082, 242278 1840, 312710, 47613, 14480, 5346, 10965, 391114 1850, 515547, 138882, 18593, 8032, 15061, 696115 1860,813669,279122,32903,23593,25492,1174779 1870,942292,419921,45468,37393,33029,1478103 1880, 1164673, 599495, 56559, 51980, 38991, 1911698 1890,1441216,838547,87050,88908,51693,2507414 1900, 1850093, 1166582, 152999, 200507, 67021, 3437202 1910,2331542,1634351,284041,430980,85969,4766883 1920, 2284103, 2018356, 469042, 732016, 116531, 5620048 1930, 1867312, 2560401, 1079129, 1265258, 158346, 6930446 1940, 1889924, 2698285, 1297634, 1394711, 174441, 7454995 1950, 1960101, 2738175, 1550849, 1451277, 191555, 7891957 1960, 1698281, 2627319, 1809578, 1424815, 221991, 7781984 1970, 1539233, 2602012, 1986473, 1471701, 295443, 7894862 1980, 1428285, 2230936, 1891325, 1168972, 352121, 7071639 1990, 1487536, 2300664, 1951598, 1203789, 378977, 7322564 2000,1537195,2465326,2229379,1332650,443728,8008278 2010, 1585873, 2504700, 2230722, 1385108, 468730, 8175133 2015,1644518,2636735,2339150,1455444,474558,8550405

nycHistPop.csv

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- Columns are separated by commas on each line.

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Structured Data



• We will use the popular Python Data Analysis Library (Pandas).

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- To use, add to the top of your file:

import pandas as pd

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- Often abbreviated: df.

K. St. John (Hunter & AMNH)

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Source: https://en.wikipedia.org/wiki/Demographics_of_New_York_City,,,,, All population figures are consistent with present-day boundaries.,,,,, First census after the consolidation of the five boroughs..... Year, Manhattan, Brooklyn, Oueens, Bronx, Staten Island, Total 1698,4937,2017,,,727,7681 1771,21863,3623,...2847,28423 1790, 33131, 4549, 6159, 1781, 3827, 49447 1800,60515,5740,6642,1755,4563,79215 1810,96373,8303,7444,2267,5347,119734 1820, 123706, 11187, 8246, 2782, 6135, 152056 1830, 202589, 20535, 9049, 3023, 7082, 242278 1840.312710.47613.14480.5346.10965.391114 1850, 515547, 138882, 18593, 8032, 15061, 696115 1860,813669,279122,32903,23593,25492,1174779 1870,942292,419921,45468,37393,33029,1478103 1880, 1164673, 599495, 56559, 51980, 38991, 1911698 1890,1441216,838547,87050,88908,51693,2507414 1900, 1850093, 1166582, 152999, 200507, 67021, 3437202 1910,2331542,1634351,284041,430980,85969,4766883 1920,2284103,2018356,469042,732016,116531,5620048 1930, 1867312, 2560401, 1079129, 1265258, 158346, 6930446 1940, 1889924, 2698285, 1297634, 1394711, 174441, 7454995 1950, 1960101, 2738175, 1550849, 1451277, 191555, 7891957 1960, 1698281, 2627319, 1809578, 1424815, 221991, 7781984 1970, 1539233, 2602012, 1986473, 1471701, 295443, 7894862 1980, 1428285, 2230936, 1891325, 1168972, 352121, 7071639 1990, 1487536, 2300664, 1951598, 1203789, 378977, 7322564 2000,1537195,2465326,2229379,1332650,443728,8008278 2010.1585873.2504700.2230722.1385108.468730.8175133 2015, 1644518, 2636735, 2339150, 1455444, 474558, 8550405

- To read in a CSV file: myVar = pd.read_csv("myFile.csv")
- Pandas has its own type, DataFrame, that is perfect for holding a sheet of data.
- Often abbreviated: df.
- It also has **Series**, that is perfect for holding a row or column of data.

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Year, Manhattan, Brooklyn, Queens, Bronx, Staten Island, Total 1698, 4937, 2017, ., 727, 7681 1771, 21863, 3623, ., 2847, 28423 1790, 33131, 4549, 6159, 1781, 3827, 49447 1800,60515,5740,6642,1755,4563,79215 1810, 96373, 8303, 7444, 2267, 5347, 119734 1820,123706,11187,8246,2782,6135,152056 1830, 202589, 20535, 9049, 3023, 7082, 242278 1840.312710.47613.14480.5346.10965.391114 1850, 515547, 138882, 18593, 8032, 15061, 696115 1860,813669,279122,32903,23593,25492,1174779 1870,942292,419921,45468,37393,33029,1478103 1880, 1164673, 599495, 56559, 51980, 38991, 1911698 1890,1441216,838547,87050,88908,51693.2507414 1900,1850093,1166582,152999,200507,67021,3437202 1910,2331542,1634351,284041,430980,85969,4766883 1920.2284103.2018356.469042.732016.116531.5620048 1930, 1867312, 2560401, 1079129, 1265258, 158346, 6930446 1940,1009924,2690205,1297634,1394711,174441,7454995 1950, 1960101, 2738175, 1550849, 1451277, 191555, 7891957 1960,1698281,2627319,1809578,1424815,221991,7781984 1970, 1539233, 2602012, 1986473, 1471701, 295443, 7894862 1980.1428285.2230936.1891325.1168972.352121.7071639 1990, 1487536, 2300664, 1951598, 1203789, 378977, 7322564 2000.1537195.2465326.2229379.1332650.443728.8008278 2010,1585873,2504700,2230722,1385108,468730,8175133 2015, 1644518, 2636735, 2339150, 1455444, 474558, 8550405

nycHistPop.csv

K. St. John (Hunter & AMNH)

import matplotlib.pyplot as plt import pandas as pd

1800,60515,5740,6642,1755,4563,79215 1810, 96373, 8303, 7444, 2267, 5347, 119734 1820,123706,11187,8246,2782,6135,152056 1830, 202589, 20535, 9049, 3023, 7082, 242278 1840.312710.47613.14480.5346.10965.391114 1850, 515547, 138882, 18593, 8032, 15061, 696115 1860,813669,279122,32903,23593,25492,1174779 1870,942292,419921,45468,37393,33029,1478103 1880, 1164673, 599495, 56559, 51980, 38991, 1911698 1890,1441216,838547,87050,88908,51693.2507414 1900,1850093,1166582,152999,200507,67021,3437202 1910,2331542,1634351,284041,430980,85969,4766883 1920.2284103.2018356.469042.732016.116531.5620048 1930,1867312,2560401,1079129,1265258,158346,6930446 1940,1009924,2690205,1297634,1394711,174441,7454995 1950, 1960101, 2738175, 1550849, 1451277, 191555, 7891957 1960,1698281,2627319,1809578,1424815,221991,7781984 1970.1539233.2602012.1986473.1471701.295443.7894862 1980,1428285,2230936,1891325,1168972,352121,7071639 1990, 1487536, 2300664, 1951598, 1203789, 378977, 7322564 2000.1537195.2465326.2229379.1332650.443728.8008278 2010,1585873,2504700,2230722,1385108,468730,8175133 2015, 1644518, 2636735, 2339150, 1455444, 474558, 8550405

nycHistPop.csv

import matplotlib.pyplot as plt import pandas as pd

pop = pd.read_csv('nycHistPop.csv', skiprows=5)

All population figures are consistent with present-day boundaries.,,,,,, First census after the consolidation of the five boroughs, , , , , Year, Manhattan, Brooklyn, Oueens, Bronx, Staten Island, Total 1698,4937,2017,.,727,7681 1771,21863,3623,.,2847,28423 1790, 33131, 4549, 6159, 1781, 3827, 49447 1800,60515,5740,6642,1755,4563,79215 1810, 96373, 8303, 7444, 2267, 5347, 119734 1820,123706,11187,8246,2782,6135,152056 1830, 202589, 20535, 9049, 3023, 7082, 242278 1840.312710.47613.14480.5346.10965.391114 1850, 515547, 138882, 18593, 8032, 15061, 696115 1860,813669,279122,32903,23593,25492,1174779 1870,942292,419921,45468,37393,33029,1478103 1880, 1164673, 599495, 56559, 51980, 38991, 1911698 1890,1441216,838547,87050,88908,51693.2507414 1900.1850093.1166582.152999.200507.67021.3437202 1910,2331542,1634351,284041,430980,85969,4766883 1920.2284103.2018356.469042.732016.116531.5620045 1930,1867312,2560401,1079129,1265258,158346,6930446 1940,1009924,2690205,1297634,1394711,174441,7454995 1950, 1960101, 2738175, 1550849, 1451277, 191555, 7891957 1960,1698281,2627319,1809578,1424815,221991,7781984 1970.1539233.2602012.1986473.1471701.295443.7894862 1980,1428285,2230936,1891325,1168972,352121,7071639 1990, 1487536, 2300664, 1951598, 1203789, 378977, 7322564 2000.1537195.2465326.2229379.1332650.443728.8008278 2010,1585873,2504700,2230722,1385108,468730,8175133 2015, 1644518, 2636735, 2339150, 1455444, 474558, 8550405

Source: https://en.wikipedia.org/wiki/Demographics of New York City.....

nycHistPop.csv

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import matplotlib.pyplot as plt import pandas as pd

pop = pd.read_csv('nycHistPop.csv', skiprows=5)

pop.plot(x="Year")

plt.show()

Source: https://en.wikipedia.org/wiki/Demographics_of_New_York_City,..., All population figures are consistent with present-day boundaries...,, First census after the consolidation of the five boroughs,...,

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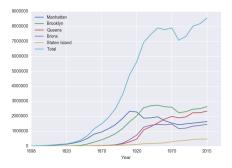
nycHistPop.csv

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import matplotlib.pyplot as plt import pandas as pd

pop = pd.read_csv('nycHistPop.csv',skiprows=5)

pop.plot(x="Year")
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nycHistPop.csv

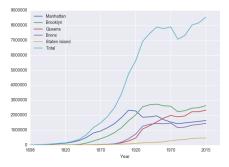
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import matplotlib.pyplot as plt import pandas as pd

pop = pd.read_csv('nycHistPop.csv',skiprows=5)

pop.plot(x="Year")
plt.show()



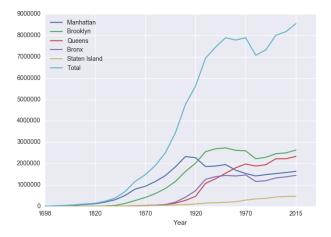
(To get above color scheme, import seaborn.)

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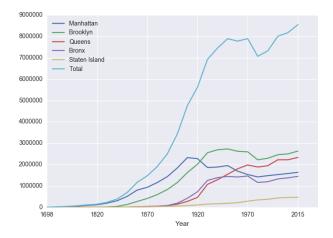
nycHistPop.csv

Nice Graph, But Doesn't Answer the Question



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Nice Graph, But Doesn't Answer the Question



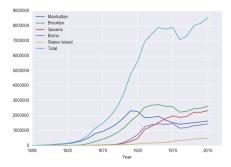
Let's survey what else you can do with pandas....

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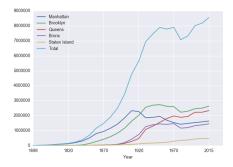
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Series in Pandas



• Series can store a column or row of a DataFrame.

Series in Pandas



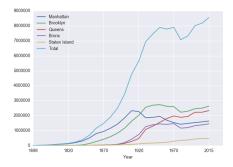
- Series can store a column or row of a DataFrame.
- Example: pop["Manhattan"] is the Series corresponding to the column of Manhattan data.

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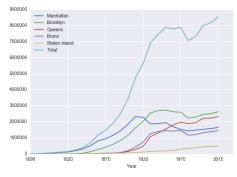
Series in Pandas



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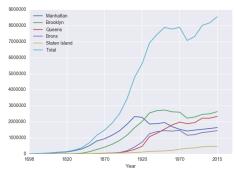
```
• Example:
    print("The largest number living in the Bronx is",\
    pop["Bronx"].max())
```

```
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```



Predict what the following will do:

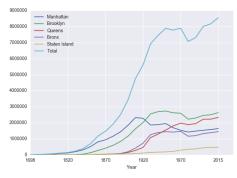
• print("Queens:", pop["Queens"].min())



Predict what the following will do:

- print("Queens:", pop["Queens"].min())
- print("S I:", pop["Staten Island"].mean())

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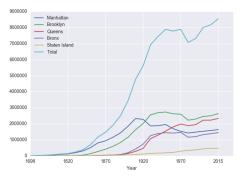
Predict what the following will do:

- print("Queens:", pop["Queens"].min())
- print("S I:", pop["Staten Island"].mean())
- print("S I:", pop["Staten Island"].std())

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Predict what the following will do:

```
• print("Queens:", pop["Queens"].min())
```

• print("S I:", pop["Staten Island"].mean())

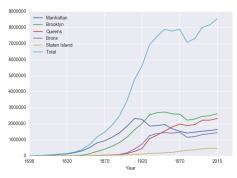
```
• print("S I:", pop["Staten Island"].std())
```

o pop.plot.bar(x="Year")

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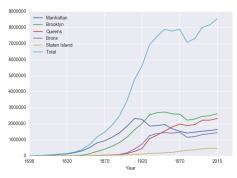
Predict what the following will do:

- print("Queens:", pop["Queens"].min())
- print("S I:", pop["Staten Island"].mean())
- print("S I:", pop["Staten Island"].std())
- o pop.plot.bar(x="Year")
- pop.plot.scatter(x="Brooklyn", y= "Total")

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Predict what the following will do:

- print("Queens:", pop["Queens"].min())
- print("S I:", pop["Staten Island"].mean())
- print("S I:", pop["Staten Island"].std())
- o pop.plot.bar(x="Year")
- pop.plot.scatter(x="Brooklyn", y= "Total")
- pop["Fraction"] = pop["Bronx"]/pop["Total"]

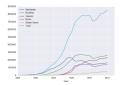
K. St. John (Hunter & AMNH)

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Predict what the following will do:

• print("Queens:", pop["Queens"].min())



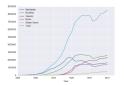
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Predict what the following will do:

● print("Queens:", pop["Queens"].min())

Minimum value in the column with label "Queens".



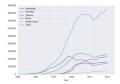
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Predict what the following will do:

- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean())



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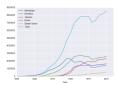


Predict what the following will do:

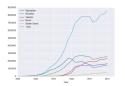
- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".

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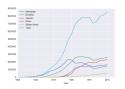
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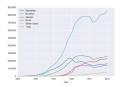
- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())



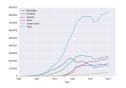
- o print("Queens:", pop["Queens"].min()) Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std()) Standard deviation of values in the column "Staten Island".



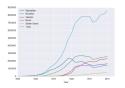
- o print("Queens:", pop["Queens"].min()) Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())
 Standard deviation of values in the column "Staten
 Island".
- op.plot.bar(x="Year")



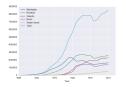
- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())
 Standard deviation of values in the column "Staten
 Island".
- pop.plot.bar(x="Year")
 Bar chart with x-axis "Year".



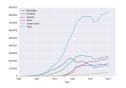
- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())
 Standard deviation of values in the column "Staten
 Island".
- pop.plot.bar(x="Year") Bar chart with x-axis "Year".
- pop.plot.scatter(x="Brooklyn", y= "Total")



- o print("Queens:", pop["Queens"].min()) Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())
 Standard deviation of values in the column "Staten
 Island".
- pop.plot.bar(x="Year") Bar chart with x-axis "Year".
- pop.plot.scatter(x="Brooklyn", y= "Total")
 Scatter plot of Brooklyn versus Total values.



- print("Queens:", pop["Queens"].min())
 Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean())
 Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std()) Standard deviation of values in the column "Staten Island".
- pop.plot.bar(x="Year") Bar chart with x-axis "Year".
- pop.plot.scatter(x="Brooklyn", y= "Total")
 Scatter plot of Brooklyn versus Total values.
- pop["Fraction"] = pop["Bronx"]/pop["Total"]

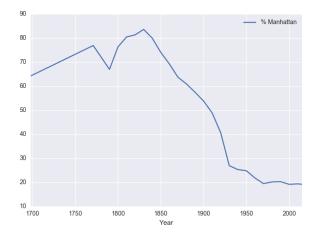


- o print("Queens:", pop["Queens"].min()) Minimum value in the column with label "Queens".
- print("S I:", pop["Staten Island"].mean()) Average of values in the column "Staten Island".
- print("S I :", pop["Staten Island"].std())
 Standard deviation of values in the column "Staten
 Island".
- pop.plot.bar(x="Year") Bar chart with x-axis "Year".
- pop.plot.scatter(x="Brooklyn", y= "Total")
 Scatter plot of Brooklyn versus Total values.
- pop["Fraction"] = pop["Bronx"]/pop["Total"] New column with the fraction of population that lives in the Bronx.

Graph the percentage of the total population that live in Manhattan:

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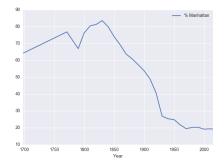
Graph the percentage of the total population that live in Manhattan:



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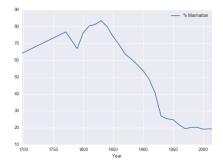
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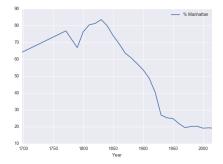
import matplotlib.pyplot as plt import pandas as pd

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import matplotlib.pyplot as plt import pandas as pd import seaborn

K. St. John (Hunter & AMNH)

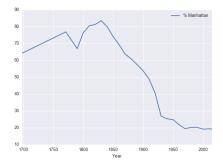
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```
import matplotlib.pyplot as plt
import pandas as pd
import seaborn
pop = pd.read_csv('nycHistPop.csv',skiprows=5)
```

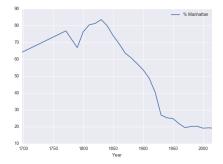
K. St. John (Hunter & AMNH)

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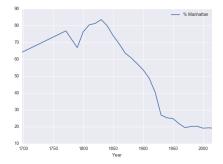
18 July 19



```
import matplotlib.pyplot as plt
import pandas as pd
import seaborn
pop = pd.read_csv('nycHistPop.csv',skiprows=5)
pop["% Manhattan"] = 100*pop["Manhattan"]/pop["Total"]
```

K. St. John (Hunter & AMNH)

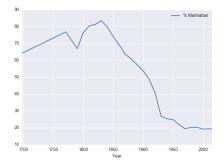
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```
import matplotlib.pyplot as plt
import pandas as pd
import seaborn
pop = pd.read_csv('nycHistPop.csv',skiprows=5)
pop["% Manhattan"] = 100*pop["Manhattan"]/pop["Total"]
pop.plot(x="Year",y="% Manhattan")
```

K. St. John (Hunter & AMNH)

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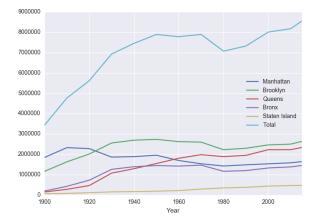
```
import matplotlib.pyplot as plt
import pandas as pd
import seaborn
pop = pd.read_csv('nycHistPop.csv',skiprows=5)
pop["% Manhattan"] = 100*pop["Manhattan"]/pop["Total"]
pop.plot(x="Year",y="% Manhattan")
plt.show()
```

```
K. St. John (Hunter & AMNH)
```

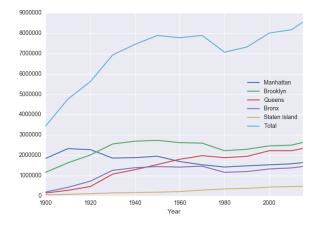
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Population Since 1900



Population Since 1900

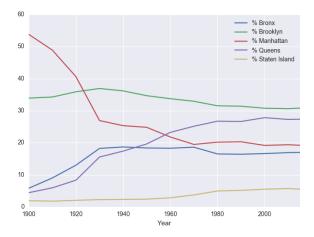


```
pop1900 = pop[pop['Year'] >= 1900]
pop1900.plot(x="Year")
plt.show()
```

K. St. John (Hunter & AMNH)

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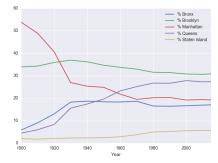
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K. St. John (Hunter & AMNH)

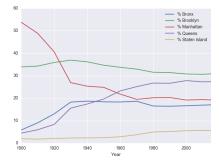
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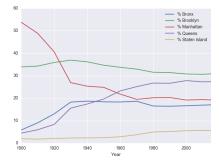
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boros = ["Bronx", "Brooklyn", "Manhattan", "Queens", "Staten Island"]

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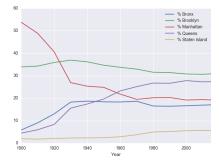


boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension

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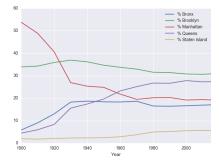


boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension
for boro in boros:

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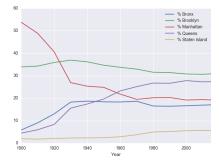


boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension
for boro in boros:
 print('Computing percentage for', boro)

```
pop["% " + boro] = 100*pop2[boro]/pop["Total"]
```

K. St. John (Hunter & AMNH)

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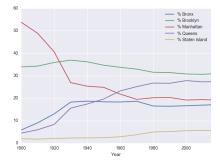


boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension
for boro in boros:
 print('Computing percentage for', boro)
 pop["% " + boro] = 100*pop2[boro]/pop["Total"]
pop = pop.drop(boros, axis = 1)

K. St. John (Hunter & AMNH)

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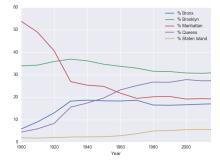


boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension
for boro in boros:
 print('Computing percentage for', boro)

```
pop["% " + boro] = 100*pop2[boro]/pop["Total"]
pop = pop.drop(boros, axis = 1)
```

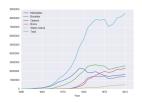
```
pop = pop.drop("Total", axis = 1)
```

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boros = ["Bronx","Brooklyn","Manhattan","Queens","Staten Island"]
percentCol = ["% "+ boro for boro in boros] #List comprehension
for boro in boros:

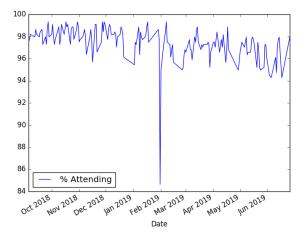
Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

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Manhattan Hunter HS, 2018-2019

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Open Data for All New Yorkers

Where can you find public Wi-Fi in your neighborhood? What kind of tree is in front of your office? Learn about where you live, work, eat, shop and play using NYC Open Data.

Search Open Data for things like 311, Buildings, Crime



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• Freely available source of data.

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- Maintained by the NYC data analytics team.

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- Will use pandas, pyplot & folium libraries to analyze, visualize and map the data.

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Search Open Data for things like 311, Buildings, Crime



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- Freely available source of data.
- Maintained by the NYC data analytics team.
- Will use pandas, pyplot & folium libraries to analyze, visualize and map the data.
- More on downloading NYC OpenData datasets after break.

K. St. John (Hunter & AMNH)

• NYC OpenData has daily population counts for schools.



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- NYC OpenData has daily population counts for schools.
- 2018-2019 data for Manhattan Hunter linked on webpage.



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- NYC OpenData has daily population counts for schools.
- 2018-2019 data for Manhattan Hunter linked on webpage. Download now to work through the exercise.

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- NYC OpenData has daily population counts for schools.
- 2018-2019 data for Manhattan Hunter linked on webpage. Download now to work through the exercise.
- Can download your own from OpenData NYC (view data and filter by "School DBN").



- NYC OpenData has daily population counts for schools.
- 2018-2019 data for Manhattan Hunter linked on webpage. Download now to work through the exercise.
- Can download your own from OpenData NYC (view data and filter by "School DBN").
- Dates need to be converted from 'YYYYMMDD' to a datetime format:

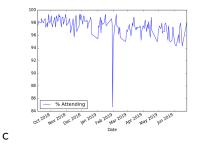


- NYC OpenData has daily population counts for schools.
- 2018-2019 data for Manhattan Hunter linked on webpage. Download now to work through the exercise.
- Can download your own from OpenData NYC (view data and filter by "School DBN").
- Dates need to be converted from 'YYYYMMDD' to a datetime format:

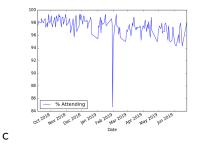
df["Date"] = pd.to_datetime(df["Date"].apply(str))

 Goal: Make a plot of daily attendance (as percentage of enrolled).





df = pd.read_csv('dailyAttendance.csv') #Read file to a dataframe



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K. St. John (Hunter & AMNH)

20 November 2019 27 / 51



df = pd.read_csv('dailyAttendance.csv') #Read file to a dataframe df["Date"] = pd.to_datetime(df["Date"].apply(str)) df["% Attending"] = 100*df["Present"]/df["Enrolled"]

K. St. John (Hunter & AMNH)

20 November 2019 27 / 51



df = pd.read_csv('dailyAttendance.csv') #Read file to a dataframe df["Date"] = pd.to_datetime(df["Date"].apply(str)) df["% Attending"] = 100*df["Present"]/df["Enrolled"] df.plot(x='Date',y="% Attending")

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• Find an interesting data set for a challenge.



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- Find an interesting data set for a challenge.
- Suggested places to look:

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- Find an interesting data set for a challenge.
- Suggested places to look:
 - ► NYC OpenData.

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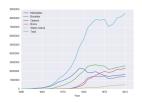
- Find an interesting data set for a challenge.
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 - Weather Underground (for historical weather data).



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- Suggested places to look:
 - ► NYC OpenData.
 - Weather Underground (for historical weather data).
 - Kaggle Open Datasets: data, code, and competitions for data science.

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Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

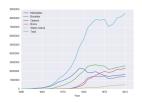
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Break



Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

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 - We will use a small version (1000 lines).



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- A simple, but very powerful, technique is binning data: grouping data into the number of occurrences for each categories.



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- A simple, but very powerful, technique is binning data: grouping data into the number of occurrences for each categories.
- Can often show patterns that individual data points do not.
- We will bin parking tickets by attributes: license plate number, car color, etc.

K. St. John (Hunter & AMNH)

Sum E	Plate	Regis	Plate	Issue	Viola	Vehic	Vehic	Issui	Stree	Stree	Stree	Veh
14471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	Ρ	27390	36290	36350	20
14471524	JCV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Р	36290	27390	13113	20
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	Ρ	36270	11710	27390	20
14471525	JGX1641	NY	PAS	06/24/2019	19	SDN	AUDI	Р	36270	11710	27390	20
14471527	GDM8069	NY	СОМ	07/06/2019	48			Ρ	31190	36310	36330	20
14471529	HXH5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Р	36270	11710	27390	20
4471533	HXM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	20
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	Р	36270	11710	27390	202
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Р	36290	11710	27390	202
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	JCA5331	NY	PAS	07/01/2019	46	SDN	ACURA	Р	36270	11710	27390	202
14471537	JFW5006	99	PAS	06/16/2019	46	SDN	HONDA	Р	36270	11710	27390	202
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Р	36270	11710	27390	202
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	202

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Sem. 1	Pizza 1	Rept. 1	Plate. 1	hour. 1	Viola. 1	Whic. 1	WHE. I	host, 1	5tres. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY.	PAS	06/28/2013	29	50%	BNIN	P	27290	36290	36350	8
14401524	(CV5522	NY.	FAS	06/25/2013	29	50N	10907	9	26290	27290	13113	20
16471125	GAMBRIS 6	NY.	PA6	0616/2018	11	NUM	ENV .		38279	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50%	400	P	36279	11710	27590	20
14401527	62148068	NY.	COM	03/06/2013	48			9	21190	36210	36330	20
16471128	K040242	NY.	PA6	06160011	- 18	NUM	NISA		38279	11710	27190	20
14471533	808360	NY	PAS	06/14/2019	40	SUON	10401	P	36290	11710	27590	20
144015331	GW10540	NY.	FAS	0614/0019	45	930N	HONDA	9	26279	11710	27290	20
16471333.	X01709	NY.	PA6	06/28/2011	48	NUM	novor		38290	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50%	50000	P	31190	40404	40484	
14401536	(CASE)1	NY.	FAS	03/01/0019	45	50N	ACURA	9	26279	11710	27290	20
16471387	(FM1006	10	PA6	0616/2018	- 18	325	HENDA		38279	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACUBA	P	36279	11710	27590	20
14401538	GM0545	NY	FAS	0615/0019	13	9.0N	NISSA	9	26272	11710	27290	20

Sem. 1	Pizza 1	Rept. 1	Plats. 1	hour. 1	Viola. 1	Whic. I	WHIC: I	host, 1	5tres. 1	\$996. I	\$016. I	Veh
1401523	1872561	NY.	PA5	06/28/2013	29	50N	BNIN	P	27299	36290	36350	20
14401524	(CV5522	NY.	FAS	06/25/2013	29	SON	novor	9	26290	27290	13113	20
16471125	GAMBRIDE	NY.	PA6	0616/2018		san	EV/V		38279	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50N	400	P	36279	11710	27590	20
14401527	62110068	NY.	COM	03/06/2019	-48			9	21190	36210	36330	20
16071128.	x0+0242	NY.	PA6	06162011		NUN	NISA		38279	11710	27190	20
14471533	8083470	NY	PAS	06/14/2019	40	SUON	10401	P	36290	11710	27590	20
14401533	QM-8540	NY.	FAS	0614-2019	-46	930N	HONDA.	9	26272	11710	27290	20
16071133.	XX81749	NY .	PA6	06/28/2011	- 48	NUN	novor		38210	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50N	50000	P	31190	40404	40484	
14401536	(CASE)1	NY.	FAS	03/21/2013	-46	50N	RCURA	9	26272	11710	27290	20
16071107	(PM1006		PA6	0616/2011		101	HENEN.		38279	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACUBA	P	36279	11710	27590	20
14401538	GM0545	NY	ENG	0615/0019	13	930N	NISSA	9	26272	11710	27290	20

• Instead of zipcode, classified by the issuing police precinct.

200

Sem. 1	Pizza 1	Rept. 1	Plate. 1	hour. 1	Viola. 1	Whic. 1	WHIC: I	host, 1	\$tree. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY	PAS	06/28/2013	29	50%	BNIN		27390	36290	36350	8
14401524	(CV5522	NY	FAS	06/25/2013	29	50N	novor	9	26290	27290	13113	20
16471125	GAMBRIDE	NY	PA6	0616/2018		NUM	EV/V		38272	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50%	400	P	36279	11710	27590	25
14401527	62148068	NY	COM	03/06/2013	48			9	21190	36210	36330	20
16471128	K040242	NY	PA6	06160011	- 44	NUM	NISA		38272	11710	27190	20
14471533	8083470	NY	PAS	06/14/2019	40	SUON	10401	P	36290	11710	27590	25
144015331	GW10540	NY	FAS	0614/0019	45	930N	HONDA	9	26279	11710	27290	20
16471333.	X01709	NY	PA6	06/28/2011	48	NUM	novor		38290	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50%	50000	P	31190	40404	40484	
14401536	(CASED)	NY	FAS	03/01/0019	45	50N	RCURA	9	26279	11710	27290	20
16471387	(FM1006	11	PA6	0616/2018		325	HENDA		38272	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACUBA	P	36279	11710	27590	25
14401538	GM0545	NY	FAS	0615/0019	13	9.0N	NISSA	9	26272	11710	27290	2

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Sem. 1	Pizza 1	Rept. 1	Plate. 1	hour. 1	Viola. 1	Whic. 1	WHIC: I	host, 1	\$tree. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY	PAS	06/28/2013	29	50%	BNIN		27390	36290	36350	8
14401524	(CV5522	NY	FAS	06/25/2013	29	50N	novor	9	26290	27290	13113	20
16471125	GAMBRIDE	NY	PA6	0616/2018		NUM	EV/V		38272	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50%	400	P	36279	11710	27590	25
14401527	62148068	NY	COM	03/06/2013	48			9	21190	36210	36330	20
16471128	K040242	NY	PA6	06160011		NUM	NISA		38272	11710	27190	20
14471533	808360	NY	PAS	06/14/2019	40	SUON	10401	P	36290	11710	27590	25
144015331	GW10540	NY	FAS	0614/0019	45	930N	HONDA	9	26279	11710	27290	20
16471333.	X01709	NY	PA6	06/28/2011	48	NUM	novor		38290	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50%	50000	P	31190	40404	40484	
14401536	(CASED)	NY	FAS	03/01/0019	45	50N	RCURA	9	26279	11710	27290	20
16471387	(FM1006	11	PA6	0616/2018	- 18	325	HENDA		38272	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACUBA	P	36279	11710	27590	25
14401538	GM0545	NY	FAS	0615/0019	13	9.0N	NISSA	9	26272	11710	27290	2

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Sem. 1	Pizza 1	Rept. 1	Plats. 1	hour. 1	Viola. 1	Whic. 1	Vehic. 1	host, 1	\$tree. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY	PA5	06/28/2013	29	50%	BVIN		27390	36290	36350	8
14401524	(CV5522	NY	FAS	06/25/2013	29	50N	TOYOF.	9	26290	27290	13113	20
16071128	GANGED 5	NY.	PA6	0616/2011	11	NUM	ENV .		38279	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50%	600	P	36279	11710	27590	25
14401527	62110068	NY	COM	03/06/2019	-48			9	21190	36210	36330	20
16071128.	x0+0242	NY.	PA6	06162011		NUM	NISA		38279	11710	27190	20
14471533	8083470	NY	PAS	06/14/2019	40	SUON	10401	P	36290	11710	27590	25
14401533	QM-8540	NY	FAS	0614-2019	-46	930N	HONDA.	9	26272	11710	27290	20
16071133.	XX81749	NY.	PA6	06/28/2011	- 48	NUM	novor		38210	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50%	50000	P	31190	40404	40484	
14401536	(CASE)1	NY	FAS	03/21/2013	-46	50N	RCURA	9	26272	11710	27290	20
16471387	743006	99	PA6	0616-0011		325	HENDA.		18272	11710	27190	20
14071537	8040634	NY	PA5	0616/2013	-45		ACURA		36279	11710	27590	3
1401536	GM0545	NY	FAS	0615-0018	13	9.0N	NSSA	9	26172	11210	27290	2

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Sem. I	Pizza 1	Rept. 1	Plats. 1	hour. 1	Viola. 1	Whic. 1	Vehic. 1	host, 1	\$tree. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY	PA5	06/28/2013	29	50N	BVIN		27390	36290	36350	8
14401524	(CV5522	NY	FAS	06/25/2013	29	SON	TOYOF.	9	26290	27290	13113	20
16471125	GAMBRIDE	NY	PA6	0616/2018		sam	ENV .		38272	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50N	600	P	36279	11710	27590	25
14401527	62110068	NY	COM	03/06/2019	-48			9	21190	36210	36330	20
16071128.	x0+0242	NY.	PA6	06162011		NUM	NISA		38279	11710	27190	20
14471533	808360	NY	PAS	06/14/2019	40	9.0N	10401	P	36290	11710	27590	25
14401533	QM-8540	NY	FAS	0614-2019	-46	930N	HONDA.	9	26272	11710	27290	20
16071333.	XX81749	NY.	PA6	06/28/2011	- 48	NUM	novor		38210	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50N	50000	P	31190	40404	40484	
14401536	(CASE)1	NY	FAS	03/25/2019	-46	50N	RCURA	9	26272	11710	27290	20
16071107	(PM1006	99	PA6	0616/2011		121	HENDA.		38279	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACURA	P	36279	11710	27590	25
\$40153L	GM0545	NY	FAS	0615-0018	13	9.0N	NSSA	9	26172	11210	27290	2

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- Sample line of CSV file:

1335632335,L040HZ,FL,PAS,06/09/2015,46,SUBN,NISSA,X,35430,14510,15710,0,0020,20,74,921167,E074,0000,1213 83 ST,,0,408,C,,BBBBBBB,ALL,ALL,RED,0,0,-,0,,,,,

K. St. John (Hunter & AMNH)

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Sem. I	Pizza 1	Rept. 1	Plats. 1	hour. 1	Viola. 1	Whic. 1	Vehic. 1	host, 1	\$tree. 1	\$996. I	\$016. I	Vet
1401523	1872561	NY	PA5	06/28/2013	29	50N	BVIN		27390	36290	36350	8
14401524	(CV5522	NY	FAS	06/25/2013	29	SON	TOYOF.	9	26290	27290	13113	20
16471125	GAMBRIDE	NY	PA6	0616/2018		sum	ENV .		38272	11710	27190	20
14071525	1001643	NY	PAS	06/24/2013	19	50N	600	P	36279	11710	27590	25
14401527	62110068	NY	COM	03/06/2019	-48			9	21190	36210	36330	20
16071128.	x0+0242	NY.	PA6	06162011		NUM	NISA		38279	11710	27190	20
14471533	8083470	NY	PAS	06/14/2019	40	9.0N	10401	P	36290	11710	27590	25
14401533	QM-8540	NY	FAS	0614-2019	-46	930N	HONDA.	9	26272	11710	27290	20
16071333.	XX81749	NY.	PA6	06/28/2011	- 48	NUM	novor		38210	11710	27190	20
14471533	0012184	NE	PAS	07/96/2019	-48	50N	50000	P	31190	40404	40484	
14401536	(CASE)1	NY	FAS	03/21/2013	-46	50N	RCURA	9	26272	11710	27290	20
16071107	(PM1006	99	PA6	0616/2011		121	HENDA.		38279	11710	27190	20
14471537	H042634	NY	PAS	0616/2013	-45		ACURA	P	36279	11710	27590	25
\$40153L	GM0545	NY	FAS	0615-0018	13	9.0N	NSSA	9	26172	11210	27290	2

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- Each entry begins with a unique identifier to look up the ticket.
- Sample line of CSV file:

 $1335632335, \texttt{L040HZ}, \texttt{FL}, \texttt{PAS}, \texttt{06}/\texttt{09}/\texttt{2015}, \texttt{46}, \texttt{SUBN}, \texttt{NISSA}, \texttt{X}, \texttt{35430}, \texttt{14510}, \texttt{15710}, \texttt{0}, \texttt{0020}, \texttt{20}, \texttt{74}, \texttt{921167}, \texttt{E074}, \texttt{0000}, \texttt{1213}, \texttt{1213}, \texttt{1214}, \texttt{$

83 ST,,0,408,C,,BBBBBBBB,ALL,ALL,RED,0,0,-,0,,,,,

 Issued on June 9, 2015 to a passenger car with Florida plates, L040HZ. The red Nissan SUV received it on W 83rd Street.

K. St. John (Hunter & AMNH)

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	IssuL.	Stree	Stree	Stree	Veh
4471523	(ET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
4471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	203
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	203
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	203
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	203
4471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
4471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	203
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	203

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	Issul	Stree	Stree	Stree	Veh
14471523	(ET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	203
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
14471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	203
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Р	36270	11710	27390	203
14471533	H0043470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	203
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	203
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Р	36270	11710	27390	203
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	203

In groups, brainstorm about how to answer:

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	IssuL.	Stree	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Р	36290	27390	13113	203
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	50
4471525	(0x1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	203
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	H0M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	203
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	203
4471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	Р	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
4471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	203
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	203

In groups, brainstorm about how to answer:

• Which car got the most tickets?

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	IssuL.	Stree	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Р	36290	27390	13113	20
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	20
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	TOYOT	P	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
4471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	20
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

In groups, brainstorm about how to answer:

- Which car got the most tickets?
- What color of car is most likely to get a ticket?

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	IssuL.	Stree	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Р	36290	27390	13113	20
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	20
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	TOYOT	P	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
4471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	20
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

In groups, brainstorm about how to answer:

- Which car got the most tickets?
- What color of car is most likely to get a ticket?
- What type of license gets the most tickets?

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Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	Issui	Stree	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	Р	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	P	36290	27390	13113	20
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	20
4471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Ρ	36270	11710	27390	20
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	20
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	P	36270	11710	27390	20

In groups, brainstorm about how to answer:

- Which car got the most tickets?
- What color of car is most likely to get a ticket?
- What type of license gets the most tickets?
- Are all states equally represented in license plates that get tickets?

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Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	Issui	Stree	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	Р	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	P	36290	27390	13113	20
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	20
4471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Ρ	36270	11710	27390	20
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	20
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	P	36270	11710	27390	20

In groups, brainstorm about how to answer:

- Which car got the most tickets?
- What color of car is most likely to get a ticket?
- What type of license gets the most tickets?
- Are all states equally represented in license plates that get tickets?
- Which location yields the most tickets?

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Sum	Plate I	Regis I	Plate I	Issue I	Viola	Vehic. 1	Vehic i	Issul	Stree I	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	Ρ	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			P	31190	36310	36330	20
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	Р	36270	11710	27390	20
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	20
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	Р	36270	11710	27390	20
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	20
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

Sum	Plate	Regis I	Plate	Issue I	Viola	Vehic. 1	Vehic. 1	Issul	Stree 1	Stree	Stree	Veh
14471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
4471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	Ρ	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			P	31190	36310	36330	20
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	Ρ	36270	11710	27390	20
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	Ρ	36270	11710	27390	20
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	20
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	P	36270	11710	27390	20

How can tell which car got the most tickets?

Sum	Plate	Regis 1	Plate	Issue 1	Viola	Vehic.	Vehic.	Issul	Stree 1	Stree :	Stree I	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
4471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	Р	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	сом	07/06/2019	48			Р	31190	36310	36330	20
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	Ρ	36270	11710	27390	20
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	TOYOT	Р	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
4471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	Р	36270	11710	27390	20
4471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	Ρ	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Р	36270	11710	27390	20
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	P	36270	11710	27390	20

How can tell which car got the most tickets?

• Need to a unique way to identify different cars.

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Sum	Plate 1	Regis i	Plate	Issue	Viola	Vehic.	Vehic. 1	Issul	Stree 1	Stree	Stree	Vet
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
4471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	сом	07/06/2019	48			P	31190	36310	36330	20
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	20
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	20
4471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
4471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	20
14471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
4471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	20
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

How can tell which car got the most tickets?

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 Luckily, cars almost always have license plates
 – unique by state.

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Sum	Plate	Regis i	Plate	Issue I	Viola	Vehic.	Vehic!	Issul	Stree 1	Stree	Stree I	Veh
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	TOYOT	Ρ	36290	27390	13113	203
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	Р	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GDM8069	NY	сом	07/06/2019	48			P	31190	36310	36330	203
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	203
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	203
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	203
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

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 (For this simple exercise, assume each license plate ID is unique— not unreasonable since every state has a different schema for assigning numbers, but to be more accurate should keep track of license plate number and issuing state.)

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Sum	Plate	Regis i	Plate	Issue I	Viola	Vehic.	Vehic!	Issul	Stree 1	Stree	Stree	Veh
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	TOYOT	Ρ	36290	27390	13113	203
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	Р	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GDM8069	NY	сом	07/06/2019	48			P	31190	36310	36330	203
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	203
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	203
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	203
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

How can tell which car got the most tickets?

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 (For this simple exercise, assume each license plate ID is unique— not unreasonable since every state has a different schema for assigning numbers, but to be more accurate should keep track of license plate number and issuing state.)
- Want to "bin" tickets by license plates ("Plate ID"),

K. St. John (Hunter & AMNH)

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Sum	Plate	Regis i	Plate	Issue	Viola	Vehic.	Vehic	Issul	Stree 1	Stree	Stree I	Veh
4471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	TOYOT	Ρ	36290	27390	13113	203
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	Р	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GDM8069	NY	сом	07/06/2019	48			P	31190	36310	36330	203
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	20
4471533	HKM3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	203
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	203
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5331	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
14471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	Ρ	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	203
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	20

How can tell which car got the most tickets?

- Need to a unique way to identify different cars.
 Luckily, cars almost always have license plates— unique by state.
 (For this simple exercise, assume each license plate ID is unique— not unreasonable since every state has a different schema for assigning numbers, but to be more accurate should keep track of license plate number and issuing state.)
- Want to "bin" tickets by license plates ("Plate ID"), and then count the size of bins.

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Sac

• Sample program:

• Sample program: import pandas as pd

```
    Sample program:
import pandas as pd
tickets = pd.read_csv('tickets.csv')
```

 Sample program: import pandas as pd tickets = pd.read_csv('tickets.csv') #Print out the data frame: print(tickets)

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 Sample program: import pandas as pd tickets = pd.read_csv('tickets.csv') #Print out the data frame: print(tickets) #Print out licence plates: print(tickets["Plate ID"])

 Sample program: import pandas as pd tickets = pd.read_csv('tickets.csv') #Print out the data frame: print(tickets) #Print out licence plates: print(tickets["Plate ID"]) #Print out plates & number of tickets each got: print(tickets["Plate ID"].value_counts())

● Sample program: import pandas as pd tickets = pd.read_csv('tickets.csv') #Print out the data frame: print(tickets) #Print out licence plates: print(tickets["Plate ID"]) #Print out plates & number of tickets each got: print(tickets["Plate ID"].value_counts()) #Print 10 worst & number of tickets:

print(tickets["Plate ID"].value_counts()[:10])

```
● Sample program:
import pandas as pd
tickets = pd.read_csv('tickets.csv')
#Print out the data frame:
print(tickets)
#Print out licence plates:
print(tickets["Plate ID"])
#Print out plates & number of tickets each got:
print(tickets["Plate ID"].value_counts())
#Print 10 worst & number of tickets:
```

print(tickets["Plate ID"].value_counts()[:10])

• For the sample data set, there were few cars that got more than a ticket a day...

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	IssuL.	Stree	Stree	Stree	Vet
4471523	(ET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
4471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BWW	P	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
4471527	GD/48069	NY	COM	07/06/2019	48			Р	31190	36310	36330	203
4471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Р	36270	11710	27390	203
4471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	P	36290	11710	27390	203
4471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
4471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	203
4471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	P	31190	40404	40404	
4471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	P	36270	11710	27390	203
4471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	203
4471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	203

Sum	Plate	Regis	Plate	Issue	Viola	Vehic	Vehic	Issul	Stree	Stree	Stree	Veh
14471523	(ET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
14471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Ρ	36290	27390	13113	20
14471525	GMK6954	NY	PAS	06/16/2019	19	SUBN	BMW	Р	36270	11710	27390	20
14471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	203
14471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	203
14471529	H00H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	P	36270	11710	27390	203
14471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	Р	36290	11710	27390	203
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	203
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	Ρ	36290	11710	27390	203
14471533	GDH2184	ME	PAS	07/06/2019	48	SDN	DODGE	P	31190	40404	40404	
14471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	Р	36270	11710	27390	203
14471537	[PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	203
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	Ρ	36270	11710	27390	203
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	Р	36270	11710	27390	203

In groups, write programs for:

Sum	Plate	Regis	Plate	Issue	Viola	Vehic.	Vehic	Issul	Stree	Stree	Stree	Vet
14471523	JET2661	NY	PAS	06/28/2019	21	SDN	BMW	P	27390	36290	36350	20
4471524	(CV6523	NY	PAS	06/28/2019	20	SDN	тоуот	Р	36290	27390	13113	20
4471525	GMR6954	NY	PAS	06/16/2019	19	SUBN	BWW	Р	36270	11710	27390	20
4471525	(GX1641	NY	PAS	06/24/2019	19	SDN	AUDI	P	36270	11710	27390	20
4471527	GDM8069	NY	COM	07/06/2019	48			Ρ	31190	36310	36330	20
14471529	H0H5242	NY	PAS	06/14/2019	46	SUBN	NISSA	Р	36270	11710	27390	20
14471533	H00M3470	NY	PAS	06/14/2019	40	SUBN	тоуот	р	36290	11710	27390	20
14471533	GWH9640	NY	PAS	06/14/2019	46	SUBN	HONDA	P	36270	11710	27390	20
14471533	HKB1769	NY	PAS	06/28/2019	40	SUBN	тоуот	P	36290	11710	27390	20
14471533	GDH2184	ME	PAS	07/05/2019	48	SDN	DODGE	Р	31190	40404	40404	
14471536	(CA5301	NY	PAS	07/01/2019	46	SDN	ACURA	Р	36270	11710	27390	20
14471537	(PW5006	99	PAS	06/16/2019	46	SDN	HONDA	P	36270	11710	27390	20
14471537	HGR2634	NY	PAS	06/16/2019	46		ACURA	P	36270	11710	27390	20
14471538	GYM7645	NY	PAS	06/15/2019	19	SUBN	NISSA	P	36270	11710	27390	20

In groups, write programs for:

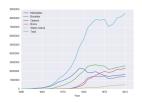
- Which car got the most tickets?
- What color of car is most likely to get a ticket?
- What type of license gets the most tickets?
- Are all states equally represented in license plates that get tickets?
- Which location yields the most tickets?

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Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

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Variations on the Theme: OpenData Film Permits

NYC OpenData

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Film Permits

Permits are generally required when asserting the exclusive use of city property, like a sidewalk, a 🔤 street, or a park. See http://www1.nvc.gov/site/mome/permits/when-permit-required.page

EventID :	EventType :	StartDateTi	EndDateTime	EnteredOn ↓ :	EventAg	ParkingHeld :	Borou
455063	Shooting Permit	12/06/2018 07:00	12/06/2018 09:00	12/05/2018 12:36	Mayor's Offic	STARR AVENUE b	Queens
454967	Shooting Permit	12/06/2018 07:00	12/06/2018 05:00	12/04/2018 09:11	Mayor's Offic	EAGLE STREET be	Brooklyn
454941	Shooting Permit	12/06/2018 07:00	12/06/2018 07:00	12/04/2018 05:44	Mayor's Offic	SOUTH OXFORD	Brooklyn
454920	Shooting Permit	12/06/2018 10:00	12/06/2018 11:59	12/04/2018 03:28	Mayor's Offic	13 AVENUE betw	Queens
454914	Shooting Permit	12/06/2018 08:00	12/06/2018 11:00	12/04/2018 03:05	Mayor's Offic	ELDERT STREET b	Brooklyn
454909	Shooting Permit	12/05/2018 08:00	12/05/2018 06:00	12/04/2018 02:45	Mayor's Offic	ELDERT STREET b	Brooklyn
454905	Shooting Permit	12/06/2018 07:00	12/06/2018 10:00	12/04/2018 02:17	Mayor's Offic	35 STREET betwe	Queens

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Example: OpenData Film Permits

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Film Permits

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Permits are generally required when asserting the exclusive use of city property, like a sidewalk, a 🔤 street, or a park. See http://www1.nyc.gov/site/mome/permits/when-permit-required.page									More Views Filter Visualize Export Discuss Embed About				
EventID :	EventType i	StartDateTi	EndDateTime	EnteredOn 4	EventAg	ParkingHeld i	Borou	Com i	Police i	Categ i	SubC i	Count i	ZipCo i
455063	Shooting Permit	12/06/2018 07:00	12/06/2018 09:00	12/05/2018 12:36	Mayor's Offic	STARR AVENUE b	Queens	2	108	Television	Episodic s	United Sta	11101
454967	Shooting Permit	12/06/2018 07:00	12/06/2018 05:00	12/04/2018 09:11	Mayor's Offic	EAGLE STREET be	Brooklyn	1	94	Television	Episodic s	United Sta	11222
454941	Shooting Permit	12/06/2018 07:00	12/05/2018 07:00	12/04/2018 05:44	Mayor's Offic	SOUTH OXFORD	Brooklyn	2, 6	76, 88	Still Photo	Not Applic	United Sta	11217, 11
454920	Shooting Permit	12/06/2018 10:00	12/05/2018 11:59	12/04/2018 03:28	Mayor's Offic	13 AVENUE betw	Queens	1, 3, 7	109, 7, 90	Film	Feature	United Sta	10002, 11
454914	Shooting Permit	12/06/2018 08:00	12/05/2018 11:00	12/04/2018 03:05	Mayor's Offic	ELDERT STREET b	Brooklyn	4, 5	104, 75, 83	Television	Episodic s	United Sta	11207, 11
454909	Shooting Permit	12/05/2018 08:00	12/05/2018 06:00	12/04/2018 02:45	Mayor's Offic	ELDERT STREET b	Brooklyn	4	83	Television	Episodic s	United Sta	11237
454905	Shooting Permit	12/06/2018 07:00	12/05/2018 10:00	12/04/2018 02:17	Mayor's Offic	35 STREET betwe	Queens	1	114	Television	Cable-epis	United Sta	11101, 11

• What's the most popular street for filming?

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Example: OpenData Film Permits

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Film Pern	nits							2	f 9	8 8 8	Q. Find in	h this Datase	et i i i i i i i i i i i i i i i i i i i
		d when asserting th w1.nyc.gov/site/mo				3		м	ore Views	ilter Visual	ize Export	Discuss Er	mbed About
EventID :	EventType i	StartDateTi	EndDateTime	EnteredOn \downarrow i	EventAg	ParkingHeld i	Borou i	Com i	Police i	Categ i	SubC i	Count i	ZipCo i
455063	Shooting Permit	12/06/2018 07:00	12/06/2018 09:00	12/05/2018 12:36	Mayor's Offic	STARR AVENUE b	Queens	2	108	Television	Episodic s	United Sta	11101
454967	Shooting Permit	12/06/2018 07:00	12/06/2018 05:00	12/04/2018 09:11	Mayor's Offic	EAGLE STREET be	Brooklyn	1	94	Television	Episodic s	United Sta	11222
454941	Shooting Permit	12/06/2018 07:00	12/05/2018 07:00	12/04/2018 05:44	Mayor's Offic	SOUTH OXFORD	Brooklyn	2, 6	76, 88	Still Photo	Not Applic	United Sta	11217, 11
454920	Shooting Permit	12/06/2018 10:00	12/05/2018 11:59	12/04/2018 03:28	Mayor's Offic	13 AVENUE betw	Queens	1, 3, 7	109, 7, 90	Film	Feature	United Sta	10002, 11
454914	Shooting Permit	12/06/2018 08:00	12/05/2018 11:00	12/04/2018 03:05	Mayor's Offic	ELDERT STREET b	Brooklyn	4, 5	104, 75, 83	Television	Episodic s	United Sta	11207, 11
454909	Shooting Permit	12/05/2018 08:00	12/05/2018 06:00	12/04/2018 02:45	Mayor's Offic	ELDERT STREET b	Brooklyn	4	83	Television	Episodic s	United Sta	11237
454905	Shooting Permit	12/06/2018 07:00	12/05/2018 10:00	12/04/2018 02:17	Mayor's Offic	35 STREET betwe	Queens	1	114	Television	Cable-epis	United Sta	11101, 11

- What's the most popular street for filming?
- What's the most popular borough?

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K. St. John (Hunter & AMNH)

Example: OpenData Film Permits

NYC OpenData

Home Data About - Learn - Alerts Contact Us Blog Q Sign In

Film Perr	nits							2	f 9	885	Q. Find in	h this Datase	st.
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455063	Shooting Permit	12/06/2018 07:00	12/06/2018 09:00	12/05/2018 12:36	Mayor's Offic	STARR AVENUE b	Queens	2	108	Television	Episodic s	United Sta	11101
454967	Shooting Permit	12/06/2018 07:00	12/06/2018 05:00	12/04/2018 09:11	Mayor's Offic	EAGLE STREET be	Brooklyn	1	94	Television	Episodic s	United Sta	11222
454941	Shooting Permit	12/06/2018 07:00	12/06/2018 07:00	12/04/2018 05:44	Mayor's Offic	SOUTH OXFORD	Brooklyn	2, 6	76, 88	Still Photo	Not Applic	United Sta	11217, 11
454920	Shooting Permit	12/06/2018 10:00	12/05/2018 11:59	12/04/2018 03:28	Mayor's Offic	13 AVENUE betw	Queens	1, 3, 7	109, 7, 90	Film	Feature	United Sta	10002, 11
454914	Shooting Permit	12/06/2018 08:00	12/05/2018 11:00	12/04/2018 03:05	Mayor's Offic	ELDERT STREET b	Brooklyn	4, 5	104, 75, 83	Television	Episodic s	United Sta	11207, 11
454909	Shooting Permit	12/05/2018 08:00	12/05/2018 06:00	12/04/2018 02:45	Mayor's Offic	ELDERT STREET b	Brooklyn	4	83	Television	Episodic s	United Sta	11237
454905	Shooting Permit	12/06/2018 07:00	12/05/2018 10:00	12/04/2018 02:17	Mayor's Offic	35 STREET betwe	Queens	1	114	Television	Cable-epis	United Sta	11101, 11

- What's the most popular street for filming?
- What's the most popular borough?
- How many TV episodes were filmed?

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454009	Shooting Fermit	12/05/2018 08:00	12/05/2018 05:00	12/04/2018 02:45	Meyor's Offic	ELDERT STREET 6	Drooklyn	4	13	Television	Ephodic s.,	United Sta	11237
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• Download the data as a CSV file and store on your computer.

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454909	Shooting Fermit	12/05/2018 08:00	12/05/2018 05:00	12/04/2018 02:45	Meyor's Offic	ELOERT STREET 6	Drooklyn	4	83	Television	Ephodic s.,	United Sta	11237
454905	Shooting Fermit	12/06/2018 07:05	13/06/2018 10:00	12/04/2018 02:17	Mayors Offic	25 STREET DODAR.	Queens		114	Television	Cable-spis	United Sta	11101, 11

• Download the data as a CSV file and store on your computer.

• Python program:

```
#CSci 127 Teaching Staff
#March 2019
#OpenData Film Permits
#Import pandas for reading and analyzing CSV data:
import pandas as pd
csvFile = "filmPermits.csv" #Name of the CSV file
tickets = pd.read_csv(csvFile)#Read in the file to a dataframe
```

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454009	Shooting Fermit	12/05/2018 08:00	12/05/2018 05:00	12/04/2018/02:45	Mayor's Offic.	ELOCRT STREET 6	Drooklyn	4	13	Television	Ephodic s.,	United Sta	11237
454905	Shooting Fermit	12/06/2018 07:05	13/06/2018 10:00	12/04/2018 02:17	Mayors Offic	25 STREET DODAR.	Queens		114	Television	Cable-spis	United Sta	11101, 11

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```

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csvFile = "filmPermits.csv" #Name of the CSV file
tickets = pd.read_csv(csvFile)#Read in the file to a dataframe
print(tickets) #Print out the dataframe
print(tickets["ParkingHeld"]) #Print out streets (multiple times)
```

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print(tickets["ParkingHeld"]) #Print out streets (multiple times)
print(tickets["ParkingHeld"].value_counts()) #Print out streets & number of times used
```

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454941	Shooting Permit	12/06/2018 07:00	12/06/2018 17:00	12/94/2018 05:44	Mayor's Offic.	SOUTH ORPORD	Draoklyn	2.6	75,88	Sil Photo	Not Applic	United Sta	11217, 11
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tickets = pd.read_csv(csvFile)#Read in the file to a dataframe
print(tickets) #Print out the dataframe
print(tickets["ParkingHeld"]) #Print out streets (multiple times)
print(tickets["ParkingHeld"].value_counts()) #Print out streets & number of times used
print(tickets["ParkingHeld"].value_counts()[:10]) #Print 10 most popular
```

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In Groups: OpenData Film Permits

NYC OpenData

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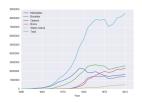
Can approach the other questions in the same way:

- What's the most popular street for filming?
- What's the most popular borough?
- How many TV episodes were filmed?

K. St. John (Hunter & AMNH)

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Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

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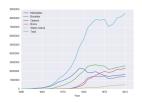
Design a Challenge



With your group, brainstorm about a design challenge that:

- An interesting publicly available data set,
- To analyze with the pandas commands we've discussed.

Outline



- Recap
- Design Challenge: NYC Population
- Variations on the Theme
- Design a Challenge
- Break
- Design Challenge: Parking Tickets
- Variations on the Theme
- Design a Challenge
- Wrap Up

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 Introduced pandas for analyzing structured data.



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- Introduced pandas for analyzing structured data.
 - Plotting, simple stats functions, and slicing.



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- Introduced pandas for analyzing structured data.
 - Plotting, simple stats functions, and slicing.
 - Didn't cover: accessing rows, joining/merging tables, applying functions, ...



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- Introduced pandas for analyzing structured data.
 - Plotting, simple stats functions, and slicing.
 - Didn't cover: accessing rows, joining/merging tables, applying functions, ...
- Used publicly available data:

- Introduced pandas for analyzing structured data.
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See you in three weeks!