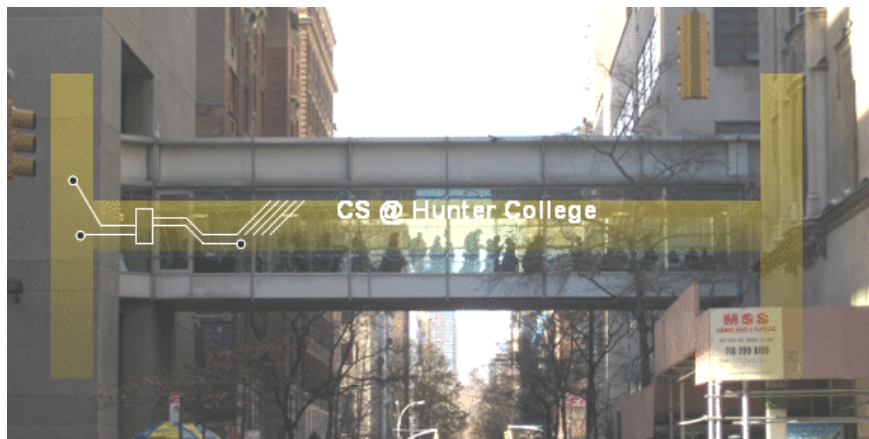


# CSci 127: Introduction to Computer Science



[hunter.cuny.edu/csci](http://hunter.cuny.edu/csci)

# Announcements



- Each lecture includes a survey of computing research and tech in NYC.

*Today: Prof. Raffi Khatchadourian (software engineering)*

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From lecture slips & recitation sections.

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*Yes, we will, since 1) it's fundamental, and 2) the same ideas are used for accessing formatted data (today's topic).*

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*Yes, we will, since 1) it's fundamental, and 2) the same ideas are used for accessing formatted data (today's topic).*
- Could you spend more time on circuits/logical expressions/truth tables/decisions?  
*We will do a bit today, but much more in the following weeks.*

# Today's Topics



- Recap: Logical Expressions & Circuits
- Accessing Formatted Data
- Preview: Functions
- Final Exam Overview

# Recap: Logical Operators

## and

in1		in2	<i>returns:</i>
False	and	False	False
False	and	True	False
True	and	False	False
True	and	True	True

# Recap: Logical Operators

## and

in1		in2	<i>returns:</i>
False	and	False	False
False	and	True	False
True	and	False	False
True	and	True	True

## or

in1		in2	<i>returns:</i>
False	or	False	False
False	or	True	True
True	or	False	True
True	or	True	True

# Recap: Logical Operators

## and

in1		in2	returns:
False	and	False	False
False	and	True	False
True	and	False	False
True	and	True	True

## or

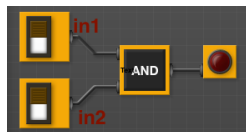
in1		in2	returns:
False	or	False	False
False	or	True	True
True	or	False	True
True	or	True	True

## not

	in1	returns:
not	False	True
not	True	False

# Logical Operators & Circuits

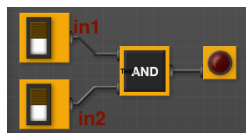
- Each logical operator (and, or, & not) can be used to join together expressions.





# Logical Operators & Circuits

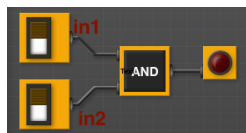
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Example: in1 and in2

# Logical Operators & Circuits

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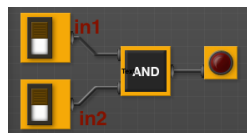


Example: in1 and in2

- Each logical operator (and, or, & not) has a corresponding logical circuit that can be used to join together inputs.

# Logical Operators & Circuits

- Each logical operator (and, or, & not) can be used to join together expressions.



Example: in1 and in2

- Each logical operator (and, or, & not) has a corresponding logical circuit that can be used to join together inputs.

Example: see image.

# Examples:

*Examples from last lecture:*

```
origin = "Indian Ocean"
winds = 100
if (winds > 74):
    print("Major storm, called a ", end="")
    if origin == "Indian Ocean" or origin == "South Pacific":
        print("cyclone.")
    elif origin == "North Pacific":
        print("typhoon.")
    else:
        print("hurricane.")

visibility = 0.2
winds = 40
conditions = "blowing snow"
if (winds > 35) and (visibility < 0.25) and \
    (conditions == "blowing snow" or conditions == "heavy snow"):
    print("Blizzard!")
```

## In Pairs or Triples:

*Predict what the code will do:*

```
x = 6
y = x % 4
w = y**3
z = w // 2
print(x,y,w,z)
x,y = y,w
print(x,y,w,z)
x = y / 2
print(x,y,w,z)
```

```
sports = ["Field Hockey", "Swimming", "Water Polo"]
mess = "Qoauxca BrletRce crcx qvBnqa ocUxk"
result = ""
for i in range(len(mess)):
    if i % 3 == 0:
        print(mess[i])
        result = result + mess[i]
print(sports[1], result)
```

- And, design a program that asks the user for an image and then displays the upper left quarter of the image.  
(First, design the pseudocode. If time, expand to a Python program.)

# Python Tutor

```
x = 6
y = x % 4
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z = w // 2
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x,y = y,w
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x = y / 2
print(x,y,w,z)
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(Demo with pythonTutor)

# Design Question

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How to approach this:

- Create a “To Do” list of what your program has to accomplish.



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- Example:
  - ① Ask user for an image name.

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- Example:
  - ① Ask user for an image name.
  - ② Read in image.

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  - ③ Figure out size of image.

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- Example:
  - ① Ask user for an image name.
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  - ④ Make a new image that's half the height and half the width.

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  - ② Read in image.
  - ③ Figure out size of image.
  - ④ Make a new image that's half the height and half the width.
  - ⑤ Display the new image.



# Structured Data

Undergraduate			
College	Full-time	Part-time	Total
Baruch	11,288	3,922	15,210
Brooklyn	10,198	4,208	14,406
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- Common to have data structured in a spread sheet.

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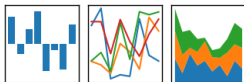
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- We will use the popular Python Data Analysis Library (**Pandas**).

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pandas

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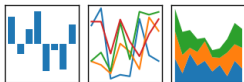


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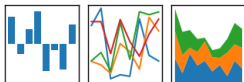
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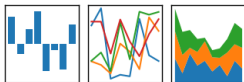


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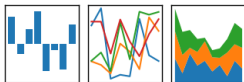


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- See end of Lab 6 for directions on downloading it to your home machine.
- To use, add to the top of your file:

```
import pandas as pd
```

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- Excel .xls files have much extra formatting.

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- Each row is a line in the file.
- Columns are separated by commas on each line.

# CSV Files

Source: [https://en.wikipedia.org/wiki/Demographics\\_of\\_New\\_York\\_City](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,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



# Reading in CSV Files

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Lehman	6,600	4,720	11,320
Medgar Evers	4,760	2,059	6,819
NYCCT	10,912	6,370	17,282
Queens	11,693	4,633	16,326
Staten Island	9,584	2,948	12,532
York	5,066	3,192	8,258

- To read in a CSV file: `myVar = pd.read_csv("myFile.csv")`

# Reading in CSV Files

Undergraduate			
College	Full-time	Part-time	Total
Baruch	11,288	3,922	15,210
Brooklyn	10,198	4,208	14,406
City	10,067	3,250	13,317
Hunter	12,223	4,500	16,723
John Jay	9,831	2,843	12,674
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- Pandas has its own type, **DataFrame**, that is perfect for holding a sheet of data.

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- Often abbreviated, `df`.

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- Often abbreviated, `df`.
- It also has **Series**, that is perfect for holding a row or column of data.

# Example: Reading in CSV Files

```
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,Queens,Bronx,Staten Island,Total
1698,4937,2017,,127,7881
1771,21883,3623,,2847,28423
1790,35131,45649,6159,1181,3827,49447
1800,40515,5740,6642,1755,4543,79215
1810,46373,8303,7444,2267,5347,119734
1820,123706,11187,8246,2782,6135,152056
1830,202589,20535,9049,3023,7082,242278
1840,312710,47613,34480,5344,10965,391114
1850,515547,138882,18593,8032,15061,696115
1860,813649,279122,32903,23593,25492,1174779
1870,942282,419901,45648,37393,33029,1478103
1880,1164673,599495,56559,51980,38991,1911698
1890,1441216,838547,87050,88908,51692,2507414
1900,1650093,1166582,132899,205507,67021,3437202
1910,2331542,1634351,284041,430989,85969,4766883
1920,2284183,2018356,448942,732018,116531,5420048
1930,1867312,2560451,1079129,1265258,150346,6904446
1940,1889924,2698285,1297634,1394711,174441,7454995
1950,1940101,2738275,1550849,1452277,191555,7893257
1960,1698281,2627319,1809578,1424815,221991,7781986
1970,1539233,2402012,1996473,1471701,295443,7094862
1980,1428285,2230936,1891325,1164872,352121,7077439
1990,1487536,2300644,1951598,1203789,378977,7322564
2000,1537195,2465326,2229379,1332450,443728,8008278
2010,1648473,2504760,2230722,1385108,448736,81751533
2015,1644518,2636738,2339155,1455444,474558,8550405
```

nycHistPop.csv

In Lab 6

# Example: Reading in CSV Files

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

```
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,Queens,Bronx,Staten Island,Total
1698,4937,2017,,727,7881
1771,21883,3623,,2847,28423
1790,35131,4548,6159,1181,3827,49447
1800,40515,5740,6642,1755,4543,79215
1810,46373,6303,7444,2267,5347,119734
1820,123706,11187,8246,2782,6135,152056
1830,202589,20535,8049,3023,7082,242278
1840,312710,47613,34480,5344,10965,391114
1850,515547,138882,18593,8032,15061,696115
1860,813649,279122,32903,23593,25492,1174779
1870,942282,419801,45468,37393,33029,1478103
1880,1164673,599495,56559,51980,38991,1911698
1890,1441216,838547,87050,88908,51692,2507414
1900,1650093,1166582,132899,200507,67021,3437202
1910,2331542,1634351,284041,430989,85969,4766883
1920,2284183,2018356,448942,732018,116531,5420048
1930,1867312,2580401,1079129,1265238,150346,6905446
1940,1889924,2698285,1297634,1394711,174441,7454395
1950,1960101,2738275,1550849,1452277,191555,7893257
1960,1698281,2627319,1809578,1424815,221991,7781986
1970,1539233,2602012,1986473,1471701,295443,7094862
1980,1428285,2230936,1891325,1168972,352121,7071439
1990,1487536,2300644,1951598,1203789,378977,7322564
2000,1537195,2465326,2229379,1332450,443728,8008278
2010,1648473,2504760,2230728,1385108,448738,8175133
2015,1644518,2636738,2339150,1455444,476558,8550405
```

nycHistPop.csv

In Lab 6

# Example: Reading in CSV Files

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

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

```
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, Queens, Bronx, Staten Island, Total  
1698,4937,2017,,127,7881  
1773,21883,3623,,2847,28423  
1790,35131,4548,6159,1181,3827,49447  
1800,40515,5740,6642,1755,4543,79215  
1810,46373,6303,7444,2267,5347,119734  
1820,123706,11187,8246,2782,6135,152056  
1830,202589,20535,8049,3023,7082,242278  
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1860,813649,279122,32903,23593,25492,1174779  
1870,942282,419801,45648,37393,33029,1478183  
1880,1164673,599495,56559,51980,38991,1911698  
1890,1441216,838547,87050,88908,51693,2507414  
1900,1850093,1166582,152899,200507,67021,3437202  
1910,2331542,1634351,284041,430989,85969,4766883  
1920,2284183,2018356,448942,732018,116531,3620348  
1930,1867312,2580451,1079129,1265258,159346,6505446  
1940,1889924,2698285,1297634,1394711,174441,7454395  
1950,1960101,2738275,1550849,1452277,191555,7893257  
1960,1698281,2627319,1809578,1424815,221991,7781984  
1970,1539233,2602012,1986473,1471701,295443,7894862  
1980,1428285,2230936,1891325,1168972,352121,7071439  
1990,1487536,2300644,1951598,1203789,378977,7322564  
2000,1531795,2465326,2229379,1332450,443728,8008278  
2010,1648473,2568760,2230728,1385108,448738,8175133  
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```

nycHistPop.csv

In Lab 6

# Example: Reading in CSV Files

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

```
pop = pd.read_csv('nycHistPop.csv', skiprows=5)
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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,Queens,Bronx,Staten Island,Total
1698,4937,2017,,727,7881
1773,21883,3623,,2847,28423
1790,35131,4548,6159,1181,3827,49447
1800,40515,5740,6642,1755,4543,79215
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1870,942282,419801,45648,37393,33029,1478103
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1900,1650093,1166582,152899,200507,67021,3437202
1910,2331542,1634351,284041,430989,85969,4766883
1920,2284193,2018356,448942,732018,116531,5420048
1930,1867312,2580451,1079129,1265258,150346,6904446
1940,1889924,2698285,1297634,1394711,174441,7454395
1950,1940101,2738275,1550849,1452277,191555,7893257
1960,1698281,2627319,1809578,1424815,221991,7781986
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2000,1531795,2465326,2229379,1332450,443728,8008278
2010,1548473,2504760,2230728,1385108,448738,8175133
2015,1644518,2636738,2339150,1455444,474558,8550405
```

nycHistPop.csv

In Lab 6

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



# Example: Reading in CSV Files

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

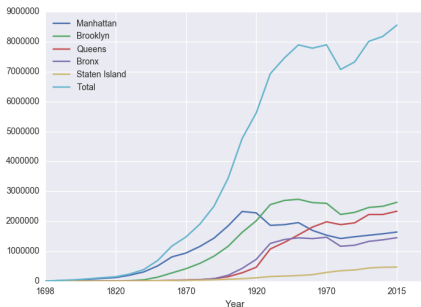
```
pop = pd.read_csv('nycHistPop.csv', skiprows=5)
```

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

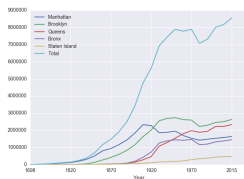
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All population figures are consistent with present-day boundaries.....
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.....
Year,Manhattan,Brooklyn,Queens,Bronx,Staten Island>Total
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1840,312710,47613,14480,5344,10965,393114
1850,515547,138882,18593,8032,15561,696115
1860,813649,279122,32903,23593,25492,1174779
1870,942292,419901,45648,37393,33529,1478103
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1890,1441216,838547,87050,88908,51693,2507414
1900,1650093,1166582,152899,20507,67021,3437202
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1980,1428285,2230936,1891325,1148972,352121,7077439
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2000,1537195,2465326,2229379,1332450,443728,8008278
2010,1648473,2504760,2230722,1385108,448730,8175133
2015,1644518,2636735,2339150,1455444,474558,8550405
```

nycHistPop.csv

In Lab 6

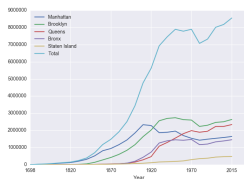


# Series in Pandas



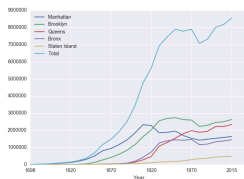
- Series can store a column or row of a DataFrame.

# 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.

# Series in Pandas



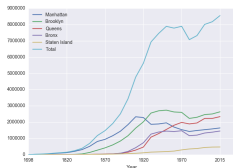
- Series can store a column or row of a DataFrame.
- Example: `pop["Manhattan"]` is the Series corresponding to the column of Manhattan data.
- Example:  

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

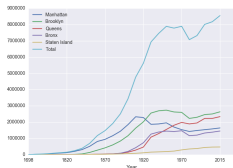
# In Pairs or Triples

Predict what the following will do:

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



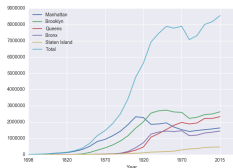
# In Pairs or Triples



Predict what the following will do:

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

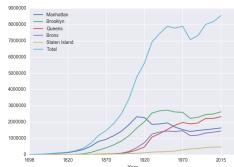
# In Pairs or Triples



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())`

# In Pairs or Triples

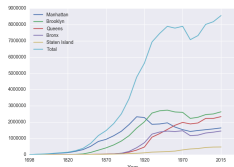


Predict what the following will do:

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



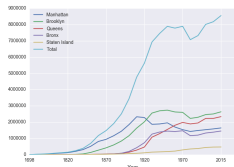
# In Pairs or Triples



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())`
- `pop.plot.bar(x="Year")`
- `pop.plot.scatter(x="Brooklyn", y="Total")`

# In Pairs or Triples



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())`
- `pop.plot.bar(x="Year")`
- `pop.plot.scatter(x="Brooklyn", y="Total")`
- `pop["Fraction"] = pop["Bronx"] / pop["Total"]`

# CS Survey Talk

Prof. Raffi Khatchadourian

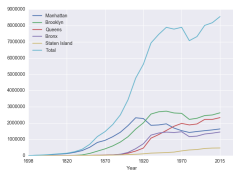


Department of Computer Science  
Hunter College & the Graduate Center  
Software Engineering

# Solutions

Predict what the following will do:

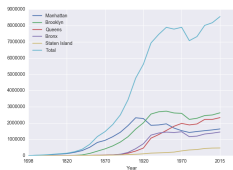
- `print("Queens:", pop["Queens"].min())`



# Solutions

Predict what the following will do:

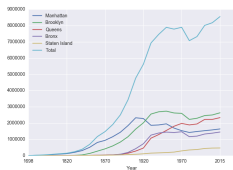
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*



# Solutions

Predict what the following will do:

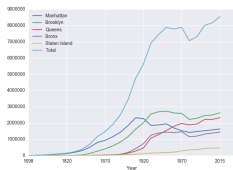
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`



# Solutions

Predict what the following will do:

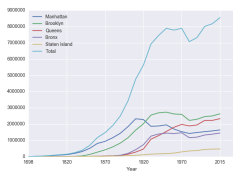
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*



# Solutions

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

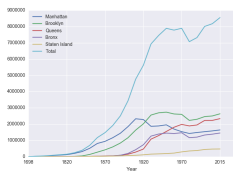




# Solutions

Predict what the following will do:

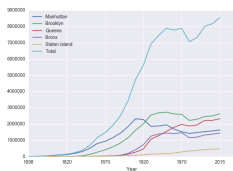
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
- `print("S I :", pop["Staten Island"].std())`  
*Standard deviation of values in the column "Staten Island".*



# Solutions

Predict what the following will do:

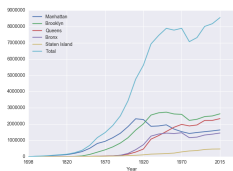
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
- `print("S I :", pop["Staten Island"].std())`  
*Standard deviation of values in the column "Staten Island".*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



# Solutions

Predict what the following will do:

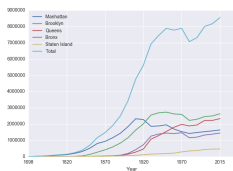
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
- `print("S I :", pop["Staten Island"].std())`  
*Standard deviation of values in the column "Staten Island".*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



# Solutions

Predict what the following will do:

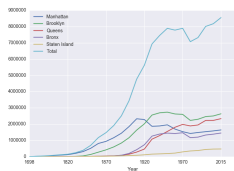
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
- `print("S I :", pop["Staten Island"].std())`  
*Standard deviation of values in the column "Staten Island".*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



# Solutions

Predict what the following will do:

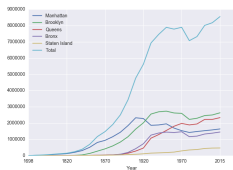
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
- `print("S I :", pop["Staten Island"].std())`  
*Standard deviation of values in the column "Staten Island".*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



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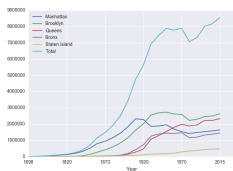
- `print("Queens:", pop["Queens"].min())`  
*Minimum value in the column with label "Queens".*
- `print("S I:", pop["Staten Island"].mean())`  
*Average of values in the column "Staten Island".*
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*Standard deviation of values in the column "Staten Island".*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



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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())`  
*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.*



# In Pairs or Triples

Write a complete Python program that reads in the file, `cunyF2016.csv`, and produces a scatter plot of full-time versus part-time enrollment.

Undergraduate			
College	Full-time	Part-time	Total
Baruch	11,288	3,922	15,210
Brooklyn	10,198	4,208	14,406
City	10,067	3,250	13,317
Hunter	12,223	4,500	16,723
John Jay	9,831	2,843	12,674
Lehman	6,800	4,720	11,320
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`cunyF2016.csv`



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- 1 *Include `pandas` & `pyplot` libraries.*
- 2 *Read in the CSV file.*
- 3 *Set up a scatter plot.*
- 4 *Display plot.*

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```
import matplotlib.pyplot as plt
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- 3 *Set up a scatter plot.*

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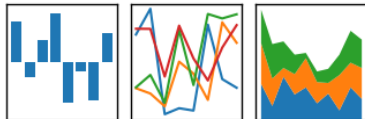
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pop.plot(x="Full-time",y="Part-time")
```
- 4 *Display plot.*  

```
plt.show()
```

# Recap

# pandas

$$y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$$

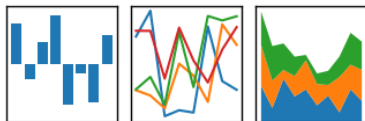


- Pandas library has elegant solutions for accessing & analyzing structured data.

# Recap

# pandas

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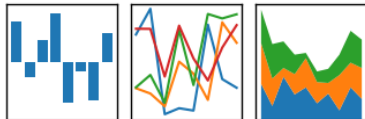


- Pandas library has elegant solutions for accessing & analyzing structured data.
- Can manipulate individual columns or rows ('Series').

# Recap

# pandas

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- Pandas library has elegant solutions for accessing & analyzing structured data.
- Can manipulate individual columns or rows ('Series').
- Has useful functions for the entire sheet ('DataFrame') such as plotting.

# Lecture Slips



- On-line lecture slips: [tinyurl.com/yc6j6ubr](https://tinyurl.com/yc6j6ubr)



# Final Prep



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- Will discuss solutions next lecture.