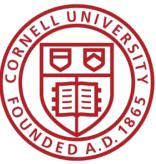


PUBPOL 2130/ INFO 3130

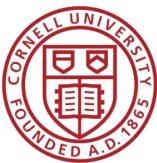
Lab 1





Logistics!

- Main GitHub Repository [here](#)
- We suggest going through the [gentle intro notebook](#) if you do not have experience with programming concepts
- You can reach out to us at
 - jrg377@cornell.edu
 - tp399@cornell.edu
- You can use Colab, Jupyter, VS code, etc.
- We're not going to do installations today, let's work on Colab if you don't have Jupyter, VS Code, etc. installed



Jan 24: Pandas

Python library used for data manipulation, analysis, and cleaning.

```
▶ instructors = pd.Series(["Laura Tach", "Moon Duchin", "Rachel Riedl", "Benjamin Soltoff"], index=["PUBPOL 2301","PUBPOL 2130","PUBPOL 2320","INFO 2951"])

print("\nPandas Series Example")
print(instructors)
```

Pandas Series Example
PUBPOL 2301 Laura Tach
PUBPOL 2130 Moon Duchin
PUBPOL 2320 Rachel Riedl
INFO 2951 Benjamin Soltoff
dtype: object

Series

```
df = pd.DataFrame({
    "id": [
        "PUBPOL 2301",
        "PUBPOL 2130",
        "PUBPOL 2320",
        "INFO 2951",
    ],
    "name": [
        "Introduction to Public Policy",
        "Data and the State: How Governments See People and Places",
        "Global Democracy and Public Policy",
        "Introduction to Data Science with R",
    ],
    "instructor": ["Laura Tach", "Moon Duchin", "Rachel Riedl", "Benjamin Soltoff"],
    "credits": [4., 4., 3., 4.],
})
df
```

Dataframe

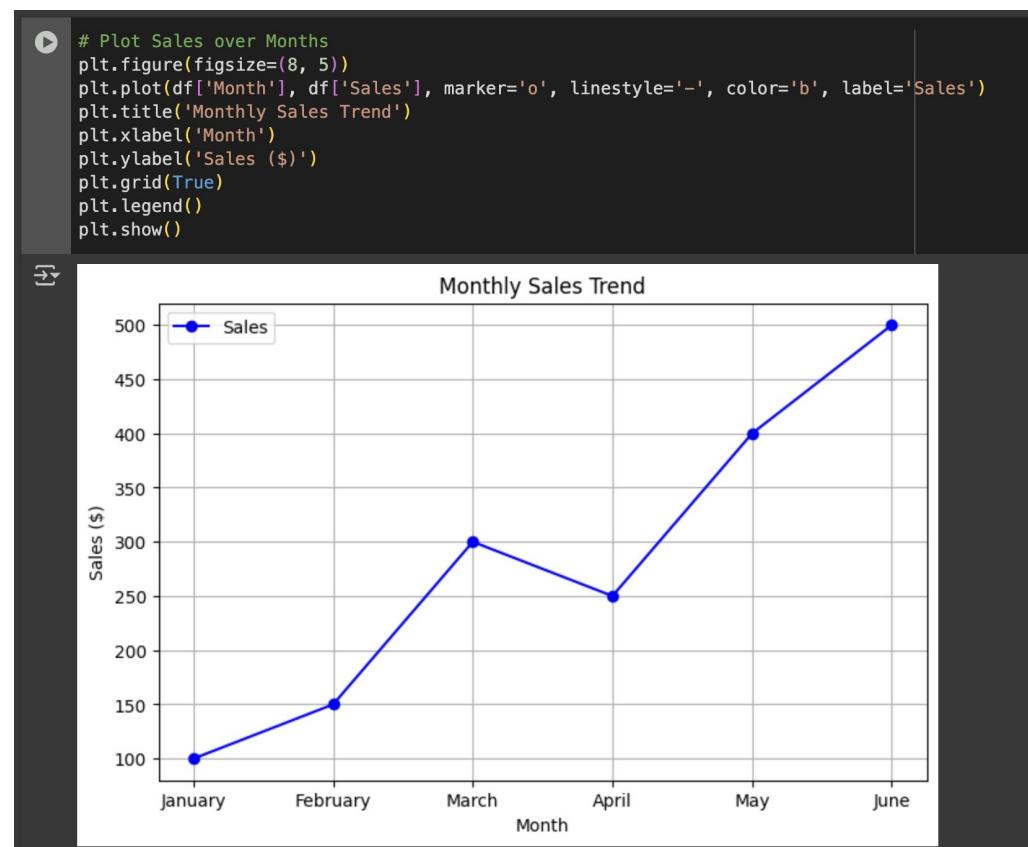
	id	name	instructor	credits
0	PUBPOL 2301	Introduction to Public Policy	Laura Tach	4.0
1	PUBPOL 2130	Data and the State: How Governments See People...	Moon Duchin	4.0
2	PUBPOL 2320	Global Democracy and Public Policy	Rachel Riedl	3.0
3	INFO 2951	Introduction to Data Science with R	Benjamin Soltoff	4.0

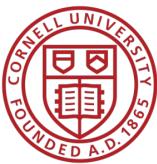


Jan 24: Matplotlib Theory

Python library used for creating static, interactive, and animated visualizations.

- Versatility
- Customization
- Integration
- Interactive Capabilities
- Export Options

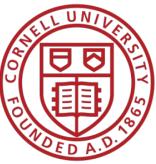




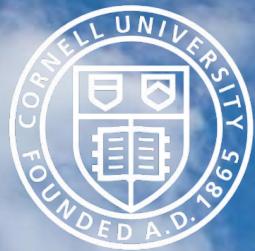
Jan 24: Matplotlib Disadvantages

- Complexity
- Verbose Syntax
- Limited Interactivity
- Performance Issues
- Default Aesthetics are Outdated





Let's start executing week1.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 2





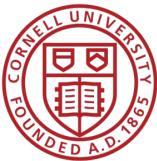
Announcements!

Weekly homework assignments:

- Will be due in 11 days
- New homework assigned on Fridays during lab
- Turn in on Gradescope

Upcoming exam on Feb. 13th:

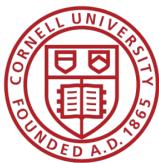
- Will be 40 minutes, in class
- Lecture on Feb. 11th – likely exam review or makeup



Announcements!

Homework Reminders:

- Don't give us code unless we ask for it!
 - *Don't turn in an .ipynb file*
 - *Turn in exports, not screenshots*
- Make sure axis labels are clear
- Include information on parameters that don't change
- Default parameters in matplotlib may not be optimal – experiment with different ones
 - *E.g., binning with histograms*



Jan 31: Census Data

United States® Census Bureau

Search Advanced Search

All Tables Maps Charts Profiles Pages

P1 | RACE

Decennial Census Universe: Total population 2020: DEC Redistricting Data (PL 94-171)

Label

	New York
Total:	20,201,249
Population of one race:	18,433,786
White alone	11,143,349
Black or African American alone	2,986,172
American Indian and Alaska Native alone	149,690
Asian alone	1,933,127
Native Hawaiian and Other Pacific Islander alone	10,815
Some Other Race alone	2,210,633
Population of two or more races:	1,767,463
Population of two races:	1,649,229
White; Black or African American	175,686
White; American Indian and Alaska Native	113,950
White; Asian	148,927
White; Native Hawaiian and Other Pacific Islander	3,890
White; Some Other Race	840,481
Black or African American; American Indian and Alaska Native	31,562
Black or African American; Asian	21,450
Black or African American; Native Hawaiian and Other Pacific Islander	3,274
Black or African American; Some Other Race	226,733
American Indian and Alaska Native; Asian	5,958
American Indian and Alaska Native; Native Hawaiian and Other Pacific Islander	564

Apps Help FAQ Feedback

1 Filter New York **4109 Results** View: 10 | 25 | 50 Download Table Data

Geographies Nation > State > County > County Subdivision > Place > ZIP Code Tabulation Area > Metropolitan/Micropolitan Statistical Area > Census Tract > Block > Block Group > All Geographies >

Topics Business and Economy > Education > Employment > Families and Living Arrangements > Government >

Search for a filter or table



Jan 31: census Python Package

- **Wrapper** for the United States Census Bureau's API
 - More information [here](#)
- Information on the Census Bureau API is [here](#) and [here](#)
 - You can request an API key [here](#)

```
from census import Census
from us import states

c = Census("MY_API_KEY")
c.acs5.get(('NAME', 'B25034_010E'),
           {'for': 'state:{}'.format(states.MD.fips)})
```

Note: you do not need an API key for querying small quantities of data, with minimal restrictions (e.g. <500 queries/day per IP)



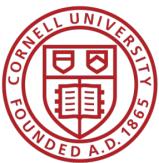
Jan 31: Exporting plots

- Tricky in Colab vs. VSCode/Jupyter
- In **matplotlib**: `plt.savefig("file_name.jpg")`
- In **Colab**:

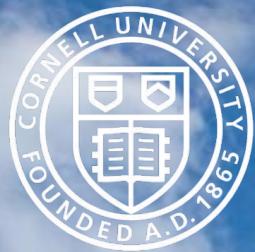
```
from google.colab import files  
plt.savefig("file_name.jpg")  
files.download("file_name.jpg")
```

- Alternatively, you can use simple scripts in Colab to save exports to your **temporary** Colab environment

```
plt.savefig("file_name.jpg", format="jpeg", dpi=95)
```



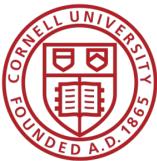
Let's start executing Week2.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 3

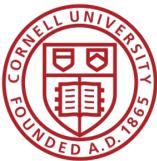




Feb 07: What Are Shapefiles?

A shapefile is a widely used **geospatial data format** for mapping locations, boundaries, and spatial relationships.

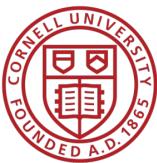
- It represents **geographic features** as points, lines, or polygons.
- **Common Uses:** Political boundaries, census tracts, roads, environmental features.
- **Shapefile Components:**
 - .shp – Stores geometry (the actual shapes).
 - .shx – Index for quick lookup.
 - .dbf – Attribute data (tabular information).



Feb 07: Census Shapefiles

Some examples of Census shapefiles

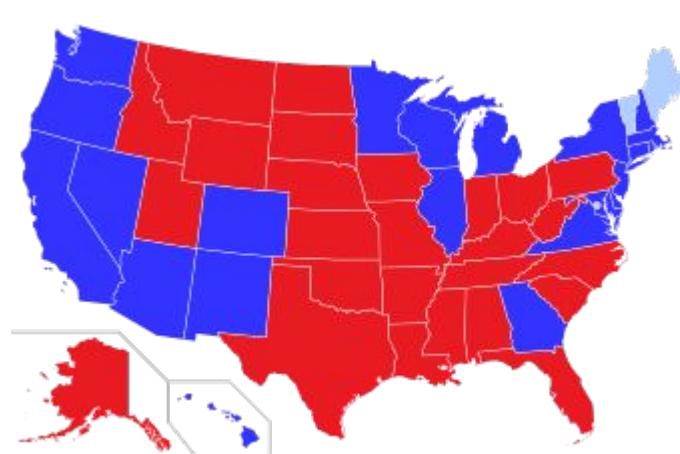
- States
- Counties and county equivalents
- County subdivisions
- Census tracts
- American Indian, Alaska Native, Native Hawaiian areas
- Tribal subdivisions
- Roads, rails, rivers
- School districts, etc.



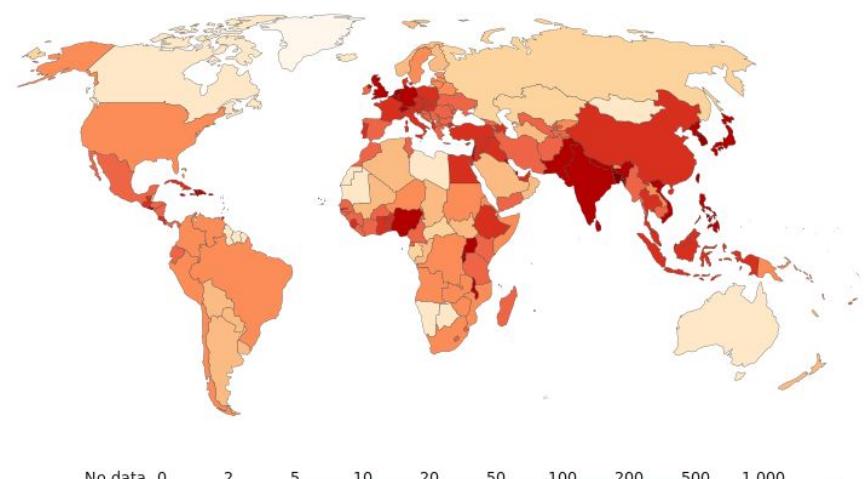
Feb 07: What Is a Choropleth Map?

A choropleth map is a thematic map where areas are shaded or colored based on data values.

- Each region (e.g., state, county) is filled with a color corresponding to a data variable (e.g., population, unemployment rate).



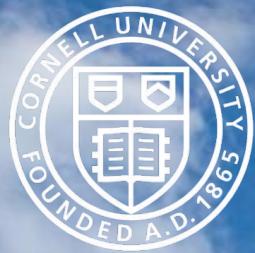
Population density, 2022
The number of people per km² of land area



Data source: HYDE (2017); Gapminder (2022); UN WPP (2022); UN FAO (2022)



Let's start executing Week3.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 4





Feb 14: Announcements

- No new homework this week
- Seems you all did great on the test (and particularly on the Python problems)
- Python literacy is a learning objective -- understanding basic syntax is important
 - Spend time on the notebooks outside of class, and ask questions!
- We plan to offer a Python learning session some time next week if you need more support
 - Please fill out this google form:
<https://tinyurl.com/2130-poll>



Feb 14: Census Blocks

Blocks:

- Statistical areas with natural boundaries (e.g., roads)
- Cover the entire U.S.
- Smallest geographic unit for demographic data

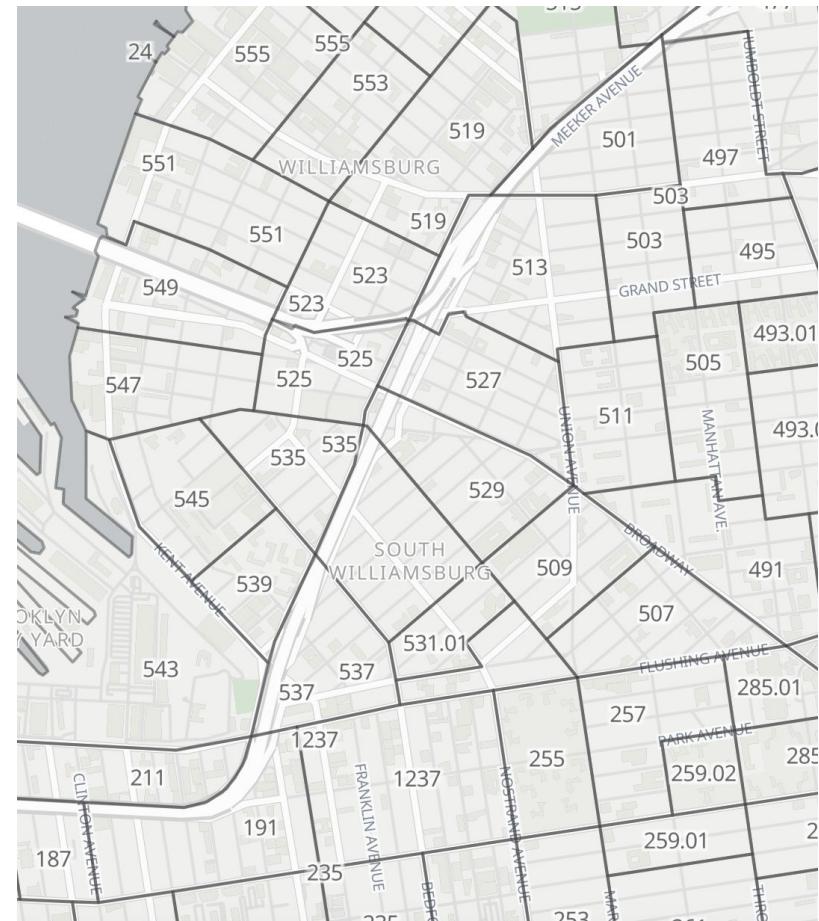


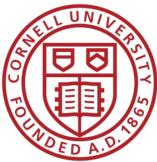


Feb 14: Census Tracts

Tracts:

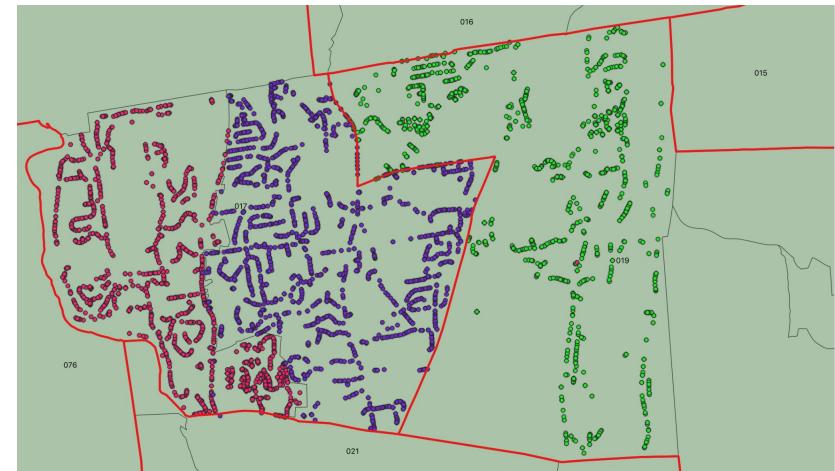
- Small, statistical subdivisions of counties
- Population between 1,200 and 8,000
- Spatial size varies widely





Feb 14: Precincts

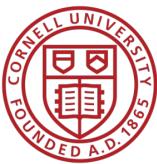
- Finest resolution of election data
- Not consistently maintained by states!
- mggg contains an open repository of precincts data





Feb 14: maup

- A geospatial toolkit for redistricting data
- Helpful for:
 - Aggregating from blocks to precincts
 - Disaggregating from precincts to blocks
 - “Prorating” data when there is no clean overlap



Feb 14: Assigning precincts to districts

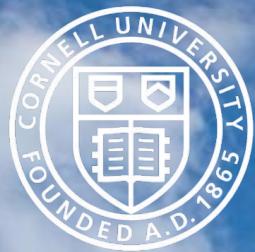
Assigning precincts to districts

The `assign` function in `maup` takes two sets of geometries called `sources` and `targets` and returns a pandas `Series`. The Series maps each geometry in `sources` to the geometry in `targets` that covers it. (Here, geometry *A* covers geometry *B* if every point of *A* and its boundary lies in *B* or its boundary.) If a source geometry is not covered by one single target geometry, it is assigned to the target geometry that covers the largest portion of its area.

```
>>> import maup
>>>
>>> precinct_to_district_assignment = maup.assign(precincts, districts)
>>> # Add the assigned districts as a column of the `precincts` GeoDataFrame:
>>> precincts["DISTRICT"] = precinct_to_district_assignment
>>> precinct_to_district_assignment.head()
0    7
1    5
2   13
3    6
4    1
dtype: int64
```



Let's start executing Week4.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 5





Logistics!

- Created a slide to guide you through joins - please refer to it
- Today's notebook requires creation of a Google sheet, and map creation in flowmap.blue using this Google sheet; in CASE your sheet creation does not work, you can use these sheet IDs instead
 - yAnUt3bQcGpokOzteRQ2iGQUK3Lyw5S0OMwOOCa0wLQ
 - 1ffUAGYyzzPn3yY-0HehKnVsLayonauOLFtSPT3cPd0A
 - 1DoFxzh7 TKj2hbW7WV7fxbt0 MjjliKhAsGY7TDv4TE



Feb 21: Flowmaps

A flow map is a type of thematic map that visualizes movement or flow of objects, people, or data from one location to another.

Uses lines/arrows to show direction and magnitude of movement.

- **Lines/Arrows:** Represent movement direction.
- **Line Thickness/Color:** Can indicate volume/intensity of movement.
- **Nodes (Start/End Points):** Origin and destination of movement.
- **Base Map:** Provides spatial context (e.g., roads, cities, regions).

What we're using for Flowmaps: [flowmap.blue](#)



Feb 21: Flowmaps Applications

Flowmaps are commonly used to visualise these patterns

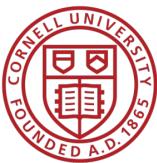
- **Migration Patterns:** People moving between cities or countries.
- **Trade Flows:** Import/export routes between regions.
- **Transportation & Traffic:** Airline routes, shipping lanes, or road traffic.
- **Energy Distribution:** Power grids, oil pipelines.
- **Internet Data Flow:** Digital connectivity between locations.



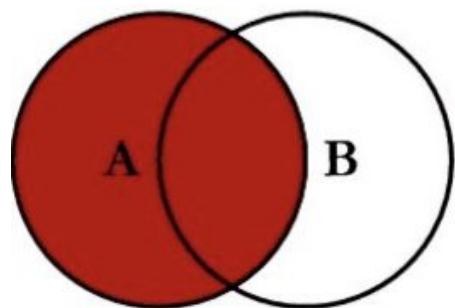
Feb 21: Flowmap example

Visualization of Bike Sharing System movements in Helsinki



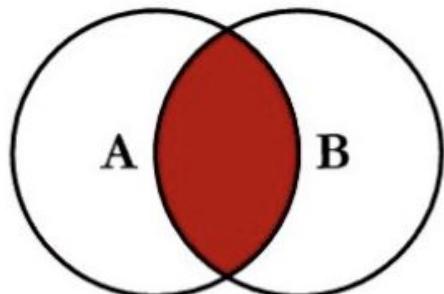


Feb 21: Joins (reference slide)



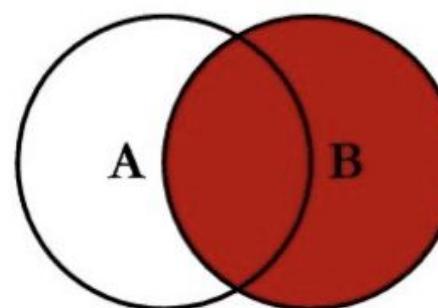
Left Join

Keeps all elements in the left table, and common elements from the right table ONLY



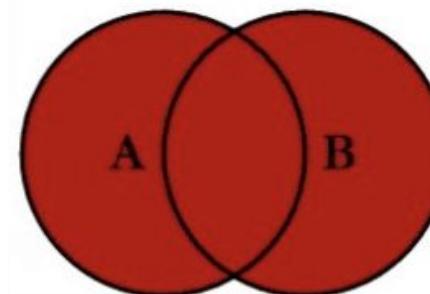
Inner Join

Keeps all elements common to BOTH tables



Right Join

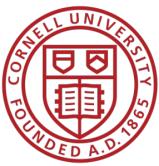
Keeps all elements in the right table, and common elements from the left table ONLY



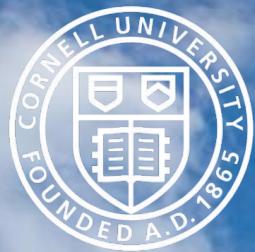
Full Outer Join

Keeps ALL elements in from both tables.

```
df_joined = left_table.join(right_table, how = "outer")
```



Let's start executing Week5.ipynb together!



PUBPOL 2130/ INFO 3130

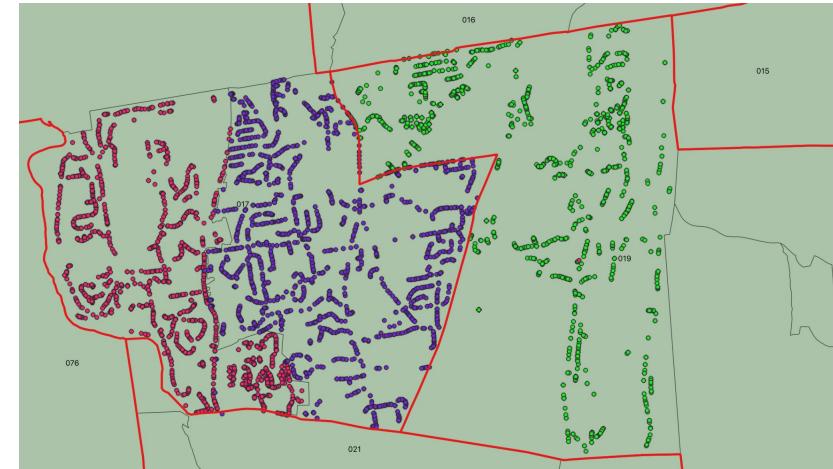
Lab 6

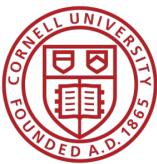




Feb 28: Precincts

- Finest resolution of election data
- Not consistently maintained by states!
- mggg contains an open repository of precincts data





Feb 28: Precincts

A partial view of an American flag is visible on the left side of the slide, showing stars and stripes.

2022 NY General Election Shapefile and
Results by Election District

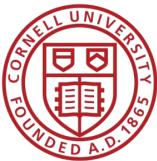
Credit and Thank You

This data would not have been possible to create and provide without the assistance of each of New York's county Boards of Elections commissioners and staffers, who were often very eager to help provide the data I was looking for; county GIS and Planning Department workers, who filled in major gaps; Derek Willis and the volunteers at Open Elections who worked diligently to clean up and standardize the election results data files I compiled for the entire state; the folks at the Redistricting Data Hub who helped verify and correct shapefile errors; and the wonderful folks at the Census Bureau and the U.S. Department of Agriculture National Resources Conservation Service and the U.S. Geological Survey, without whom the shapefile would have been impossible to create.



Feb 28: maup

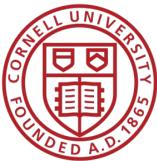
- A geospatial toolkit for redistricting data
- Potential uses:
 - Assigning precincts to districts,
 - Aggregating block data to precincts,
 - Disaggregating data from precincts down to blocks,
 - Prorating data when units do not nest neatly, and
 - Fixing topological issues, overlaps, and gaps



Feb 28: maup.assign()

```
blocks_to_precincts_assignment = maup.assign(blocks, pcts)
```

- Takes in two sets of geometries
- Returns a Pandas series with the assignments
- Use cases:
 - Assign blocks to precincts
 - Assign precincts to districts



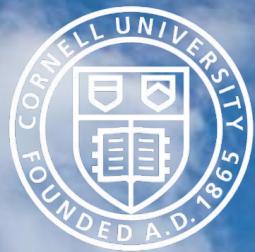
Feb 28: maup.prorate()

```
maup.prorate(overlapping_blocks, data_to_porate, weights)
```

- Problem:
 - You have precincts with some election results data
 - You also have different precincts (e.g., redistricting)
- We can **prorate** data from the old precincts to the new precincts by weighting the data proportional to the overlapping population.
 - Disaggregate to Census blocks
 - Reaggregate to new precincts



Let's start executing Week6.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 7





Announcements!

Weekly homework assignments:

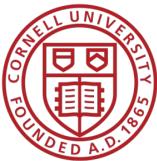
- Homework will reflect the following submission timelines in Gradescope going forward
 - 7 days till submission deadline
 - +4 slip days
- i.e. you will still have up to 11 days to submit your homework!



Mar 07: PUMS Data

PUMS = Public Use Microdata Sample

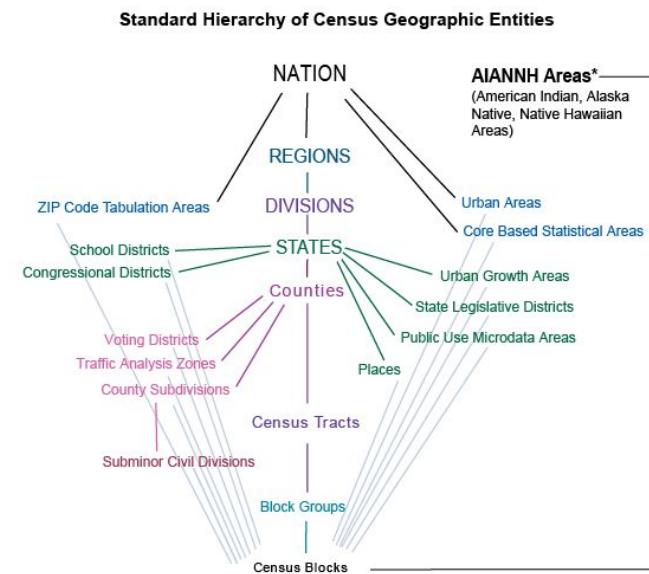
- Provided by the Census Bureau
- Individual or household level information
 - Age
 - Race
 - Gender
 - Income
 - Employment
 - Housing variables

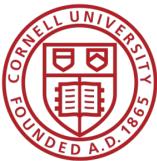


Mar 07: PUMS Data

PUMS = Public Use Microdata Sample

- What makes it “microdata”
 - ACS provides data at a Block Group or Tract level (most granular geography)
 - PUMS provides data at
 - An individual level, OR
 - A household level
 - Geographic granularity: PUMAs





Mar 07: PUMS Data

New York City PUMAs and Community Districts

Public Use Microdata Areas (PUMAs) approximate
NYC Community Districts (CDs).

- PUMAs
- Community District boundaries
- Joint Interest Areas (JIAs) e.g. parks and airports

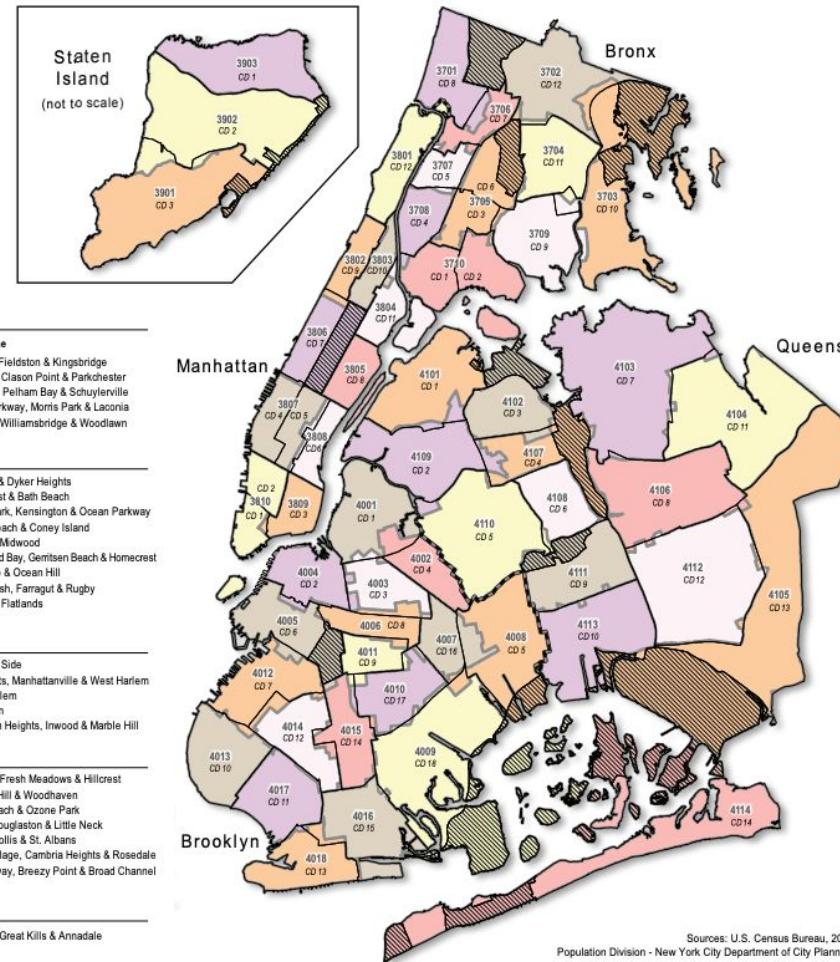
Bronx	
CD	PUMA PUMA Name
1 & 2	3710 Hunts Point, Longwood & Melrose
3 & 6	3705 Belmont, Crotona Park East & East Tremont
4	3708 Concourse, Highbridge & Mount Eden
5	3707 Morris Heights, Fordham South & Mount Hope
7	3706 Bedford Park, Fordham North & Norwood
	8 3701 Riverdale, Fieldston & Kingsbridge
	9 3709 Castle Hill, Clason Point & Portchester
	10 3703 Co-op City, Pelham Bay & Schuylerville
	11 3704 Pelham Parkway, Morris Park & Laconia
	12 3702 Wakefield, Williamsbridge & Woodlawn

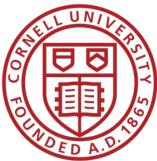
Brooklyn	
CD	PUMA PUMA Name
1	4001 Greenpoint & Williamsburg
2	4004 Brooklyn Heights & Fort Greene
3	4003 Bedford-Stuyvesant
4	4002 Bushwick
5	4008 East New York & Starrett City
6	4005 Park Slope, Carroll Gardens & Red Hook
7	4012 Sunset Park & Windsor Terrace
8	4006 Crown Heights North & Prospect Heights
9	4011 Crown Heights So., Prospect Lefferts & Wingate
	10 4013 Bay Ridge & Dyker Heights
	11 4017 Bensonhurst & Bath Beach
	12 4014 Borough Park, Kensington & Ocean Parkway
	13 4018 Brighton Beach & Coney Island
	14 4015 Flatbush & Midwood
	15 4016 Sheepshead Bay, Gerritsen Beach & Homecrest
	16 4007 Brownsville & Ocean Hill
	17 4010 East Flatbush, Faragut & Rugby
	18 4009 Canarsie & Flatlands

Manhattan	
CD	PUMA PUMA Name
1 & 2	3810 Battery Park City, Greenwich Village & Soho
3	3809 Chinatown & Lower East Side
4 & 5	3807 Chelsea, Clinton & Midtown Business District
6	3808 Murray Hill, Gramercy & Stuyvesant Town
7	3806 Upper West Side & West Side
	8 3805 Upper East Side
	9 3802 Hamilton Hts, Manhattanville & West Harlem
	10 3803 Central Harlem
	11 3804 East Harlem
	12 3801 Washington Heights, Inwood & Marble Hill

Queens	
CD	PUMA PUMA Name
1	4101 Astoria & Long Island City
2	4109 Sunnyside & Woodside
3	4102 Jackson Heights & North Corona
4	4107 Elmhurst & South Corona
5	4110 Ridgewood, Glendale & Middle Village
6	4108 Forest Hills & Rego Park
7	4103 Flushing, Murray Hill & Whitestone
	8 4106 Briarwood, Fresh Meadows & Hillcrest
	9 4111 Richmond Hill & Woodhaven
	10 4113 Howard Beach & Ozone Park
	11 4104 Bayside, Douglaston & Little Neck
	12 4112 Jamaica, Hollis & St. Albans
	13 4105 Queens Village, Cambria Heights & Rosedale
	14 4114 Far Rockaway, Breezy Point & Broad Channel

Staten Island	
CD	PUMA PUMA Name
1	3903 Port Richmond, Stapleton & Mariner's Harbor
2	3902 New Springville & South Beach
	3 3901 Tottenville, Great Kills & Annadale





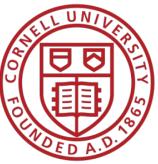
Mar 07: PUMS Data

- Where to access PUMS data

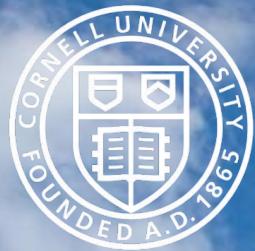
- Census Bureau:

<https://www.census.gov/programs-surveys/acs/microdata/access.html>

- IPUMS: <https://usa.ipums.org/usa/>
 - Census Bureau data portal: <https://data.census.gov/cedsci/>



Let's start executing Week7.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 8





Announcements!

Prelim & Final Exam Dates!

- Next prelim is March 25th
- Potentially optional third exam (announced on Tuesday)
- Final deliverable due Wed, May 14th @ 12:00 PM
- HW6 Warmup Question Change! (See Gradescope)

Reminder:

- Today's data is **sensitive!** Do not distribute it – you will need to attest that you have destroyed it at the end of the semester.



Mar 14: OPTN Data

- OPTN = Organ Procurement & Transplantation Network
- Data is available [here](#)
- Contains pre- and post-transplant information on:
 - Waiting list candidates
 - donor/recipient matches
 - Deceased and living donors
- We use [STAR \(Standard Transplant Analysis and Research\) files](#)



Mar 14: Seaborn

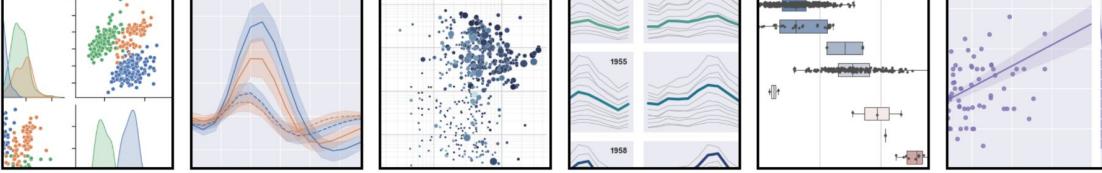
- Today's notebook also uses **seaborn**

 [seaborn](#)

Installing Gallery Tutorial API Releases Citing FAQ

🔍 🔍 📁 🐦

seaborn: statistical data visualization



Seaborn is a Python data visualization library based on [matplotlib](#). It provides a high-level interface for drawing attractive and informative statistical graphics.

For a brief introduction to the ideas behind the library, you can read the [introductory notes](#) or the [paper](#). Visit the installation page to see how you can download the package and get started with it. You can browse the [example gallery](#) to see some of the things that you can do with seaborn, and then check out the [tutorials](#) or [API reference](#) to find out how.

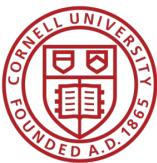
To see the code or report a bug, please visit the [GitHub repository](#). General support questions are most at home on [stackoverflow](#), which has a dedicated channel for seaborn.

Contents

[Installing](#)
[Gallery](#)
[Tutorial](#)
[API](#)
[Releases](#)
[Citing](#)
[FAQ](#)

Features

- **New** Objects: [API](#) | [Tutorial](#)
- Relational plots: [API](#) | [Tutorial](#)
- Distribution plots: [API](#) | [Tutorial](#)
- Categorical plots: [API](#) | [Tutorial](#)
- Regression plots: [API](#) | [Tutorial](#)
- Multi-plot grids: [API](#) | [Tutorial](#)
- Figure theming: [API](#) | [Tutorial](#)
- Color palettes: [API](#) | [Tutorial](#)



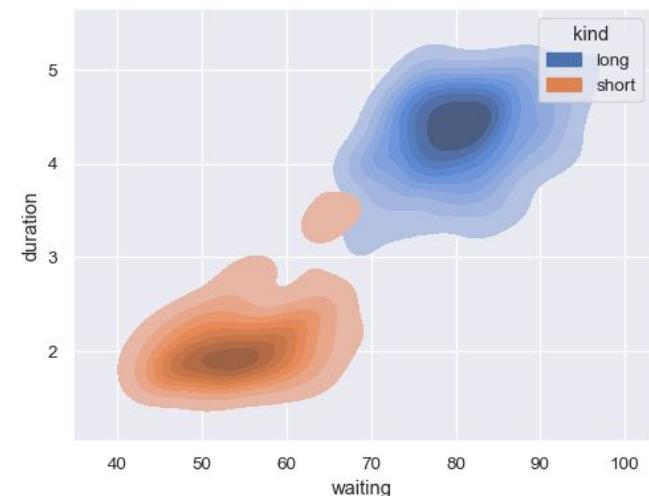
Mar 14: Seaborn: kdeplot

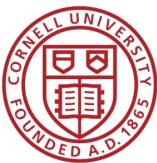
seaborn.kdeplot

```
seaborn.kdeplot(data=None, *, x=None, y=None, hue=None, weights=None,
palette=None, hue_order=None, hue_norm=None, color=None, fill=None,
multiple='layer', common_norm=True, common_grid=False, cumulative=False,
bw_method='scott', bw_adjust=1, warn_singular=True, log_scale=None, levels=10,
thresh=0.05, gridsize=200, cut=3, clip=None, legend=True, cbar=False, cbar_ax=None,
cbar_kws=None, ax=None, **kwargs) #
```

Plot univariate or bivariate distributions using kernel density estimation.

A kernel density estimate (KDE) plot is a method for visualizing the distribution of observations in a dataset, analogous to a histogram. KDE represents the data using a continuous probability density curve in one or more dimensions.





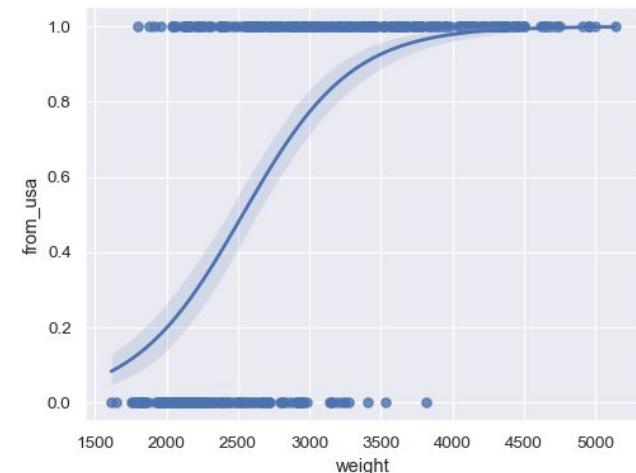
Mar 14: Seaborn: regplot

seaborn.regplot

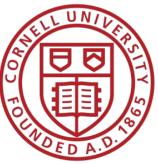
```
seaborn.regplot(data=None, *, x=None, y=None, x_estimator=None, x_bins=None,  
x_ci='ci', scatter=True, fit_reg=True, ci=95, n_boot=1000, units=None, seed=None,  
order=1, logistic=False, lowess=False, robust=False, logx=False, x_partial=None,  
y_partial=None, truncate=True, dropna=True, x_jitter=None, y_jitter=None,  
label=None, color=None, marker='o', scatter_kws=None, line_kws=None, ax=None) #
```

Plot data and a linear regression model fit.

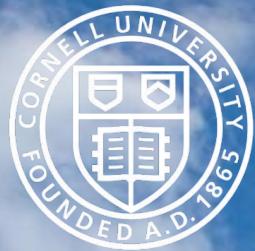
There are a number of mutually exclusive options for estimating the regression model. See the [tutorial](#) for more information.



For binary outcomes



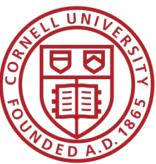
Let's start executing Week8.ipynb together!



PUBPOL 2130/ INFO 3130

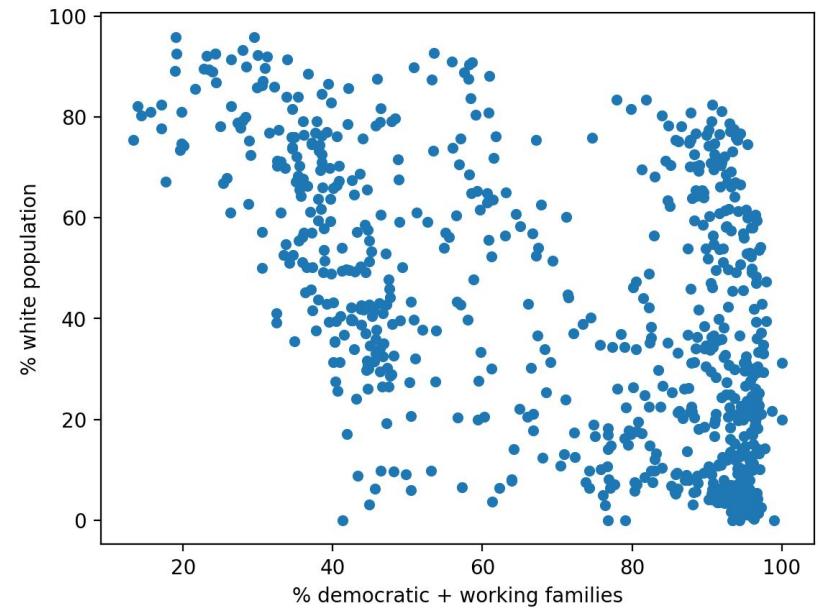
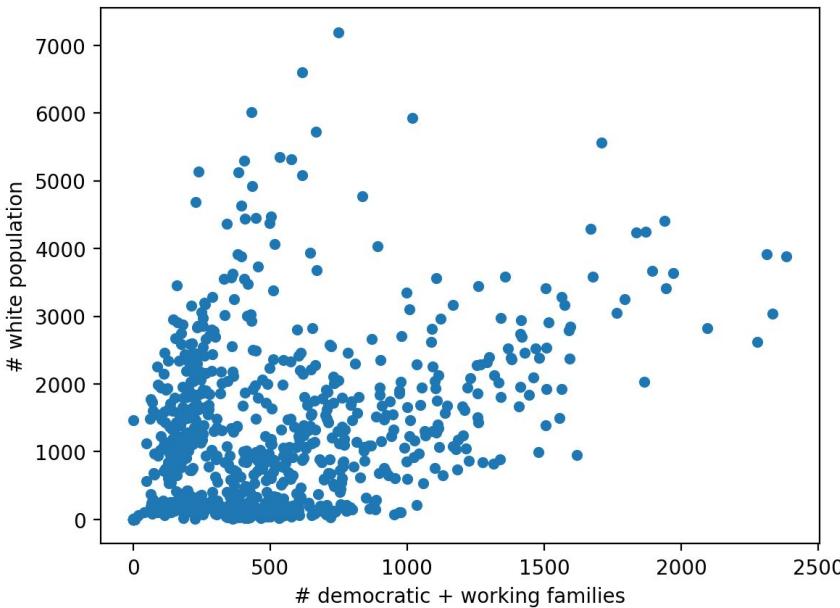
Lab 9

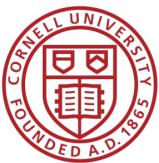




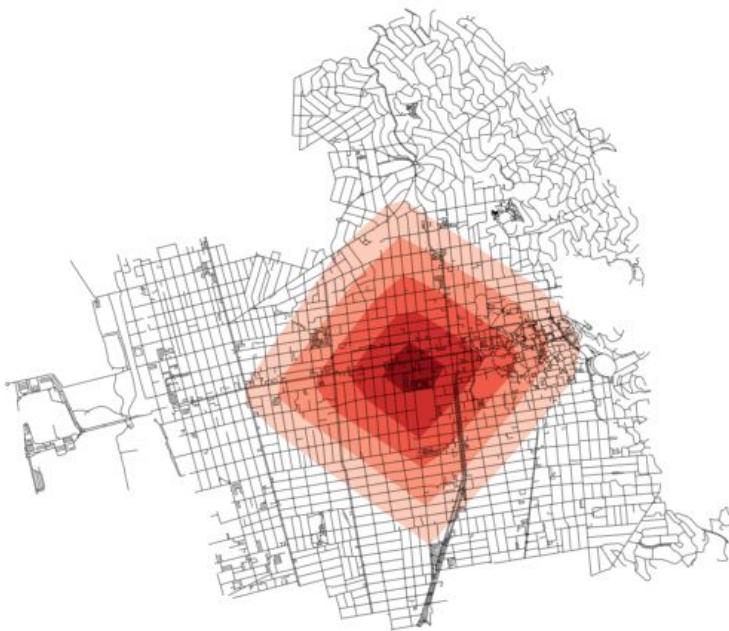
Announcements!

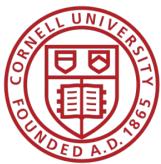
- Upcoming prelim on Tuesday!
- Review session Saturday
- Shares vs. counts – normalization is important!





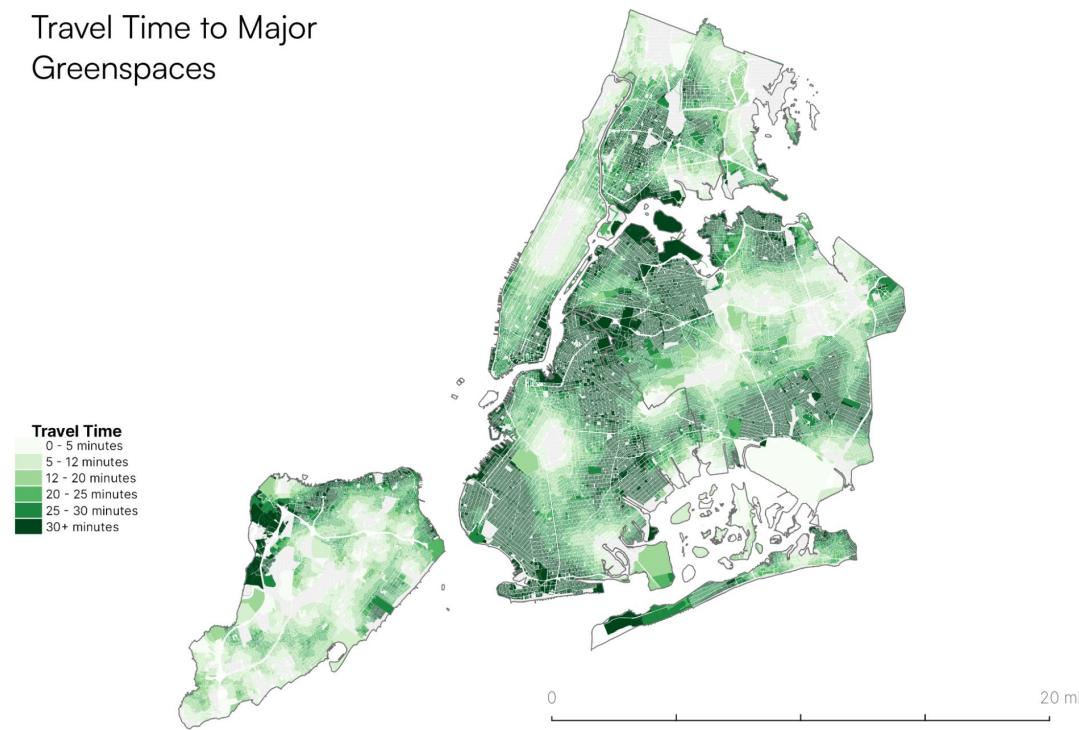
Mar 21: Isochrones!





Mar 21: Walk to a Park Initiative

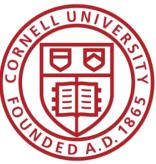
Travel Time to Major
Greenspaces



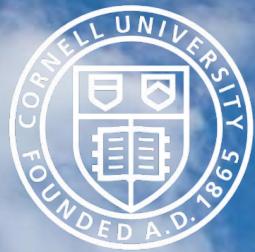


Mar 21: OSMnx

- Python package for working with street networks
- Uses Openstreetmap
- Can download and model walking, driving, or biking networks for different places



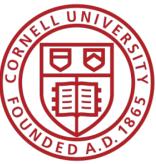
Let's start executing Week9.ipynb together!



PUBPOL 2130/ INFO 3130

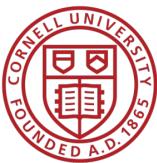
Lab 10





Announcements!

- No homework for spring break!

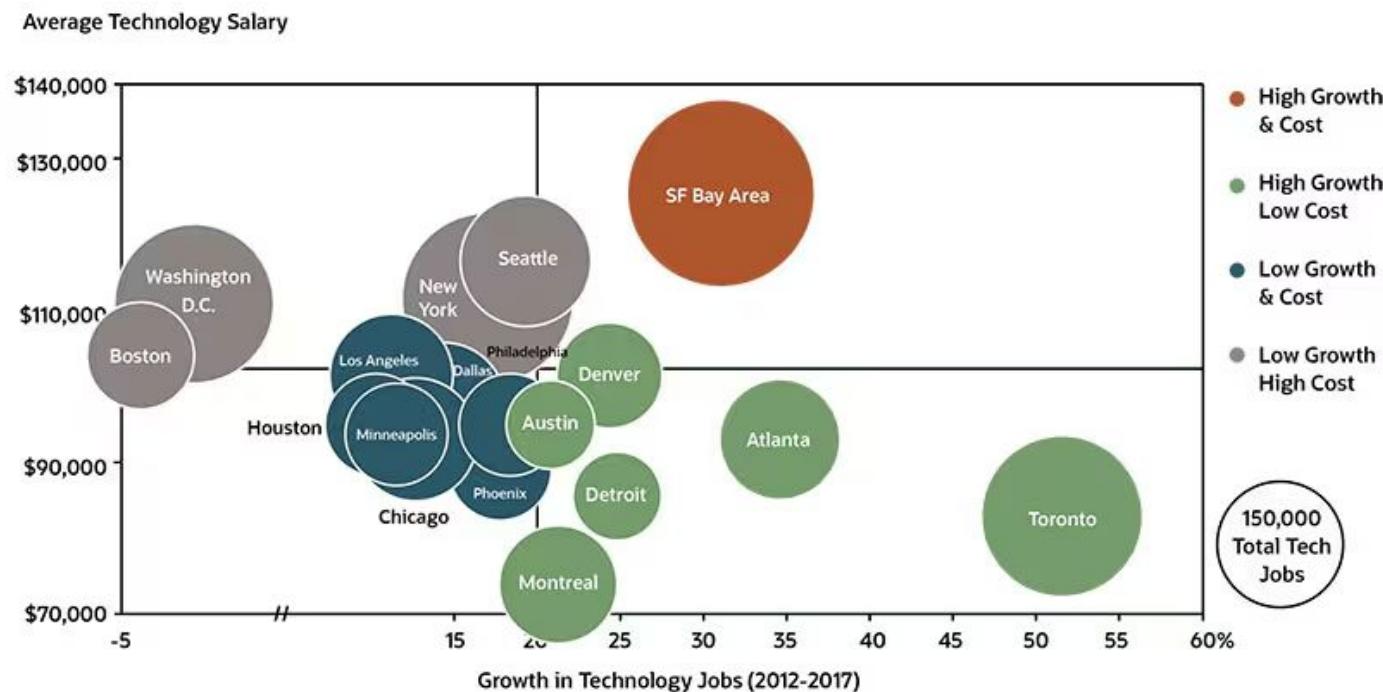


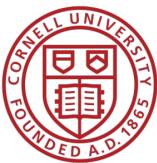
Mar 28: Bubble Charts

- Bubble size indicates a metric

Technology Markets In North America

While San Francisco is still the largest market for technology jobs, Toronto is the fastest growing and it offers companies significantly lower wage costs.

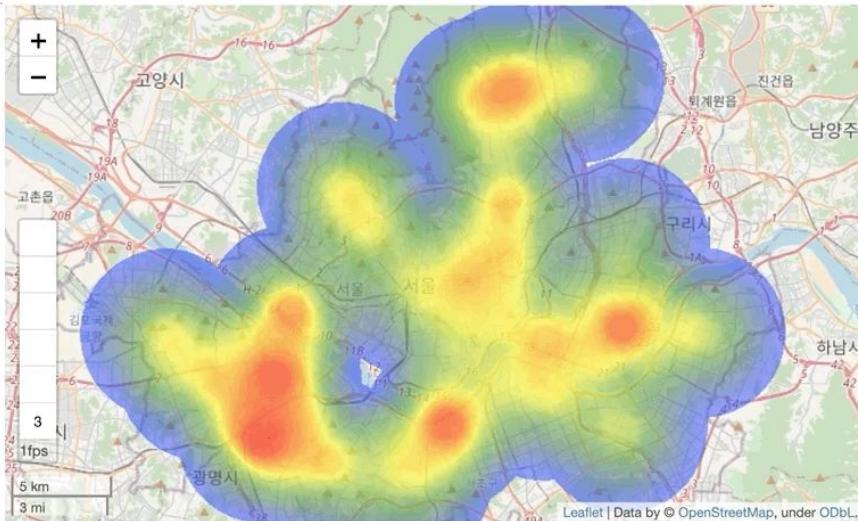




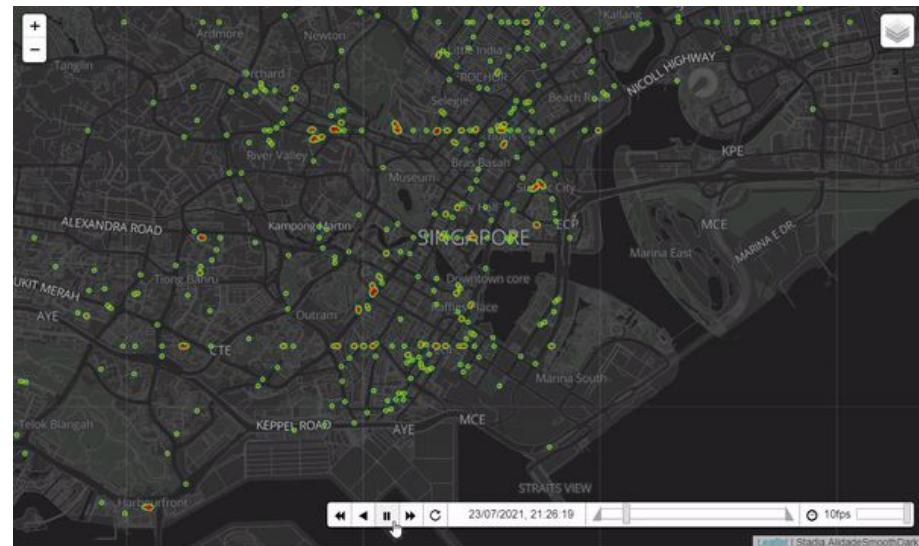
Mar 28: Folium Heat Maps with Time

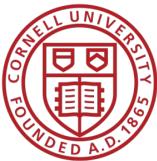
- Visualize how event intensity changes over time

Pollution in Seoul



Taxis in Singapore



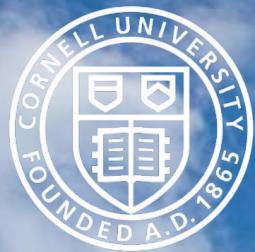


Mar 28: Today's data

- Vera Institute of Justice ([GitHub repo](#))
 - Incarcerations data for all states
- San Francisco's Open Data Portal ([DataSF](#))
- [Chicago Data Portal](#)



Let's start executing Week10.ipynb together!



PUBPOL 2130/ INFO 3130

Lab 11





Apr 11: Redlining

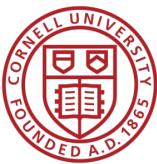
Discriminatory practice of denying mortgages, insurance, or investment in certain neighborhoods—primarily Black or immigrant communities—based on demographics.

- Institutionalized by the Home Owners' Loan Corporation (HOLC)



Apr 11: Home Owners' Loan Corporation

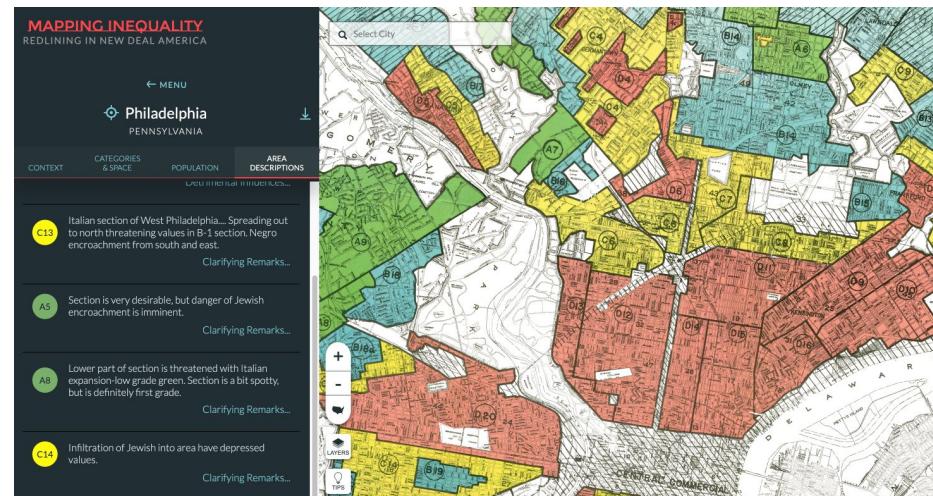
- Created in 1933 as part of the New Deal during the Great Depression
- Purpose
 - Refinance mortgages to prevent foreclosure
 - Stabilize housing markets
 - Expand home ownership
- How
 - HOLC assigned grades to areas



Apr 11: HOLC Grading

- Labeled Black, immigrant, and non-white neighborhoods as “blighted” or “hazardous”
- Banks and insurers refused loans to residents in these areas
- Goal: Preserve white-only, “desirable” neighborhoods

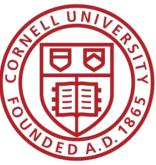
Grade	Color	Label
A	Green	“Best”
B	Blue	“Still Desirable”
C	Yellow	“Definitely Declining”
D	Red	“Hazardous”





Apr 11: Data

- Historic HOLC Maps from [Mapping Inequality Project](#)
- Historic Redlining Indicator (HRI) from [ICPSR \(2020 release\)](#)
- Social Vulnerability Index from CDC ([documentation](#))
- ACS 2022 Census Data



Let's start executing Week11.ipynb together!