

Recap of Unemployment Rates Internal Project

Fall 2022: Michael Yip, Susan Yang

This Semester's Timeline

Week Plan

Week 1: 09/19 - 09/30

- Introduction to project
- Brief review of Python

Week 2: 10/03 - 10/10

- Introduction to **NumPy**, **Pandas**

Week 3: 10/10 - 10/14

- Visualization using **Matplotlib**

Week 4: 10/22 - 10/30

- Brief Intro to **Seaborn**
- Basic **API call** to retrieve data

Week 5: 10/31 - 11/7

- **Regression: Linear & Logistic**
- **K-means**

Week 6: 11/8 - 11/15

- **Support Vector Machines**

NO MEETING WEEK

- Enjoy your break!

Week 8: 11/28 - 12/2

- Design your own final project!
- Prepare for final deliverable

Week 9: 12/2

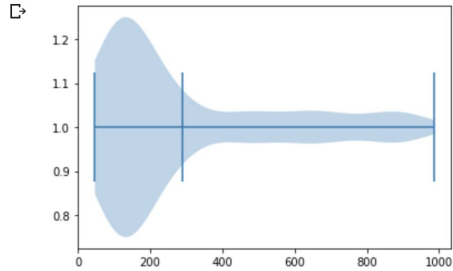
- **Final Presentation**



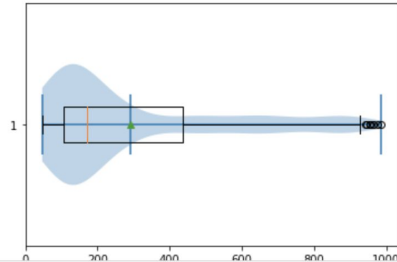
Lab highlights

Week 3 Matplotlib

```
plt.violinplot(stock["Price"], vert = False, showmeans = True)  
plt.show()
```



```
[ ] plt.boxplot(x = stock["Price"], vert = False, showmeans = True)  
plt.violinplot(stock["Price"], vert = False, showmeans = True)  
plt.show()
```



We added some more methods of **beautifying** the graphs, dealing with **overplotting**, labels **overlapping**, and **log&sqrt** plot, etc. We also introduced more plots types including **Stack plot** and **Stem plot**.



Week 4 API calls

```
def pull_data (series_id):  
    # Specify json as content type to return  
    headers = {'Content-type': 'application/json'}  
  
    # Submit the list of series as data  
    data = json.dumps({"seriesid": [series_id], "startyear": "2003", "endyear": "2022", "registrationkey":  
  
    # Post request for the data  
    p = requests.post('https://api.bls.gov/publicAPI/v2/timeseries/data/', data=data, headers=headers)  
    # change p into json type for further querying  
    json_data = json.loads(p.text)  
  
    # query data  
    results = json_data["Results"][0]["series"][0]["data"]  
    return results
```

For `json_data["Results"][0]["series"][0]["data"]`, feel free to run the `pull_data` lines by lines and print out `json_data` to see we query in this way

Further query data

```
[ ] df = [pd.DataFrame() for i in range(20)] # range(20) because we have 20 survey IDs selected  
data = pd.DataFrame() # create an empty DataFrame  
for j in range(20):  
    results = pull_data(series_list[j]) # api call for each survey ID  
    for i in np.arange(0, len(results), 1): # might need to change based on years you choose  
        year = results[i]["year"]  
        month = results[i]["period"][1:] # this is to delete the first letter "M" in every month data  
        month_name = results[i]["periodName"]  
        date_list = year + "-" + month + "-01"
```

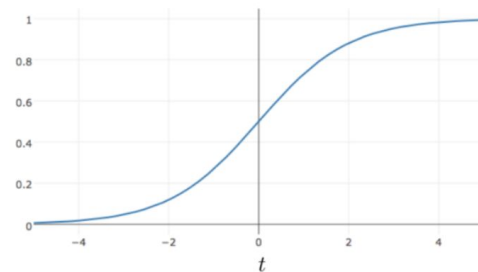
We talked about what API is, why retrieving data from API, and how to query a retrieved JSON file. With sample code for US Bureau of Labor Statistics and another example for Youtube API provided.



Week 6 Logistic Regression

Properties of a Logistic Function

The logistic function is a type of **sigmoid**, a function with the following properties.



Importance of Classification

Our motivation for performing logistic regression was to predict **categorical labels**. Specifically, we were looking to perform **binary classification**, i.e. classification where our outputs are 1 or 0.

win or lose

disease or no disease

spam or ham

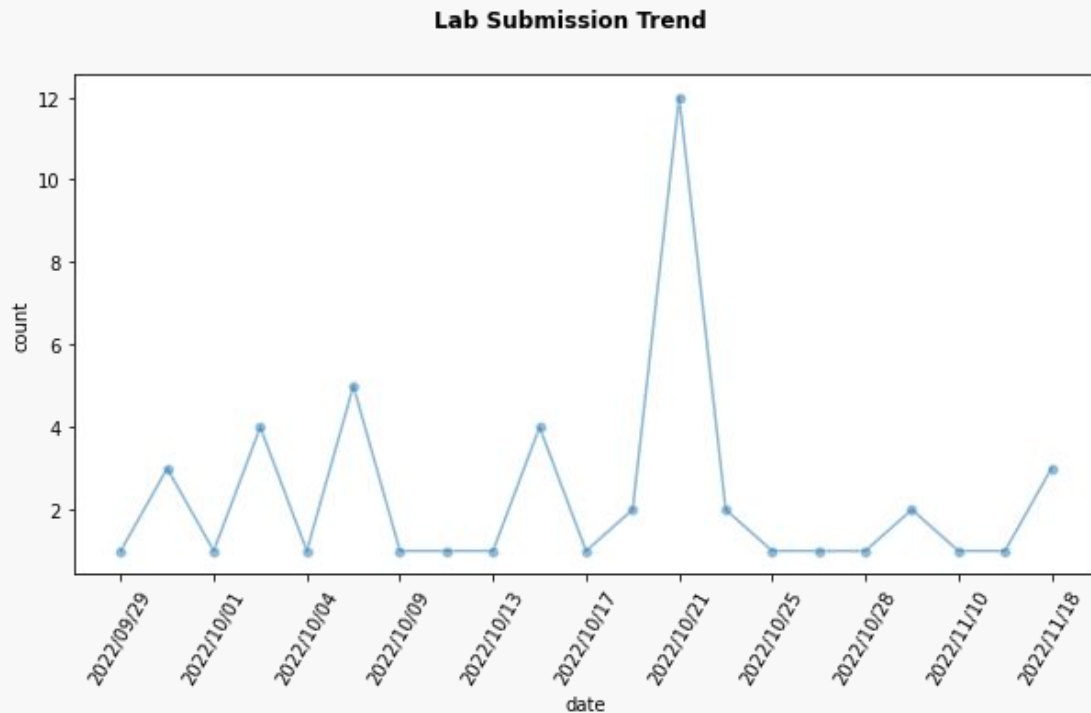
However, the **output of logistic regression is a continuous value** in the range $[0, 1]$, which we interpret as a probability – specifically, $P(Y = 1|x)$.

In order to **classify** – that is, to predict a 1 or 0 – we pair our logistic model with a **decision rule**, or **threshold**.



Feedback for members
Thank you for the great job!

Submission and Completeness



In total we received about 50 submissions. Most people finished labs on NumPy, Pandas, Matplotlib, and Seaborn.

Pretty satisfying results!



Submission and Completeness

Members Feedback and Attitude

Only 23 Feedbacks collected and mostly on first two weeks



The feedback from first two weeks are mostly **positive** with **meaningful** feedback/advice, members are happy to have a **zoom** option. Though we still receive few abstention of final projects.



Work from members

Final Group Presentation

Group	Order	Member
1	1	Janelle Correa, Jiayin Lin, Jennifer Ly
2	2	Daniel Hwang, William Lee, Bianca Rein Del Rosario
3	3	Nadine Ratinho, Meng-Han Wu
4	5	Arjun Balaji, Nick Erwin Zhiyang Chia
5	4	Yesenia Morales, Sinead McCaffery

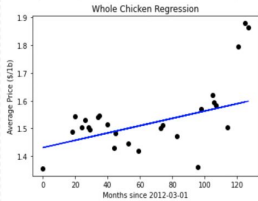
We have five groups and
all groups presented in
the end!!!

**Big shout out to all of
our members!**

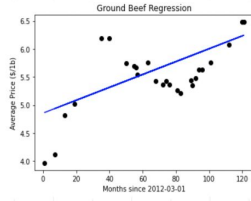
Goodjob 👍



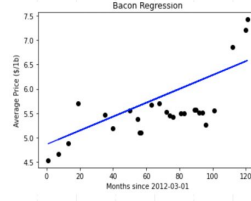
Some highlights



$$y = 0.0013x + 1.4312$$



$$y = 0.0115x + 4.8552$$

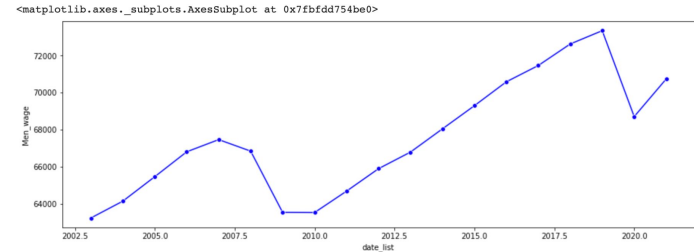


$$y = 0.0142x + 4.8608$$

The coefficient and thus the rate of growth in price for **chicken** is a **full order of magnitude** lower than its beef and pork counterparts

This 2nd line graph compares the wages of men over the years 2003-2021. The wages now range from 64,000-72,000. We can see a fairly similar shaped line, still with the slight dips in the middle and the end. By looking at the line graph the average wages of men are about half of the total wages which should be expected.

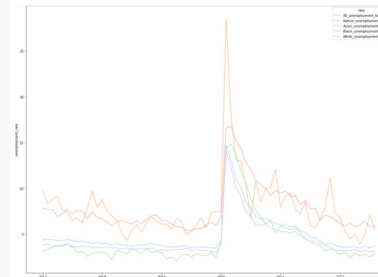
```
[ ] plt.figure(figsize=(15,5))
sns.lineplot(x = "date_list", y = "Men_wage", data = wage_df, marker = 'o', color = "blue")
```



Conclusion

- Okun's law and coefficient show that one percent drop of unemployment gap will cause 1.667 percent of output gap drop in the United State
- We can calculate any country or place Okun coefficient through linear regression
- We can now use the linear regression to predict the output or GNP of the given year by using the unemployment rate.

How did unemployment rates change?



Native Unemployment rate has the highest increase and reached the highest value of 29%.

All other races increased as well, but only to a maximum of 15-16%.

```
# finding max value
max_pos = plotting_data['unemployment_rate'].idxmax()
plotting_data.iloc[max_pos]
```

unemployment_rate	rate	year
234	28.6	Native_unemployment_rate 2020-04-01

Presentation links:

Group 1: <https://colab.research.google.com/drive/1xGPuXtW79-hj9qkM2WK5aiRuKJXfP0Op?usp=sharing>
https://docs.google.com/presentation/d/1UH7jM27FaUsvM1bpj_Ei2BjEG38LVp8d5JB6FbnWfQw/edit#slide=id.g1a6389126ac_0_49

Group 2:

<https://docs.google.com/presentation/d/1ab8ikVOYLpH-q0796Qx90M9Fp1LrFeme60BeRZC7kmY/edit?usp=sharing>

Group 3:

https://docs.google.com/presentation/d/1l4j_MBctDbKnmeuGwVULnpSnkRdLC1Tb9W7KeZRJ_4Y/edit#slide=id.g8ea72f4a77_6_125

Group 4:

https://docs.google.com/presentation/d/1Z1mZnMecSwL5UE84okfCS8qhrnEfShRM_86pCeqbXU/edit?usp=sharing

Group 5: <https://colab.research.google.com/drive/17yxPA3tHjzYzGNnXx7BlIsAgRfBu6hu5?usp=sharing>

Presentation from project member: Jiayin Lin

https://docs.google.com/presentation/d/1UH7jM27FaUsvM1bpj_Ei2BjEG38LVp8d5JB6FbnWfQw/edit#slide=id.g1a6389126ac_0_49

Thanks for coming!
Hope you enjoyed our
semester.