# Project 1: Data Analysis of Restaurants in Washington DC

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#### **Outline:**

A client wants to open a restaurant in Washington DC. What type of restaurant should they open and where in DC?

#### Our Analysis will answer **4 questions**:

- What is the most competitive restaurant type in each DC Quadrant?
- What is the average restaurant quality for restaurant types in each DC Quadrant?
- What are the best graded restaurants in DC? Do they reflect their demographics?
- Food trucks vs Restaurants, what to pick?

## What is the most competitive restaurant type in each DC Quadrant?

This data will help us find out which restaurant type may have been exhausted in DC and which restaurant types have opportunity to grow



## What is the average/best restaurant quality for restaurant types in each DC Quadrant?

This data will help identify if there is a particular restaurant type that may be associated with lower food quality. If so, this may present an opportunity to stand out.



Do ethnic groups in DC impact restaurant types in each DC Quadrant?

With this question, we can identify any trends that may be related to DC's ethnicity. This will help us understand if a particular group of people impact the amount of a restaurant type in DC. If there is a trend, the new restaurant may have to serve those groups of people in someway with the new restaurant.



#### Understanding the problem: Methods used

#### What is needed?

- Total number of restaurants in DC
- Ratings of
   Restaurants type
   ratings
- Description of ethnic groups in DC

#### How do we find that?

- Find a good data source
- Determine quadrants
- Zip codes/Coordinates

#### Treatment + Analysis

- Loading data
- Combining fata
- Plot data
- Analyse

Sources of Informations

## Retrieving data: Food

#### Yelp



#### Pros:

- Detailed
- Developer friendly
- Includes both
   Api and CSV
   files

#### Cons:

- No format
- Only allows 1000 data maximum per entry

#### **United States Census**



# Retrieving data: Population

#### Pros:

- Very detailed
- Trusted

#### Cons:

- Cryptic Api calls
- Time consuming
- Does not include good coordinates for lower state level

#### DC Health



# Retrieving data: Population

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- CSV files included

## Processing data

#### **Loading data: Ison and CSV**

Json

```
#Call
data = json.loads(req.text)
print(json.dumps(data, indent=4))
    "businesses": [
            "id": "6lBhCjv-imBpa3IYKVSCCA",
            "alias": "maxs-kosher-grill-washington",
            "name": "Max's Kosher Grill",
            "image url": "https://s3-media4.fl.yelpcdn.com/bphoto/YF130-FV2LuRLy3vueGaCA/o.jpg",
            "is closed": false,
            "url": "https://www.yelp.com/biz/maxs-kosher-grill-washington?adjust creative=1m4MG5a37UVgnQoLAVWITw&utm campaign
=yelp api v3&utm medium=api v3 business search&utm source=1m4MG5a37UVgnQoLAVWITw",
            "review count": 5,
            "categories": [
                    "alias": "kosher",
                    "title": "Kosher"
                    "alias": "falafel",
                    "title": "Falafel"
```

#### Loading data: Json and CSV

**CSV** 

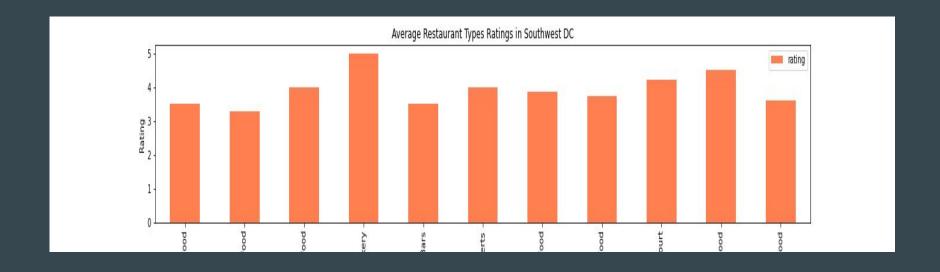
```
#Southwest Ethnicity Data
zip_20024_Hisp = pd.read_csv('DC_Health_Data/Hispanic_Latino_Population_by_Race_Zip_Code_20024.csv')
zip_20024_Non = pd.read_csv('DC_Health_Data/Non_Hispanic_Latino_by_Race_Zip_Code_20024.csv')
zip_20032_Hisp = pd.read_csv('DC_Health_Data/Hispanic_Latino_Population_by_Race_Zip_Code_20032.csv')
zip_20032_Non = pd.read_csv('DC_Health_Data/Non_Hispanic_Latino_by_Race_Zip_Code_20032.csv')

#Southeast Data
zip_20003_Hisp = pd.read_csv('DC_Health_Data/Hispanic_Latino_Population_by_Race_Zip_Code_20003.csv')
zip_20003_Non = pd.read_csv('DC_Health_Data/Non_Hispanic_Latino_by_Race_Zip_Code_20003.csv')
Ward_8_Hisp = pd.read_csv('DC_Health_Data/Hispanic_Latino_Population_by_Race_Ward_Ward_8.csv')
Ward_8_Non = pd.read_csv('DC_Health_Data/Non_Hispanic_Latino_by_Race_Ward_Ward_8.csv')
```

#### Framing the data: Pandas

	name	category	latitude	longitude	rating	address	State	Zipcode	initial
0	Surfside	Tacos	38.943537	-77.077791	3.5	4200 Wisconsin Ave NW	DC	20015	NW
2	Tabla	Georgian	38.931088	-77.023394	5.0	3227 Georgia Ave NW	DC	20010	NW
3	Viet Chopsticks	Vietnamese	38.945732	-77.064860	4.5	4304 Connecticut Ave NW	DC	20008	NW
5	Rosemary Bistro Cafe	French	38.955090	-77.070230	5.0	5010 Connecticut Ave NW	DC	20008	NW
6	Uptown Market	Grocery	38.947138	-77.064944	4.5	4465 Connecticut Ave NW	DC	20008	NW
	652	750.	932		9.07	829	7.2		113
934	Po Boy Jim	Bars	38.916570	-77.024275	3.5	1934 9th St NW	DC	20001	NW
936	Pi Pizzeria	Pizza	38.897154	-77.024709	4.0	910 F St NW	DC	20004	NW
940	LiLLiES Restaurant & Bar	Mediterranean	38.928427	-77.054125	3.5	2915 Connecticut Ave NW	DC	20008	NW
941	Angelico Pizzeria	Pizza	38.948699	-77.079876	3.5	4529 Wisconsin Ave NW	DC	20016	NW
947	Cafe Saint-Ex	American (New)	38.915423	-77.031673	3.5	1847 14th St NW	DC	20009	NW

## **Plotting Data:**



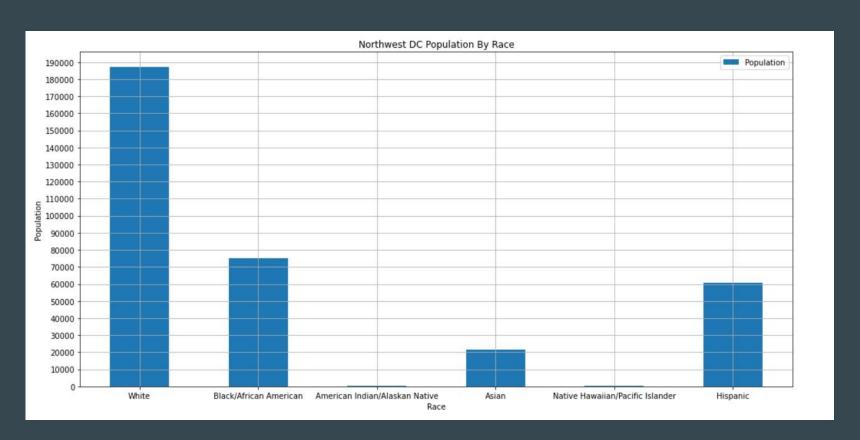
#### Analysis: Pandas, logic and a lot of courage

100.440.00
rating
count 20.000000
mean 4.078740
std 0.275229
min 3.500000
<b>25</b> % 3.948005
50% 4.067254
<b>75</b> % 4.220395
max 4.600000

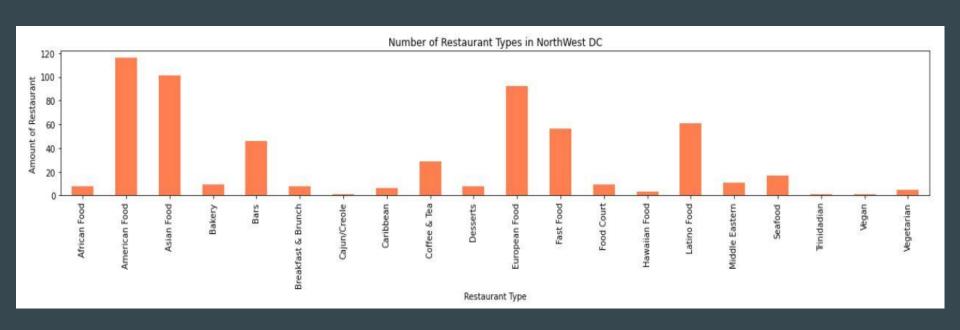
American Food	169
Asian Food	142
European Food	112
Fast Food	91
Latino Food	77
Bars	61
Coffee & Tea	38
Seafood	33
Food Court	24
African Food	19
Breakfast & Brunch	15
Bakery	13
Desserts	13
Middle Eastern	12
Caribbean	10
Vegetarian	5
Hawaiian Food	5
Vegan	4
Cajun/Creole	3
Trinidadian	1
Name: category, dtype:	int64

## Results

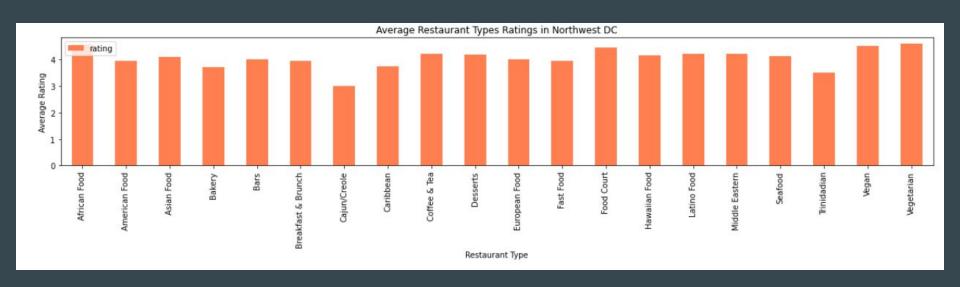
### Northwest



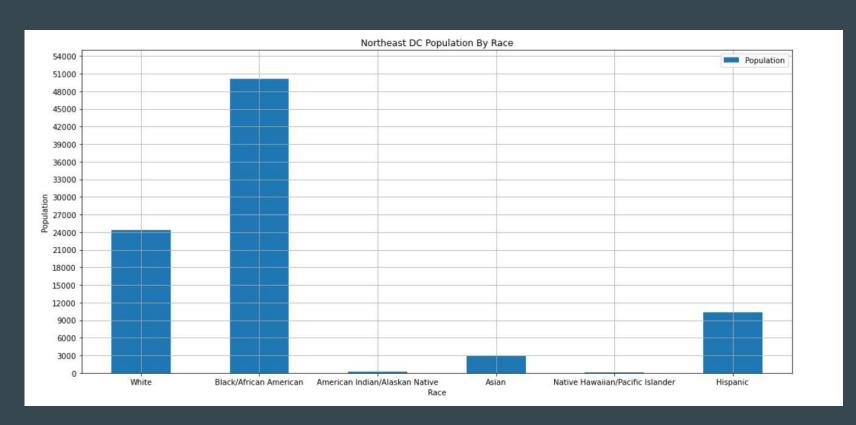
## Northwest



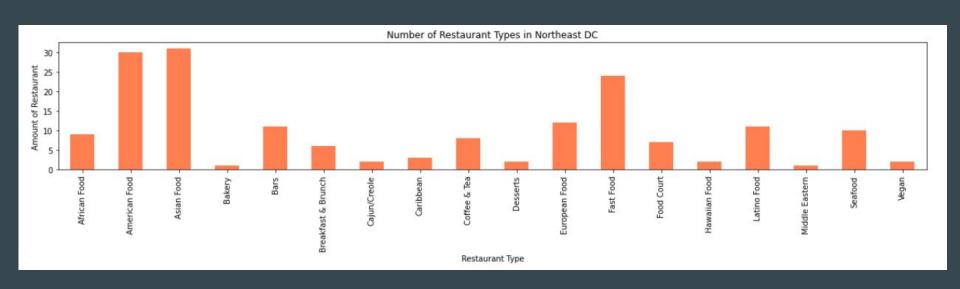
## Northwest



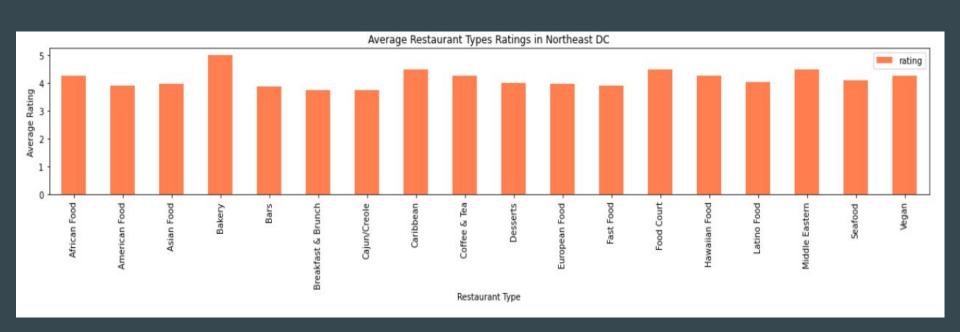
### Northeast



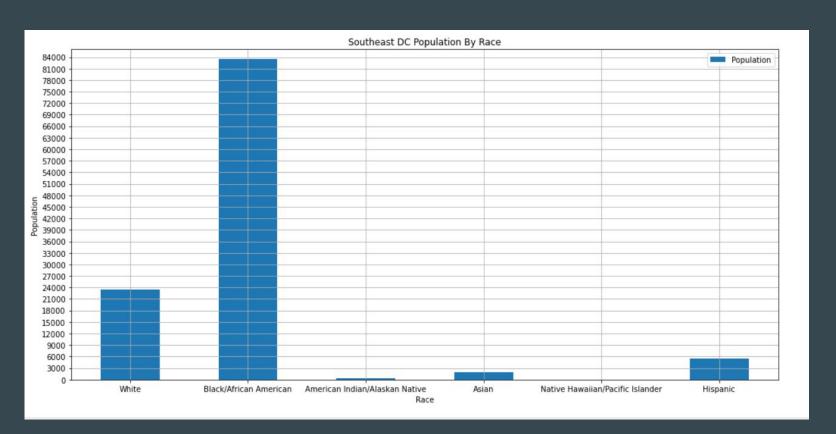
## Northeast



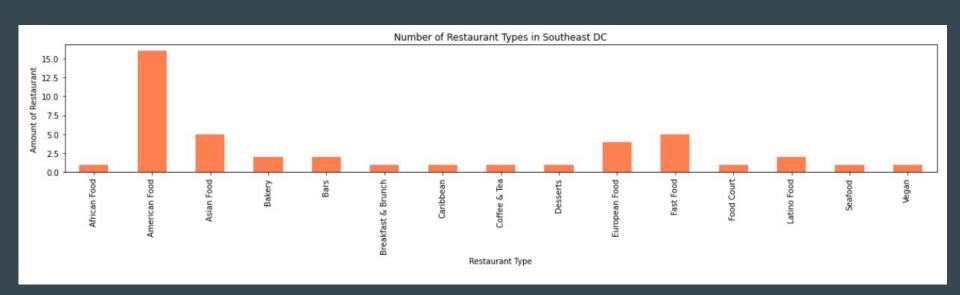
### Northeast



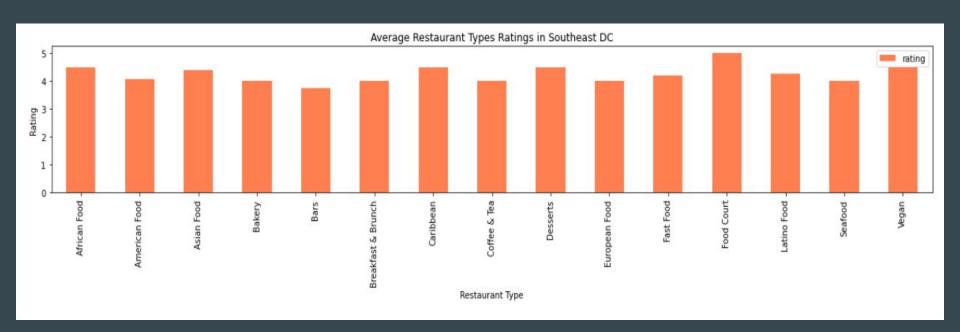
### Southeast



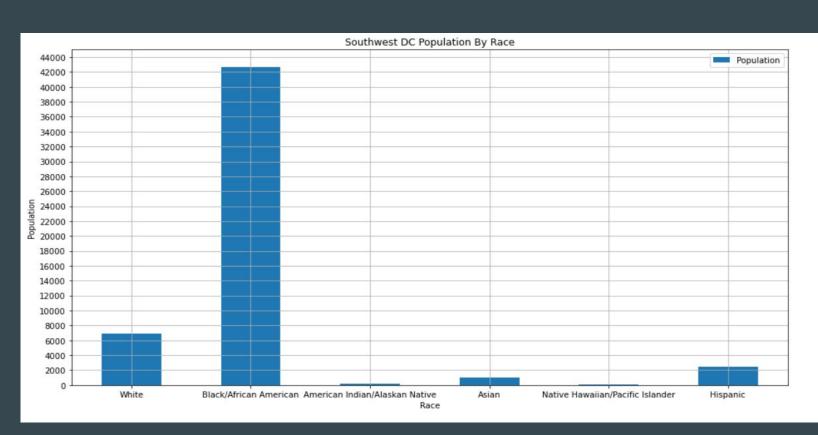
### Southeast



## Southeast



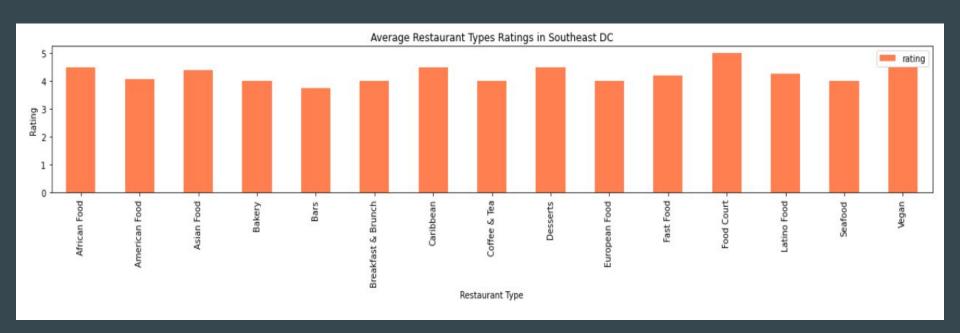
### Southwest



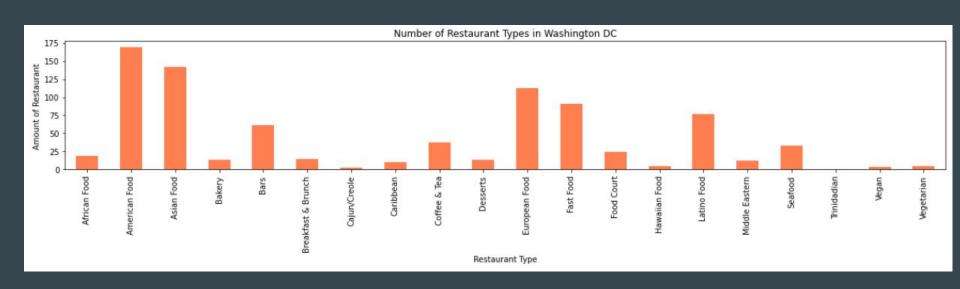
## Southwest



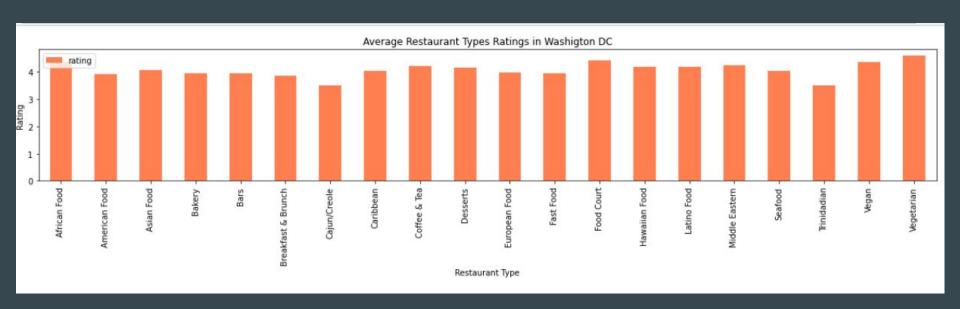
### Southwest



## Washington DC



## **Washington DC**



#### Trend analysis

#### **Findings**

- Most competitive restaurant types across each quadrants are American and Asian restaurants.
- Least competitive market for restaurant types are foreign and niche restaurants.
- All restaurant types have an average rating of **3 stars** or more but the restaurant types in Southwest DC have lower average ratings than any other quadrant.
- Despite the competitive nature of Asian restaurants, there are few Asians living in DC compared to White or African American. African Americans are majority in every DC quadrant except in Northwest DC which is majority white. In Northwest DC, there were more european restaurants than any other quadrant.



Food

Food

4

3.93

91

Fast

Food

Food

#### **Food Trucks**

- Food trucks do not have a fixed location
- Can sell everywhere and attract tourists!
- Higher average rating than all restaurants with a 4.34 out of 5!



#### Insights:

There is more research to be done to build a stronger recommendation.

We didn't anticipate African Americans being the majority in every quadrant of DC except

Northwest

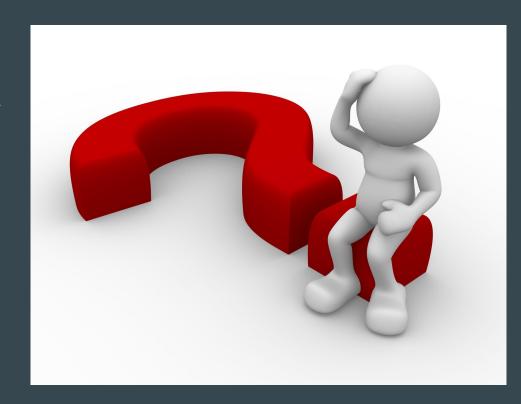
An interesting find is that Asian food is just as competitive as American Restaurants

Another find is that Southeast DC is the only location without a much diversity in restaurant types but yet as the highest amount of African americans.



#### Problems encountered:

- The Ethnicity Data was only limited to Zip codes and Wards which made it difficult to get a very accurate representation of the four quadrants. Some wards and zip code spanned across multiple quadrants.
- Troubleshooting our code was quite a challenge!



## Questions?

## Thank you for listening!

