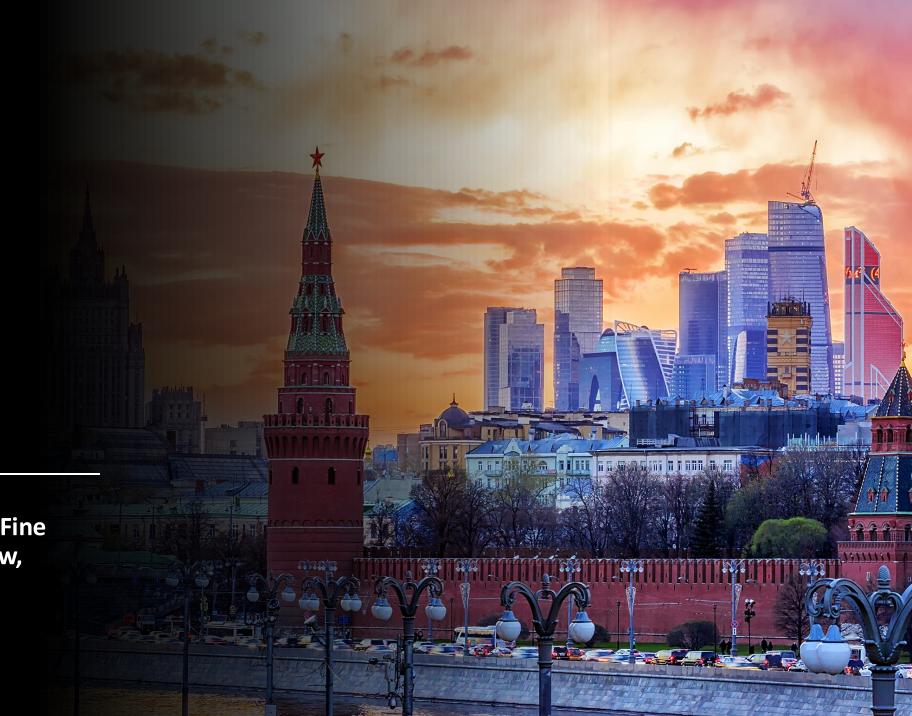
# IBM Data Science Capstone

Construction of a European Fine Dining Restaurant in Moscow, Russia





### Fine Dining: A Definition

- Fine dining restaurants are a very high class of restaurants which feature prominent décor made of high-quality materials, high prices on the food, and an etiquette that must be followed by staff and visitors alike.
- Fine dining has been around since the late 18th century and is very popular in Western Europe and parts of North America.

### **Business Problem Definition**



- This project will focus on a restaurateur trying to open a high-end fine dining restaurant in Moscow, Russia.
- The city districts will be clustered into groups
- The best group should be one that has many gastronomical venues and plenty of sources of fresh ingredients, thus maximizing revenue and minimizing costs on food

### **Data Sources**

- Our source of data is the following Wikipedia
  - page: <a href="https://en.wikipedia.org/wiki/Cat">https://en.wikipedia.org/wiki/Cat</a> egory:Districts of Moscow
- We will also use geocoder to get geocoordinate data and Foursquare to get venue information







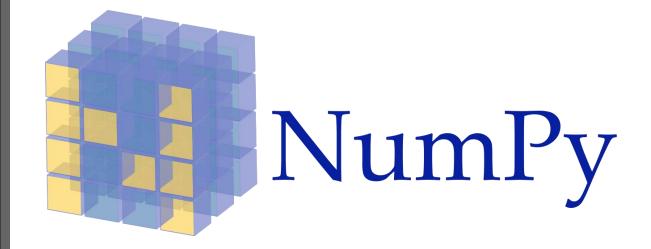
**FOURSQUARE** 

### Data Processing Tools

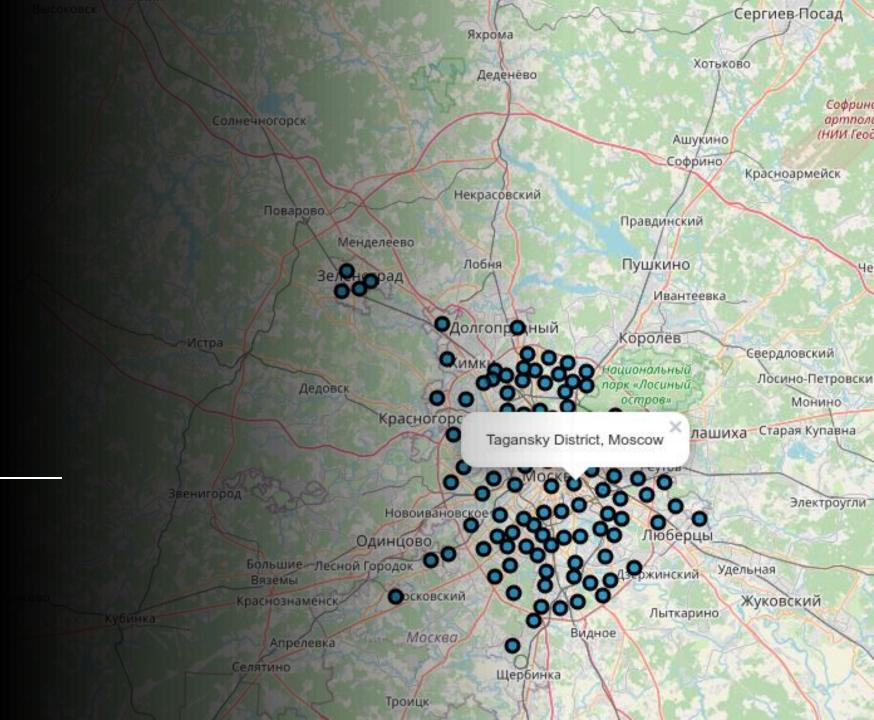
- We will use Pandas to parse and store the data
- NumPy will be used to perform some operations on the data
- We will use Scikit-Learn to analyse it by using the K-means algorithm







# Districts of Moscow



#### Ивантеевка Черкизово Фрязин Долгопру вій Королёв Шёлково Мытищи Химки Национальный парк «Лосиный остров» Красногорск Востаный Балашиха Clustered **Districts** Москва Новоивановское Люберцы The result of clustering the districts Котельники Красково of Moscow into 11 Different districts Малаховка Дзержинский Октябрьский админи ративный Развилка Лыткарино Коммунарка Видное

## Clustered District DataFrames

Some resulting DataFrames after clustering

# Cluster 2 | cluster 1 | enoscow\_final\_df.loc(moscow\_final\_df['label'] | == 1, moscow\_final\_df.columns[list(range(2, moscow\_final\_df.shape[1]))]]

1st Nost | 2nd Most | 2nd Mo

[407] # Cluster |

cluster\_8 = moscow\_final\_df.loc[moscow\_final\_df['label'] == 7, moscow\_final\_df.columns[list(range(2, moscow\_final\_df.shape[1]))]]
cluster\_8[0:20]

[407]:		1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue	11th Most Common Venue	Common	13th Most Common Venue	14th Most Common Venue	15th Most Common Venue	label	District
	11	Cosmetics Shop	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	7	Bogorodskoye District
	31	Cosmetics Shop	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm		Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	7	Khoroshyovsky District
	63	Cosmetics Shop	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm		Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	7	Novo- Peredelking District
	113	Cosmetics Shop	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	7	Yasenevo District
	118	Cosmetics Shop	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm		Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	7	Zapadnoye Degunino District

[413]: # Cluster 1

cluster\_0 = moscow\_final\_df.loc(moscow\_final\_df('label') == 0, moscow\_final\_df.columns[list(range(2, moscow\_final\_df.shape[1])))]
cluster\_0[0:5]

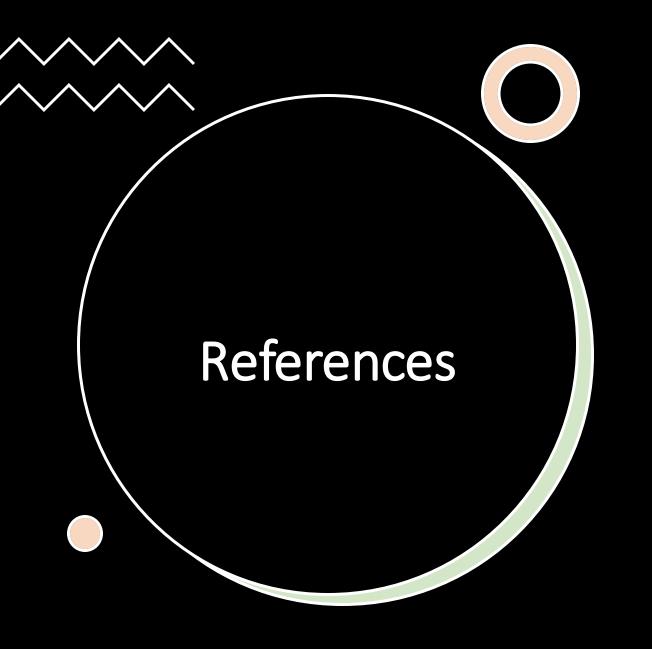
[413]:		1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue	11th Most Common Venue	12th Most Common Venue	13th Most Common Venue	14th Most Common Venue	15th Most Common Venue	label	District
	1	Health Food Store	Zoo Exhibit	Film Studio	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Fish Market	Escape Room	Flower Shop		Food Court	Forest	Fountain	0	Akademichesky District
	2	Liquor Store	Film Studio	Event Space	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Fish Market	Frozen Yogurt Shop	Flower Shop	Food & Drink Shop	Food Court	Forest	Fountain	0	Alexeyevsky District
	4	Convenience Store	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Escape Room	Food & Drink Shop	Food Court	Forest	Fountain	0	Babushkinsky District
	5	Pet Store	Film Studio	Event Space	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Fish Market	Czech Restaurant	Flower Shop	Food & Drink Shop	Food Court	Forest	Fountain	0	Basmanny District
	6	Salon / Barbershop	Zoo Exhibit	Fast Food Restaurant	Event Space	Exhibit	Fabric Shop	Farm	Farmers Market	Film Studio	Electronics Store	Fish Market	Flower Shop	Food & Drink Shop	Food Court	Forest	0	Begovoy District

	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue	11th Most Common Venue	12th Most Common Venue	13th Most Common Venue	14th Most Common Venue	15th Most Common Venue	label	District
14	Bakery	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower	Food & Drink Shop	Food Court	Forest	Fountain	French Restaurant	10	Chertanovo Severnoye District
27	Bakery	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Food & Drink Shop	Food Court	Forest	Fountain	French Restaurant	10	Ivanovskoye District
30	Bakery	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Food & Drink Shop	Food Court	Forest	Fountain	French Restaurant	10	Khoroshyovo- Mnyovniki District
53	Bakery	Zoo Exhibit	Fish Market	Exhibit	Fabric Shop	Farm	Farmers Market	Fast Food Restaurant	Film Studio	Flower Shop	Food & Drink Shop	Food Court	Forest	Fountain	French Restaurant	10	Metrogorodok District



### Conclusion

 Based on the above results, the best cluster of districts would be CLUSTER NUMBER 2 due to a high amount of expensive sushi restaurants that would guarantee wealthy clientele as well an abundance of food sources such as farms



- Pandas Documentation: https://pandas.pydata.org/pandasdocs/stable/index.html
- Scikit Learn Documentation:
   https://scikit learn.org/stable/auto\_examples/clust
   er/plot\_kmeans\_silhouette\_analysis.h
   tml
- Silhouette Analysis: https://scikitlearn.org/stable/auto\_examples/clust er/plot\_kmeans\_silhouette\_analysis.h tml



Thank you!