

# Clustering Paris and Marseille districts: restaurant location

Wojciech Stach  
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# Outline

- ▶ Background
- ▶ Problem Statement
- ▶ Data
- ▶ Methods
- ▶ Results
- ▶ Conclusions

# Background

- ▶ Business location can have huge impact on profitability
  - ▶ Especially important for restaurants as customers often use proximity as a very important factor in making decisions where to eat
  - ▶ Knowledge of neighborhoods is essential to shape menu and service to meet clients' needs
- ▶ This project focuses on two largest cities in France: Paris and Marseille
  - ▶ 20 districts (arrondissements) in Paris
  - ▶ 16 districts (arrondissements) in Marseille

# Problem Statement and Audience

## ► Problem

- Perform districts analysis in Paris and Marseille to identify similar and dissimilar districts in terms of types of venues

## ► Audience

- New/future business owners
  - to help finding a good district for their business
  - to choose the restaurant type and menu that best address the clients' needs
- Existing business owners
  - To help choosing location when expanding or moving their business to a different location

# Data

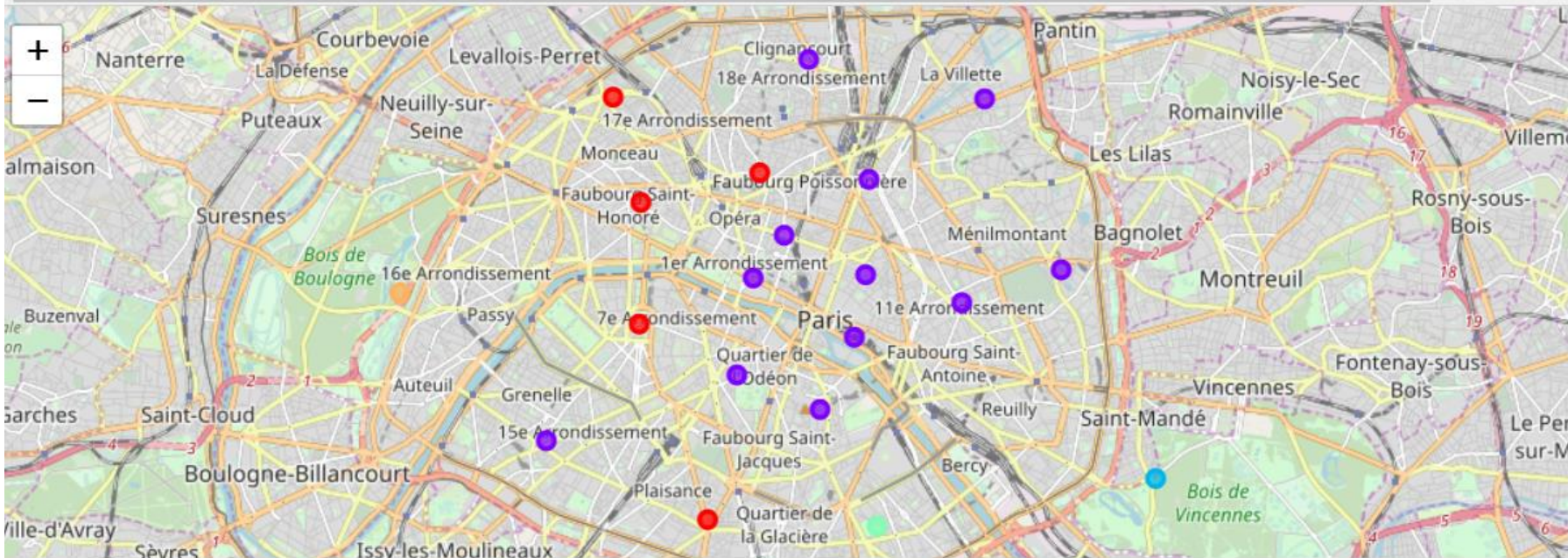
- ▶ The following datasets were used
  - ▶ <https://public.opendatasoft.com/explore/dataset/arrondissements-millesimes0/table/>
    - ▶ Data for 20 districts in Paris and 16 districts in Marseille
      - ▶ names
      - ▶ longitude and latitude
  - ▶ <https://developer.foursquare.com/>
    - ▶ community-based venues data

# Methods

- ▶ K-means clustering
  - ▶ Paris districts
  - ▶ Marseille districts
  - ▶ Paris & Marseille districts together
- ▶ Elbow method to find optimal number of clusters for each case
- ▶ Mark clusters on maps
- ▶ Find common characteristics of districts from the same clusters

# Results - Paris

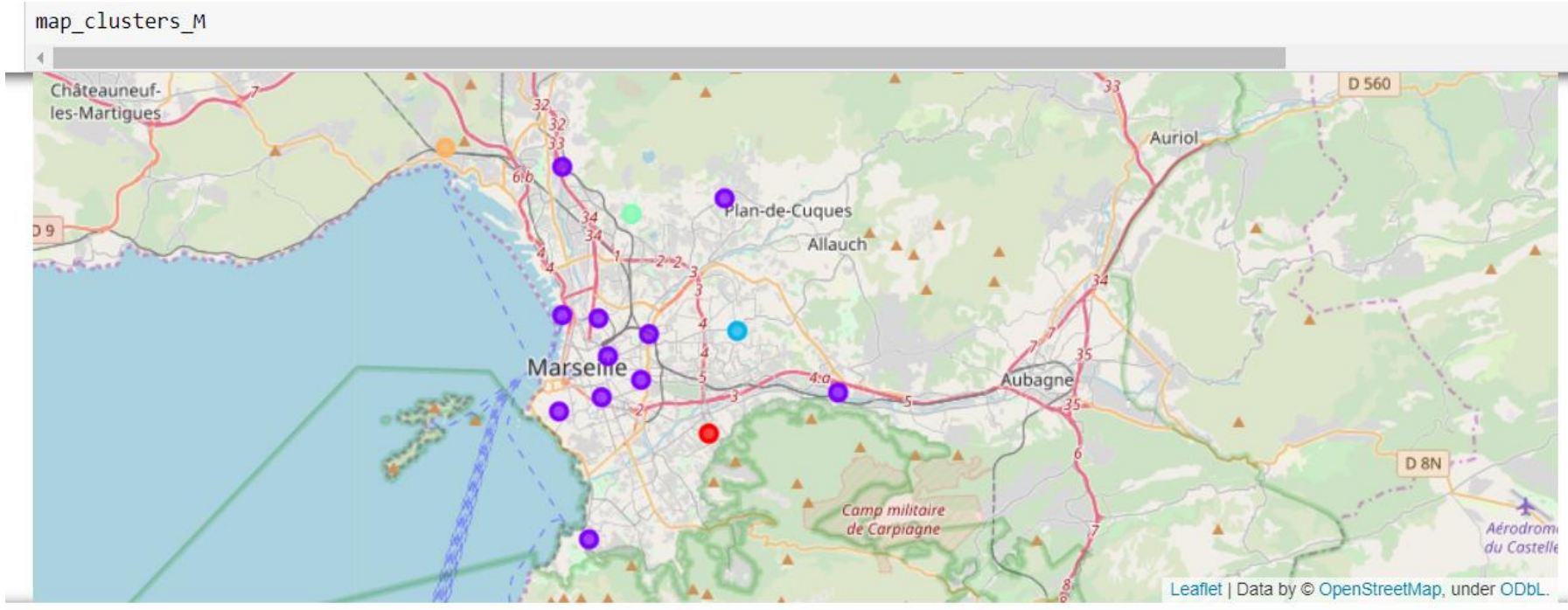
map\_clusters\_P



- Cluster 0 - red: Central-West districts of Paris (major venues: French restaurants and hotels)
- Cluster 1 - purple: Central-East districts of Paris and 15th district (that is in the West of Paris). Similar profile to Central-East district except for non-French restaurants showing quite high in terms of most popular venues
- Cluster 2 - blue: residential districts. Main venues include markets, stores, zoo
- Cluster 3 - green: multicultural districts, specifically including lots of Asian restaurants
- Cluster 4 - orange: upscale districts. Main venues include plaza, pool, lake.



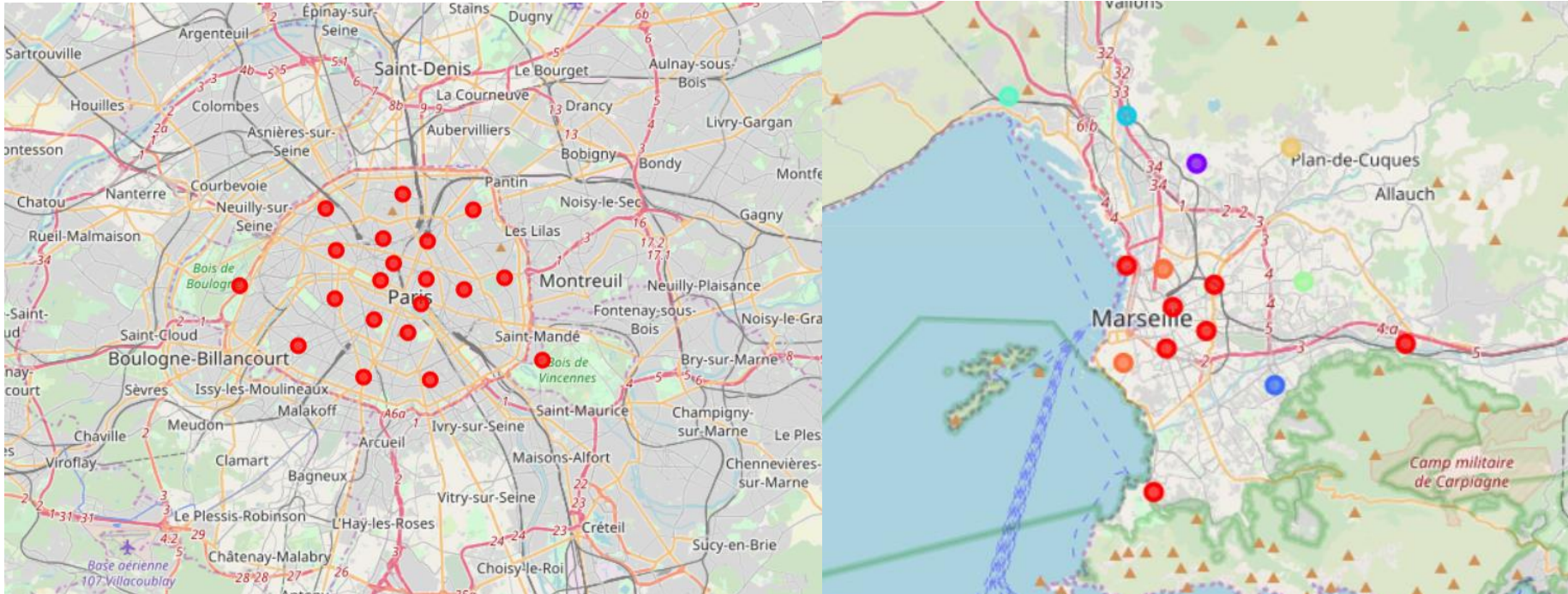
# Results - Marseille



- Cluster 1 - purple: Center area of Marseille. Most popular venues include hotels and restaurants. Interestingly, districts 8<sup>th</sup>, 11<sup>th</sup>, 13<sup>th</sup>, and 15<sup>th</sup> that are further from the center of Marseille are in this cluster, too
- Clusters 0, 2, 3, 4 - other colors: Other venues are more popular, including shops, tram stations, etc. More knowledge about the city is necessary to provide better distinctions between these clusters



# Results - combined Paris (left) & Marseille (right)



- Cluster 0 - red: Includes all districts from Paris and seven districts from Marseille (all of them from cluster 1 in Marseille clustering). This suggests that all Paris districts are quite like each other when compared to Marseille districts
- Clusters 1-7: Outskirts of Marseille. More knowledge about the city is necessary to provide better distinctions between these clusters

# Conclusions and Future Work

- ▶ Clustering is capable of grouping districts based on multiple features
  - ▶ Paris clusters match intuitive classification of central districts vs. outskirts
    - ▶ Interestingly, the 15<sup>th</sup> district show more similarities to the central districts
  - ▶ Marseille clusters show similar clustering in terms of central districts
    - ▶ 8<sup>th</sup>, 11<sup>th</sup>, 13<sup>th</sup>, and 15<sup>th</sup> districts are more similar to the central districts despite being geographically closer to the outskirts
  - ▶ Paris and Marseille clusters show that Paris clusters are very similar to each other when compared with Marseille
  - ▶ It would be very beneficial to have a local expert to provide more insights and explanations of the clustering results
- ▶ The clustering analyses can be helpful in making decisions about choosing the location for a restaurant
  - ▶ Looking at the most popular venues and similar districts can provide interesting information
- ▶ Future work may include adding population, socio-economic as well as demographic data to provide more comprehensive analysis