A decorative horizontal bar with a teal segment on the left and an orange segment on the right.

Clustering Techniques for On-Demand Transport Data: a case study

Pedro Afonso, Ana Alves

Pedro Afonso

a21240004@isec.pt¹



Biography



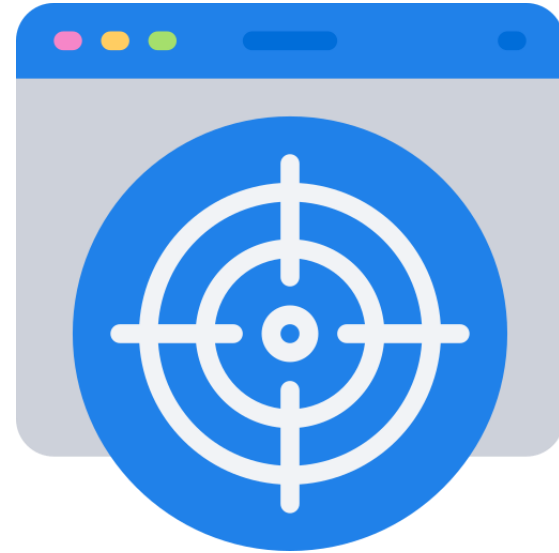
- Carlos Pedro Marques Afonso
- 23 years old
- Graduated in Computer Engineering from the Polytechnic Institute of Coimbra, more specifically at the Instituto Superior de Engenharia de Coimbra
- Finishing the Master's degree in Informatics and Systems, software development branch
- Computer Engineer at The Municipality of Estarreja



Table of contents

- Motivations
- Goals
- Related Work
- Clustering
 - K-Means
 - DBSCAN
- Comparison of Techniques
- Case Study
- Conclusions
- Future work

**Transport On Demand
has a great impact on
urban mobility**





Motivations

- Reduce the number of vehicles needed
- Reduce the amount of pollution needed
- Decrease the difficulties in arranging parking
- Increase the flow of transit routes



Goals

- Analyse existing clustering algorithms, namely K-Means and DBSCAN
- Choose the best one to use in the case study
- Apply the algorithm to a sample dataset and evaluate it to define pickup zones that can help us improve transport routes



Overview of Related Work

- DBSCAN was already used to group other types of geo-referenced data apart from transport requests
- It was found to be the most used for on-demand transport domain



Grouping of people

- There are financial incentives and methods of road organizations to group as many people in each vehicle as:
 - **High-occupancy vehicle lanes**
 - **High-occupancy toll lanes**
 - **Slugging lines**



Clustering

Partitioned

K-Means

Hierarchical

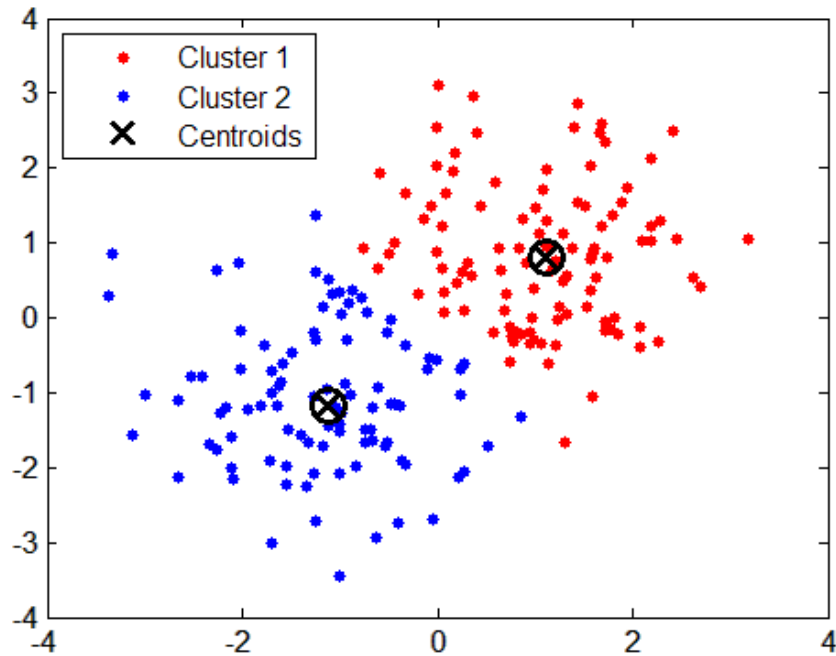
Density-based

DBSCAN

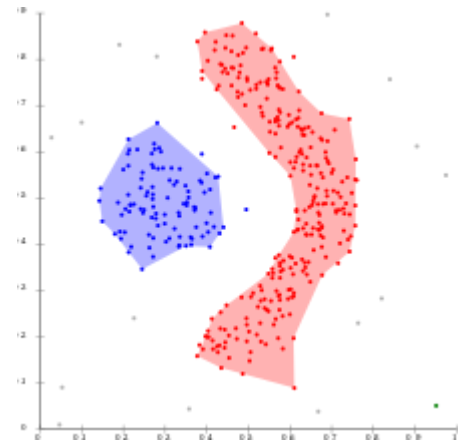
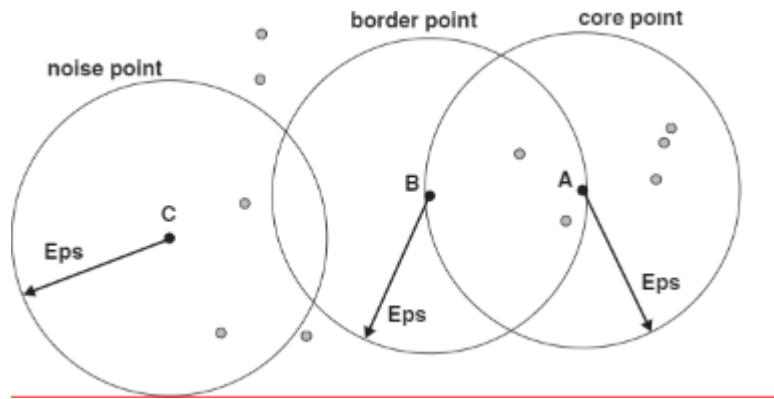
Grid



K-Means



DBSCAN





Technique comparison

- Feature A: ability to identify clusters with random shapes
- Feature B: ability to identify clusters in datasets with high data volume
- Feature C: good performance in obtaining results
- Feature D: ability to deal with noise
- Feature E: parameterization/initial configuration of the algorithm
- Feature F: Handle numeric values

Quoted from the thesis of “Análise de dados e Machine Learning na Mobilidade Urbana” by João Pedro Fernandes Simões (<https://comum.rcaap.pt/bitstream/10400.26/29858/1/Joao-Pedro-Fernandes-simoes.pdf>)

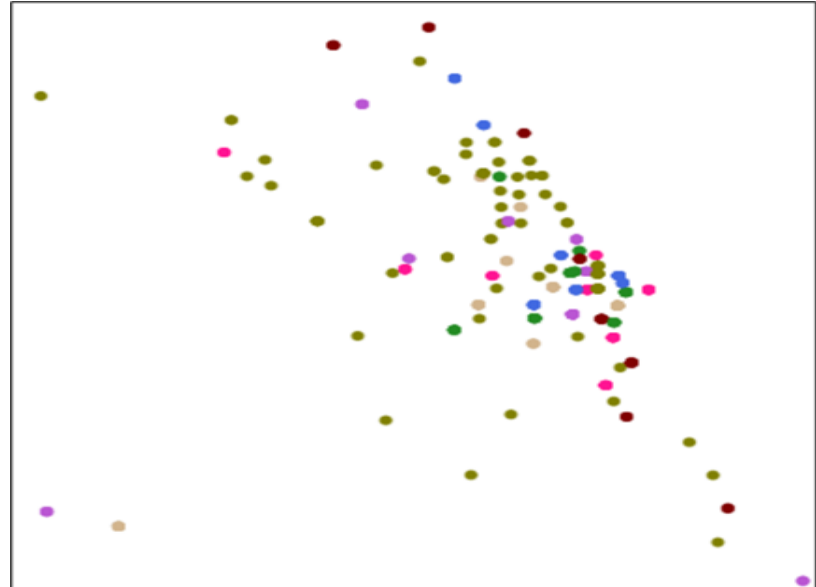
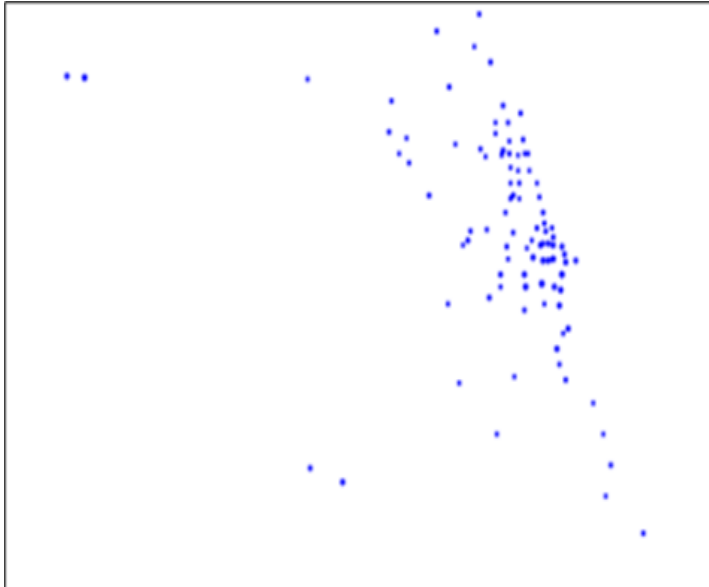


Technique comparison

Algorithm	A	B	C	D	E
K-Means	X	X	X	X	✓
DBSCAN	✓	✓	✓	✓	✓



Case Study – Clustering pickup locations



https://github.com/pedroafonsoo/clustering_case_study_industrial_seminars/blob/master/dbscan_case_study_dbscan.ipynb



Conclusions

- DBSCAN algorithm is the most appropriate to aggregate transport requests compared to K-Means
- By applying the algorithm with a real data subset, we obtained the set of associated clusters that define pickup points, with a strong silhouette value, indicating quality in the result



Future work

- **Challenge: There should be a limit in the maximum number of elements in each cluster!**
 - Because in a vehicle there is a certain number of passengers and in a cluster, there can be more passengers than the maximum capacity that the vehicle allows



Future work

- Combine the DBSCAN clustering technique with Constrained K-Means
- Constrained K-Means restrict the capacity of the minimum and maximum number of points for each cluster and at the same time guarantee the optimization of the distance between the points



Thanks!