

# **Customer Clustering**

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#### Springboard Capstone 3

In this project we will focus on clustering customers by similarity. This step precedes the handling of individual clusters for various purposes.

# **Project Walkthrough**

#### Importing Data

# This dataset was As in almost all cases, downloaded from datasets do not come Kaggle.com and was optimized for the machine

imported using pandas' "read\_csv"

method.

**Exploratory Data Analysis** 

learning algorithm and

should be visualized by

the data scientist before

being engineered.

#### **Feature Engineering**

# After visualizing the data, a data scientist gains an understanding of which transformations to apply to the data to emphasize trends and relationships.

#### **Data Standardization**

Many machine learning algorithms require data to be standardized before being fed as input.

Especially distance-based algorithms such as K-Means, which we will be using in this project.

#### Modeling

In this final step, we will evaluate the performance of the model prioritizing the sum of squared errors and using the silhouette score to supplement our evaluation.

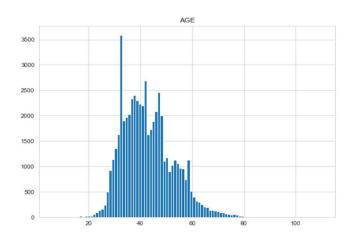
# **Quick View of the Dataset**

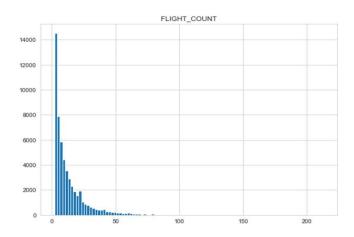
|   | AGE  | FLIGHT_COUNT | AVG_INTERVAL | MAX_INTERVAL |
|---|------|--------------|--------------|--------------|
| 0 | 39.0 | 1.945910     | 109.000000   | 274          |
| 1 | 60.0 | 2.944439     | 33.666667    | 126          |
| 2 | 39.0 | 2.995732     | 33.368421    | 124          |
| 3 | 37.0 | 2.397895     | 37.500000    | 80           |
| 4 | 29.0 | 1.386294     | 53.000000    | 132          |

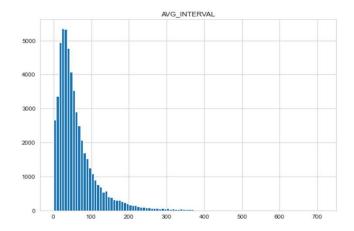
| Age                   | Flight Count                               | Average<br>Interval  | Max Interval  |  |
|-----------------------|--|--|---|--|
| Age of the individual | Total Flight<br>Count of the<br>individual | Average<br>number of days<br>an individual<br>does not fly | Maximum<br>number of days<br>an individual<br>has not flown |  |

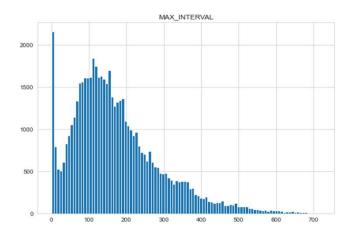
|              | count   | mean       | std        | min      | 25%       | 50%        | 75%        | max        |
|--------------|---------|------------|------------|----------|-----------|------------|------------|------------|
| AGE          | 55000.0 | 42.701400  | 9.773425   | 6.000000 | 35.000000 | 42.000000  | 48.000000  | 110.000000 |
| FLIGHT_COUNT | 55000.0 | 2.159785   | 0.897792   | 0.693147 | 1.386294  | 2.079442   | 2.772589   | 5.361292   |
| AVG_INTERVAL | 55000.0 | 62.979631  | 64.506386  | 0.000000 | 25.166667 | 44.600000  | 77.500000  | 714.000000 |
| MAX_INTERVAL | 55000.0 | 172.201745 | 117.382011 | 0.000000 | 90.000000 | 149.000000 | 230.000000 | 719.000000 |

## **Feature Distributions**

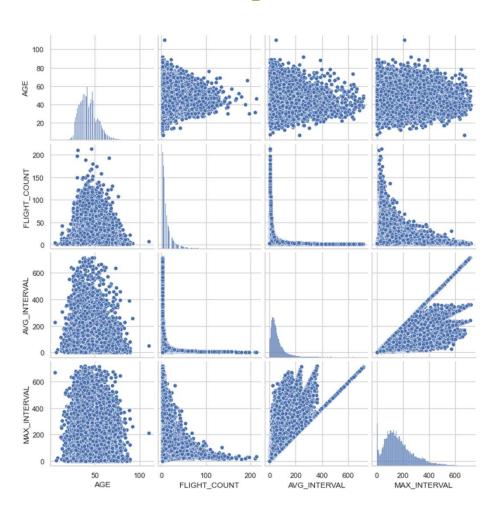




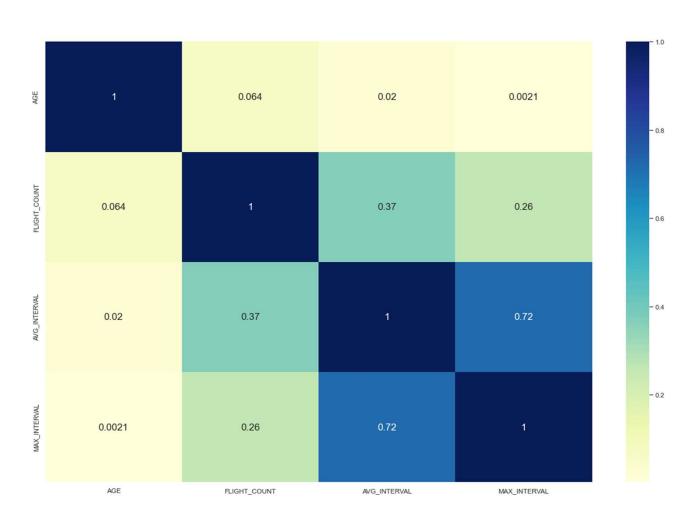




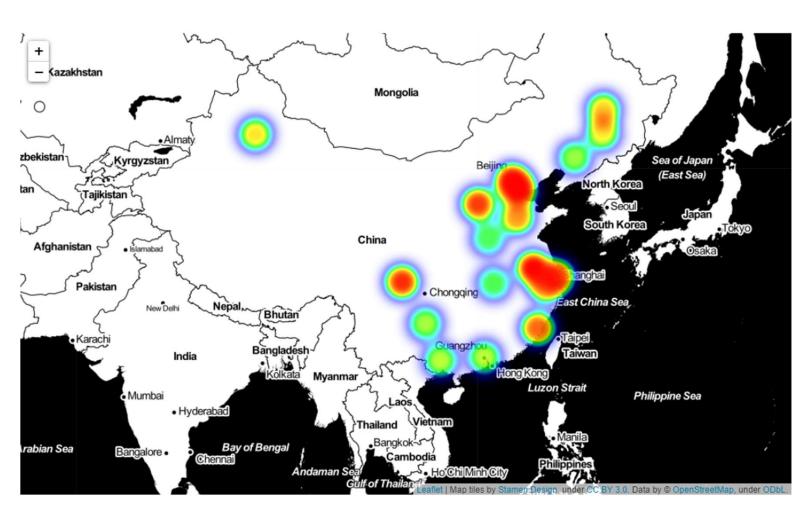
# **Pairplot**



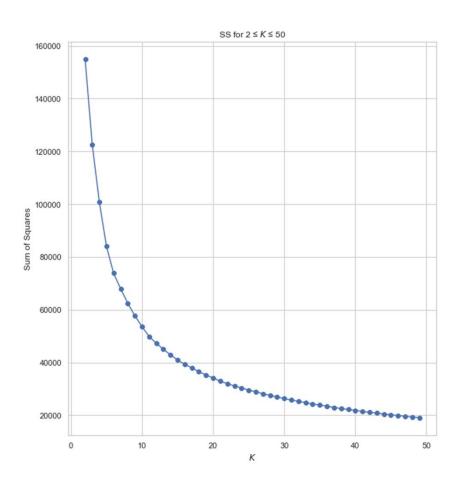
### **Feature Correlations**

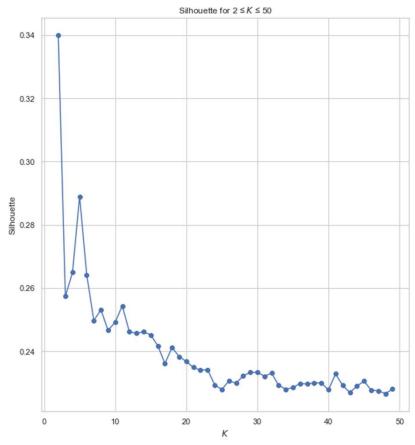


# **Most Frequent Flyers**

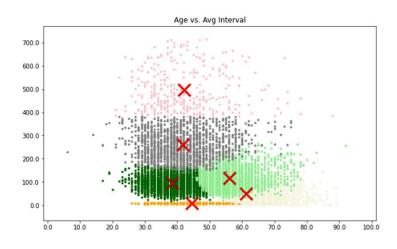


## **Model Performance Evaluation**





# **Resulting Clusters!**



K-Means created some awesome clusters that visibly separated individuals in a way that can be understood.

