

## **Dataset Overview**

Using the Google Landmarks Dataset V2, curated a sub-dataset based on two buildings: Terminal Tower and the Flatiron Building.

### Total amount of images

Terminal Towers: 256 images Flatiron Building: 170 images











## My Implementation Method .....

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### 1. Curating a Dataset

Google Landmarks Dataset V2, Terminal Towers + Flatiron Building

## 4. Extracting RGB values

Took the RGB values from each image

### 2. Creating Directories

Separated the data into a training set and a testing set

### 5. Reshaping array

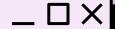
Transforming from a 3D array into a 2D array

### 3. Resizing the data

Preprocessing testing sets for both classes in for loops so images are the same size

### 6. Creating a model

Trained a k-means model to cluster the data into two groups



## **Evaluation Methodology**

The amount of true labels in the training predictions was 272 images with 203 that matched.

The amount of true labels in the testing predictions was 85 images with 62 that matched.





## Methodology



### **Training + Testing accuracy**

Checked how many of the predictions matched the actual label of the image. Accuracy: number of correct images/total images

### Analyzing the clusters visually

I graphed the data points to see the shape of the two clusters. When the clusters overlap less it means that k-means is more accurate.



### Results

Balanced Dataset = Equal # of Images in Each Class Unbalanced Dataset = One Class had more images

The unbalanced dataset has a higher accuracy percentage the the balanced as it has more images to train on.

	Unbalanced	Balanced
Training Accuracy	75.3%	74.1%
Testing Accuracy	74.6%	72.9%

#### **Unbalanced dataset**

```
correct = 0
correct = 0
                                                for i in range(85):
for i in range(340):
                                                  if test_predictions[i] == true labels[i]:
  if train predictions[i] == true labels[i]:
                                                    correct += 1
    correct += 1
print("Total correct value:", correct)
                                                print("Total correct value:", correct)
accuracy = correct/len(train predictions)*100
                                                accuracy = correct/len(test predictions)*100
accuracy = round(accuracy,3)
                                                accuracy = round(accuracy,3)
print("Training Accuracy:", accuracy,"%")
                                                print("Testing Accuracy:", accuracy,"%")
Total correct value: 256
                                                Total correct value: 63
Training Accuracy: 75.294 %
                                                Testing Accuracy: 74.118 %
```

#### **Balanced dataset**

```
correct = 0
for i in range(272):
    if train_predictions[i] == true_labels[i]:
        correct += 1
print("Total correct value:", correct)
accuracy = correct/len(train_predictions)*100
accuracy = round(accuracy,3)
print("Training Accuracy:", accuracy,"%")

Total correct value: 203
Accuracy: 74.632 %
```

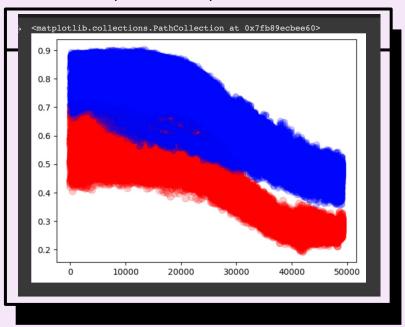
```
correct = 0
for i in range(85):
   if test_predictions[i] == true_labels[i]:
      correct += 1
print("Total correct value:", correct)
accuracy = correct/len(test_predictions)*100
accuracy = round(accuracy,3)
print("Testing Accuracy:", accuracy,"%")
Total correct value: 62
Testing Accuracy: 72.941 %
```





## Results

An Example of a Graph of the Clusters



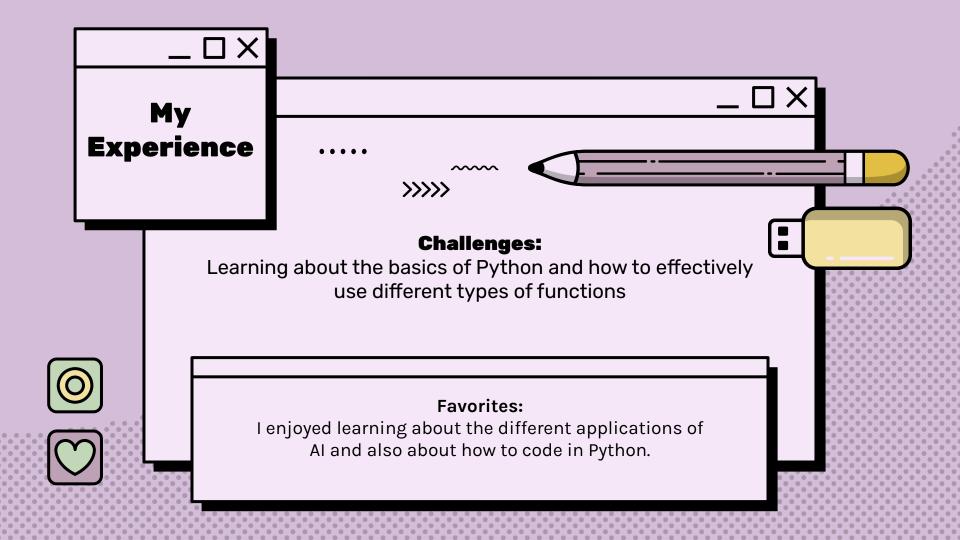
The datapoints partially overlap in the center



### **Demo**

The model is able to classify whether the image is from Terminal Tower or from Flatiron Building.







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### **What I've Learned**

- Understanding the impacts of Al and how it works
- Different types of libraries (Machine Learning Libraries)
- Experimenting with Python, Jupyter notebook, and Google Collab



