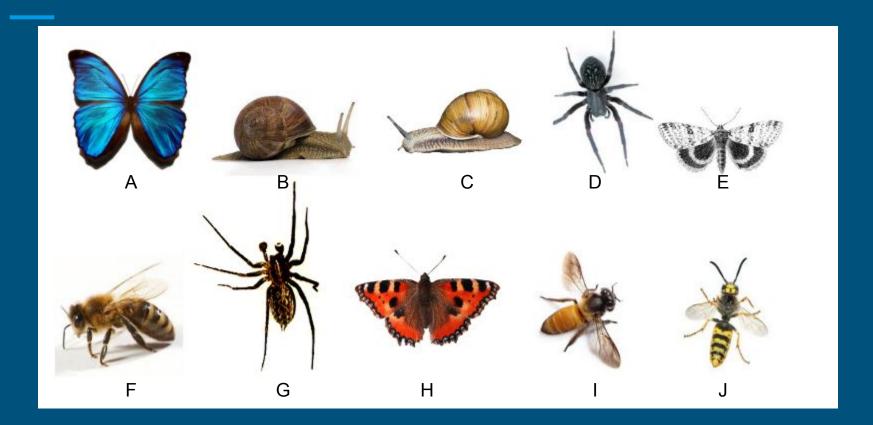
# Clustering

Sam D. Lojacono, Seth Adams, Kiana Herr, David Moste, Joel Bianchi, Maxwell Yearwood

# How many different groups of creepy-crawlies do you see?

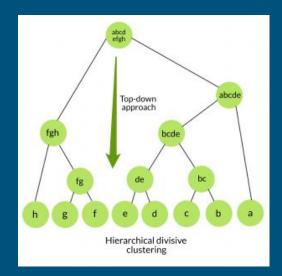


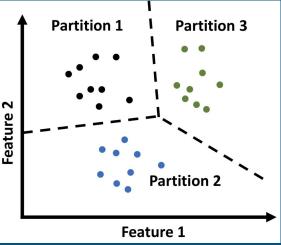
#### What is Clustering?

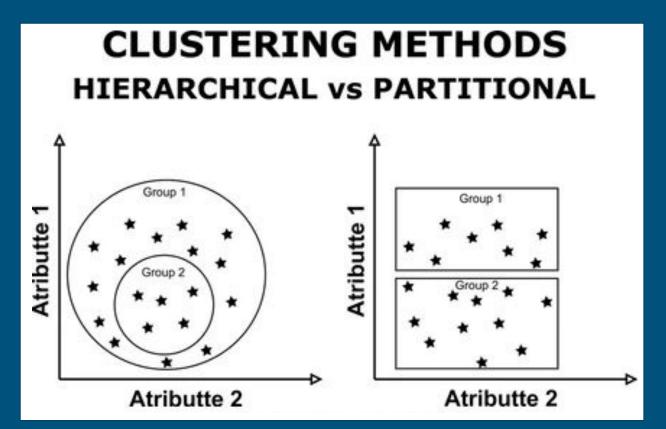
- Clustering is the process of dividing a dataset into groups in such a way that data points in the same cluster are similar to each other and different from those in other clusters
- Clustering is used in machine learning and data science to group similar data points together.
- The goal of clustering is to identify patterns in the data and group them into meaningful clusters.

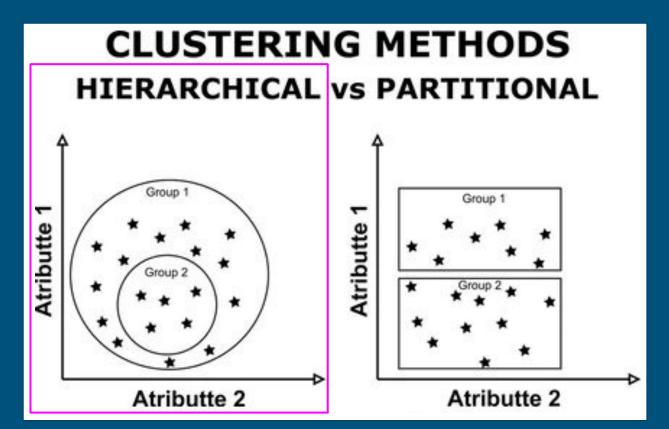
Clustering

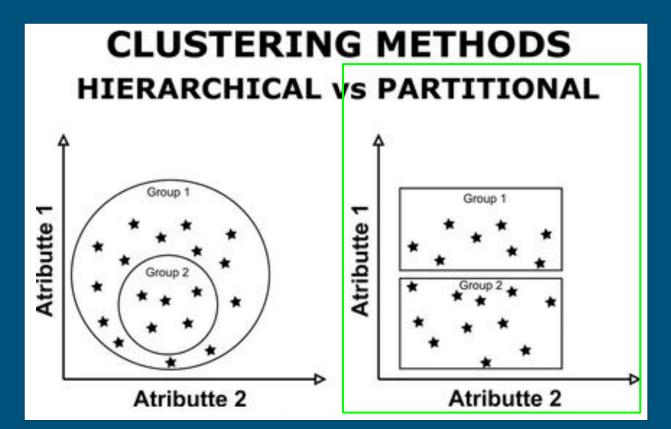
- There are two main types of clustering algorithms – hierarchical clustering and partitional clustering.
- Hierarchical clustering involves creating a tree-like structure of clusters, where smaller clusters are merged or nestled into larger ones.
- **Partitional clustering** involves dividing the data points into non-overlapping clusters, where each data point belongs to exactly one cluster.





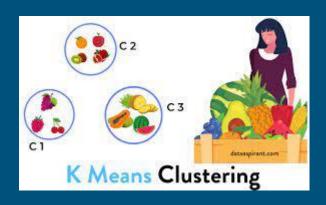


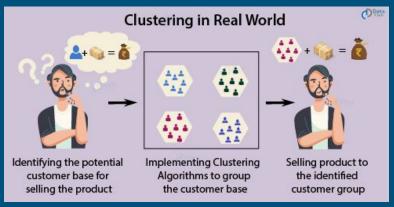




#### Applications of Clustering

- Image segmentation and compression
- Customer segmentation in marketing
- Document clustering in natural language processing
- Anomaly detection in cybersecurity
- Simple and efficient technique for finding groups of similar observations in large and high-dimensional data sets



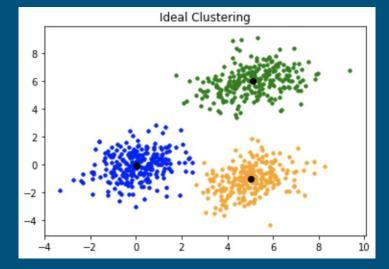


#### What is k-Means Clustering?

- A method of partitioning observations into clusters
- Clusters are characterized by their centroids, which serve as markers for the members of the cluster

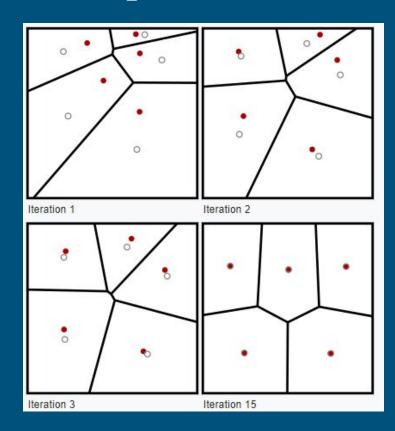
Each observation belongs to the cluster with the nearest mean (cluster)

centers or centroid)



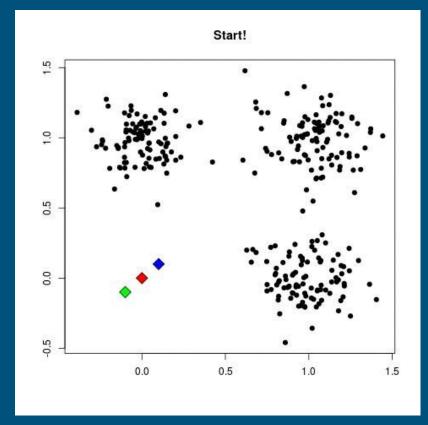
# The Iterative Refinement Technique

- The most common algorithm for k-means clustering
- Also known as "Lloyd's algorithm" or "naïve k-means"
- Alternates between assigning data points to the nearest cluster and recomputing the cluster means
- Converges when the centroids no longer change



# Steps to the K-means Clustering Algorithm

- Choose the number of clusters to form (k)
- Randomly select k observations and assign them as the initial centroids
- Assign each observation to the nearest centroid using a distance measure
- Recalculate the centroids of each cluster as the mean of all observations in that cluster
- Repeat steps 3 and 4 until convergence



#### Limitations of k-Means Clustering

- Does not guarantee to find the optimum solution
- Determining the optimal number of clusters can be subjective or based on external criteria



## "Notebooks": Popular Data Science Tools

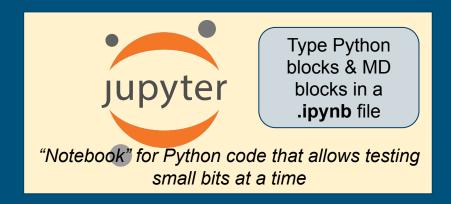


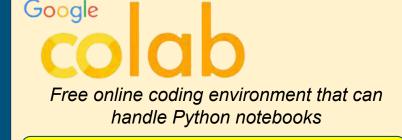
Type Python in a .py file

Language we are programming in



Free online coding environment for Python

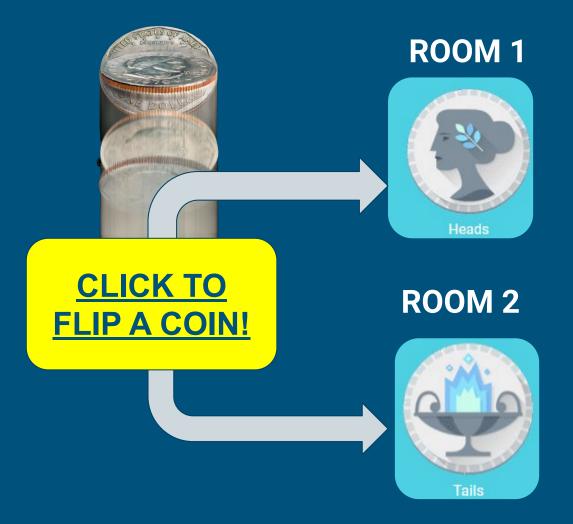




Code-Along Starter Code Notebook

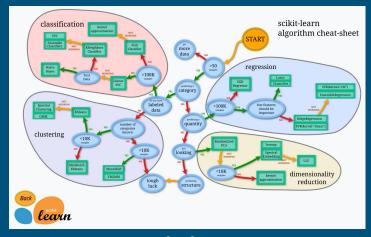
#### CODE-ALONG

# TO DO LIST 1. Flip a Coin 2. Join a Room 3. Open Copy of Starter Code Notebook 4. Code along!

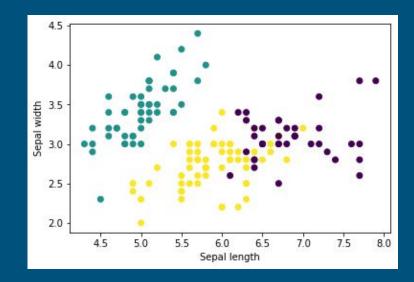


#### Intro to Scikit Learn

- Machine learning package for python
- Heavily used in data science
- Has built-in methods for various classification, regression, and clustering algorithms
- Also has built-in methods for the undersides of data science as well (preprocessing)

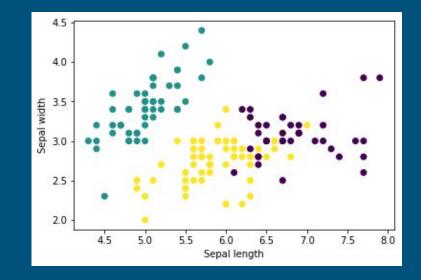


```
# Import required libraries
       from sklearn.datasets import load_iris
       from sklearn.cluster import KMeans
       import matplotlib.pyplot as plt
       # Load the Iris dataset
       iris = load iris()
       X = iris.data
       # Create an implementation of kmeans
11
       kmeans = KMeans(n clusters=3)
       labels = kmeans.fit_predict(X)
12
13
       # Plot the results
14
15
       plt.scatter(X[:, 0], X[:, 1], c=labels)
       plt.xlabel('Sepal Length')
       plt.ylabel('Sepal width')
17
       plt.show()
18
19
```



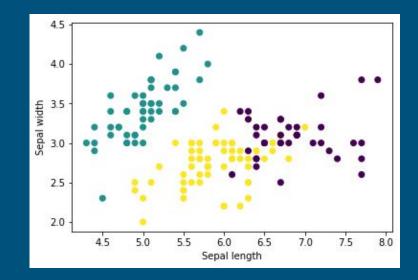
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These lines import both the kmeans algorithm and a prepackaged dataset from scikit learn



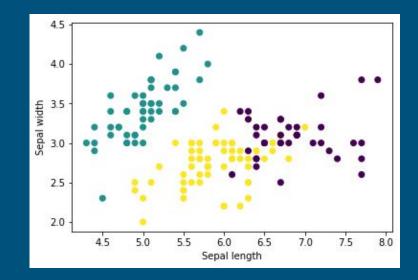
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These lines bring the iris dataset into the working code



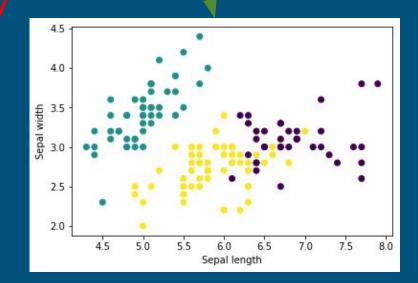
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```

These lines of code run the entire algorithm we just worked on...



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These lines print out the plot shown below



# Homework Options

Click Here for Homework