Species Segmentation with Cluster Analysis Data Science and Business Analytics intern -The spark foundation By- Srishti Chopra Import the relevant libraries In [1]: import numpy as np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns sns.set() from sklearn.cluster import KMeans Load the data Load data from the csv file: 'iris_dataset.csv'. In [2]: # Load the data data = pd.read csv('iris dataset.csv') data Out[2]: sepal_length sepal_width petal_length petal_width 0 5.1 3.5 1.4 0.2 4.9 3.0 1.4 0.2 2 4.7 3.2 1.3 0.2 4.6 3.1 1.5 0.2 5.0 3.6 1.4 0.2 145 6.7 3.0 5.2 2.3 146 6.3 2.5 5.0 1.9 147 6.5 3.0 5.2 2.0 148 6.2 3.4 5.4 2.3 149 5.9 3.0 5.1 1.8 150 rows × 4 columns Plot the data In [3]: # Create a scatter plot based on two corresponding features plt.scatter(data['sepal length'], data['sepal width']) plt.xlabel('Lenght of sepal') plt.ylabel('Width of sepal') plt.show() 4.5 4.0 3.5 3.0 2.5 2.0 4.5 5.0 6.0 Lenght of sepal Standardize the variables In [4]: # import some preprocessing module from sklearn import preprocessing # scale the data for better results x scaled = preprocessing.scale(data) x scaled array([[-9.00681170e-01, 1.03205722e+00, -1.34127240e+00, Out[4]: -1.31297673e+00],[-1.14301691e+00, -1.24957601e-01, -1.34127240e+00, -1.31297673e+00], [-1.38535265e+00, 3.37848329e-01, -1.39813811e+00, -1.31297673e+00],[-1.50652052e+00, 1.06445364e-01, -1.28440670e+00, -1.31297673e+00],[-1.02184904e+00, 1.26346019e+00, -1.34127240e+00, -1.31297673e+00], [-5.37177559e-01, 1.95766909e+00, -1.17067529e+00, -1.05003079e+00], [-1.50652052e+00, 8.00654259e-01, -1.34127240e+00, -1.18150376e+00], [-1.02184904e+00, 8.00654259e-01, -1.28440670e+00,-1.31297673e+00],[-1.74885626e+00, -3.56360566e-01, -1.34127240e+00, -1.31297673e+00],[-1.14301691e+00, 1.06445364e-01, -1.28440670e+00, -1.44444970e+00],[-5.37177559e-01, 1.49486315e+00, -1.28440670e+00, -1.31297673e+00],[-1.26418478e+00,8.00654259e-01, -1.22754100e+00, -1.31297673e+00],[-1.26418478e+00, -1.24957601e-01, -1.34127240e+00, -1.44444970e+00],[-1.87002413e+00, -1.24957601e-01, -1.51186952e+00, -1.44444970e+00],[-5.25060772e-02, 2.18907205e+00, -1.45500381e+00, -1.31297673e+00], [-1.73673948e-01, 3.11468391e+00, -1.28440670e+00, -1.05003079e+00],[-5.37177559e-01, 1.95766909e+00, -1.39813811e+00, -1.05003079e+00], [-9.00681170e-01, 1.03205722e+00, -1.34127240e+00, -1.18150376e+00], [-1.73673948e-01, 1.72626612e+00, -1.17067529e+00, -1.18150376e+00],[-9.00681170e-01, 1.72626612e+00, -1.28440670e+00, -1.18150376e+00],[-5.37177559e-01, 8.00654259e-01, -1.17067529e+00,-1.31297673e+00],[-9.00681170e-01, 1.49486315e+00, -1.28440670e+00, -1.05003079e+00], [-1.50652052e+00, 1.26346019e+00, -1.56873522e+00, -1.31297673e+00],[-9.00681170e-01, 5.69251294e-01, -1.17067529e+00,-9.18557817e-01, [-1.26418478e+00,8.00654259e-01, -1.05694388e+00, -1.31297673e+00],[-1.02184904e+00, -1.24957601e-01, -1.22754100e+00,-1.31297673e+00], [-1.02184904e+00, 8.00654259e-01, -1.22754100e+00, 1.05003079e+00], [-7.79513300e-01, 1.03205722e+00, -1.28440670e+00, -1.31297673e+00], [-7.79513300e-01, 8.00654259e-01, -1.34127240e+00, -1.31297673e+00], [-1.38535265e+00, 3.37848329e-01, -1.22754100e+00, -1.31297673e+00], [-1.26418478e+00, 1.06445364e-01, -1.22754100e+00, -1.31297673e+00], [-5.37177559e-01, 8.00654259e-01, 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7.05892939e-01, 3.96171883e-01], [1.28034050e+00, 3.37848329e-01, 1.10395287e+00, 1.44795564e+00], [-2.94841818e-01, -5.87763531e-01, 6.49027235e-01, 1.05353673e+00], [2.24968346e+00, -5.87763531e-01, 1.67260991e+00, 1.05353673e+00], [5.53333275e-01, -8.19166497e-01, 6.49027235e-01, 7.90590793e-01], [1.03800476e+00, 5.69251294e-01, 1.10395287e+00, 1.18500970e+00], [1.64384411e+00, 3.37848329e-01, 1.27454998e+00, 7.90590793e-01], [4.32165405e-01, -5.87763531e-01, 5.92161531e-01, 7.90590793e-01], [3.10997534e-01, -1.24957601e-01, 6.49027235e-01, 7.90590793e-01], [6.74501145e-01, -5.87763531e-01, 1.04708716e+00, 1.18500970e+00], [1.64384411e+00, -1.24957601e-01, 1.16081857e+00, 5.27644853e-01], [1.88617985e+00, -5.87763531e-01, 1.33141568e+00, 9.22063763e-01], [2.49201920e+00, 1.72626612e+00, 1.50201279e+00, 1.05353673e+00], [6.74501145e-01, -5.87763531e-01, 1.04708716e+00, 1.31648267e+00], [5.53333275e-01, -5.87763531e-01, 7.62758643e-01, 3.96171883e-01], [3.10997534e-01, -1.05056946e+00, 1.04708716e+00, 2.64698913e-01], [2.24968346e+00, -1.24957601e-01, 1.33141568e+00, 1.44795564e+00], [5.53333275e-01, 8.00654259e-01, 1.04708716e+00, 1.57942861e+00], [6.74501145e-01, 1.06445364e-01, 9.90221459e-01, 7.90590793e-01], [1.89829664e-01, -1.24957601e-01, 5.92161531e-01, 7.90590793e-01], 1.06445364e-01, 9.33355755e-01, [1.28034050e+00, 1.18500970e+00], [1.03800476e+00, 1.06445364e-01, 1.04708716e+00, 1.57942861e+00], [1.28034050e+00, 1.06445364e-01, 7.62758643e-01, 1.44795564e+00], [-5.25060772e-02, -8.19166497e-01, 7.62758643e-01, 9.22063763e-01], 3.37848329e-01, 1.21768427e+00, [1.15917263e+00, 1.44795564e+00], [1.03800476e+00, 5.69251294e-01, 1.10395287e+00, 1.71090158e+00], [1.03800476e+00, -1.24957601e-01, 8.19624347e-01, 1.44795564e+00], [5.53333275e-01, -1.28197243e+00, 7.05892939e-01, 9.22063763e-01], [7.95669016e-01, -1.24957601e-01, 8.19624347e-01, 1.05353673e+00], [4.32165405e-01, 8.00654259e-01, 9.33355755e-01, 1.44795564e+00], [6.86617933e-02, -1.24957601e-01, 7.62758643e-01, 7.90590793e-01]]) Take Advantage of the Elbow Method **WCSS** In [5]: wcss = []# 'cl num' is a that keeps track the highest number of clusters we want to use the WCSS method for. We have it cl num = 10for i in range (1,cl num): kmeans= KMeans(i) kmeans.fit(x scaled) wcss_iter = kmeans.inertia wcss.append(wcss iter) C:\Users\COMAP\Downloads\Anaconda\lib\site-packages\sklearn\cluster\ kmeans.py:881: UserWarning: KMeans is know n to have a memory leak on Windows with MKL, when there are less chunks than available threads. You can avoid i t by setting the environment variable OMP_NUM_THREADS=1. warnings.warn([600.00000000000003, Out[5]: 223.73200573676343, 140.96581663074699, 114.42970777082239, 91.12763977985172, 80.24972235577128, 71.3007056411862, 63.01282906393713, 54.60041897786318] The Elbow Method In [6]: # Finding the optimum number of clusters for k-means classification number clusters = range(1,cl num) # Plotting the results onto a line graph, # `allowing us to observe 'The elbow' plt.plot(number_clusters, wcss) plt.title('The Elbow Method') plt.xlabel('Number of clusters') plt.ylabel('Within-cluster Sum of Squares') Text(0, 0.5, 'Within-cluster Sum of Squares') The Elbow Method 600 Squares 500 ō 400 Within-cluster Si 300 100 Number of clusters **Understanding the Elbow Curve** Based on the Elbow Curve, 3 seem the most likely. 3 Clusters In [7]: $kmeans_3 = KMeans(3)$ kmeans_3.fit(x_scaled) KMeans(n_clusters=3) Out[7]: In [8]: clusters 3 = data.copy() clusters_3['cluster_pred']=kmeans_3.fit_predict(x_scaled) In [9]: plt.scatter(clusters_3['sepal_length'], clusters_3['sepal_width'],\ c= clusters_3 ['cluster_pred'], cmap = 'rainbow') <matplotlib.collections.PathCollection at 0x25133492fa0> Out[9]: 4.5 4.0 3.5 3.0 2.5 2.0