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In [1]: import os
        from skimage import io,color,exposure,filters
        import numpy as np
        import pandas as pd
        from sklearn.decomposition import PCA
        import warnings
        warnings.filterwarnings("ignore")
        def angle(dx, dy):
         return np.mod(np.arctan2(dy, dx), np.pi)
        direct = "C:\\Users\\mohan\\Desktop\\Cropped"
        breeds=os.listdir(direct)
        images = []
        dog breed = []
        for index, breed in enumerate(breeds):
            img path = os.path.join(direct, breed)
            for image in os.listdir(img path):
                src path = os.path.join(img path, image)
                img = io.imread(src path)
                img = color.rgb2gray(img)
                img = angle(filters.sobel h(img), filters.sobel v(img))
                hist, = exposure.histogram(img, nbins=36)
                hist = hist / np.sum(hist) # normalization added
                images.append(hist)
                dog breed.append(index)
        images = np.array(images)
        dog breed = np.array(dog breed)
        dr=PCA(2)
        images=dr.fit transform(images)
In [2]: from sklearn.cluster import KMeans, BisectingKMeans, SpectralClustering
        kmeans variants = {
            'Random': {'init': 'random'},
            'KMeans++': {'init': 'k-means++'},
            'BisectingKMeans': {'init': 'random'},
            'SpectralClustering': {}
        labels dict = {}
        for variant, params in kmeans variants.items():
            if variant == 'SpectralClustering':
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clustering = SpectralClustering(n clusters=4, random state=42)
            elif variant == 'BisectingKMeans':
                 clustering = BisectingKMeans(n clusters=4, random state=42)
            else:
                clustering = KMeans(n clusters=4, random state=42, **params)
            labels = clustering.fit predict(images)
            labels dict[variant] = labels
        Random = labels dict['Random']
        kmeans = labels dict['KMeans++']
        bisecting = labels dict['BisectingKMeans']
        spectralclustering = labels dict['SpectralClustering']
In [3]: from sklearn.cluster import DBSCAN
        dbscan = DBSCAN(eps=0.02, min samples=3).fit(images).labels
        print("For DBSCAN clustering with 4 clusters:")
        print("
                   eps = 0.02")
        print("
                   min samples = 3")
        print("were used.")
       For DBSCAN clustering with 4 clusters:
           eps = 0.02
           min samples = 3
       were used.
In [4]: from sklearn.cluster import AgglomerativeClustering
        linkage methods = ['single', 'complete', 'average', 'ward']
        results = {}
        for method in linkage methods:
            clustering = AgglomerativeClustering(n clusters=4, linkage=method)
            labels = clustering.fit predict(images)
            results[method] = labels
        Agglomerative single = results['single']
        Agglomerative complete = results['complete']
        Agglomerative average = results['average']
        Agglomerative ward = results['ward']
In [5]: from sklearn.metrics import fowlkes mallows score
        clustering methods = {
             'Random': Random,
             'kmeans': kmeans,
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'bisecting': bisecting,
            'spectralclustering': spectralclustering,
            'dbscan': dbscan,
            'Agglomerative single': Agglomerative single,
            'Agglomerative complete': Agglomerative complete,
            'Agglomerative average': Agglomerative average,
            'Agglomerative ward': Agglomerative ward
        for method name, labels in clustering methods.items():
            score = fowlkes mallows score(dog breed, labels)
            print(f'{method name}: {score}')
       Random: 0.2975270506119972
       kmeans: 0.29909666651045397
       bisecting: 0.2972809339865332
       spectralclustering: 0.31521169137383054
       dbscan: 0.4874562084130845
       Agglomerative single: 0.4943827347937293
       Agglomerative complete: 0.36713456124640975
       Agglomerative average: 0.48907882180388346
       Agglomerative ward: 0.32235438804477073
In [6]: from sklearn.metrics import silhouette score
        clustering methods = {
            'Random': Random,
            'kmeans': kmeans,
            'bisecting': bisecting,
            'spectralclustering': spectralclustering,
            'dbscan': dbscan,
            'Agglomerative single': Agglomerative single,
            'Agglomerative complete': Agglomerative complete,
            'Agglomerative average': Agglomerative average,
            'Agglomerative ward': Agglomerative ward
        for method name, labels in clustering methods.items():
            score = silhouette score(images, labels)
            print(f'{method name}: {score}')
```

Random: 0.40305783333055095 kmeans: 0.4055704996457822 bisecting: 0.37036373178557724

spectralclustering: 0.24084357806466253

dbscan: 0.7308506454338473

Agglomerative single: 0.809639336741712 Agglomerative complete: 0.4137738008851826 Agglomerative average: 0.5539758547312447 Agglomerative ward: 0.3934205143234151

In []: '''Rank the methods from the best to the worst for our dataset based on Fowlkes-Mallows index Agglomerative single, Agglomerative average, Agglomerative complete, Agglomerative ward'''

'''Rank the methods from the best to the worst for our dataset based on Silhouette Coefficient. Agglomerative single, Agglomerative average, Agglomerative complete, Agglomerative ward'''