```
In [6]: import os
        directory = os.getcwd()
        images = r'cropped_images'
        import warnings
        warnings.filterwarnings("ignore")
In [2]: # as in assignment 2 added normalization
        img=[]
        hist=[]
        breed=[]
        class_labels = {'n02085936-Maltese_dog':0,'n02096294-Australian_terrier':1,'n021065
        for folder in os.listdir(images):
            path = os.path.join(images, folder)
            if os.path.isdir(path):
                Images = os.listdir(path)
                crop_images = [image for image in Images if image.lower().endswith(('.jpg')
                for image in crop_images:
                    src_path = os.path.join(path, image)
                    img.append(src path)
                    breed.append(class_labels[folder])
        import numpy as np
        from skimage import filters
        from skimage import data, exposure, img_as_float,color
        def angle(dx, dy):
         """Calculate the angles between horizontal and vertical operators."""
         return np.mod(np.arctan2(dy, dx), np.pi)
        from skimage import filters
        from skimage import io
        for image in img:
            dog_img
                     = io.imread(image)
            dog_img=color.rgb2gray(dog_img)
            angle_sobel = angle(filters.sobel_h(dog_img),filters.sobel_v(dog_img))
            Hist,_=exposure.histogram(angle_sobel, nbins=36)
            hist.append(Hist/np.sum(Hist))
        hist=np.array(hist)
        breed=np.array(breed)
        #PCA
        from sklearn.decomposition import PCA
```

```
pca = PCA(n_components=2)
         edgehist=pca.fit_transform(hist)
In [7]: from sklearn.cluster import KMeans, Bisecting KMeans, Spectral Clustering, Agglomerative
         from sklearn.metrics import silhouette_score,fowlkes_mallows_score
         model dict={"Random" :KMeans(init="random", n clusters=4),
                 "KMeans++":KMeans(init="k-means++",n_clusters=4),
                 "BisectingKmeans":BisectingKMeans(init="random",n_clusters=4),
                 "SpectralClustering":SpectralClustering(n clusters=4),
                  "Agglomerate":AgglomerativeClustering(n_clusters=4) }
         algorith=[]
         silhouette=[]
         fowlkesmallows=[]
         for method,model in model_dict.items():
             if method == "Agglomerate":
                 for link in ['single','complete', 'average','ward']:
                     model=AgglomerativeClustering(n_clusters=4,linkage=link)
                     predict=model.fit_predict(edgehist)
                     algorith.append(link)
                     silhouette.append(silhouette_score(edgehist,predict))
                     fowlkesmallows.append(fowlkes_mallows_score(breed,predict))
             else:
                 model=model.fit(edgehist)
                 algorith.append(method)
                 silhouette.append(silhouette_score(edgehist,model.labels_))
                 fowlkesmallows.append(fowlkes_mallows_score(breed,model.labels_))
In [14]: #DBSCAN
         from sklearn.cluster import DBSCAN
         model = DBSCAN(eps=0.01, min_samples=3).fit(edgehist)
         pred = model.labels_
         n_clusters_ = len(set(pred)) - (1 if -1 in pred else 0)
         print("Estimated number of clusters: %d" % n_clusters_)
```

## algorith.append('DBSCAN') silhouette.append(silhouette score(edgehist,pred)) fowlkesmallows.append(fowlkes\_mallows\_score(breed,pred))

Estimated number of clusters: 4

## DBSCAN parameters used eps as 0.01 and min\_samples as 3 in order to get 4 clusters

```
In [16]: import pandas as pd
         algorithm_scores=pd.DataFrame({'method': algorith, 'fowlkes-mallows': fowlkesmallow
```

## best to worst based on fowlkes score

```
algorithm_scores.sort_values('fowlkes-mallows',ascending=False)
```

Out[17]:		method	fowlkes-mallows	silhouette
	4	single	0.506164	0.698595
	6	average	0.505364	0.504076
	8	DBSCAN	0.497707	0.262266
	5	complete	0.356079	0.437289
	3	SpectralClustering	0.346667	0.001596
	0	Random	0.310303	0.433507
	1	KMeans++	0.310303	0.433507
	2	BisectingKmeans	0.305513	0.396617
	7	ward	0.299091	0.414192

## best to worst silhouette score

In [18]: algorithm\_scores.sort\_values('silhouette',ascending=False)

Out[18]:		method	fowlkes-mallows	silhouette
	4	single	0.506164	0.698595
	6	average	0.505364	0.504076
	5	complete	0.356079	0.437289
	0	Random	0.310303	0.433507
	1	KMeans++	0.310303	0.433507
	7	ward	0.299091	0.414192
	2	BisectingKmeans	0.305513	0.396617
	8	DBSCAN	0.497707	0.262266
	3	SpectralClustering	0.346667	0.001596

In [ ]: ### https://scikit-learn.org/stable/modules/clustering.html