In [1]:
```
#Imports
import xml.etree.ElementTree as ET
import os
import glob
import random
import numpy as np
import matplotlib.pyplot as plt
import skimage as ski
import sklearn
from skimage import filters
from skimage import data
from skimage import io
from skimage.color import rgb2gray
from pathlib import Path
from PIL import Image
import sklearn.metrics
from sklearn.decomposition import PCA
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
import json
```

In [2]:
```
def get_bounding_boxes(annot):
    xml = annot
    tree = ET.parse(xml)
    root = tree.getroot()
    objects = root.findall('object')
    bbox = []
    for o in objects:
        bndbox = o.find('bndbox')
        xmin = int(bndbox.find('xmin').text)
        ymin = int(bndbox.find('ymin').text)
        xmax = int(bndbox.find('xmax').text)
        ymax = int(bndbox.find('ymax').text)
        bbox.append((xmin, ymin, xmax, ymax))
    return bbox
```

In [3]:
```
def crop_images(image_files, annotation_files, target_directory):
    for i in range(len(image_files)):
        bbox = get_bounding_boxes(annotation_files[i])
        dog_image_path = image_files[i]
        im = Image.open(dog_image_path)

        for j, box in enumerate(bbox):
            im2 = im.crop(box)
            im2 = im2.resize((128, 128), Image.LANCZOS)

            new_path = dog_image_path.replace('data/images', target_directory)
            new_path = new_path.replace('.jpg', f'-{j}.jpg')
            im2 = im2.convert('RGB')

            head, _ = os.path.split(new_path)
            Path(head).mkdir(parents=True, exist_ok=True)
```

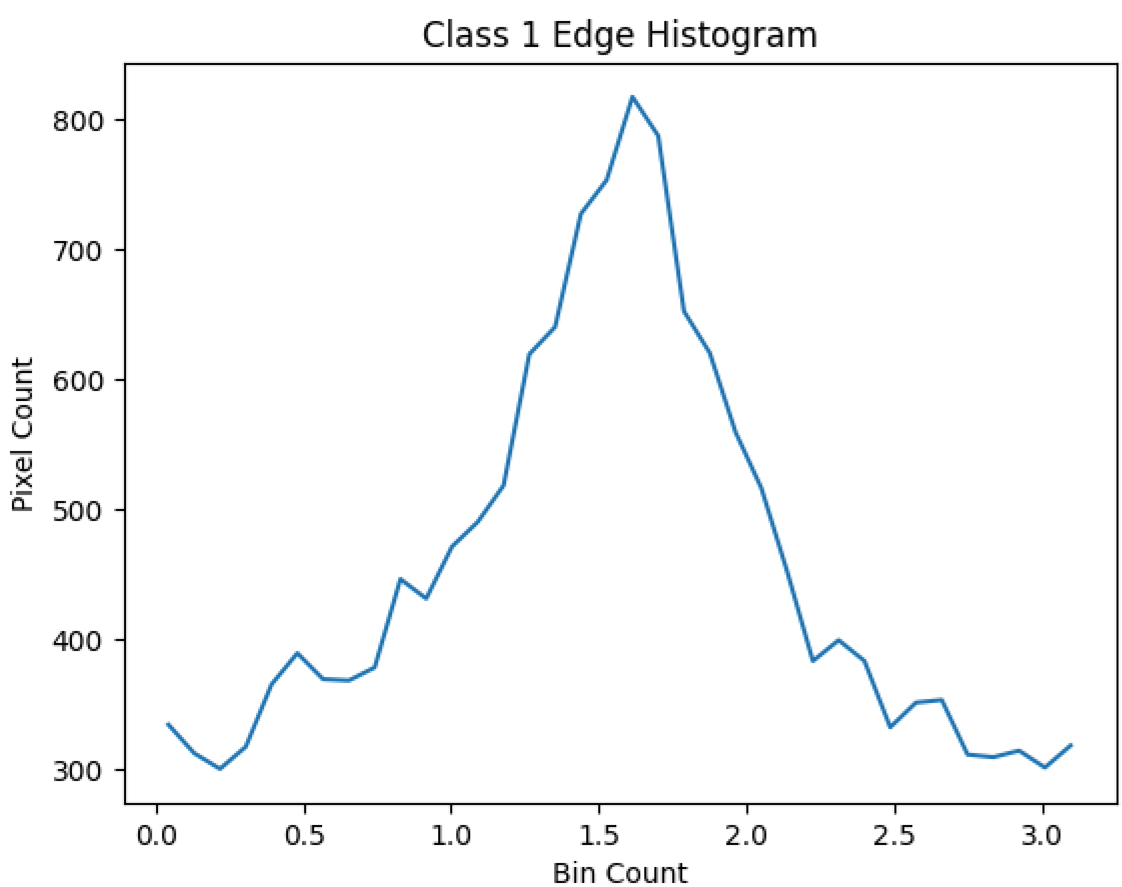
```
            im2.save(new_path)
```

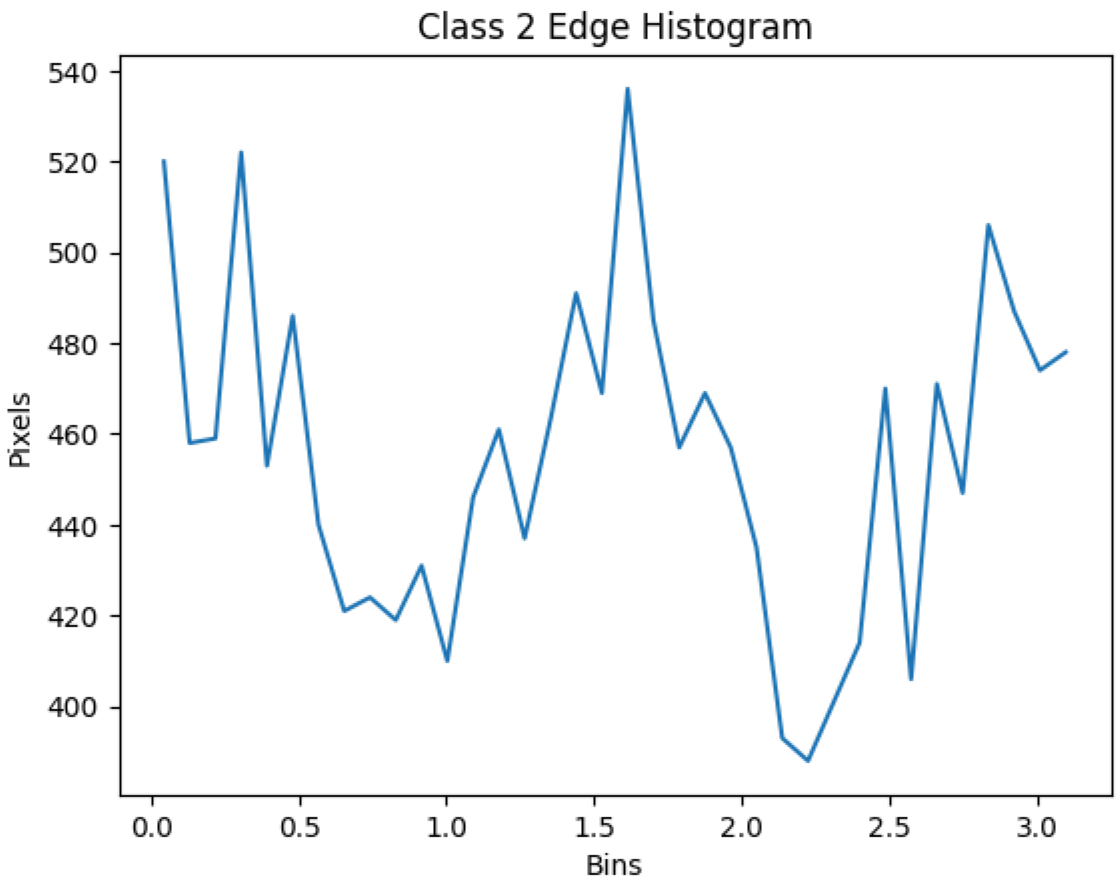
In [4]:
```
def angle(dx, dy):
    return np.mod(np.arctan2(dy,dx), np.pi)
```

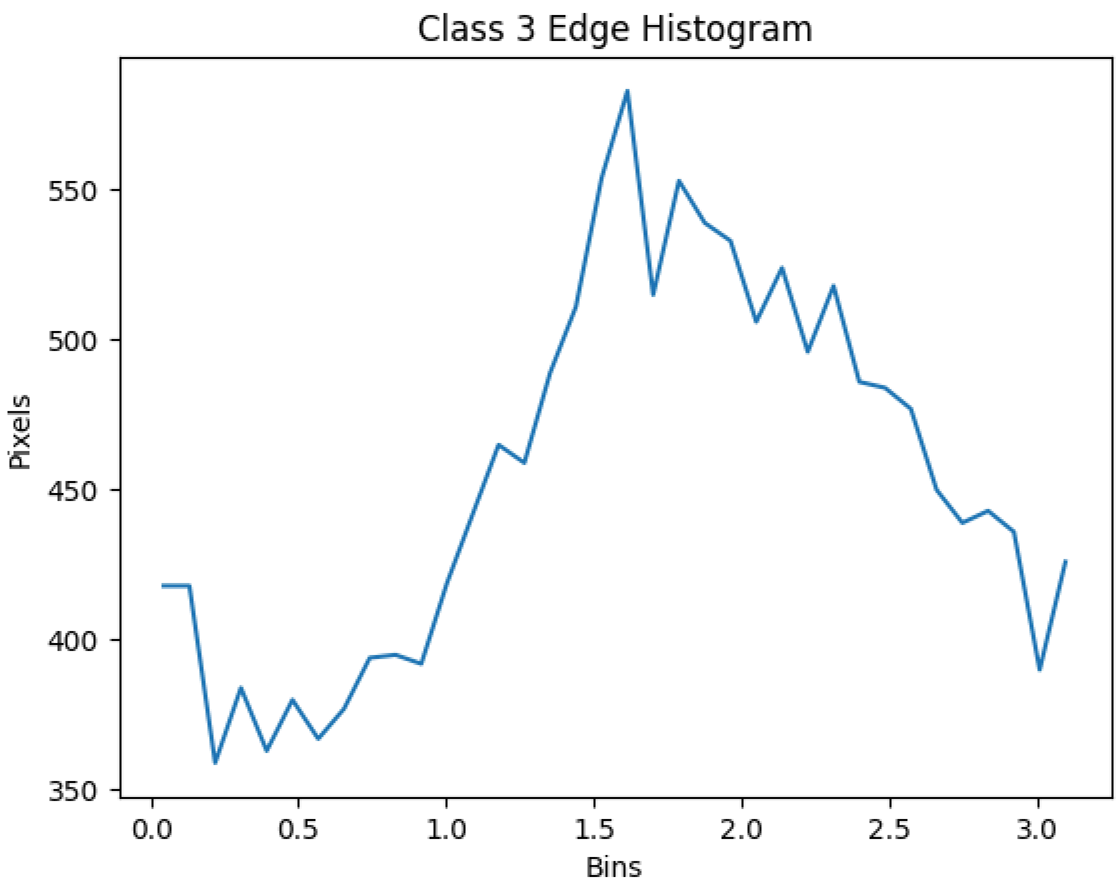
In [5]:
```
#Crop the images using the boundaries
crop_images(glob.glob(os.path.join(r'data/images/n02093647-Bedlington_terrier','*.j
            glob.glob(os.path.join(r'data/annotations/n02093647-Bedlington_terrier'
            './data/cropped/class1')
crop_images(glob.glob(os.path.join(r'data/images/n02099849-Chesapeake_Bay_retriever
            glob.glob(os.path.join(r'data/annotations/n02099849-Chesapeake_Bay_retr
            './data/cropped/class2')
crop_images(glob.glob(os.path.join(r'data/images/n02100735-English_setter','*.jpg')
            glob.glob(os.path.join(r'data/annotations/n02100735-English_setter', '*
            './data/cropped/class3')
crop_images(glob.glob(os.path.join(r'data/images/n02116738-African_hunting_dog','*.
            glob.glob(os.path.join(r'data/annotations/n02116738-African_hunting_dog
            './data/cropped/class4')
```

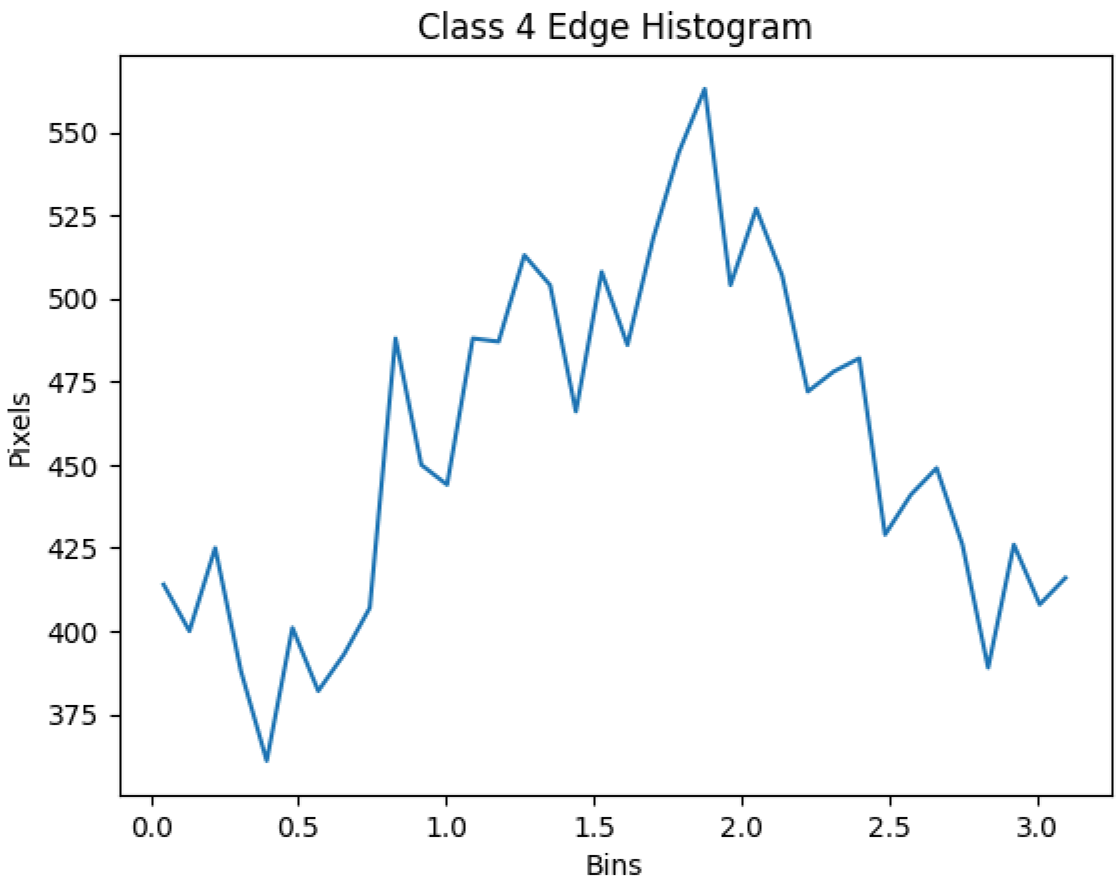
In [6]:
```
#Feature Extraction: Edge Histogram and Similarity Measurements
#Choosing one random Image from each class
randomImage1= random.choice(glob.glob(os.path.join(r'data/cropped/class1/n02093647-
randomImage2 = random.choice(glob.glob(os.path.join(r'data/cropped/class2/n02099849
randomImage3 = random.choice(glob.glob(os.path.join(r'data/cropped/class3/n02100735
randomImage4 = random.choice(glob.glob(os.path.join(r'data/cropped/class4/n02116738

#Convert Each Random Image to Grayscale
grayScale1 = rgb2gray(io.imread(randomImage1))
grayScale2 = rgb2gray(io.imread(randomImage2))
grayScale3 = rgb2gray(io.imread(randomImage3))
grayScale4 = rgb2gray(io.imread(randomImage4))

#Calculate angle sobel for each image
angleSobel1 = angle(filters.sobel_h(grayScale1),filters.sobel_v(grayScale1))
angleSobel2 = angle(filters.sobel_h(grayScale2),filters.sobel_v(grayScale2))
angleSobel3 = angle(filters.sobel_h(grayScale3),filters.sobel_v(grayScale3))
angleSobel4 = angle(filters.sobel_h(grayScale4),filters.sobel_v(grayScale4))

#Edge Histogram
histogram1, bin1 = ski.exposure.histogram(angleSobel1, nbins=36)
histogram2, bin2 = ski.exposure.histogram(angleSobel2, nbins=36)
histogram3, bin3 = ski.exposure.histogram(angleSobel3, nbins=36)
histogram4, bin4 = ski.exposure.histogram(angleSobel4, nbins=36)

#Plot the images with their corresponding edge histogram value
#Class1
plt.figure()
plt.plot(bin1, histogram1)
plt.xlabel("Bin Count")
plt.ylabel("Pixel Count")
plt.title("Class 1 Edge Histogram")
plt.fill()
plt.show()
#Class2
```

```
plt.figure()
plt.plot(bin2, histogram2)
plt.xlabel("Bins")
plt.ylabel("Pixels")
plt.title("Class 2 Edge Histogram")
plt.fill()
plt.show()
#Class3
plt.figure()
plt.plot(bin3, histogram3)
plt.xlabel("Bins")
plt.ylabel("Pixels")
plt.title("Class 3 Edge Histogram")
plt.fill()
plt.show()
#Class4
plt.figure()
plt.plot(bin4, histogram4)
plt.xlabel("Bins")
plt.ylabel("Pixels")
plt.title("Class 4 Edge Histogram")
plt.fill()
plt.show()
```




Class 2 Edge Histogram
540
520
500
480
460
440
420
400
Pixels
0.0
0.5
1.0
1.5
2.0
2.5
3.0
Bins


Class 3 Edge Histogram
550
500
450
400
350
Pixels
0.0
0.5
1.0
1.5
2.0
2.5
3.0
Bins

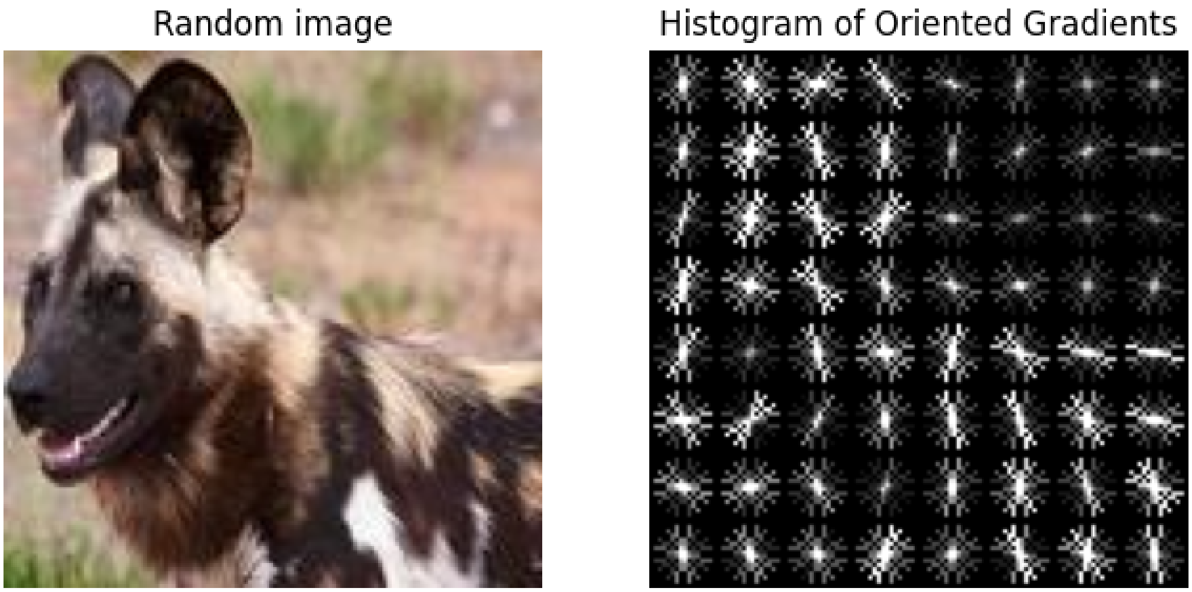


In [7]:
```
#Pick 2 edge histograms from the 4 you have constructed and perform histogram betwe
#We will use histogram 2 and histogram 3 for this example

#Euclidean Distance
euclideanDistance = sklearn.metrics.pairwise.euclidean_distances([histogram2],[hist
#Manhattan Distance
manhattanDistance = sklearn.metrics.pairwise.manhattan_distances([histogram2],[hist
#Cosine
cosineDistance = sklearn.metrics.pairwise.cosine_distances([histogram2],[histogram3
print(f"Euclidean Distance: {euclideanDistance[0][0]:.2f}")
print(f"Manhattan Distance: {manhattanDistance[0][0]:.2f}")
print(f"Cosine Distance: {cosineDistance[0][0]:.2f}")
```

```
Euclidean Distance: 421.14
Manhattan Distance: 2138.00
Cosine Distance: 0.01
```

In [12]:
```
#Histogram of Oriented Gradient(Hog)
#Picking one image to compute its HOGH descriptors
fd, image = ski.feature.hog(io.imread(randomImage4),orientations = 8,pixels_per_cel
fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(8, 4), sharex=True, sharey=True)

ax1.axis('off')
ax1.imshow(io.imread(randomImage4), cmap=plt.cm.gray)
ax1.set_title('Random image')

ax2.axis('off')
ax2.imshow(ski.exposure.rescale_intensity(image, in_range=(0, 10)), cmap=plt.cm.gra
```

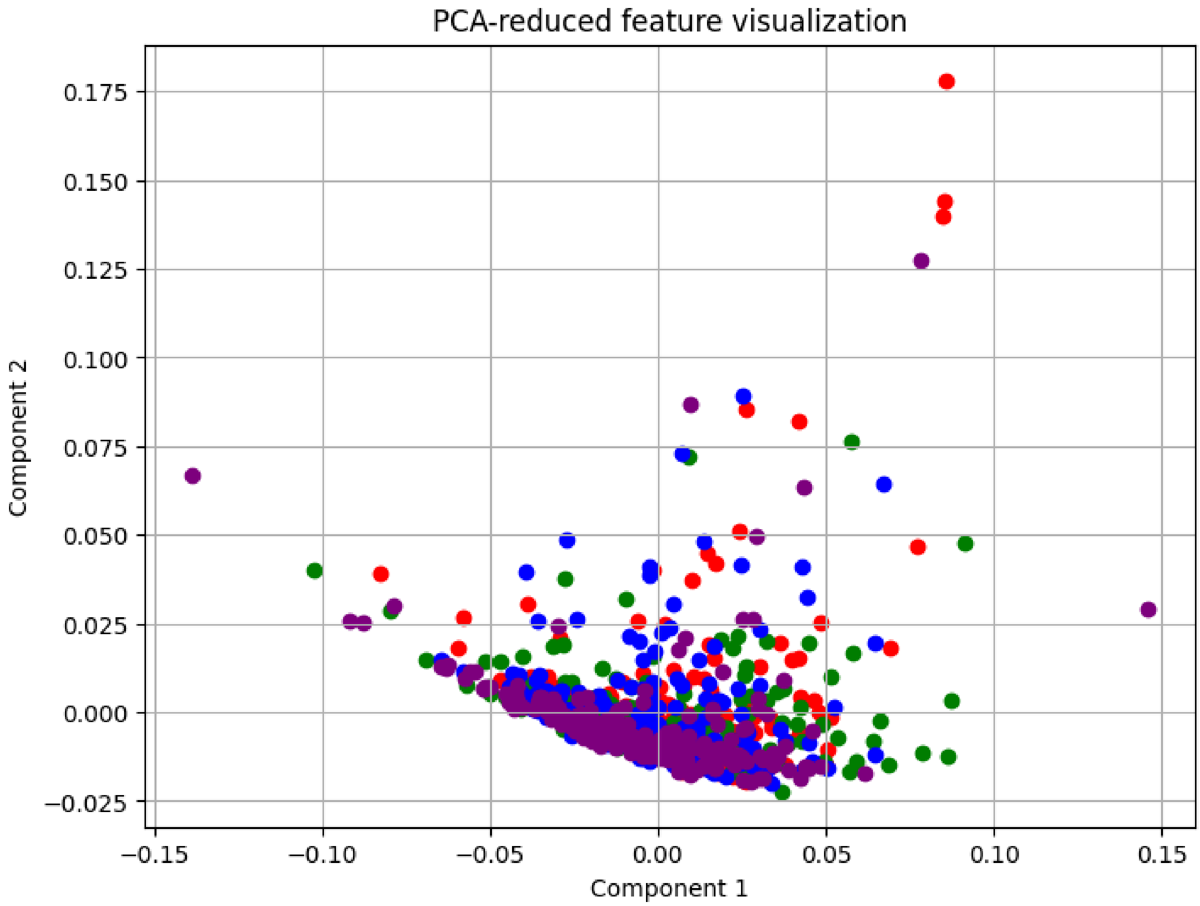
```
ax2.set_title('Histogram of Oriented Gradients')
plt.show()
```



In [15]:
```
#Dimensionality reduction (using Principal Component Analysis, PCA)
#Convert all the images from all classes into edge histograms
class1Directory = 'data/cropped/class1/n02093647-Bedlington_terrier'
class2Directory = 'data/cropped/class2/n02099849-Chesapeake_Bay_retriever'
class3Directory = 'data/cropped/class3/n02100735-English_setter'
class4Directory = 'data/cropped/class4/n02116738-African_hunting_dog'

masterDirectory = [class1Directory, class2Directory, class3Directory, class4Directo
masterHistograms = []
masterLabels = []

for class_idx, class_dir in enumerate(masterDirectory):
    for filename in os.listdir(class_dir):
        if filename.endswith((".jpg")):
            image_path = os.path.join(class_dir, filename)
            image = io.imread(image_path)
            image = rgb2gray(image)
            sobel_h = filters.sobel_h(image)
            sobel_v = filters.sobel_v(image)
            angles = angle(sobel_h, sobel_v)
            hist, nbin = ski.exposure.histogram(angles, nbins=36)
            hist = hist / np.sum(hist)
            masterHistograms.append(hist)
            masterLabels.append(class_idx)
```

In [16]:
```
#Dimensionality reduction from 36 to 2
histograms = np.array(masterHistograms)
labels = np.array(masterLabels)
pca = PCA(n_components=2)
histograms_pca = pca.fit_transform(histograms)
plt.figure(figsize=(8, 6))
colors = ['red', 'green', 'blue', 'purple']
for i in range(4):
    class_data = histograms_pca[labels == i]
```

```
    plt.scatter(class_data[:, 0], class_data[:, 1], color=colors[i], label=f'Class
plt.title('PCA-reduced feature visualization')
plt.xlabel('Component 1')
plt.ylabel('Component 2')
plt.grid(True)
plt.show()
```

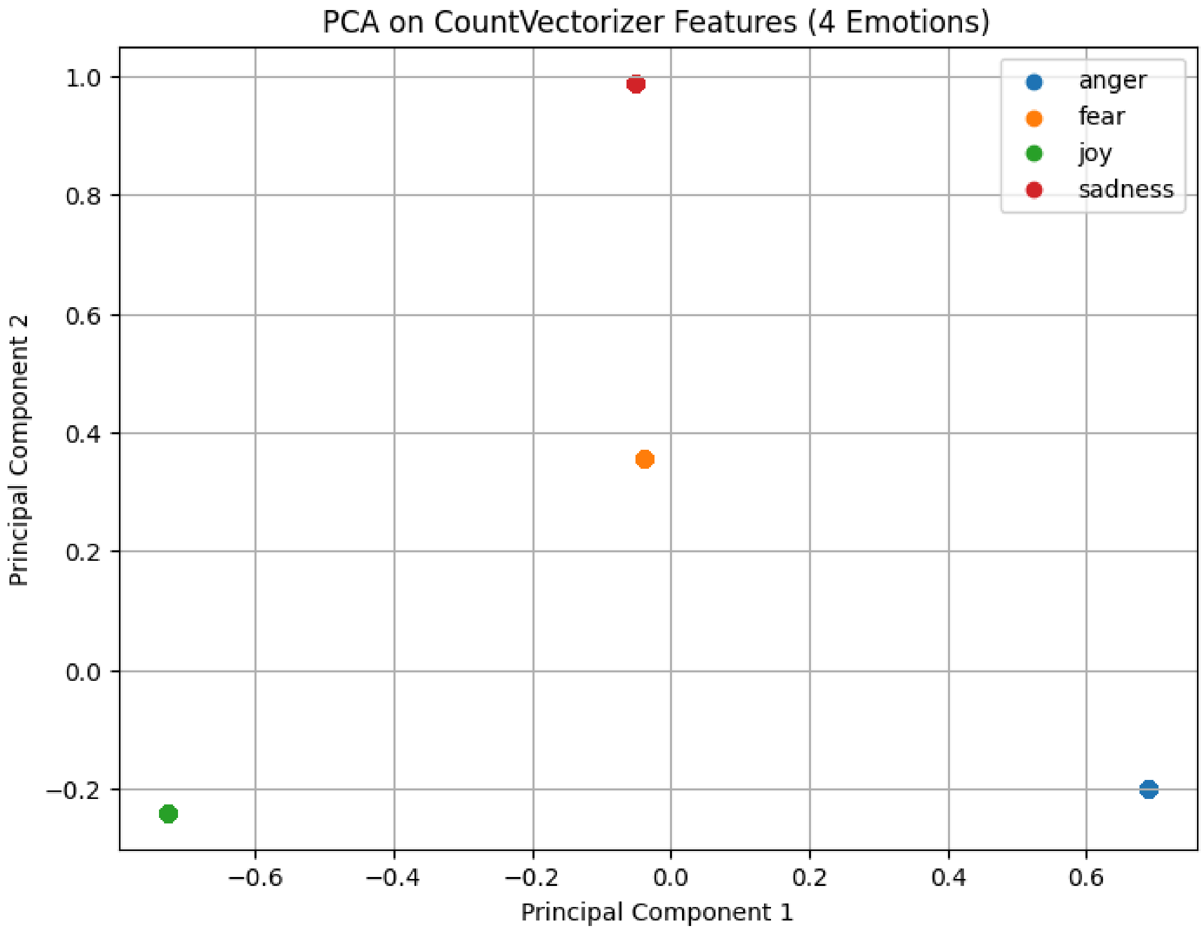


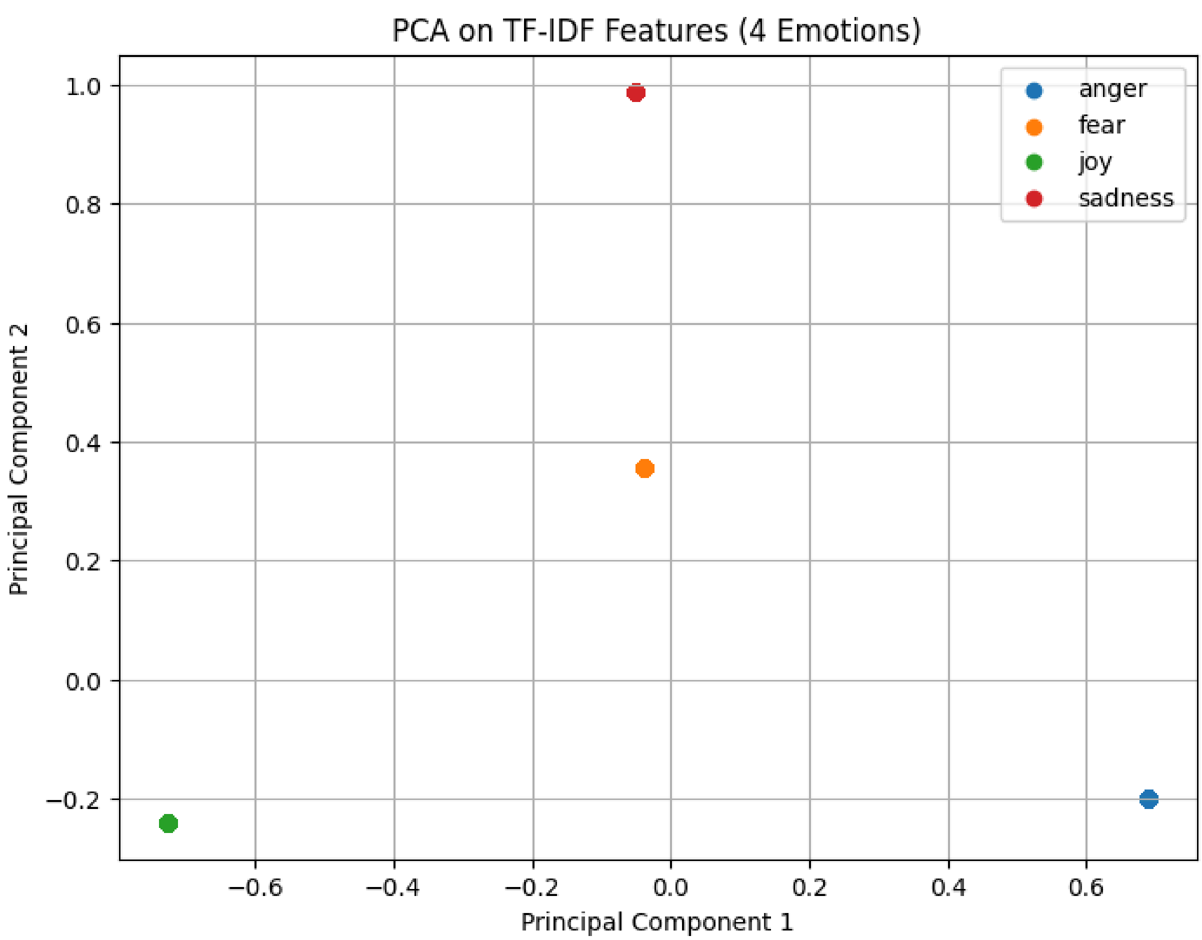
How many classes are visually separable (i.e., non-overlapping)? There seems to be no classes visually seperable. All 4 classes seem to be overlapping.

In [18]:
```
#4 Classes of emotions I have selected
selected_emotions = ['anger', 'joy', 'sadness', 'fear']
json_file_path = 'train.json'
labels = []

with open(json_file_path, 'r') as file:
    for line in file:
        data = json.loads(line)
        tweet_labels = [emotion for emotion in selected_emotions if data.get(emotio
        if tweet_labels:
            labels.append(tweet_labels[0])
count = CountVectorizer()
tfidf = TfidfVectorizer()
countFeatures = count.fit_transform(labels)
tfidfFeatures = tfidf_vectorizer.fit_transform(labels)
print(f"Dimensionality of countvectorizer is  {countFeatures.shape}")
print(f"Dimensionality of TF-IDF: {tfidfFeatures.shape}")
```

```
tfidfFeatures = tfidf_vectorizer.fit_transform(labels)

pca = PCA(n_components=2)
count_features_2d = pca.fit_transform(count_features.toarray())
tfidf_features_2d = pca.fit_transform(tfidf_features.toarray())

unique_labels = np.unique(labels)
plt.figure(figsize=(8, 6))
for class_label in unique_labels:
    indices = [i for i, label in enumerate(labels) if label == class_label]
    plt.scatter(count_features_2d[indices, 0], count_features_2d[indices, 1], label

plt.title('CountVectorizer')
plt.xlabel('Component 1')
plt.ylabel('Component 2')
plt.legend()
plt.grid(True)
plt.show()

plt.figure(figsize=(8, 6))
for class_label in unique_labels:
    indices = [i for i, label in enumerate(labels) if label == class_label]
    plt.scatter(tfidf_features_2d[indices, 0], tfidf_features_2d[indices, 1], label

plt.title('TF-IDF Features')
plt.xlabel('Component 1')
plt.ylabel('Component 2')
plt.legend()
plt.grid(True)
plt.show()
```

```
Dimensionality of countvectorizer is  (2701, 4)
Dimensionality of TF-IDF: (2701, 4)
```


PCA on CountVectorizer Features (4 Emotions)
anger
fear
joy
sadness
Principal Component 1
Principal Component 2
1.0
0.8
0.6
0.4
0.2
0.0
−0.2
−0.6
−0.4
−0.2
0.0
0.2
0.4
0.6


PCA on TF-IDF Features (4 Emotions)
anger
fear
joy
sadness
Principal Component 1
Principal Component 2
1.0
0.8
0.6
0.4
0.2
0.0
−0.2
−0.6
−0.4
−0.2
0.0
0.2
0.4
0.6

Plot the 2D points using four different colors for data from the four classes (see Figure 1) for both token count features and tf-idf features in two separate plots.

How many classes are visually separable (i.e., non-overlapping) for both plots? All the classes are visually seperable.