## **Image Process and Machine Learning**

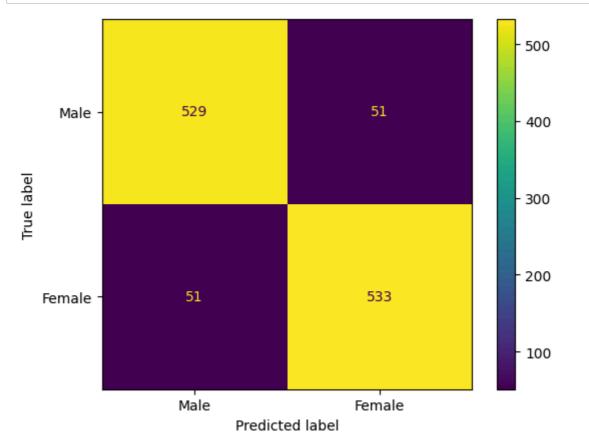
Dataset in use: <a href="https://www.kaggle.com/datasets/cashutosh/gender-classification-dataset/data">https://www.kaggle.com/datasets/cashutosh/gender-classification-dataset/data</a>)

dataset/data (<a href="https://www.kaggle.com/datasets/cashutosh/gender-classification-dataset/data">https://www.kaggle.com/datasets/cashutosh/gender-classification-dataset/data</a>)

```
In [1]: import os
        import cv2
        import random
        import pickle
        import logging
        import numpy as np
        from tqdm import tqdm
        from sklearn import metrics
        from skimage import feature
        from sklearn.svm import SVC
        import matplotlib.pyplot as plt
In [2]: import warnings
        warnings.filterwarnings("ignore")
In [3]: def metric_report(actual, predicted):
            acc = metrics.accuracy_score(actual, predicted)
            precision = metrics.precision_score(actual, predicted)
            recall = metrics.recall_score(actual, predicted)
            f1 = metrics.f1 score(actual, predicted)
            return (acc, precision, recall, f1)
In [4]: ext='jpg'
In [5]: def image reader(file path, in use, ext=ext):
            images = []
            files = os.listdir(file_path)
            random.shuffle(files)
            files = files[:int(len(files)*in use)]
            for file in tqdm(files, desc = 'Reading Images'):
                if file.endswith(ext):
                    img_path = os.path.join(file_path, file)
                    img = cv2.imread(img path, 0)
                    images.append(img)
            return images
In [6]: |male_dir = './dataset/Training/male'
        female_dir = './dataset/Training/female'
        male_test_dir = './dataset/Validation/male'
        female test dir = './dataset/Validation/female'
```

## **Histogram of Oriented Gradients**

```
In [12]: def hog_predict(model_instance, target, r_shape=(64, 128)):
             if os.path.isfile(target): logging.debug(f"{target} is a single image.")
             img resized = cv2.resize(target, r shape)
             hist = hog.compute(img_resized)
             hist = hist.flatten()
             pred = model_instance.predict(hist.reshape(1, -1))[0]
             return pred
In [13]: data_hog_male, labels_hog_male = hog_pattern_extractor(hog, train_male_image
         data_hog_female, labels_hog_female = hog_pattern_extractor(hog, train_female
         100%
         41/5941 [00:01<00:00, 4439.48it/s]
         100%
         10/5810 [00:01<00:00, 4422.32it/s]
In [14]: | data = np.vstack((data_hog_male, data_hog_female))
         labels = np.hstack((labels_hog_male, labels_hog_female))
In [15]: model_hog = SVC(kernel='linear', random_state=42)
         model_hog.fit(data, labels)
Out[15]:
                          SVC
                                            (https://scikit-
         SVC(kernel='linear', random_state=42)
         <
         actual_m, predicted_m = hog_detector(model_hog, test_male_images, 0)
In [16]:
         actual_f, predicted_f = hog_detector(model_hog, test_female_images, 1)
         SVM Inference: 100%
         580/580 [00:04<00:00, 129.63it/s]
         SVM Inference: 100%
         584/584 [00:04<00:00, 142.42it/s]
         actual = np.hstack((actual_m, actual_f))
In [17]:
         predicted = np.hstack((predicted_m, predicted_f))
In [18]: | metric_report(actual, predicted)
Out[18]: (0.9123711340206185,
          0.9126712328767124,
          0.9126712328767124,
          0.9126712328767124)
In [19]: confusion matrix = metrics.confusion matrix(actual, predicted)
```



```
In [21]: fname = 'hog_svm_model.sav'
# pickle.dump(model_hog, open(fname, 'wb'))
In [22]: loaded_model = pickle.load(open(fname, 'rb'))
In [23]: # Asagidaki img_path degiskenine istenen goruntulerin pathi verilir ve hog i
    img_path = './test_images/064943.jpg.jpg'
    img = cv2.imread(img_path, 0)
    actual = int(os.path.basename(img_path).startswith('1'))
    pred = hog_predict(loaded_model, img)
    if actual == pred:
        print('Gender Detected!')
```

Gender Detected!

## **Local Binary Patterns**

```
In [24]: | class LocalBinaryPatterns:
             def __init__(self, numPoints, radius):
                  self.numPoints = numPoints
                  self.radius = radius
             def describe(self, image, eps=1e-7):
                  lbp = feature.local_binary_pattern(image, self.numPoints, self.radiu
                  (hist, _) = np.histogram(lbp.ravel(),
                                           bins=np.arange(0, self.numPoints + 3),
                                           range=(0, self.numPoints + 2))
                  hist = hist.astype("float")
                  hist /= (hist.sum() + eps)
                  return hist
In [25]: | def lbp_detector(model, target_images, gender, ext=ext):
             actual, predicted = [], []
             for img in tqdm(target_images, desc='SVM Inference'):
                  hist = desc.describe(img)
                  pred = model.predict(hist.reshape(1, -1))[0]
                  actual.append(gender)
                  predicted.append(pred)
             return actual, predicted
In [26]: def lbp_predict(model_instance, target):
             hist = desc.describe(target)
             hist = hist.flatten()
              pred = model instance.predict(hist.reshape(1, -1))[0]
              return pred
In [27]: desc = LocalBinaryPatterns(20, 5)
In [28]: male histograms = np.load('male lbp histograms.npy')
         female histograms = np.load('female lbp histograms.npy')
In [29]: | data = np.vstack((male histograms, female histograms))
         labels = np.hstack( (np.zeros(len(male_histograms)), np.ones(len(female_hist
In [30]: model lbp = SVC(kernel='linear', random state=42)
         model lbp.fit(data, labels)
Out[30]:
                           SVC
                                              (https://scikit-
                                                arn.org/1.4/modules/generated/sklearn.svm.SVC.
          SVC(kernel='linear', random_state=42)
```

```
actual_m, predicted_m = lbp_detector(model_lbp, test_male_images, 0)
In [31]:
         actual_f, predicted_f = lbp_detector(model_lbp, test_female_images, 1)
         SVM Inference: 100%
         580/580 [00:03<00:00, 174.89it/s]
         SVM Inference: 100%
         584/584 [00:03<00:00, 171.71it/s]
In [32]: | actual = np.hstack((actual_m, actual_f))
         predicted = np.hstack((predicted_m, predicted_f))
In [33]: metric_report(actual, predicted)
Out[33]: (0.6821305841924399,
          0.7026515151515151,
          0.6352739726027398,
          0.6672661870503597)
In [34]: confusion_matrix = metrics.confusion_matrix(actual, predicted)
In [35]: cm_display = metrics.ConfusionMatrixDisplay(confusion_matrix = confusion_mat
         cm_display.plot()
         plt.show()
                                                                              400
                Male
                                423
                                                         157
                                                                              350
          True label
                                                                              300
                                                                             - 250
                                213
                                                         371
              Female
                                                                             - 200
                                Male
                                                       Female
                                       Predicted label
In [36]: fname = 'lbp svm model.sav'
         pickle.dump(model_lbp, open(fname, 'wb'))
In [37]: loaded model = pickle.load(open(fname, 'rb'))
```

```
In [38]: # Asagidaki img_path degiskenine istenen goruntulerin pathi verilir ve lbp i
    img_path = './test_images/064943.jpg.jpg'
    img = cv2.imread(img_path, 0)
    actual = int(os.path.basename(img_path).startswith('1'))
    pred = lbp_predict(loaded_model, img)
    if actual == pred:
        print('Gender Detected!')
```

Gender Detected!

## **Scale Invariant Feature Transform**

```
In [39]: sift = cv2.SIFT_create()
In [40]: def descriptor extractor(image):
             keypoints, descriptors = sift.detectAndCompute(image, None)
             return keypoints, descriptors
In [41]: def preprocessing(img, r_shape):
             img = cv2.resize(img, r_shape, interpolation=cv2.INTER_AREA)
             return img
In [42]: def sift_pattern_extractor(desc: cv2.HOGDescriptor, images, gender, num_desc
             data = []
             labels = []
             for img in tqdm(images):
                 img_resized = preprocessing(img, r_shape)
                 keypoints, descriptors = descriptor_extractor(img_resized)
                 if(len(descriptors) < num_descriptors):</pre>
                      continue
                 random_indices = random.sample(range(len(descriptors)), num_descript
                 random_descriptors = [descriptors[i] for i in random_indices]
                 data.append(random descriptors)
                 labels.append(gender)
             return data, labels
```

```
In [44]: #data_sift_male, labels_sift_male = sift_pattern_extractor(sift, train_male_
#data_sift_female, labels_sift_female = sift_pattern_extractor(sift, train_female)
```

```
In [45]: #data = np.vstack((data_sift_male, data_sift_female))
#labels = np.hstack((labels_sift_male, labels_sift_female))
```

```
In [46]: ## Bag of Words will be implemented (feedback)
```