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In [2]: import csv
        from PIL import Image
        import numpy as np
        import os
        import string
        import pickle
        from sklearn.preprocessing import StandardScaler
        from sklearn.pipeline import Pipeline
        from sklearn.model selection import train test split
        from sklearn.metrics import classification_report
        from sklearnex import patch sklearn
        from sklearn.svm import LinearSVC
```

1 Define function

from matplotlib import pyplot as plt

```
In [4]: |# load image from csv and return image on folder
        def csv to image(csv path,csv image shape initialized, image shape target, image folder path):
            something to note:
            - don't include the header on the csv data
            - each row save the image where the first column is the label followed the pixel value
            - this function just good at minimize image not to enlarge, since interpolation used are good for minimize
            args:
            csv path : where the csv that will be loaded saved
            csv image shape : image shape saved on the csv
            image shape target : size of the image you want as the output
            1.1.1
            count = 1
            #get list alphabet
            Alphabet Mapping List = list(string.ascii uppercase)
            # Make folder of each alphabet, if there no folder with a specific alphabel listed on Alphabel Mapping list
            for alphabet in Alphabet Mapping List:
                path = image folder path + '/' + alphabet
                if not os.path.exists(path):
                    os.makedirs(path)
            with open(csv path, newline='') as csvfile:
                # make a object that can help to iterate over lines of CSV file
                reader = csv.reader(csvfile, delimiter=',', quotechar='|')
                # Iterate over lines
                for row in reader:
                    # in each row column 1 is the label and saved in index 0. so we need indexing to row 0 to get the la
                    digit Name = row[0]
                    # in each row, column 2 to the end is the pixel of the image saved
                    image array = np.asarray(row[1:])
                    #reshaping the array from (784,) to (28,28) since image should be 2 dimension atleast
                    image array = image array.reshape(csv image shape initialized)
                    # buld image object
```

```
new_image = Image.fromarray(image_array.astype('uint8'))
# Resize to the Image to the size we want
new_image = new_image.resize(image_shape_target)

#convert integer value being alphabel, example, if zero then "A", if 1 then "B"
label = Alphabet_Mapping_List[int(digit_Name)]

image_Path = image_folder_path + '/' + label + '/' + str(label) + '-' + str(count) + '.png'
new_image.save(image_Path)
count = count + 1

if count % 1000 == 0:
    print ("Images processed: " + str(count))
```

4

```
In [5]: #load image from folder and return stacked numpy array for handwritten recognition
        def load image(directory, image folder path=(24, 24)):
            1.1.1
            args:
            - directory : directory path of all image folder saved
            - image folder path : target image you want as return
            return:
            stacked image aray with the label
            total image = 0
            label names list = [folder for folder in os.listdir(directory)]
            #Calculate how many image to make multidimensional Array
            for label name in label names list :
                total image += len(os.listdir(os.path.join(directory,label name)))
            #Define array of image and target
            stack of image = np.empty((total image,image shape target[0],image shape target[1]))
            stack of label = np.empty(total image)
            temp = total image
            for i, label name in enumerate(label names list) :
                image class path = os.path.join(directory,label name)
            for image in os.listdir(image class path) :
                image = os.path.join(image class path,image)
                #Open Image
                image = Image.open(image)
                #Resize Image
                resized image = image.resize(image shape target)
                resized image array = np.asarray(resized image)
                #Append to Array
                stack of image[total image-temp] = resized image array
                stack of label[total image-temp] = i
                temp -= 1
            return (stack of image, stack of label)
```

```
In [6]: # Load Image from csv and return stacked numpy array for handwritten recognition
        def csv to array(csv path,csv image shape initialized, image shape target):
            something to note:
            - don't include the header on the csv data
            - each row save the image where the first column is the label followed the pixel value
            - this function just good at minimize image not to enlarge, since interpolation used are good for minimize
            args:
            csv path: where the csv that will be loaded saved
            csv image shape initialized : image shape saved on the csv
            image shape target : size of the image you want as the output
            1.1.1
            count = 1
            X stacked = []
            y stacked = []
            with open(csv path, newline='') as csvfile:
                # make a object that can help to iterate over lines of CSV file
                reader = csv.reader(csvfile, delimiter=',', quotechar='|')
                # Iterate over lines
                for row in reader:
                    # in each row column 1 is the label and saved in index 0. so we need indexing to row 0 to get the la
                    y_stacked.append(int(row[0]))
                    # in each row, column 2 to the end is the pixel of the image saved
                    image array = np.asarray(row[1:])
                    #reshaping the array from (784,) to (28,28) since image should be 2 dimension atleast
                    image array = image array.reshape(csv image shape initialized)
                    # buld image object
                    new image = Image.fromarray(image array.astype('uint8'))
                    # Resize to the Image to the size we want
                    new image = new image.resize(image shape target)
                    # Append to the list
                    X stacked.append(np.array(new image))
```

```
X_stacked = np.array(X_stacked)
y_stacked = np.array(y_stacked)
return X_stacked, y_stacked
```

2 Data Preparation

```
In [13]: # Run this code to convert the CSV file to image
         csv to image(csv path,csv image shape initialized, image shape target, image folder path)
         KeyboardInterrupt
                                                    Traceback (most recent call last)
         Input In [13], in <cell line: 2>()
               1 # Run this code to convert the CSV file to image
         ----> 2 csv to image(csv path,csv image shape initialized, image shape target, image folder path)
         Input In [9], in csv to image(csv path, csv image shape initialized, image shape target, image folder path)
              49 label = Alphabet Mapping List[int(digit Name)]
              51 image_Path = image_folder_path + '/' + label + '/' + str(label) + '-' + str(count) + '.png'
         ---> 52 new image.save(image Path)
              53 count = count + 1
              55 if count % 1000 == 0:
         File /usr/local/lib/python3.9/dist-packages/PIL/Image.py:2320, in Image.save(self, fp, format, **params)
            2317
                         fp = builtins.open(filename, "w+b")
            2319 try:
         -> 2320
                     save handler(self, fp, filename)
            2321 except Exception:
            2322
                     if open fp:
         File /usr/local/lib/python3.9/dist-packages/PIL/PngImagePlugin.py:1388, in save(im, fp, filename, chunk, save
         _all)
            1385 chunk(fp, b"IEND", b"")
            1387 if hasattr(fp, "flush"):
         -> 1388
                     fp.flush()
         KeyboardInterrupt:
```

2.1 Load Data

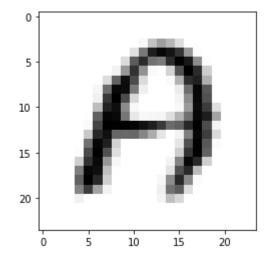
----> 2 X, y = image_to_array(image_folder_path, image_size=(24, 24))

NameError: name 'image_to_array' is not defined

In [12]: # Run this code to Load code from csv and save to array for Handwritten recognition
X, y = csv_to_array(csv_path,csv_image_shape_initialized, image_shape_target)

In [13]: # show the image
plt.imshow(X[2], cmap= 'binary')

Out[13]: <matplotlib.image.AxesImage at 0x198d3d45990>



3 Modeling

3.1 Flattening The Image

```
In [14]: # Image need to flattened before classify
# (num_of_image,width, height) -> (num_of_image, width*height)
X = X.reshape(X.shape[0], X.shape[1]* X.shape[2])
```

3.1.1 Train Test Splitting

```
In [15]: # Splitting between train and test
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 123321)
```

3.1.2 Pipelining

```
In [17]: # Activate Intel Extension so the algorith can run faster
patch_sklearn()

# define the algorithm
linearsvc = LinearSVC(random_state=0, tol=1e-5)

# so the step each image will follow this step before can be trained or inference
svc_steps = [
    ('scaler', StandardScaler()),
    ('classify', linearsvc)
]

# build pipeline from the svc_steps
svc_pipe = Pipeline(svc_steps)
```

Intel(R) Extension for Scikit-learn* enabled (https://github.com/intel/scikit-learn-intelex)

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

4 Evaluation

```
In [19]: # Predict the test data
y_pred = svc_pipe.predict(X_test)
```

4.1 Precision, Recall, F1-score and Accuracy

```
print(classification_report(y_test, y_pred))
        precision
                                recall f1-score
                                                support
                         0.83
                 0
                                 0.80
                                          0.81
                                                   2751
                 1
                         0.81
                                 0.66
                                          0.73
                                                   1737
                                 0.89
                                          0.88
                  2
                         0.87
                                                   4760
                  3
                         0.75
                                 0.67
                                          0.71
                                                   2030
                         0.80
                                          0.78
                  4
                                 0.75
                                                   2267
                  5
                                          0.85
                         0.92
                                 0.78
                                                    241
                 6
                         0.80
                                          0.75
                                 0.71
                                                   1143
                         0.71
                                          0.69
                  7
                                 0.66
                                                   1407
                 8
                         0.86
                                 0.82
                                          0.84
                                                    230
                                          0.71
                 9
                         0.76
                                 0.67
                                                   1706
                         0.78
                                          0.73
                 10
                                 0.69
                                                   1101
                         0.92
                                          0.93
                 11
                                 0.95
                                                   2284
                         0.85
                                          0.87
                 12
                                 0.88
                                                   2449
                 13
                         0.78
                                 0.78
                                          0.78
                                                   3805
                         0.89
                                          0.92
                 14
                                 0.96
                                                  11611
                         0.86
                                          0.88
                 15
                                 0.91
                                                   3872
                 16
                         0.79
                                 0.72
                                          0.75
                                                   1164
                                 0.74
                 17
                         0.77
                                          0.76
                                                   2282
                 18
                         0.92
                                 0.93
                                          0.92
                                                   9705
                         0.90
                                 0.94
                                          0.92
                 19
                                                   4486
                         0.83
                                          0.84
                 20
                                 0.85
                                                   5792
                 21
                         0.85
                                 0.88
                                          0.87
                                                    844
                         0.79
                                          0.77
                 22
                                 0.76
                                                   2131
                 23
                         0.84
                                 0.76
                                          0.80
                                                   1329
                         0.81
                                          0.81
                 24
                                 0.81
                                                   2173
                                          0.83
                 25
                         0.86
                                 0.80
                                                   1191
                                          0.85
                                                  74491
           accuracy
           macro avg
                         0.83
                                 0.80
                                          0.81
                                                  74491
```

0.85

0.85

74491

5 Save the model

0.85

weighted avg

```
In [17]: filename = 'svc_model.sav'
In []: # save the model to disk
pickle.dump(svc_pipe, open(filename, 'wb'))
```

6 Load the model

```
In [18]: # load the model from disk
svc_pipe = pickle.load(open(filename, 'rb'))
```