Task

Kindly download the dataset of Fruits 360 from Kaggle
https://www.kaggle.com/datasets/moltean/fruits
a.Convert all the images which are in the training folder into 1D with labels.
b. Apply Classification models on that dataset and show which one performs best on their best parame ters.

Then write the deployment code of that model.

```
In [20]: import os
    import numpy as np
    import pandas as pd
    import cv2
    from sklearn.model_selection import train_test_split
    from sklearn.preprocessing import LabelEncoder
    from sklearn.svm import SVC
    from sklearn.metrics import classification_report ,accuracy_score
    import matplotlib.pyplot as plt
    from sklearn.ensemble import RandomForestClassifier
```

```
In [2]: print(os.listdir("../OpenCvDataSet/fruits-360-original-size/fruits-360-original-size/Training"))
```

```
['apple_6', 'apple_braeburn_1', 'apple_crimson_snow_1', 'apple_golden_1', 'apple_golden_2', 'apple_golden_
3', 'apple_granny_smith_1', 'apple_hit_1', 'apple_pink_lady_1', 'apple_red_1', 'apple_red_2', 'apple_red_3',
'apple_red_delicios_1', 'apple_red_yellow_1', 'apple_rotten_1', 'cabbage_white_1', 'carrot_1', 'cucumber_1',
'cucumber_3', 'eggplant_violet_1', 'pear_1', 'pear_3', 'zucchini_1', 'zucchini_dark_1']
```

```
In [3]: hs = [ ython Languages\Navttac AI course\OpenCvDataSet\fruits-360-original-size\fruits-360-original-size\Training\approx ython Languages\Navttac AI course\OpenCvDataSet\fruits-360-original-size\fruits-360-original-
```

```
In [4]: ### This file will
        dataset =[]
        for i in folder paths:
            folder name = os.path.basename(i)
            for file name in os.listdir(i):
                image path = os.path.join(i, file name)
                if os.path.isfile(image path): # Only consider files
                    # Load the image using OpenCV
                    image = cv2.imread(image path, cv2.IMREAD GRAYSCALE)
                    # If the image was successfully loaded
                    if image is not None:
                        # Resize the grayscale image to 250X250 pixels
                        resized_image = cv2.resize(image, (250, 250))
                        # Flatten the image and append each pixel as a separate feature along with the label to the d
                        flattened image = resized image.flatten().tolist()
                        dataset.append(flattened image + [folder name])
```

```
In [5]: ## convert the matrix into dataFrame and change last column name into Label

img_df = pd.DataFrame(data =dataset)
img_df.rename(columns={img_df.iloc[:, -1].name : "label"},inplace=True)
img_df
```

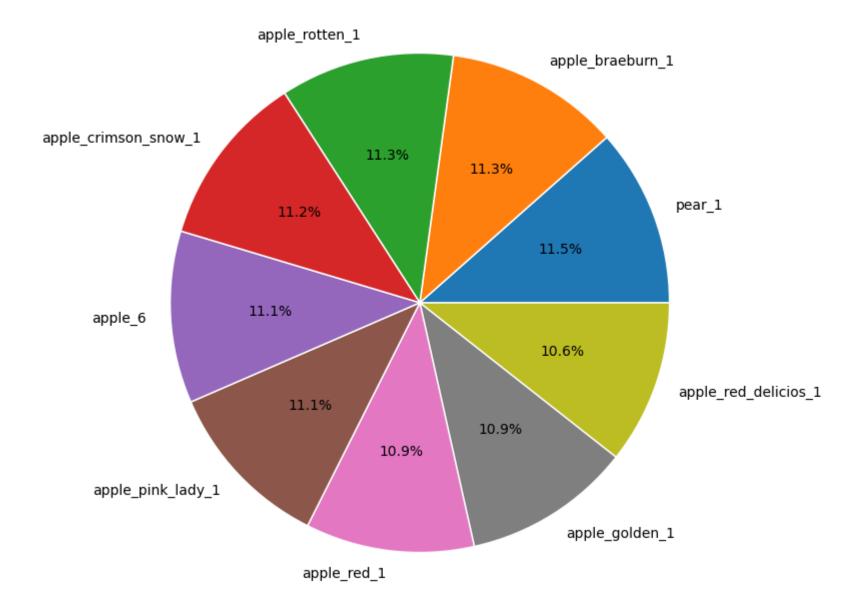
Out[5]:

	0	1	2	3	4	5	6	7	8	9	 62491	62492	62493	62494	62495	62496	62497	62498	62499	label
0	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	apple_6
1	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	apple_6
2	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	apple_6
3	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	apple_6
4	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	apple_6
2823	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	pear_1
2824	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	pear_1
2825	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	pear_1
2826	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	pear_1
2827	255	255	255	255	255	255	255	255	255	255	 255	255	255	255	255	255	255	255	255	pear_1

2828 rows × 62501 columns

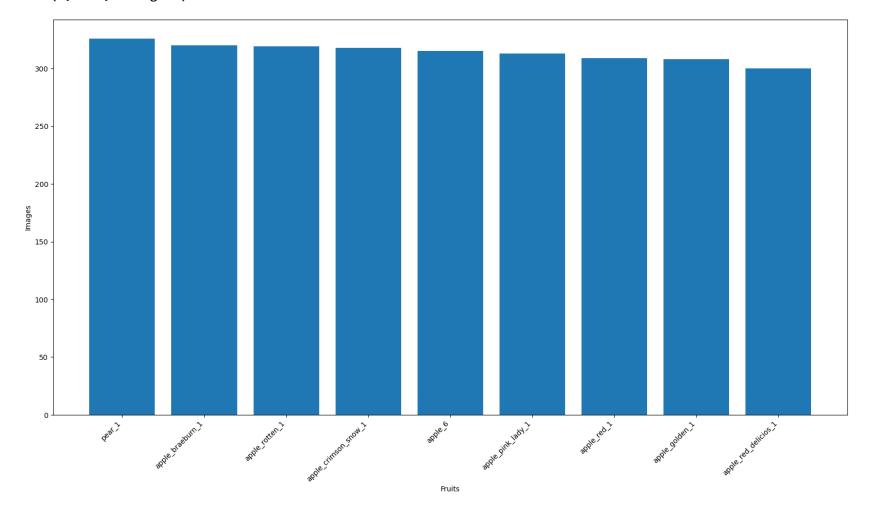
```
In [6]: ## Count the Label value in row
        y = img_df.iloc[: , -1]
        y.value_counts()
Out[6]: pear_1
                                326
        apple_braeburn_1
                                320
        apple_rotten_1
                                319
        apple_crimson_snow_1
                                318
        apple_6
                                315
        apple_pink_lady_1
                                313
        apple_red_1
                                309
        apple_golden_1
                                308
        apple_red_delicios_1
                                300
        Name: label, dtype: int64
```

```
In [8]: ##
plt.figure(figsize=(8,8))
plt.pie(y.value_counts(),labels=y.value_counts().index,autopct='%1.1f%%',explode = [0.01 for i in range(9)]);
```



```
In [9]:
    plt.figure(figsize=(20,10))
    plt.bar(y.value_counts().index, y.value_counts().values ,)
    plt.xticks(rotation=45, ha='right');
    plt.xlabel("Fruits")
    plt.ylabel("Images")
```

Out[9]: Text(0, 0.5, 'Images')



```
In [10]: ## Split data into x and Y
         x = img_df.drop("label" , axis =1)
         y = img df.iloc[:,-1]
In [11]: | ## scale the Values in Range of 0 to 1 & label encode the value of y
         x \text{ scaled} = x/255
         encoder = LabelEncoder()
         y encoded = encoder.fit transform(y)
         encoder.classes
Out[11]: array(['apple_6', 'apple_braeburn_1', 'apple_crimson_snow_1',
                 'apple golden 1', 'apple pink lady 1', 'apple red 1',
                 'apple red delicios 1', 'apple rotten 1', 'pear 1'], dtype=object)
In [12]: ## Train Test split
         x_train ,x_test ,y_train ,y_test = train_test_split(x_scaled , y_encoded ,test_size = 0.2 ,random_state = 4
In [21]: ## Model selection
         clf = RandomForestClassifier()
         clf.fit(x train ,y train)
Out[21]:
          ▼ RandomForestClassifier
          RandomForestClassifier()
In [22]: | ##Basic Evalution
         print("svc.score",clf.score(x test ,y test))
         y preds = clf.predict(x test)
         print("Accuracy score", accuracy score(y test ,y preds))
         svc.score 1.0
         Accuracy score 1.0
```

```
In [23]: ### Classification Report
print(classification_report(y_test , y_preds))
```

```
precision
                            recall f1-score
                                                 support
           0
                    1.00
                              1.00
                                         1.00
                                                      63
                                                      70
           1
                    1.00
                              1.00
                                         1.00
           2
                    1.00
                              1.00
                                         1.00
                                                      60
           3
                    1.00
                               1.00
                                         1.00
                                                      54
           4
                    1.00
                              1.00
                                         1.00
                                                      72
           5
                    1.00
                              1.00
                                         1.00
                                                      62
           6
                                                      52
                    1.00
                              1.00
                                         1.00
           7
                                                      69
                    1.00
                              1.00
                                         1.00
           8
                    1.00
                              1.00
                                         1.00
                                                      64
                                         1.00
                                                     566
    accuracy
                                         1.00
                                                     566
   macro avg
                    1.00
                               1.00
weighted avg
                    1.00
                              1.00
                                         1.00
                                                     566
```

```
In [47]: ### DepLoyment: Take a picture of a

inputDataSet = []
input_img = cv2.imread("../OpenCvDataSet/fruits-360-original-size/fruits-360-original-size/Test/apple_crimson

resized_input_image = cv2.resize(input_img, (250, 250))
flattened_input_image = resized_input_image.flatten().tolist()
inputDataSet.append(flattened_input_image)
```

```
In [48]: ##Scale the image before converting it into dataFrame
         inp = np.array(inputDataSet) / 255
         deploy df = pd.DataFrame(inp)
         deploy df
Out[48]:
                                                9 ... 62490 62491 62492 62493 62494 62495 62496 62497 62498 62499
          0 1.0 1.0 1.0 1.0 1.0 1.0 1.0 1.0 1.0 ...
                                                        1.0
                                                              1.0
                                                                    1.0
                                                                          1.0
                                                                                1.0
                                                                                      1.0
                                                                                             1.0
                                                                                                   1.0
                                                                                                         1.0
                                                                                                               1.0
         1 rows × 62500 columns
In [43]:
         encoder.classes_
Out[43]: array(['apple_6', 'apple_braeburn_1', 'apple_crimson_snow_1',
                 'apple_golden_1', 'apple_pink_lady_1', 'apple_red_1',
                 'apple_red_delicios_1', 'apple_rotten_1', 'pear_1'], dtype=object)
In [45]: ## predict the Fruit
         labels = {i:label for i,label in enumerate(encoder.classes )}
         y = svc.predict(deploy df)
         print("The Fruit is",labels[y[0]])
         The Fruit is apple crimson snow 1
```