## 

Obtaining dependency information for opencv-python from https://files.pythonhosted.org/packages/66/82/564168a349148298aca281e342551404ef 5521f33fba17b388ead0a84dc5/opencv\_python-4.10.0.84-cp37-abi3-macosx\_1 1\_0\_arm64.whl.metadata (https://files.pythonhosted.org/packages/66/8 2/564168a349148298aca281e342551404ef5521f33fba17b388ead0a84dc5/opencv python-4.10.0.84-cp37-abi3-macosx 11 0 arm64.whl.metadata)

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Requirement already satisfied: numpy>=1.21.2 in ./anaconda3/lib/pytho n3.11/site-packages (from opencv-python) (1.24.3)

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Installing collected packages: opencv-python Successfully installed opencv-python-4.10.0.84

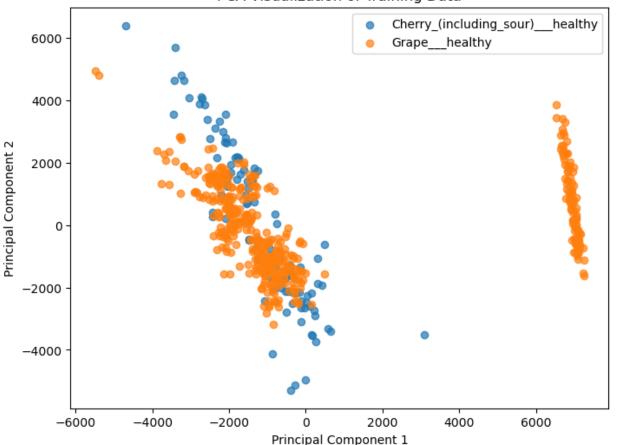
from sklearn.metrics import confusion\_matrix

## In [35]: import numpy as np import cv2 from sklearn.decomposition import PCA from sklearn.svm import SVC from sklearn.model\_selection import train\_test\_split from sklearn.metrics import classification\_report import os import matplotlib.pyplot as plt from tensorflow.keras.models import Sequential from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Der from tensorflow.keras.utils import to\_categorical import seaborn as sns

```
In [36]: # --- Step 1: Load Plant Image Dataset ---
         def load_images_from_folder(folder_path, target_size=(128, 128)):
             """Load images from a folder, resize them, and convert to grayscal
             images = []
             labels = []
             for label in os.listdir(folder_path):
                 label_path = os.path.join(folder_path, label)
                 if os.path.isdir(label path):
                     for file in os.listdir(label_path):
                         file path = os.path.join(label path, file)
                         if file.lower().endswith(('.jpg', '.jpeg', '.png')):
                             img = cv2.imread(file path, cv2.IMREAD GRAYSCALE)
                             if img is not None:
                                 img = cv2.resize(img, target size)
                                 images.append(img)
                                 labels.append(label)
                             else:
                                 print(f"Warning: Unable to load image {file_pa
             if len(images) == 0:
                 raise ValueError("No valid images found in the dataset folder.
             return np.array(images), np.array(labels)
In [37]: # Load your dataset here
         data_folder = "/Users/jayant/Documents/MA 544"
         images, labels = load_images_from_folder(data_folder)
In [38]: # Flatten the images for numerical processing
         X = images.reshape(images.shape[0], -1)
         # Encode labels as integers
         from sklearn.preprocessing import LabelEncoder
         label_encoder = LabelEncoder()
         y = label encoder.fit transform(labels)
In [39]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.
In [40]: # Compute PCA for dimensionality reduction
         pca = PCA(n_components=50) # Reduce to 50 principal components
         X train pca = pca.fit transform(X train)
         X test pca = pca.transform(X test)
```

```
In [41]: PCA visualization
ef plot_pca(X_pca, y, title="PCA Visualization"):
    """Visualize the PCA-reduced data."""
    plt.figure(figsize=(8, 6))
    for i, label in enumerate(np.unique(y)):
        plt.scatter(X_pca[y == label, 0], X_pca[y == label, 1], label=laplt.title(title)
    plt.xlabel("Principal Component 1")
    plt.ylabel("Principal Component 2")
    plt.legend()
    plt.show()
lot_pca(X_train_pca, y_train, title="PCA Visualization of Training Data
```

## PCA Visualization of Training Data



```
In [42]: # Use Support Vector Machine (SVM) for classification
svm_classifier = SVC(kernel='linear', random_state=42)
svm_classifier.fit(X_train_pca, y_train)
```

```
Out[42]: 

SVC

SVC(kernel='linear', random_state=42)
```

```
In [44]: # Generate a classification report
    print("Classification Report:")
    print(classification_report(y_test, y_pred, target_names=label_encoder

# --- Step 6: Detect Disease in a New Image ---
    def predict_disease(image_path):
        """Predict the disease for a given plant image."""
        img = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
        if img is None:
            raise ValueError(f"Unable to read the image at {image_path}. Coing = cv2.resize(img, (128, 128))
        img_flattened = img.flatten().reshape(1, -1)
        img_pca = pca.transform(img_flattened)
```

return label\_encoder.inverse\_transform(prediction)[0]

prediction = svm\_classifier.predict(img\_pca)

## Classification Report:

In [43]: # --- Step 5: Evaluate the Model ---

y\_pred = svm\_classifier.predict(X\_test\_pca)

pport	precision	recall	f1-score	su
pport				
Cherry_(including_sour)health 18	0.74	0.78	0.76	
Grapehealth	o.97	0.96	0.96	
accurac	cy		0.94	
macro av	g 0.85	0.87	0.86	
141 weighted av	g 0.94	0.94	0.94	
141				

```
In [45]: # Example usage
  new_image_path = "/Users/jayant/Documents/MA 544/Cherry_(including_sour result = predict_disease(new_image_path)
  print(f"The Plant is: {result}")
```

The Plant is: Cherry\_(including\_sour)\_\_\_healthy

```
In [46]: # Preprocess images for CNN
         images_cnn = images / 255.0 # Normalize pixel values to [0, 1]
         labels_cnn = to_categorical(y) # One-hot encode labels
         # Split data for CNN
         X_train_cnn, X_test_cnn, y_train_cnn, y_test_cnn = train_test_split(im
         # Define the CNN model
         cnn_model = Sequential([
             Conv2D(32, (3, 3), activation='relu', input_shape=(128, 128, 1)),
             MaxPooling2D((2, 2)),
             Conv2D(64, (3, 3), activation='relu'),
             MaxPooling2D((2, 2)),
             Flatten(),
             Dense(128, activation='relu'),
             Dropout(0.5),
             Dense(len(np.unique(y)), activation='softmax') # Output layer for
         ])
         cnn_model.compile(optimizer='adam', loss='categorical_crossentropy', m
```

```
In [47]: # Train the CNN
       cnn_history = cnn_model.fit(
           X_train_cnn, y_train_cnn,
           validation_data=(X_test_cnn, y_test_cnn),
           epochs=10,
           batch_size=32
       Epoch 1/10
       18/18 [============= ] - 3s 132ms/step - loss: 0.6607
       - accuracy: 0.7815 - val_loss: 0.4213 - val_accuracy: 0.8723
       Epoch 2/10
       - accuracy: 0.8117 - val loss: 0.3510 - val accuracy: 0.8723
       Epoch 3/10
       18/18 [============== ] - 2s 134ms/step - loss: 0.4225
       - accuracy: 0.8295 - val_loss: 0.3149 - val_accuracy: 0.9078
       Epoch 4/10
       18/18 [============== ] - 2s 131ms/step - loss: 0.3240
       - accuracy: 0.8810 - val_loss: 0.1924 - val_accuracy: 0.9078
       Epoch 5/10
       18/18 [=============== ] - 2s 132ms/step - loss: 0.2038
       - accuracy: 0.9147 - val_loss: 0.1360 - val_accuracy: 0.9645
       Epoch 6/10
       18/18 [=============== ] - 2s 132ms/step - loss: 0.1543
       - accuracy: 0.9485 - val loss: 0.1186 - val accuracy: 0.9645
       Epoch 7/10
       18/18 [============== ] - 2s 132ms/step - loss: 0.1059
       - accuracy: 0.9556 - val_loss: 0.1089 - val_accuracy: 0.9433
       Epoch 8/10
       18/18 [============== ] - 2s 133ms/step - loss: 0.0977
       - accuracy: 0.9663 - val_loss: 0.0829 - val_accuracy: 0.9716
       Epoch 9/10
       - accuracy: 0.9840 - val loss: 0.0888 - val accuracy: 0.9716
       Epoch 10/10
       18/18 [============== ] - 2s 135ms/step - loss: 0.0567
       - accuracy: 0.9876 - val_loss: 0.0697 - val_accuracy: 0.9787
In [48]: # Evaluate the CNN
       cnn_loss, cnn_accuracy = cnn_model.evaluate(X_test_cnn, y_test_cnn)
       print(f"CNN Test Accuracy: {cnn_accuracy:.2f}")
       5/5 [=========== ] - 0s 30ms/step - loss: 0.0697 -
       accuracy: 0.9787
```

CNN Test Accuracy: 0.98

```
In [49]: # Prediction using CNN
         def predict_disease_cnn(image_path):
             """Predict the disease for a given plant image using CNN."""
             img = cv2.imread(image_path, cv2.IMREAD_GRAYSCALE)
             if img is None:
                 raise ValueError(f"Unable to read the image at {image_path}. (
             img = cv2.resize(img, (128, 128)).astype("float32") / 255.0
             img = np.expand_dims(img, axis=(0, -1)) # Add batch and channel d
             prediction = cnn model.predict(img)
             return label encoder.inverse transform([np.argmax(prediction)])[0]
In [50]: # Example usage
         result cnn = predict disease cnn(new image path)
         print(f"The predicted plant disease using CNN is: {result_cnn}")
         1/1 [======= ] - 0s 31ms/step
         The predicted plant disease using CNN is: Cherry (including sour) h
         ealthy
In [51]: # Function to plot confusion matrix
         def plot_confusion_matrix(y_true, y_pred, class_names, model_name="Mod
             cm = confusion_matrix(y_true, y_pred)
             plt.figure(figsize=(10, 8))
             sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=cla
             plt.xlabel("Predicted")
             plt.ylabel("Actual")
             plt.title(f"Confusion Matrix for {model name}")
```

plt.show()

In [52]: # For SVM Model
y\_pred\_svm = svm\_classifier.predict(X\_test\_pca) # Predictions from SV
plot\_confusion\_matrix(y\_test, y\_pred\_svm, class\_names=label\_encoder.cl

