Face Recognition Using SVM

AIM:

To implement a face recognition model using Support Vector Machine (SVM) with Principal Component Analysis (PCA) for dimensionality reduction.

ALGORITHM:

- **Step 1:** Load the Labeled Faces in the Wild (LFW) dataset.
- **Step 2:** Flatten the face images into 1D feature vectors.
- Step 3: Normalize the data using StandardScaler.
- **Step 4:** Split the dataset into training and testing sets (80% train, 20% test).
- **Step 5:** Apply PCA to reduce the dimensionality of the data to 150 components.
- **Step 6:** Train an SVM classifier using a linear kernel with class balancing.
- **Step 7:** Predict the labels for the test data using the trained SVM model.
- **Step 8:** Calculate and display the accuracy of the model.
- Step 9: Display a confusion matrix to evaluate the model's performance.
- **Step 10:** Test the model with a sample image and show the predicted label.

SOURCE CODE:

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.datasets import fetch_lfw_people
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.decomposition import PCA
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, confusion_matrix

# Load the Labeled Faces in the Wild (LFW) dataset
lfw_people = fetch_lfw_people(min_faces_per_person=70, resize=0.4)
X = lfw_people.images # Face images (Gray-scale)
y = lfw_people.target # Person labels
target_names = lfw_people.target_names # Names of people

# Flatten images for SVM input (Convert 2D images to 1D feature vectors)
```

```
n samples, h, w = X.shape
X = X.reshape(n samples, h * w)
# Normalize data
scaler = StandardScaler()
X = scaler.fit transform(X)
# Split data (80% training, 20% testing)
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
# Apply PCA (Principal Component Analysis) for dimensionality reduction
n components = 150 # Reduce features to 150 dimensions
pca = PCA(n components=n components, whiten=True)
X train pca = pca.fit transform(X train)
X test pca = pca.transform(X test)
# Train SVM classifier
svm_classifier = SVC(kernel="linear", class_weight="balanced", probability=True)
svm classifier.fit(X train pca, y train)
# Test the model
y pred = svm classifier.predict(X test pca)
# Calculate accuracy
accuracy = accuracy score(y test, y pred)
print(f"Face Recognition Model Accuracy: {accuracy * 100:.2f}%")
# Display Confusion Matrix
conf matrix = confusion matrix(y test, y pred)
plt.figure(figsize=(6, 5))
sns.heatmap(conf matrix, annot=True, fmt="d", cmap="Blues", xticklabels=target names,
yticklabels=target names)
plt.xlabel("Predicted Label")
plt.ylabel("True Label")
plt.title("Confusion Matrix - Face Recognition")
plt.show()
# Test with a sample image
sample idx = 5 # Choose any index from test set
plt.imshow(lfw people.images[sample idx], cmap="gray")
plt.title(f'Actual: {target names[y test[sample idx]]} \nPredicted:
{target names[y pred[sample idx]]}")
plt.axis("off")
plt.show()
```

OUTPUT:

| Confusion Matrix - Face Recognition | | | | | | | | | |
|-------------------------------------|---------|----------|----------|----------|---------|---------|-----------|--|-------|
| Ariel Sharon - | 6 | 0 | 3 | 2 | 0 | 0 | 0 | | - 100 |
| Colin Powell - | 1 | 38 | 3 | 4 | 1 | 0 | 0 | | - 80 |
| ld Rumsfeld - | 1 | 2 | 17 | 1 | 0 | 0 | 1 | | - 60 |
| rge W Bush - | 0 | 8 | 4 | 102 | 2 | 2 | 1 | | |
| d Schroeder - | 0 | 2 | 0 | 1 | 16 | 0 | 0 | | - 40 |
| ugo Chavez - | 0 | 1 | 1 | 0 | 2 | 8 | 1 | | - 20 |
| Tony Blair - | 0 | 2 | 0 | 2 | 2 | 0 | 21 | | |
| | haron - | Powell - | nsfeld - | / Bush - | oeder - | havez - | y Blair - | | - 0 |

Actual: George W Bush Predicted: George W Bush



RESULT: The face recognition model achieved an accuracy of 80.62%. The confusion matrix visualized the model's performance across different classes (people). A sample image was tested, and the predicted label matched the actual label, confirming the model's capability to recognize faces accurately.