## week8

## October 18, 2024

```
[1]: import numpy as np
     import matplotlib.pyplot as plt
     n samples = int(input("Enter the number of data samples: "))
     n_features = int(input("Enter the number of features (columns): "))
     print("\nNow, enter the data matrix row by row (each row should contain values⊔
      ⇔separated by space):")
     data = []
     for i in range(n_samples):
         row = list(map(float, input(f"Enter values for row {i+1}: ").split()))
         data.append(row)
     X = np.array(data)
     print("\nData Matrix (X):\n", X)
     mean_vector = np.mean(X, axis=0)
     print("\nMean Vector:\n", mean_vector)
     X_centered = X - mean_vector
     print("\nCentered Data Matrix (X_centered):\n", X_centered)
     covariance_matrix = np.cov(X_centered.T)
     print("\nCovariance Matrix:\n", covariance_matrix)
     eigenvalues, eigenvectors = np.linalg.eig(covariance_matrix)
     print("\nEigenvalues:\n", eigenvalues)
     print("\nEigenvectors:\n", eigenvectors)
     sorted_indices = np.argsort(eigenvalues)[::-1]
     sorted_eigenvalues = eigenvalues[sorted_indices]
     sorted_eigenvectors = eigenvectors[:, sorted_indices]
     print("\nSorted Eigenvalues:\n", sorted_eigenvalues)
     print("\nSorted Eigenvectors (corresponding to sorted eigenvalues):\n", _
      ⇔sorted_eigenvectors)
```

```
unit_eigenvectors = sorted_eigenvectors / np.linalg.norm(sorted_eigenvectors,_
 ⇒axis=0)
print("\nUnit Eigenvectors:\n", unit_eigenvectors)
selected_vector = unit_eigenvectors[:, 0].reshape(-1, 1)
print("\nSelected Eigenvector for 1D PCA (Principal Component 1):\n",,,
 ⇔selected_vector)
X_pca_1d = X_centered.dot(selected_vector)
print("\n1D Principal Component Projection Values:\n", X_pca_1d)
plt.figure(figsize=(8, 6))
plt.scatter(X[:, 0], X[:, 1], color='blue', label='Original Data', alpha=0.6)
line_vector = selected_vector.flatten()
line_start = mean_vector - 3 * line_vector
line_end = mean_vector + 3 * line_vector
plt.plot([line_start[0], line_end[0]], [line_start[1], line_end[1]],__

color='red', label='Principal Component 1', linewidth=2)

for i in range(n_samples):
    projection = mean_vector + X_pca_1d[i] * line_vector
    plt.plot([X[i, 0], projection[0]], [X[i, 1], projection[1]], 'k--', alpha=0.
 ⇒5)
    plt.scatter(projection[0], projection[1], color='green', alpha=0.7)
plt.xlabel('Feature 1')
plt.ylabel('Feature 2')
plt.title('2D to 1D PCA Projection')
plt.legend()
plt.grid(True)
plt.show()
```

Now, enter the data matrix row by row (each row should contain values separated by space):

```
Data Matrix (X):
[[ 4. 11.]
[ 8. 4.]
[13. 5.]
[ 7. 14.]]

Mean Vector:
[8. 8.5]
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```
Centered Data Matrix (X_centered):
 [[-4.
         2.5]
 [0. -4.5]
 [5. -3.5]
 Γ-1.
        5.5]]
Covariance Matrix:
 [[ 14. -11.]
 [-11. 23.]]
Eigenvalues:
 [ 6.61513568 30.38486432]
Eigenvectors:
 [[-0.83025082 0.55738997]
 [-0.55738997 -0.83025082]]
Sorted Eigenvalues:
 [30.38486432 6.61513568]
Sorted Eigenvectors (corresponding to sorted eigenvalues):
 [[ 0.55738997 -0.83025082]
 [-0.83025082 -0.55738997]]
Unit Eigenvectors:
 [[ 0.55738997 -0.83025082]
 [-0.83025082 -0.55738997]]
Selected Eigenvector for 1D PCA (Principal Component 1):
 [[ 0.55738997]
 [-0.83025082]]
1D Principal Component Projection Values:
 [[-4.30518692]
 [ 3.73612869]
 [ 5.69282771]
 [-5.12376947]]
```

