pca

July 2, 2024

1 Principal Component Analysis (PCA)

1.0.1 Imports

```
[15]: import numpy as np from sklearn.decomposition import PCA
```

1.0.2 Implementation

```
[58]: def pca(data, k):
    # Center data
    data_centered = data - np.mean(data, axis = 0)
    # Compute covariance matrix
    n = data.shape[1]
    covariance_matrix = np.empty([n, n])
    for i in range(n):
        for j in range(n):
            covariance_matrix[i, j] = covariance(data_centered[:, i],__
 →data_centered[:, j])
    # Perform eigen decomposition of covariance matrix and sort in descending_
 \hookrightarrow order to choose top k principal components
    eigen_vals, eigen_vecs = np.linalg.eig(covariance_matrix)
    sorted_indices = np.argsort(eigen_vals)[::-1]
    eigen_vals = eigen_vals[sorted_indices][:k]
    eigen_vecs = eigen_vecs[:, sorted_indices][:, :k]
    # Project onto new feature space using top k principal components
```

```
return np.dot(data_centered, eigen_vecs)
```

1.0.3 Create Test Dataset

```
[59]: data = np.random.randn(30, 5)
```

1.0.4 Compare Implementation VS. Library Function

```
[61]: # Perform PCA using sklearn.decomposition.PCA
pca_sklearn = PCA(1)
transformed_data_sklearn = pca_sklearn.fit_transform(data)
# Perform PCA using your custom function
transformed_data_custom = pca(data, 1)
# Compare the results
print("Sklearn PCA result:")
print(transformed_data_sklearn)
print("\nCustom PCA result:")
print(transformed_data_custom)
```

Sklearn PCA result:

- [[1.38549345]
- [0.59693986]
- [0.47308285]
- [-0.37803657]
- [0.7151092]
- [1.82261136]
- [1.66109513]
- [-0.07122455]
- [-1.58217617]
- [0.14388147]
- [0.84908191]
- [-1.77249329]
- [-0.71744546]
- [0.01458373]
- [-1.10829776]
- [-0.14099715]
- [-0.87321643]
- [-0.38054763]
- [1.24638152]
- [-0.838617]
- [1.54380084]
- [-1.66461489]
- [-1.08581505]
- [0.06909277]
- [1.75389033]

- [1.49958998]
- [-0.56610264]
- [-1.08607467]
- [-1.10891903]
- [-0.4000561]]

Custom PCA result:

- [[1.38549345]
- [0.59693986]
- [0.47308285]
- [-0.37803657]
- [0.7151092]
- [1.82261136]
- [1.66109513]
- [-0.07122455]
- [-1.58217617]
- [0.14388147]
- [0.84908191]
- [-1.77249329]
- [-0.71744546]
- [0.01458373]
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- [-0.14099715]
- [-0.87321643]
- [-0.38054763]
- [1.24638152]
- [-0.838617]
- [1.54380084]
- [-1.66461489]
- [-1.08581505]
- [0.06909277]
- [1.75389033]
- [1.49958998]
- [-0.56610264]
- [-1.08607467]
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- [-0.4000561]]