

1.

(a)  $X_{\text{mean}} = \text{np.mean}(X, 0) = [1.5 \quad 2.5 \quad 3]$

(b)  $X_s = X - X_{\text{mean}}$

$$Q = \frac{1}{4} * \text{np.dot}(X_s.T, X_s) = \begin{bmatrix} 1.25 & 0.25 & -1.25 \\ 0.25 & 0.75 & 0.75 \\ -1.25 & 0.75 & 2.75 \end{bmatrix}$$

(c)  $w, v = \text{np.linalg.eig}(Q)$

$$w = [3.56166464, 1.1733803, 0.01495506]$$

$$v = \begin{bmatrix} [-0.45056922, -0.66677184, -0.59363515], \\ [0.19247228, -0.72187235, 0.66472154], \\ [0.87174641, -0.18524476, -0.45358856] \end{bmatrix}$$

(d)  $\text{coefficient} = \text{np.dot}(X_s, v)$

$$\begin{bmatrix} [-2.95145599, -0.17610969, -0.0888421], \\ [1.37104342, -1.69406159, 0.0198819], \\ [-0.30682473, 0.78694448, 0.19125108], \\ [1.8872373, 1.0832268, -0.12229089] \end{bmatrix}$$

(e)  $X_{\text{hat}} = X_{\text{mean}} + \text{np.dot}(\text{coefficient}, v.T)$

$$\begin{bmatrix} [3.0000000e+00, 2.0000000e+00, 1.0000000e+00], \\ [2.0000000e+00, 4.0000000e+00, 5.0000000e+00], \\ [1.0000000e+00, 2.0000000e+00, 3.0000000e+00], \\ [8.8817842e-16, 2.0000000e+00, 5.0000000e+00] \end{bmatrix}$$

(f)  $X_{\text{hat}} = X_{\text{mean}} + \text{np.dot}(\text{coefficient}[:, :2], v[:, :2].T)$

$$\begin{bmatrix} [2.94726021, 2.05905526, 0.95970224], \\ [2.0118026, 3.98678407, 5.0090182], \\ [1.11353336, 1.87287129, 3.0867493], \\ [-0.07259617, 2.08128939, 4.94453025] \end{bmatrix}$$

(g)  $\text{np.sum}(\text{coefficient}[:, :2]**2) == \text{np.sum}(\text{np.sum}((X - X_{\text{hat}})**2, 1))$

2.

(a)  $\alpha_1 = v_1^T(x - \mu) = \frac{1}{\sqrt{2}} (1, 1, 0)(1, 3, 2) = 2\sqrt{2}$   
 $\alpha_2 = v_2^T(x - \mu) = \frac{1}{\sqrt{2}} (1, -1, 0)(1, 3, 2) = -\sqrt{2}$

(b)  $\hat{x} = \mu + 2\sqrt{2}\alpha_1 - \sqrt{2}\alpha_2 = (2, 3, 0)$

(c) approximation error = 4

3.

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=42)
```

```
mu, V = PCA(X_train)
```

```
ncomp = V.shape[0]
```

```
acc = []
```

```
for n in range(n_components):
```

```
    clf = Classifier()
```

```
    Z = V[:n]
```

```
    clf.fit(Z,y)
```

```
    yhat = clf.predict(X_test)
```

```
    acc.append(np.mean((y_test - yhat)**2))
```

```
index = np.argmax(acc)
```

```
opt_num = index + 1
```

4.

```
Y = reshape(X, (1000, 784))
```

```
pca = PCA(n_components = 5)
```

```
pca.fit(Y[:500,:])
```

```
Z = pca.transform(Y[500:,:])
```

```
yhat = pca.inverse_transform(Z)
```

5.

```
Xmean = np.mean(X,0)
```

```
Xs = X - Xmean[None,:]
```

```
U,S, Vtr = svd(Xs, full_matrices = False)
```

```
lam = S**2
```

```
pov = np.cumsum(lam) / np.sum(lam)
```

```
numbers = np.where(pov > 0.9)[0][0] + 1
```

```
Xhat = (U[:, :numbers] * S[None, :numbers]).dot(Vtr[:numbers,:]) + Xmean
```