

Q5)

We have to maximise $f^T C f$ given f is a unit vector.

To solve, $J = \sum_{i=1}^N \|x_i - \bar{x} - (e(x_i - \bar{x}))e\|_2$ is minimum.

$$J = \sum_{i=1}^N -(e^T(x_i - \bar{x}))^2 + \sum_{i=1}^N \|x_i - \bar{x}\|_2 = -e^T S e + \sum_{i=1}^N \|x_i - \bar{x}\|_2 \quad S = (N-1)C$$

Minimizing J wrt e is similar to maximizing $e^T S e$ wrt e as other term does not have e .

Also $e^T e = 1$. So new equation is,

$$Z = e^T S e - \lambda(e^T e - 1)$$

Derivative wrt to e and setting to 0 gives,

$$S e = \lambda e$$

This implies that e is an eigen vector of S and $e^T S e$ gives the eigen value of S .

For maximizing $e^T C e$, eigen-vector corresponding to highest eigen value is selected.

Similarly for maximizing $f^T C f$, eigen-vector corresponding to second highest eigen value is selected as e is not equal to f .

Since f and e are the eigen-vector of C .

$$C e = \lambda_1 e \quad \dots(1)$$

$$C f = \lambda_2 f \quad \dots(2)$$

Multiply eq 1 by f^T and eq2 by e^T .

$$f^T C e = f^T \lambda_1 e$$

$$e^T C f = e^T \lambda_2 f$$

Since C is a symmetric matrix, $f^T C e = (f^T C e)^T = e^T C f$.

$$\text{So, } f^T \lambda_1 e = \lambda_2 e^T f.$$

Since $\lambda_1 \neq \lambda_2$, $e^T f = 0$.

Hence, f and e are orthogonal.