

Q.5]

$$X = \{x_1, x_2, \dots, x_N\}$$

e = first eigen vector (highest eigenvalue)

f is perpendicular to e
i.e. $f^T e = 0$

$$\text{Also } f^T f = 1 \quad [\text{normalized}]$$

C = Covariance matrix of f

So, the ^{problem} can be formulated as:

$$\begin{aligned} \max f^T C f & \quad [\text{objective function}] \\ \text{s.t. } f^T f = 1 & \quad [\text{constraints}] \\ \text{and } f^T e = 0 & \end{aligned}$$

$$\therefore E(f) = f^T C f - \lambda_1 (f^T f - 1) - \lambda_2 f^T e$$

Differentiating wrt. f ,

$$\frac{\partial E}{\partial f} = C f - \lambda_1 f - \lambda_2 e = 0$$

$$\therefore e^T C f - \lambda_1 e^T f - \lambda_2 e^T e = 0$$

Now, $e^T e = 1$

$$\text{and } e^T C f = \lambda_1 e^T f = 0$$

$$\begin{aligned} \text{because } C e = \lambda_1 e & \Rightarrow e^T C = \lambda_1 e^T \\ \Rightarrow e^T C f & = \lambda_1 e^T f = 0 \end{aligned}$$

Hence, $\lambda_2 = 0$

$$\Rightarrow e^t C f = \lambda_1 e^t f$$

$$\Rightarrow C f = \lambda_1 e e^t f$$

$$\Rightarrow C f = \lambda_1 f$$

$\Rightarrow f$ is also an eigenvector of C

Since $f^t e = 0$, f and e have distinct eigenvalues

\Rightarrow it must correspond to second highest eigenvalue. $\lambda_1 = f^t C f$