

Problem Statement

Consider a set of N vectors $X = x_1, x_2, \dots, x_N$ each in R^d , with average vector \bar{x} . We have seen in class that the direction e such that $\|x_i - \bar{x} - (e \cdot (x_i - \bar{x})e)\|^2$ is minimized, is obtained by maximizing $e^T C e$, where C is the covariance matrix of the vectors in X . This vector e is the eigenvector of matrix C with the highest eigenvalue. Prove that the direction f perpendicular to e for which $f^T C f$ is maximized, is the eigenvector of C with the second highest eigenvalue. For simplicity, assume that all non-zero eigenvalues of C are distinct and that $\text{rank}(C) > 2$ [10 points]

Consider the direction f st. it is perpendicular to e
& has unit norm

$$\Rightarrow f^T e = e^T f = 0$$
$$\|f\|^2 = f^T f = 1$$

To max. $f^T C f$

or min $-f^T C f$, we'll use Lagrange Multiplier method.

$$J(f) = -f^T C f + \lambda(f^T f - 1) + \mu(f^T e)$$

$$\frac{\partial J(f)}{\partial f} = -2 C f + 2 \lambda f + \mu e = 0$$

On Premultiplying by f^T :

$$-2 f^T C f + 2 \lambda f^T f + \mu f^T e = 0$$

$$\lambda = f^T C f \quad \left(\text{as } \begin{matrix} f^T f = 1 \\ f^T e = 0 \end{matrix} \right)$$

Premult by e^T

$$-2 e^T C f + 2 \lambda (e^T f) + \mu e^T e = 0$$

$$\mu = 2 e^T C f$$

$$\Rightarrow 2Cf = 2f^T C f^2 + 2e^T C f e$$

$$Cf = (f^T C f) f + 2(e^T C e)^T f e$$

$$Cf = (f^T C f) f + 2(Ce)^T f e$$

$$Cf = (f^T C f) f + 2\lambda e^T f e$$

where λ = largest eigv.
corresponding to e

$$Cf = (f^T C f) f + 2\lambda(0)e$$

$$\Rightarrow Cf = (f^T C f) f$$

$\Rightarrow f$ is eigenvector of C . (Let the eigenvalue be λ_0)

$$\Rightarrow Cf = \lambda_0 f$$

$$\Rightarrow f^T C f = \lambda_0$$

Thus to max. $f^T C f$, λ_0 (eigenvector) needs to be as high as possible
 $\Rightarrow \lambda_0$ is second highest eigenvector.