

1 Answer

We need to prove that after the eigen-vector corresponding to largest eigen-value is chosen the next best choice is to choose the eigen-vector with second largest eigen value. Let \mathbf{e} , be the eigen-vector with largest eigen-value. Let \mathbf{f} be a unit vector perpendicular to \mathbf{e} . Then according to question we want to minimize the cost function:

$$J(\mathbf{f}) = \sum_{i=1}^N \|b_i \mathbf{f} - (\mathbf{x}_i - \mathbf{x} - a_i \mathbf{e})\|^2$$

where, $a_i = \mathbf{e}^T(\mathbf{x}_i - \mathbf{x})$; $b_i = \mathbf{f}^T(\mathbf{x}_i - \mathbf{x})$; \mathbf{x} =average value of \mathbf{x}_i

$$J(\mathbf{f}) = \sum_{i=1}^N \|b_i \mathbf{f}\|^2 + \sum_{i=1}^N \|\mathbf{x}_i - \mathbf{x} - a_i \mathbf{e}\|^2 - 2 \sum_{i=1}^N b_i \mathbf{f}^T(\mathbf{x}_i - \mathbf{x} - a_i \mathbf{e})$$

$$J(\mathbf{f}) = \sum_{i=1}^N b_i^2 + \sum_{i=1}^N \|\mathbf{x}_i - \mathbf{x} - a_i \mathbf{e}\|^2 - 2 \sum_{i=1}^N \|b_i \mathbf{f}\|^2$$

$$\implies J(\mathbf{f}) = - \sum_{i=1}^N b_i^2 + \sum_{i=1}^N \|\mathbf{x}_i - \mathbf{x} - a_i \mathbf{e}\|^2$$

(Since, \mathbf{f} is a unit vector, $\mathbf{f}^T \mathbf{e} = 0$, $\mathbf{f}^T(\mathbf{x}_i - \mathbf{x}) = b_i$)

We want to minimize $J(\mathbf{f})$ and to minimize it we should maximize the first term in the latter equation.

$$\sum_{i=1}^N b_i^2 = \sum_{i=1}^N \mathbf{f}^T(\mathbf{x}_i - \mathbf{x})(\mathbf{x}_i - \mathbf{x})^T \mathbf{f}$$

We define $\sum_{i=1}^N (\mathbf{x}_i - \mathbf{x})(\mathbf{x}_i - \mathbf{x})^T = \mathbf{S}$, where $\mathbf{S} = (N-1)\mathbf{C}$ and \mathbf{C} is the covariance matrix. Thus the above equation becomes,

$$\sum_{i=1}^N b_i^2 = \mathbf{f}^T \mathbf{S} \mathbf{f}$$

Thus, we need to maximize the above equation under the constraint, $\mathbf{f}^T \mathbf{f} = 1$. Using Lagrange's multiplier method we get,

$$\begin{aligned} G(\mathbf{f}) &= \mathbf{f}^T \mathbf{S} \mathbf{f} - \lambda(\mathbf{f}^T \mathbf{f} - 1) \\ G'(\mathbf{f}) &= \mathbf{S} \mathbf{f} - \lambda \mathbf{f} = 0 \\ \implies \mathbf{S} \mathbf{f} &= \lambda \mathbf{f} \end{aligned}$$

Thus, \mathbf{f} needs to be an eigenvector of \mathbf{S} . Rearranging the equation gives,

$$\mathbf{f}^T \mathbf{S} \mathbf{f} = \lambda$$

Thus, to maximize $\mathbf{f}^T \mathbf{S} \mathbf{f}$ we need maximum eigen-value. But we have already chosen the largest eigenvalue and its corresponding vector, thus the next best option is the second largest eigenvalue. Hence, proved that \mathbf{f} , which is perpendicular to \mathbf{e} , is the eigenvector corresponding to the second largest eigenvalue.

1.1 Another approach

We wanted the maximize $\mathbf{f}^T \mathbf{S} \mathbf{f}$. A more intuitive solution which also gives a physical interpretation is as follows:

$$\mathbf{f}^T \mathbf{S} \mathbf{f} = \mathbf{f}^T (\mathbf{S} \mathbf{f})$$

Let $\mathbf{f}^T \mathbf{S} = \mathbf{c}$, and the above equation becomes $\mathbf{f}^T \mathbf{c}$. The dot product of two vectors is maximized (under the constraint of constant magnitude of vectors) when angle between the two is 0° and thus \mathbf{c} should be $\alpha \mathbf{f}$, where α is some constant. Written mathematically,

$$\mathbf{S} \mathbf{f} = \alpha \mathbf{f}$$

Thus, \mathbf{f} is an eigenvector of \mathbf{S} matrix.