

QUESTION 3

CS663 (DIGITAL IMAGE PROCESSING) ASSIGNMENT 4

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Question 3

Problem 1

Consider a set of N vectors $\mathcal{X} = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N\}$ each in \mathbb{R}^d , with average vector $\bar{\mathbf{x}}$. We have seen in class that the direction \mathbf{e} such that $\sum_{i=1}^N \|\mathbf{x}_i - \bar{\mathbf{x}} - (\mathbf{e} \cdot (\mathbf{x}_i - \bar{\mathbf{x}}))\mathbf{e}\|^2$ is minimized, is obtained by maximizing $\mathbf{e}^t \mathbf{C} \mathbf{e}$, where \mathbf{C} is the covariance matrix of the vectors in \mathcal{X} . This vector \mathbf{e} is the eigenvector of matrix \mathbf{C} with the highest eigenvalue. Prove that the direction \mathbf{f} perpendicular to \mathbf{e} for which $\mathbf{f}^t \mathbf{C} \mathbf{f}$ is maximized, is the eigenvector of \mathbf{C} with the second highest eigenvalue. For simplicity, assume that all non-zero eigenvalues of \mathbf{C} are distinct and that $\text{rank}(\mathbf{C}) > 2$. Extend the derivation to handle the case of a unit vector \mathbf{g} which is perpendicular to both \mathbf{e} and \mathbf{f} which maximizes $\mathbf{g}^t \mathbf{C} \mathbf{g}$. [10 points]

SECTION 1

Proof(a)

PROOF We aim to maximize the expression $\mathbf{f}^t \mathbf{C} \mathbf{f}$, subject to the constraint that \mathbf{f} is orthogonal to \mathbf{e} . We will use the method of Lagrange multipliers. The Lagrangian for this optimization problem is as follows:

$$\mathcal{L}(f) = f^t C f - \lambda_1(f^t f - 1) - \lambda_2(f^t e)$$

Here, λ_1 and λ_2 are Lagrange multipliers introduced to enforce constraints. The constraint $\mathbf{f}^t \mathbf{f} = 1$ ensures that \mathbf{f} is a unit vector, and $\mathbf{f}^t \mathbf{e} = 0$ enforces that \mathbf{f} is perpendicular to \mathbf{e} .

Now to find the maximum of $\mathbf{f}^t \mathbf{C} \mathbf{f}$, we set the gradient of \mathcal{L} with respect to \mathbf{f} equal to zero:

$$\nabla \mathcal{L}(f) = 2Cf - 2\lambda_1 f - 2\lambda_2 e$$

We set this gradient to zero to maximize \mathcal{L} .

$$Cf - \lambda_1 f - \lambda_2 e = 0 \tag{1.1}$$

Now, we premultiply both sides by \mathbf{e} to remove the \mathbf{e} term from the equation:

$$\mathbf{e}^t \mathbf{C} \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 \mathbf{e}^t \mathbf{e} = 0$$

since $\mathbf{e}^t \mathbf{e} = 1$ i.e, \mathbf{e} is unit vector, therefore

$$\mathbf{e}^t \mathbf{C} \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 = 0 \quad (1.2)$$

Remark

From equation (1.2), we can infer that $\lambda_2 = 0$. This can be shown from the following equations:

$$\mathbf{e}^t \mathbf{C} \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 = 0$$

Since $\mathbf{e}^t \mathbf{e} = 1$, we can write

$$\mathbf{e}^t \mathbf{C} \mathbf{e} \mathbf{e}^t \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 = 0$$

We also know that $\mathbf{C} \mathbf{e} = \lambda \mathbf{e}$ since \mathbf{e} is the eigenvector of matrix \mathbf{C} with the highest eigenvalue(λ). Thus

$$\mathbf{e}^t \lambda \mathbf{e} \mathbf{e}^t \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 = 0$$

$$\lambda \mathbf{e}^t \mathbf{f} - \lambda_1 \mathbf{e}^t \mathbf{f} - \lambda_2 = 0$$

Also, \mathbf{f} is perpendicular to \mathbf{e} , therefore $\mathbf{e}^t \mathbf{f} = 0$ which gives:

$$\lambda_2 = 0 \quad (1.3)$$

Now using equations (1.1) and (1.3), we can write

$$\mathbf{C} \mathbf{f} = \lambda_1 \mathbf{f} \quad (1.4)$$

This result (1.4) establishes that \mathbf{f} is an eigenvector of \mathbf{C} with eigenvalue λ_1 .

Moreover, we assumed that all eigenvalues of \mathbf{C} are distinct and

$\mathbf{f}^t \mathbf{C} \mathbf{f} = \lambda_1$ (this can be seen by premultiplying equation (1.4) by \mathbf{f}^t) which means that λ_1 is the second highest eigenvalue.

□

This concludes the formal proof that the direction \mathbf{f} , which is perpendicular to \mathbf{e} and maximizes $\mathbf{f}^t \mathbf{C} \mathbf{f}$, corresponds to the eigenvector of \mathbf{C} with the second highest eigenvalue, provided that all non-zero eigenvalues of \mathbf{C} are distinct.

SECTION 2

Proof(b)

PROOF We aim to maximize $\mathbf{g}^t \mathbf{C} \mathbf{g}$ subject to the constraints that \mathbf{g} is orthogonal to both \mathbf{e} and \mathbf{f} . We will use the method of Lagrange multipliers. The Lagrangian for this optimization problem is as follows:

$$\mathcal{L}(g) = g^t C g - \lambda_3(g^t g - 1) - \lambda_4(g^t e) - \lambda_5(g^t f)$$

Here, λ_3 , λ_4 and λ_5 are Lagrange multipliers.

- $\mathbf{g}^t \mathbf{g} = 1$ ensures that \mathbf{g} is a unit vector
- $\mathbf{g}^t \mathbf{e} = 0$ enforces that \mathbf{g} is perpendicular to \mathbf{e}
- $\mathbf{g}^t \mathbf{f} = 0$ enforces that \mathbf{g} is perpendicular to \mathbf{f}

Now to find the maximum, we set the gradient of \mathcal{L} with respect to \mathbf{g} equal to zero:

$$\nabla \mathcal{L}(g) = Cg - \lambda_3 g - \lambda_4 e - \lambda_5 f = 0 \quad (2.1)$$

Now, we premultiply both sides of the equation by \mathbf{e} :

$$e^t C g - \lambda_3 e^t g - \lambda_4 e^t e - \lambda_5 e^t f = 0$$

Since $\mathbf{e}^t \mathbf{e} = 1$:

$$e^t C g - \lambda_3 e^t g - \lambda_4 - \lambda_5 e^t f = 0 \quad (2.2)$$

Remark From equation (2.2), we can infer that $\lambda_4 = 0$. This can be concluded by the following series of equations:

$$e^t C e e^t g - \lambda_3 e^t g - \lambda_4 - \lambda_5 e^t f = 0$$

$$e^t \lambda e e^t g - \lambda_3 e^t g - \lambda_4 - \lambda_5 e^t f = 0$$

$$\lambda e^t g - \lambda_3 e^t g - \lambda_4 - \lambda_5 e^t f = 0$$

Now since \mathbf{g} , \mathbf{e} and \mathbf{f} are mutually perpendicular, therefore $\mathbf{e}^t \mathbf{g} = 0$ and $\mathbf{e}^t \mathbf{f} = 0$. Hence we get

$$\lambda_4 = 0 \quad (2.3)$$

Similarly, premultiplying both sides of the gradient equation by \mathbf{f} , we get

$$\lambda_1 f^t g - \lambda_3 f^t g - \lambda_4 f^t e - \lambda_5 = 0$$

$$\lambda_5 = 0 \quad (2.4)$$

Now using equations (2.1), (2.3) and (2.4), we can write

$$\mathbf{C}\mathbf{g} = \lambda_3\mathbf{g} \tag{2.5}$$

This result (2.5) establishes that \mathbf{g} is an eigenvector of \mathbf{C} with eigenvalue λ_3 .

Moreover, we assumed that all eigenvalues of \mathbf{C} are distinct and $\mathbf{g}^t\mathbf{C}\mathbf{g} = \lambda_3$ (this can be seen by premultiplying equation (2.5) by \mathbf{g}^t) which means that λ_3 is the third highest eigenvalue. \square

This concludes the formal proof that the direction \mathbf{g} , which is perpendicular to both \mathbf{e} and \mathbf{f} and maximizes $\mathbf{g}^t\mathbf{C}\mathbf{g}$, corresponds to the eigenvector of \mathbf{C} with the third highest eigenvalue, provided that all non-zero eigenvalues of \mathbf{C} are distinct.