

# Assignment 4 Question 5 Report

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## 1 Framing the problem

The problem given is an optimization problem. We want to maximize  $f^T C f$  where  $C$  is the covariance matrix.

However, we have the constraints:-

- $f^T f = 1$
- $f^T e = 0$ ,  $e$  is the eigen vector of  $C$  with the largest eigen value.

Here  $e$  is the eigen vector of  $C$  with the largest eigen value.

## 2 Using Lagrange Multipliers

Now we can apply the formulation of Lagrange multipliers to solve this problem. We have to maximize the cost function:-

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

subject to:-

- $f^T f = 1$
- $f^T e = 0$ ,  $e$  is the eigen vector of  $C$  with the largest eigen value.

## 3 Optimization

Differentiating  $J(f)$  with respect to  $f$  and equating to zero, we get,

$$2f^T C - 2\lambda_1 f^T - \lambda_2 e^T = 0$$

Now, we can take transpose on both sides. Seeing that  $C^T$  is symmetric, we have,

$$Cf = \lambda_1 f + \frac{\lambda_2}{2} e$$

If we pre-multiply by  $e^T$ , we get an expression for  $\lambda_2$  as  $e^T f = 0$ .

$$\lambda_2 = 2e^T C f$$

Substituting this back,

$$Cf = \lambda_1 f + 2e^T C f e - - - (1)$$

But we have that  $Ce = \mu e$ . Taking transpose on both sides, we get,

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

Substituting this into (1),

$$Cf = \lambda_1 f + 2\mu e^T f e$$

Again, using  $e^T f = 0$ , we get,

$$Cf = \lambda_1 f$$

Hence  $f$  is an eigen vector of  $C$ . We have to maximize  $f^T C f$ . But since  $Cf = \lambda_1 f$ , we have,

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

Above we used  $f^T f = 1$

We want  $\lambda_1$  to be as large as possible. But  $e$  has the largest eigen value and it is perpendicular to  $f$ . Since the eigen values are unique(given in the question),  $f$  will have the second largest eigen value.

Hence proved.

## 4 Conclusions

Hence we have proved the given statement.