## **Exercise 2**

We are given the first principal component direction e, which is the eigen vector of S (a constant multiple of the covariance matrix C, i.e. S = (N-1)C) with the largest eigenvalue, say  $\lambda_e$  and is obtained by maximizing  $Var(e^Tx) = e^TSe$ . Let f be the direction perpendicular to e which maximizes  $f^TSf$ , subject to the conditions that f is a unit vector and is perpendicular to e. The Lagrangian for this problem can then be written as:

$$\mathcal{L} = \mathbf{f}^T \mathbf{S} \mathbf{f} - a(\mathbf{f}^T \mathbf{f} - 1) - b(\mathbf{f}^T \mathbf{e} - 0)$$

Derivative of the Lagrangian can then be computed as:

$$\frac{d}{d\mathbf{f}}(\mathcal{L}) = \frac{d}{d\mathbf{f}}(\mathbf{f}^T \mathbf{S} \mathbf{f} - a(\mathbf{f}^T \mathbf{f} - 1) - b(\mathbf{f}^T \mathbf{e} - 0)) = 2\mathbf{S} \mathbf{f} - 2a\mathbf{f} - b\mathbf{e} = 0$$

Pre-multiplying the equation above by  $e^T$ , we get:

$$e^T S f - a e^T f - b e^T e = 0$$

The second term in the equation above is zero because e and f are perpendicular. The first term in the expression above is  $e^T S f$ . This can further be manipulated as follows:

$$e^T S f = (S^T e)^T f$$

Since C, and thus S, is a symmetric matrix, i.e.  $S^T = S$ , we have:

$$e^T S f = (S e)^T f$$

Now, we also know that e is an eigen vector of S. Therefore,  $Se = \lambda_e e$ . Substituting this in the expression above, we have:

$$e^T S f = (\lambda_{e} e)^T f = \lambda_{e} e^T f = 0$$

The last equality follows from the fact that e and f are perpendicular.

Therefore, the first and the second terms in the original equation both are zero. Thus, we are left with  $be^Te=0$ . Since e is a unit vector, we have  $e^Te=1$ . Thus the only way that this equation is satisfied is if we have b=0.

Now, coming back to the expression for the derivative of Lagrangian, substituting b = 0, we have:

$$2\mathbf{S}\mathbf{f} - 2a\mathbf{f} = 0$$

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

Therefore, f is an eigenvector of S, and hence C. Now, we also want to maximize the variance captured in the direction of this eigenvector. So, we want to maximize the following expression:

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

In the first equality, we have used the fact that f is an eigenvector of S with an eigenvalue of  $\lambda_f$  and in the last equality, we have used the fact that f is a unit vector. Therefore, we want to maximize  $\lambda_f$ . Since  $\lambda_e$  is the largest eigenvalue corresponding to the eigenvector e,  $\lambda_f$  has to be necessarily the second largest eigenvalue.