

# Week 2 Graded assignment

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## Common data for Questions 1 and 2

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A function  $k$  is defined as follows.

$$k : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$$
$$k(x_1, x_2) = x_1^T x_2$$

### Question 1

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#### Statement

Is  $k$  a valid kernel?

#### Options

(a)

Yes

(b)

No

#### Answer

(a)

#### Solution

Let  $\phi$  be an identity transformation that is

$$\phi : \mathbb{R}^d \rightarrow \mathbb{R}^d$$
$$\phi(x) = x$$

It is clear by the definition of given kernel that

$$k(x_1, x_2) = \phi(x_1)^T \phi(x_2)$$

It implies that  $k$  is a valid kernel.

### Question 2

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If  $k$  is the valid kernel, we apply it to the three-dimensional dataset to run the kernel PCA. Select the correct options.

## Options

(a)

We cannot run the PCA as  $k$  is not a valid kernel.

(b)

It will be the same as PCA with no kernel.

(c)

It will be the same as the polynomial transformation of degree 2 and then run the PCA.

(d)

It will be the same as PCA with a third-degree polynomial kernel.

## Answer

(b)

## Solution

We have seen (in question 1) that  $k$  corresponds to the identity transformation. It implies that applying kernel and running PCA is same as standard PCA on the given dataset.

## Question 3

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### Statement

Consider ten data points lying on a curve of degree two in a two-dimensional space. We run a kernel PCA with a polynomial kernel of degree two on the same data points. Choose the correct options.

## Options

(a)

The transformed data points will lie on a 5-dimensional subspace of  $\mathbb{R}^6$ .

(b)

The transformed data points will lie on a 6-dimensional subspace of  $\mathbb{R}^{10}$ .

(c)

There will be some  $w \in \mathbb{R}^6$  that all of the data points are orthogonal to.

(d)

There will be some  $w \in \mathbb{R}^{10}$  that all of the data points are orthogonal to.

## Answer

(a), (c)

## Solution

Since we are applying the polynomial kernel of degree two on the 2D dataset, the dataset will be transformed into a 6D feature space. (verify)

And the dataset is given to lying on a curve of degree two, the transformed dataset will live in the linear subspace of  $\mathbb{R}^6$ . and therefore, there will be some  $w \in \mathbb{R}^6$  that all of the data points are orthogonal to.

## Question 4

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### Statement

Which of the following matrices can not be appropriate matrix  $K = X^T X$  for some data matrix  $X$ ?

### Options

(a)

$$\begin{bmatrix} 1 & 8 \\ 8 & -1 \end{bmatrix}$$

(b)

$$\begin{bmatrix} 1 & 8 \\ 8 & 1 \end{bmatrix}$$

(c)

$$\begin{bmatrix} 1 & 8 \\ -8 & 1 \end{bmatrix}$$

(d)

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

## Answer

(a), (b) and (c)

## Solution

We know that  $K$  matrix must be symmetric and positive semi-definite.

All the given matrices are symmetric.

We need to check whether they are positive semi-definite or not.

For that, we will check the eigenvalues of the matrices and if all the eigenvalues are non-negative, then the matrix is positive semi-definite.

Option (a).

$$A = \begin{bmatrix} 1 & 8 \\ 8 & -1 \end{bmatrix}$$

Let  $\lambda$  be the eigenvalue of  $A$ , then

$$\begin{aligned} |A - \lambda I| &= 0 \\ \begin{vmatrix} 1 - \lambda & 8 \\ 8 & -1 - \lambda \end{vmatrix} &= 0 \\ (1 - \lambda)(-1 - \lambda) &= 64 \\ \lambda^2 - 1 &= 64 \\ \lambda &= \pm\sqrt{65} \end{aligned}$$

Since  $A$  has a non-negative eigenvalue,  $A$  is not a positive semi-definite matrix.

Similarly, check for all the options.

## Question 5

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### Statement

A function  $k$  is defined as

$$\begin{aligned} k : \mathbb{R}^2 \times \mathbb{R}^2 &\rightarrow \mathbb{R} \\ k(x_1, x_2) &= (x_1^T x_2)^2 \end{aligned}$$

Is  $k$  a valid kernel?

### Options

(a)

Yes

(b)

No

### Answer

(a)

### Solution

The given function is

$$\begin{aligned} k : \mathbb{R}^2 \times \mathbb{R}^2 &\rightarrow \mathbb{R} \\ k(x_1, x_2) &= (x_1^T x_2)^2 \end{aligned}$$

Let  $x_1 = [a_1, a_2]^T$  and  $x_2 = [b_1, b_2]^T$  then

$$\begin{aligned}
k(x_1, x_2) &= ([a_1, a_2][b_1, b_2]^T)^2 \\
&= (a_1b_1 + a_2b_2)^2 \\
&= a_1^2b_1^2 + 2a_1b_1a_2b_2 + a_2^2b_2^2 \\
&= [a_1^2, \sqrt{2}a_1a_2, a_2^2][b_1^2, \sqrt{2}b_1b_2, b_2^2]^T \\
&= \phi(x_1)^T \phi(x_2)
\end{aligned}$$

where  $\phi : \mathbb{R}^2 \rightarrow \mathbb{R}$  such that  $\phi([a_1, a_2]^T) = [a_1^2, \sqrt{2}a_1a_2, a_2^2]^T$

It means there exists a transformation mapping  $\phi$  such that  $k(x_1, x_2) = \phi(x_1)^T \phi(x_2)$ .  
Therefore,  $k$  is a valid kernel.

## Question 6

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### Statement

Kernel PCA was run on the four data points  $[1, 2]^T$ ,  $[2, 3]^T$ ,  $[2, -3]^T$ , and  $[4, 4]^T$  with the polynomial kernel of degree 2. What will be the shape of the matrix  $K$ ? Notations are used as per lectures.

### Options

(a)

$2 \times 2$

(b)

$4 \times 4$

(c)

$6 \times 6$

(d)

None of the above

### Answer

(b)

### Solution

The  $K$  matrix is defined as  $X^T X$  where  $X$  is a data matrix of shape  $(d, n)$ . That is  $K$  matrix is of shape  $(n, n)$  where  $n$  is a number of examples.

It is given that  $n = 4$ . Therefore, shape of  $K$  matrix is  $(4, 4)$

## Question 7

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## Statement

Find the element at the index  $(2, 3)$  of the matrix  $K$  defined in Question 6. Take the points in the same order.

## Options

(a)

−4

(b)

16

(c)

13

(d)

196

## Answer

(b)

## Solution

The polynomial kernel of degree 2 is given by

$$k(x_1, x_2) = (x_1^T x_2 + 1)^2$$

The  $(2, 3)$ th element of  $K$  matrix will be  $k(x_2, x_3)$ .

$$k(x_2, x_3) = ([2, 3][2, -3]^T + 1)^2 = (-5 + 1)^2 = 16$$

## Question 8

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### Statement

A dataset containing 200 examples in four-dimensional space has been transformed into higher dimensional space using the polynomial kernel of degree two. What will be the dimension of transformed feature space?

### Answer

15 (No range required)

### Solution

Let the features be  $x_1, x_2, x_3$ , and  $x_4$ . After the transformation of degree two, features will be

$1, x_1, x_2, x_3, x_4, x_1x_2, x_1x_3, x_1x_4, x_2x_3, x_2x_4, x_3x_4, x_1^2, x_2^2, x_3^2$ , and  $x_4^2$ . So, the dimension of transformed feature space will be 15.

## Question 9

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### Statement

Let  $x_1, x_2, \dots, x_n$  be  $d$ -dimensional data points ( $d > n$ ) and  $X$  be the matrix of shape  $d \times n$  containing the data points. The  $k^{th}$  largest eigenvalue and corresponding unit eigenvector of  $X^T X$  is  $\lambda$  and  $\alpha_k$ , respectively. What will be the projection of  $x_i$  on the  $k^{th}$  principal component?

### Options

(a)

$$x_i^T \alpha_k$$

(b)

$$\frac{x_i^T \alpha_k}{\lambda}$$

(c)

$$\frac{x_i^T X \alpha_k}{\sqrt{\lambda}}$$

(d)

$$\frac{x_i^T X \alpha_k}{\sqrt{n\lambda}}$$

### Answer

(c)

### Solution

If the  $k^{th}$  largest eigenvalue and corresponding unit eigenvector of  $X^T X$  is  $\lambda$  and  $\alpha_k$ , respectively, the  $k^{th}$  principal component will be  $\frac{X \alpha_k}{\sqrt{\lambda}}$ .

Therefore, the projection of the point  $x_i$  on the  $k^{th}$  principal component will be  $\frac{x_i^T X \alpha_k}{\sqrt{\lambda}}$ .

## Question 10

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### Statement

Let  $k_1$  and  $k_2$  be two valid kernels. Is  $3k_1 + 5k_2$  a valid kernel?

### Options

(a)

Yes

(b)

No

## Answer

(a)

## Solution

Let  $k_1$  and  $k_2$  be two valid kernels defined on  $\mathbb{R}^d \times \mathbb{R}^d$ , it implies that they satisfies the following two properties

(1)  $k_1$  and  $k_2$  are symmetric functions that is

$$k_1(x_1, x_2) = k_1(x_2, x_1) \quad \forall x_1, x_2 \in \mathbb{R}^d \quad \text{and}$$

$$k_2(x_1, x_2) = k_2(x_2, x_1) \quad \forall x_1, x_2 \in \mathbb{R}^d$$

(2) For any  $x \in \mathbb{R}^n$ , we have

$$x^T K_1 x \geq 0 \quad \text{and}$$

$$x^T K_2 x \geq 0$$

where,

$K_1 = [k_1(x_i, x_j)]_{n \times n}$  and  $K_2 = [k_2(x_i, x_j)]_{n \times n}$  are  $K$  matrices corresponding to kernels  $k_1$  and  $k_2$ , respectively.

Assume that  $k = 3k_1 + 5k_2$

To show:  $k(x_1, x_2) = k(x_2, x_1) \quad \forall x_1, x_2 \in \mathbb{R}^d$

and

$x^T K x \geq 0 \quad \forall x \in \mathbb{R}^n$ , where  $K = [k(x_i, x_j)]_{n \times n}$  is  $K$  matrices corresponding to kernel  $k$

Now,

$$\begin{aligned} k(x_1, x_2) &= 3k_1(x_1, x_2) + 5k_2(x_1, x_2) \\ &= 3k_1(x_2, x_1) + 5k_2(x_2, x_1) \\ &= k(x_2, x_1) \end{aligned}$$

and

$$\begin{aligned} x^T K x &= x^T (3K_1 + 5K_2) x \\ &= 3x^T K_1 x + 5x^T K_2 x \geq 0 \end{aligned}$$

It implies that  $3k_1 + 5k_2$  is a valid kernel.