

EN3150 Assignment 04: Kernel Methods

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1 Kernel Methods

1. Suppose that input space to feature space mapping (projection) is given by the following function.

$$\Phi(x) = (1, \sqrt{2}x, x^2). \quad (1)$$

For a one-dimensional input space show that the corresponding kernel function is $k(x, z) = (1 + xz)^2$.

2. Express the kernel function provided above for a two-dimensional input space, where $\mathbf{x} = (x_1, x_2)$ and $\mathbf{z} = (z_1, z_2)$.
3. What is the mapping function $\Phi(\mathbf{x})$ for the kernel function provided above in the above question (Q2).
4. Consider the kernel $k = (1 + \mathbf{x}^T \mathbf{z})^2$. For the following dataset, determine the kernel matrix (gram matrix). Is this a valid kernel for this data set? justify your answer.

Sample index	Data sample ($\mathbf{x} = (x_1, x_2)$)	Feature 1 (x_1)	Feature 2 (x_2)
1	\mathbf{x}_1	1	5
2	\mathbf{x}_2	3	4
3	\mathbf{x}_3	4	2
4	\mathbf{x}_4	10	12

Note:- The kernel matrix (gram matrix) is given by

$$\mathbf{G} = \begin{bmatrix} k(\mathbf{x}_1, \mathbf{x}_1) & k(\mathbf{x}_1, \mathbf{x}_2) & \cdots & k(\mathbf{x}_1, \mathbf{x}_N) \\ k(\mathbf{x}_2, \mathbf{x}_1) & k(\mathbf{x}_2, \mathbf{x}_2) & \cdots & k(\mathbf{x}_2, \mathbf{x}_N) \\ \vdots & \vdots & \ddots & \vdots \\ k(\mathbf{x}_N, \mathbf{x}_1) & k(\mathbf{x}_N, \mathbf{x}_2) & \cdots & k(\mathbf{x}_N, \mathbf{x}_N) \end{bmatrix} \quad (2)$$

5. Use the code given in listing 1 to generate data.

```
import numpy as np
import matplotlib.pyplot as plt
from sklearn.datasets import make_circles
# Generate data with make_circles
np.random.seed(5)
X, y = make_circles(n_samples=500, factor=0.3, noise=0.1)
```

Listing 1: Data generation.

- (a) Plot the scatter plot of the data (`plt.scatter(X[:, 0], X[:, 1], c=y, cmap=plt.cm.Paired)`).
- (b) Use the following mapping to map two-dimensional space to three-dimensional space (feature space). This feature space set is known as projected set. Visualize the projected set.

$$\Phi: \mathbf{x} = (x_1, x_2) \rightarrow \Phi(\mathbf{x}) = (x_1, x_2, x_1^2 + x_2^2) \in \mathbb{R}^3.$$

- (c) Now change the "factor=0.5" and observe the visualize the feature space in 3-d. What changes can you observe?
For the same data, use this mapping: $\mathbf{x} = (x_1, x_2) \rightarrow (x_1^2, x_2^2, x_1x_2)$. Is this mapping better than the previous mapping?
- (d) Run linear SVC (`svm.SVC(kernel='linear')`) on the original dataset which is generated based on listing 1 and the projected set using the mapping given in 5b and report the classification accuracies for both cases.

2 Additional Resources

- 1. [Scikit-learn SVM documentation](#)
- 2. [Scikit-learn Custom Kernel](#)
- 3. [Stanford CS229 Lecture notes](#)

Submission

- Upload a report and your codes as a zip file named as "EN3150_your_indexno_A04.zip". Include the index number and the name within the report as well.
- The interpretation of results and the discussion are important in the report.
- An extra penalty of 10% is applied for late submission.
- Plagiarism will be checked and in cases of plagiarism, an extra penalty of 10% will be applied.