

1.1.

Given that $x = (x_1, x_2)^T$

$$\phi(x) = (x_1^2, \sqrt{2} x_1 x_2, x_2^2)^T$$

We know that $K(x, z) = \phi(x) \cdot \phi(z)$

$$K(x, z) = (x_1^2, \sqrt{2} x_1 x_2, x_2^2)^T \cdot (z_1^2, \sqrt{2} z_1 z_2, z_2^2)^T$$

$$= (x_1^2 z_1^2 + \sqrt{2} x_1 x_2 \sqrt{2} z_1 z_2 + x_2^2 z_2^2)^T$$

$$K(x, z) = (x_1 z_1 + x_2 z_2)^2$$

$$K(x, z) = (x \cdot z)^2$$

1.2

A. If we map the input vector to the feature space and then do the dot product on the mapped features. We have

$$\phi(x) \cdot \phi(z) = (x_1^2, \sqrt{2} x_1 x_2, x_2^2)^T \cdot (z_1^2, \sqrt{2} z_1 z_2, z_2^2)^T$$

$$= x_1^2 z_1^2 + \sqrt{2} x_1 x_2 \sqrt{2} z_1 z_2 + x_2^2 z_2^2$$

In the above equation we have $3 + 5 + 3 = 11$ multiplications and 2 additions.

B. If we use the equation I got in question 1

$$K(x, z) = (x_1 z_1 + x_2 z_2)^2$$

If we consider $m = x_1 z_1 + x_2 z_2$ then we have $K(x, z) = m \cdot m$ and in m we have 2 multiplications and 1 addition, so for $K(x, z)$ we will have 3 multiplications and 1 addition, where 2 multiplications are needed for m and one multiplication needed for squaring m .

Actually both A and B refer to the same equation A is expansion of equation B so we actually should get the same no of multiplications and additions for both.

2.1.

$K(x, x')$ can be written as $\text{floor}(\frac{x \cdot x'}{\|x\| \|x'\|})$ which is 1 when $x=x'$ else 0. $\phi(x)$ should be of the form $[1 \ 0 \ 0 \dots]$ it should be an m dimensional vector there only one element is 1 and rest all are zero, only then we can have $\phi(x) \cdot \phi(x')$ as 1 when $x=x'$ else 0. The closest function I can get for this scenario is $\phi(x) = \frac{x}{\|x\|}$, since dimension of X is not given I am assuming it to be m -dimension.

2.2

The given kernel is a linear kernel since it is represented as $K(x, z) = x \cdot z$ and since this is a valid kernel we will have a linear separator which can classify the given data. We know that $w \cdot x$ can be written as $w \cdot x = a_1 K(x_1, x) + a_2 K(x_2, x) + \dots$ if $K(x, x')$ is $x \cdot x'$ then $w \cdot x$ will be a linear equation and hence we get a linear separator.

2.3

This will be a bad idea when m is very large or dimension of X is smaller than m . It will be computationally expensive and also will be hard to find a function which maps accordingly.

3.

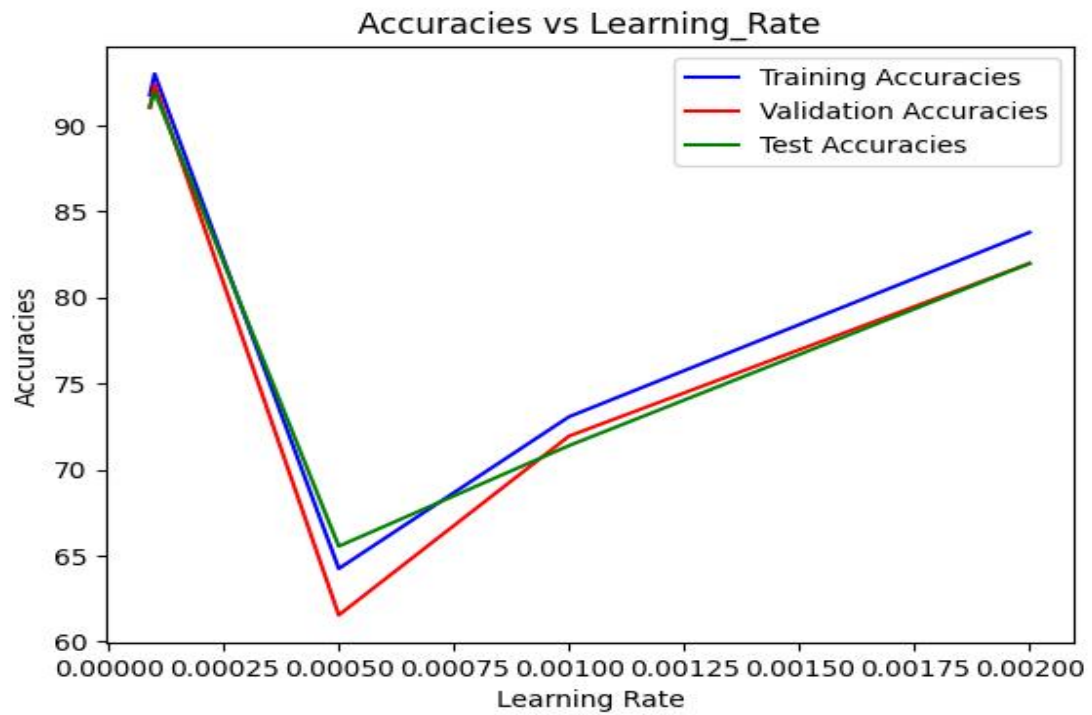
Optimal Hyper Parameters are
Learning rate = 0.0001
No of epochs = 50
Regularization = 0.85

Training Accuracy = 92.531
Validation Accuracy = 91.630
Testing Accuracy = 92.061

The code for the same is attached

The plots are as follows

Accuracies vs Learning rate



Accuracies vs Regularization Constant

