CS5691: Tutorial 7

## Question 1

Consider a Support Vector Machine and the following training data from two categories:

category	x1	x2
1	1	1
1	2	2
1	2	0
2	0	0
2	1	0
2	0	1

#### Part A

Plot these six training points, and construct by inspection the weight vector for the optimal hyperplane.

### Part B

What are the support vectors?

## Part C

Will our solution change if we delete point  $\ensuremath{\mathsf{D}}$  .

# Question 2a True/False

In theory, a Gaussian kernel SVM can model any complex separating hyperplane.

# Question 2b True/False

For every kernel function used in a SVM, one can obtain a equivalent closed form basis expansion.

# Question 2c True/False

Overfitting in an SVM is a function of number of support vectors

#### Kernel

Consider 1 dimensional data points  $x_1=-1, x_2=0, x_3=2$  , with labels -1 , 1 , -1 respectively.

## Part A

For the basis function  $\phi(x) = \begin{bmatrix} 1 \\ x^2 \end{bmatrix}$  find the kernel K.

### Part B

Use the basis expansion to see the data becomes linearly separable.

Suppose we have kernels k1 and k2. Then show that  $k(x,z)=\alpha k1(x,z)+\beta k2(x,z)$  for  $\alpha,\beta\geq 0$  is a valid kernel

# Acknowledgment

The question for SVM has been taken from Duda and Hart PRML book.