## Machine Learning 1

### 1 State TRUE/FALSE (Justify if false) - 10 points

- a K-means solves clustering problem
- b K-means is supervised learning algorithm
- c Early stopping is used for regularization of neural network
- d Backpropagation is algorithm used for regularization of neural network
- e Cross validation is used to compare different algorithms

f

g

h

i

j

## 2 Principal Component Analysis - 10 points

- a Given a plot. Draw the first PC. (3 points)
- b In above plot, if data is reduced to one dimension, is the given data linearly separable? (3 points)
- c Advantage of applying PCA before using SVM (4 points)

#### 3 Overfitting and Regularization - 10 points

a

b

 $^{\rm c}$ 

# 4 SVM, Convex Functions and Hyperplanes - 20 points

- a Given linear soft-margin SVM with squared hinge loss function. Compute the gradient of the square hinge-loss SVM. (5 points)
- b Show that the above SVM is a convex problem. (5 points)
- c Why it is important to know if a function is convex? (3 points)
- d (Given a plot) Explain if the problem can be solved with hard margin SVM. (2 points)
- e Draw two hyperplanes in above plot(one using hinge loss and other using squared hinge loss). Explain the decision. (5 points)

### 5 Regression, SGD - 20 points

- a Given optimization problem for regression. Explain how value of alpha(co-efficient of w) affects regularization procedure. (4 points)
- b Computing wJ(w) of the above problem. (4 points)
- c Compute closed form solution for the above (4 points)
- d SGD for regression problem (8 points)

### 6 Convolutional Neural Network - 15 points

- a Pseudocode for CNN with Max Pooling (6 points)
- b Given Matrix M, stride = 2, apply Max pooling to get the output matrix (3 points)
- c Use of Dimensionality reduction to avoid overfitting while solving binary classification problem using deep neural network (6 points)

## 7 Kernel - 15 points

- a Kernel Trick and use in Machine Learning (3 points)
- b Let  $\mathbf{k}_1$  and  $\mathbf{k}_2$  be kernels. Prove that

$$k_3(y, x) = (\alpha k_1(y, x) + \beta k_2(y, x))^n$$

Hint: Use Hadamard Theorem(Product of two positive semi definite matrices is also positive semi definite) (4 points)

c Value of n when kernel matrix K is singular (8 points)