

CSE471: Statistical Methods in AI

Monsoon 2016

Assignment #1: LDFs and Neural Networks

Due: Before 5:00pm on 14/09/2016

General Instructions:

- Assignment can be implemented in Matlab/Octave, Python, C/C++, R .
- Ensure that submitted assignment is your original work. Please do not copy any part from any source including your friends, seniors and/or the internet. If any such attempt is caught then serious actions including an **F grade in the course** is possible.
- A single pdf file needs to be uploaded to the Courses Portal. The file should contain your answers as well as the code you have written and its output.
- Include the assignment number, your name and roll number at the top-left of the first page of your submission.
- Your grade will depend on the correctness of answers and output. In addition, due consideration will be given to the clarity and details of your answers and the legibility and structure of your code.

Problem 1

Implement Perceptron-based Linear Discriminant Functions. The data set to be used for the exercises given below is the following sample set comprising a two-class problem.

$$\omega_1 = [(2, 7); (8, 1); (7, 5); (6, 3); (7, 8); (5, 9); (4, 5)]$$

$$\omega_2 = [(4, 2); (-1, -1); (1, 3); (3, -2); (5, 3.25); (2, 4); (7, 1)]$$

Implement the following algorithms

- A. Single-sample perceptron
 - B. Single-sample perceptron with margin
 - C. Relaxation algorithm with margin
 - D. Widrow-Hoff or Least Mean Squared (LMS) Rule
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- I. In each case, plot the data points in a graph (e.g. red: class- ω_1 and blue: class- ω_2) and also show the weight vector \mathbf{a} learnt from all of the above algorithms in the same graph (labeling clearly to distinguish different solutions).
 - II. Run each of the above algorithms for various initial values of the weight vector, and comment on the dependence of convergence time (run-time) on initialization.
 - III. Similarly explore the effect of adding different margins on the final solution as well as on the convergence (run-) time for algorithms (B) and (C).

- IV. In case of LMS, add extra points to this data for making it linearly non-separable and show how LMS provides an acceptable decision boundary with some classification error.
- V. As part of the submission include the code for each of the algorithms along with a small report that explains the algorithms, implementation details, the results and their analysis.

Problem 2

Implement a simple supervised, feed-forward, back-propagation network with sigmoid activation function for the problem of optical character recognition for any three digits between 0 and 9. Please do not use any standard NN library and write your code from scratch.

- Data: Use any three digits between 0 and 9 from the optdigits data set that comes from the [UCI Machine Learning Repository](#).
- Preprocessing: Digitize and down-sample images (to 8x8 or so).
- Classifier: Try few configurations for a 3-Layer Neural Network with varying number of hidden units and two output units and report learned weights from backpropagation done using training data.
- Report: Draw representative neural network architecture and approximately 2-3 pages of writeup on your experiments and results including the learned weights in a table and intermediate output of hidden layers for each of the three characters after learning is over.