

Mid-sem

EE698V – Machine Learning for Signal Processing

Instructions:

- Use Python 3 for coding
- Use only numpy library. No other libraries allowed.
- Submit your ipynb notebooks. Your submission will be evaluated based on the 4 functions explained below. You can define any number of extra functions, classes and variables you wish.
- The runtime of your code should be restricted to 5 minutes on Google colab CPU.
- For any questions, ask on MookIT forum.

```
class MidSem:
    def __init__(self):
        ...
    def generateData(self, params):
        ...
        return X_ki
    def trainWithMSE(self, X_ki, Yd_k):
        ...
        return w
    def trainBP(self, X_ki, Nepochs=100, eta=0.1):
        ...
        return w
    def evaluate(self, X_i, t_i, w):
        ...
        return CM, accuracy, Y_i
```

Q1. Generate Data (10 marks)

Write the function "generateData" that generates random points that are Gaussian distributed. The input "params" is a list of lists: $[[\mu_1, \sigma_1, N_1], [\mu_2, \sigma_2, N_2], [\mu_3, \sigma_3, N_3], \dots]$, where μ and σ are numpy vectors of mean and covariance matrix of the Gaussian, respectively, and N is the number of samples from that Gaussian. The shape of μ is $(M,)$ and that of σ is (M,M) , and $\text{len}(\text{params})$ is K . An example is like this

```
params = [[np.array([2.5, -1.3]), np.ones((2,2)), 5], \
          [np.array([-1.9, 3.4]), .5*np.eye(2), 4]]
```

The output "X_ki" should again be a list of numpy arrays, each array being of shape (N,M) . Here, k is the class index and i is the sample index for that class.

Hint: If you need data normalization, you can define class variables.

Q2. Linear Binary Classifier with Least Squared Solution (20 Marks)

Write the function "trainWithMSE" that trains a linear model $\mathbf{y}=\mathbf{w}\mathbf{x}$ using least squared solution. The input X_k is defined above and Yd_k is a list of class targets with $\text{len}(Yd_k)$ as 2; e.g. $Yd=[-1,1]$.

The output \mathbf{w} is a numpy array of **weights** (+ *bias*) of the linear model with shape $(\mathbf{M}+1,)$.

Q3. Linear Classifier with Softmax and Backpropagation (30 Marks)

Write the function "trainBP" that trains a model $\mathbf{y}=\text{softmax}(\mathbf{w}\mathbf{x})$ using backpropagation. The input $Nepochs$ (int) is the number of epochs for training and η is the learning rate (float). Feel free to change their default values. Here, X_k and \mathbf{w} are as defined above. The targets Yd_k can be derived from index k in X_k ; there are K classes.

Q4. Evaluation (20 Marks)

Write the function "evaluate" that tests the trained model. The input X_i is a numpy array of shape (N,M) , t_i is a numpy array of targets of shape $(N,)$ and \mathbf{w} is as defined above. The output CM is the confusion matrix, i.e., a numpy array of shape (K,K) ; $accuracy$ is the accuracy of the model; and Y_i are the predictions corresponding to X_i , i.e., each row is $\mathbf{y}=\mathbf{w}\mathbf{x}$ or $\mathbf{y}=\text{softmax}(\mathbf{w}\mathbf{x})$ and shape of Y_i is $(N,)$ or (N,K) respectively.

Note: This function should be able to evaluate the models of both Q2 and Q3.