

# Assignment – Machine Learning

**Deadline:** Dec 7, before 23:59 PM

**Submission:** answer the following questions.

## Exercise 1 [1 point]:

Consider a binary classification problem on data set  $\{(\mathbf{x}_i, y_i)\}_{i=1}^N$  with raw inputs  $\mathbf{x}_i \in \mathbb{R}^D$  and outputs  $y_i \in \{0, 1\}$ . You decide to transform your raw features to get a new dataset  $\{(\phi(\mathbf{x}_i), y_i)\}_{i=1}^N$ . Specifically, you choose  $\phi(\mathbf{x}) = \mathbf{A}\mathbf{x}$  for some  $\mathbf{A} \in \mathbb{R}^{M \times D}$  with dimension  $M > D$ .

If the original dataset is *not* linearly separable, might the transformed dataset be linearly separable? YES/NO then explain (1 sentence).

## Exercise 2 [4 points]:

Consider a 3-class SVM trained with one-vs-rest strategy. Suppose that the linear discriminant for each class  $k \in \{1, 2, 3\}$  is defined, respectively, by:

$$\begin{aligned} \mathbf{w}_1 &= (2, \frac{1}{2}) & b_1 &= 2 \\ \mathbf{w}_2 &= (3, -3) & b_2 &= \frac{9}{2} \\ \mathbf{w}_3 &= (-1, -\frac{3}{2}) & b_3 &= -2 \end{aligned}$$

- [3 marks] What values do the three linear discriminants  $y_1, y_2, y_3$  take for input  $\mathbf{x} = (-\frac{1}{2}, -2)$ ? Show your calculations for all three numbers.
- [1 mark] What class would the combined model predict? Explain.

### **Exercise 3 [4 points]:**

- a. [1 mark] Random forests are bagged decision trees. TRUE or FALSE?
- b. [1 mark] Compared to training a single model on all the training data, does bagging tend to reduce the *training error*, the *testing error*, or *both*?
- c. [2 marks] Consider training set:

```
>>> X = [[0, 0],  
         [1, 10],  
         [2, 20],  
         [3, 30]]  
>>> y = [0, 1, 2, 3]
```

Suppose we draw a bootstrap sample of  $(\mathbf{X}, \mathbf{y})$ , giving:

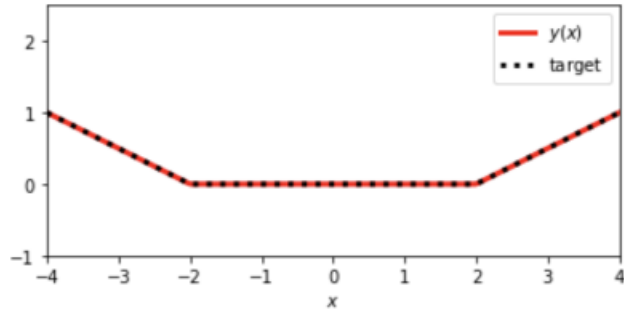
```
>>> X_bootstrap  
[[3, 30],  
 [1, 10],  
 [3, 30],  
 [0, 0]]  
>>> y_bootstrap  
[3, 1, 3, 0]
```

What *sample\_weight*, when applied to the original training set, would be equivalent to training on this bootstrap sample? i.e., give 4 numbers:

```
estimator.fit(X_bootstrap, y_bootstrap)      # bootstrap  
estimator.fit(X, y, sample_weight=[?, ?, ?, ?]) # same
```

### Exercise 4 [6 points]:

Part (c) pertains to the target function below, and the neural network  $y(x)$  that manages to approximate it (in this case, exactly).



- [1 mark] What is the specific role of backpropagation? Brief sentence.
- [2 marks] Suppose input  $x$  has 50 features and you are doing 10-way classification. How many parameters would a 50-100-20-10 sigmoid network contain? Show calculations (hidden layer 1 + hidden layer 2 + output layer).
- [3 marks] Give specific values for the parameters of the 1-2-1 ReLU neural network that matches the  $y(x)$  from the figure. The architecture is:

$$z_1 = \text{ReLU}(w_1 x + b_1)$$

$$z_2 = \text{ReLU}(w_2 x + b_2)$$

$$y = w_3 z_1 + w_4 z_2 + b_3$$

You can type `w1 = ...` to indicate the value of  $w_1$ , etc.

### **Exercise 5 [5 points]:**

Part (c) is about the function below:

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{32}x^2} + \frac{3}{4\sqrt{2\pi}} e^{-\frac{1}{2}x^2 + 4x - 4}$$

(Here symbol  $\pi \approx 3.14$  is a constant, not a component mixture weight.)

- a. [1 mark] Are Gaussian mixture models (GMMs) *parametric* or *non-parametric*?
- b. [1 mark] In the EM algorithm, how are the  $\mu_k$  typically initialized in practice?
- c. [3 marks] Is  $f(x)$  a valid Gaussian mixture density? YES/NO then
  - If YES, give the six parameters  $(w_1, \mu_1, \sigma_1)$  and  $(w_2, \mu_2, \sigma_2)$  of the model, where  $w$  is mixture weight,  $\mu$  is mean, and  $\sigma$  is standard deviation.
  - If NO, explain why it's not a Gaussian mixture density (1 sentence).