

Machine Learning BLG527E, Jan 10, 2013, Final Exam.

1	2	3	4	Total

Duration: 120 minutes.

Write your answers neatly in the space provided for them. Write your name on each sheet. Books and notes are closed. Good Luck!

QUESTIONS

QUESTION1) [30 points, 5 points each] (use at most 3 sentences per question, use of formulas, drawings etc. to better express yourself is encouraged.)

- Compare Adaboost and bagging.
- What is ROC and AUC?
- How do you use momentum and adaptive learning rate for training a multilayer perceptron?
- What is the difference between k-means clustering and Gaussian Mixture Model (GMM) clustering?
- Compare backward and forward feature selection.
- Compare Mahalanobis and Euclidean distance.

QUESTION2) [40 points, 10 points each] (use at most 10 sentences per question, use of formulas, drawings etc. to better express yourself is encouraged.)

- Assume that you have K classifiers which are independent and i th classifier has generalization error of E_i . Show that the average of these classifiers has less generalization error. Why, in practice, classifier averaging doesn't always result in less generalization error?
- What are the three basic problems that you can solve using an hmm? Explain each of problem and how it is solved briefly.
- How does 5x2 cross validation work? Under which circumstances do you use leave one out, 10-fold or 5x2 cross validation?
- How do you use Parzen windows for density estimation, classification and regression?

QUESTION3) [20points]

	Actual	$g_1(x)$	$g_2(x)$	$g_3(x)$	$g_4(x)$	$g_5(x)$	$f_1(x)$	$f_2(x)$	$f_3(x)$	$f_4(x)$	$f_5(x)$
Train	0	0	0	0	0	0	0	0	0	1	1
	1	0	1	0	0	1	1	1	1	0	0
	0	0	0	1	1	0	0	1	1	0	0
	1	0	1	0	1	0	1	0	1	0	1
Test	?	1	1	1	1	1	1	0	1	0	1
	?	0	0	0	0	0	0	1	0	1	0
	?	0	1	0	1	0	1	0	1	0	1
	?	1	1	1	1	1	0	1	0	1	0

You are given a dataset which contains 4 training inputs and output, 4 test inputs. You have two different types of classifiers $f()$ and $g()$, you train and test 5 different configurations of $f()$ and $g()$ (for example you might choose different parameters of the classifiers). The training and test outputs that are obtained for each input are shown in the table. Would you choose a classifier of kind $f()$ or $g()$ for this task? Clearly explain the reason behind your choice?

[Hint: if M samples (data sets) $X_i = \{x_i^t, r_i^t\}$, $i=1, \dots, M$ are used to fit $g_i(x)$, $i=1, \dots, M$

Name and Student ID:

Signature:

$$\text{Bias}^2(g) = \frac{1}{N} \sum_t [\bar{g}(x^t) - f(x^t)]^2$$

$$\text{Variance}(g) = \frac{1}{NM} \sum_t \sum_i [g_i(x^t) - \bar{g}(x^t)]^2$$

$$\bar{g}(x) = \frac{1}{M} \sum_i g_i(x)$$

]

QUESTION4) [10points]

Compute the change in v and w (i.e. Δv and Δw) if you modified the error function for a neural network as follows.

$$E(\mathbf{W}, \mathbf{v} | \mathcal{X}) = \frac{1}{2} \sum_t (r^t - y^t)^2 + \|\mathbf{v}\|^2$$

Hint: Without the regularization term, you would have:

$$y_i = \mathbf{v}_i^T \mathbf{z} = \sum_{h=1}^H v_{ih} z_h + v_{i0} \quad z_h = \text{sigmoid}(\mathbf{w}_h^T \mathbf{x})$$

$$\Delta v_h = \sum_t (r^t - y^t) x_h^t$$

$$\begin{aligned} \Delta w_{hj} &= -\eta \frac{\partial E}{\partial w_{hj}} = -\eta \sum_t \frac{\partial E}{\partial y^t} \frac{\partial y^t}{\partial z_h^t} \frac{\partial z_h^t}{\partial w_{hj}} \\ &= -\eta \sum_t -(r^t - y^t) v_h z_h^t (1 - z_h^t) x_j^t \\ &= \eta \sum_t (r^t - y^t) v_h z_h^t (1 - z_h^t) x_j^t \end{aligned}$$