

**The goal of this homework teach you about over fitting and cross validation.**

You are to implement regularized polynomial curve fitting.

Examples  $(x_n, y_n)$ ,  $n = 1..N$

$w_i$  is the weight for  $x_n^i$ , ( $0 \leq i \leq 19$ )

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$$\mathbf{w}^* = \operatorname{argmin}_{\mathbf{w}} \left( \sum_n \underbrace{\left( \sum_{i=0}^{19} w_i x_n^i - t_n \right)^2}_{y_n} + \lambda \sum_{i=0}^{19} w_i^2 \right)$$

Let's rewrite the above in matrix notation:  $\mathbf{X}$  is a  $20 \times N$  matrix, where  $N$  is the number of examples and the  $n$ th column of  $\mathbf{X}$  is  $(1, x_n, x_n^2, \dots, x_n^{19})^\top$ .  $\mathbf{t}$  is the vector of outputs  $\mathbf{t} = (t_1, t_2, \dots, t_n)^\top$ .

$$\mathbf{w}^* = \operatorname{argmin}_{\mathbf{w}} (\|\mathbf{X}^\top \mathbf{w} - \mathbf{t}\|^2 + \lambda \|\mathbf{w}\|^2)$$

This is solved by differentiation:

$$\mathbf{X}(\mathbf{X}^\top \mathbf{w}^* - \mathbf{t}) + \lambda \mathbf{w}^* = \mathbf{0}$$

$$(\mathbf{X}\mathbf{X}^\top + \lambda \mathbf{I})\mathbf{w}^* - \mathbf{X}\mathbf{t} = \mathbf{0}$$

$$\mathbf{w}^* = (\mathbf{X}\mathbf{X}^\top + \lambda \mathbf{I})^{-1} \mathbf{X}\mathbf{t}$$

- We will provide a training set (hw/hw1/train.txt) and a test set (hw/hw1/test.txt). Each row contains one value of  $x$  and the corresponding value of  $t$ , separated by space. Use 10-fold cross validation to find the best choice of  $\lambda$  and report the loss on the test set.
- Write a roughly 3 page summary of what you did:
  - at least one plot
  - report the best value of  $\lambda$  and the test error.
- Note that you need to find a suitable way to discretize  $\lambda$ .

**We encourage you to work in groups of up to three.**

**Every group is to do their own work.**

Hint: First implement a simple split of the train.txt into train and validation set. After that implement 10-fold cross validation.

Extra credit:

- Modify the above solution so that the bias term/weight is not regularized. Justify your answer.

- Show variance bars for the losses of the different holdouts.
- Implement leave one out cross validation and show that in some sense this is better than 10-fold cross validation.

Remember you to look at the test score only once to report results!