

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

You are to implement regularized polynomial curve fitting.

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

w_i is the weight for x_n^i , $(0 \leq i \leq 9)$

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$$\mathbf{w}^* = \operatorname{argmin}_{\mathbf{w}} \left(\sum_n \underbrace{\left(\sum_{i=0}^9 w_i x_n^i - t_n \right)^2}_{y_n} + \lambda \sum_{i=0}^9 w_i^2 \right)$$

Let's rewrite the above in matrix notation: \mathbf{X} is a $10 \times N$ matrix, where N is the number of examples and the n th column of \mathbf{X} is $(1, x_n, x_n^2, \dots, x_n^9)^\top$. \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 ([train.txt](#)) and a test set ([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 λ 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 λ and the test error.
- Note that you need to find a suitable way to discretize λ .

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

Every group is to do their own work. Same score for all members of a group.

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!

Strong suggestion: Do this homework in Matlab, Octave or R.