

CSE 5819 Assignment #4

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This assignment is organized as follows:

1. ChatGPT - pg 1–4
2. Problems - pg 5–7
3. Coding - attached as a separate .ipynb file

1 Part 1: ChatGPT Self-Learning (20pts)

You can start from the following prompts, but you need to create subsequent questions that attempt to understand the basic concepts of machine learning.

- 1.1 What is overfitting? What is underfitting?**
- 1.2 How does logistic regression handle overfitting?**
- 1.3 How does least squares handle overfitting?**
- 1.4 What regularizers can be used to regularize machine learning algorithms?**
- 1.5 How to compute gradient of the least squares loss?**
- 1.6 How to compute gradient of the logistic loss?**

2 Problems

2.1 Question One

[Gradient Descent]

[20pts] You have a univariate function you wish to minimize, $f(w) = 5(w - 11)^4$. Suppose you wish to perform gradient descent with constant step size $\alpha = 1/40$. Starting with $w_0 = 13$, perform 2 steps of gradient descent. What are w_0, \dots, w_2 ?

2.2 Question Two

[Logistic Regression]

In this problem, we are going to assume the same notation setup in class. For logistic regression, we model the class probability by

$$P(y|\mathbf{x}_i) = \sigma(y(\mathbf{w}^T \mathbf{x}_i))$$

where we define the *logistic function* $\sigma : \mathbb{R} \rightarrow \mathbb{R}$ as

$$\sigma(s) = \frac{1}{1 + e^{-s}}$$

(Note: We dropped the bias term b since we can always absorb the bias into \mathbf{w} by representing \mathbf{x} in homogeneous coordinates.)

1. [20 pts] Prove that the given σ is a properly defined probabilistic model. In other words, prove that $P(y = +1|\mathbf{x}_i) + P(y = -1|\mathbf{x}_i) = 1$

(Hint: use the relationship between $\sigma(s)$ and $\sigma(-s)$.)

2.3 Question Three

2. [20 pts] Show that the gradient of the log likelihood function, namely, $\nabla_w \log P(\mathbf{y}|X, \mathbf{w})$ is

$$\sum_{i=1}^n (1 - \sigma(y_i(\mathbf{w}^T \mathbf{x}_i))) y_i \mathbf{x}_i$$

where y_i is a scalar, and \mathbf{x}_i is a vector of descriptors, and $\mathbf{y} = [y_1, y_2, \dots, y_n]^T$ is a vector, and $X = [\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n]$ is a matrix indicating all data.