#### CS 760: Machine Learning

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Homework 3: Logistic Regression

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### Problem 3.1

- (a) I used good ol' fashioned guess and check. I set the iterations to 3000 and printed out the weights after every 100 iterations. Once the values started to level off, I considered that convergence
- (b) It took me about 2000 iterations to converge to a respectable value, although extending it to 3000 refined it just a bit more.
- (c) I performed min-max normalization to avoid running into a RuntimeWarning (when the exponential got WAY too big), so my results for  $\hat{\theta}$  were:

 $\left[-0.10924571, 0.03981236, -0.77742697, -0.01827487, 0.002808\right]$ 

- (d) The maximum log-likelihood of  $\hat{\theta}$  is: -0.6700727101571831
- (e) From Theorem 6.2 in the Logistic Regression notes, we can see that  $\hat{\theta} \xrightarrow{d} \mathcal{N}(\theta^*, I_{\theta^*}^{-1})$  where the Fisher Information is shown as:

 $I_{\theta^*} = \sum_{i=1}^{N} \frac{e^{-\theta^{*T} \mathbf{x}_i}}{(1 + e^{-\theta^{*T} \mathbf{x}_i})^2} \mathbf{x}_i \mathbf{x}_i^\mathsf{T}$ 

#### Problem 3.2

- (a) Borrowing from the Logistic Regression notes again, we can see that the MLE of the log-odds  $\hat{\omega} := \hat{\theta}^\mathsf{T} \mathsf{x}$  where  $\hat{\theta}$  are the true parameters,  $\theta^*$ .
- (b) Furthermore, the asymptotic distribution of  $\hat{\omega}$  is defined as  $\hat{\omega} \xrightarrow{d} \mathcal{N}(\theta^{*\mathsf{T}}\mathbf{x}, \mathbf{x}^{\mathsf{T}}I_{\theta^{*}}^{-1}\mathbf{x})$

## Problem 3.3

- (a)
- (b)
- (c)

# Problem 3.4

- (a)
- (b)
- (c)