

## Homework 6: The Cornerstone of AI: Gradient Descent

**DO NOT POLLUTE!** AVOID PRINTING, OR PRINT 2-SIDED MULTIPAGE.

In this homework you will implement gradient descent and use it to construct a logistic regression AI system.

To this end, we will use the titanic dataset (`titanic_data.csv`), containing the following information about 887 passengers: 1) whether they survived or not (1 = survived, 0 = deceased), 2) passenger class, 3) gender (0 = male, 1 = female), 4) age, 5) number of siblings/spouses aboard, 6) number of parents/children aboard, and 7) fare:

	Passenger 1	Passenger 2	Passenger 3	...	Passenger 887
<b>Survived</b>	<b>0</b>	<b>1</b>	<b>1</b>	...	<b>0</b>
Passenger Class	3	1	3	...	3
Gender	0	1	1	...	0
Age	22	38	26	...	32
Siblings/Spouses	1	1	0	...	0
Parents/Children	0	0	0	...	0
Fare	7.25	71.2833	7.925	...	7.75

Our goal is to construct an AI system that determines/predicts whether an individual would survive or not.

Each problem is worth 50 points.

**Problem 1.** Deliver python code to load this dataset into data vectors  $\mathbf{x}_i$  and their corresponding responses  $y_i$ .

**Problem 2.** Recall that logistic regression uses the following log-likelihood as a loss function:

$$\ell(\mathbf{w}) := - \sum_{i=1}^N \log \left[ \left( \frac{1}{1 + e^{-\mathbf{w}^T \mathbf{x}_i}} \right)^{y_i} \left( \frac{1}{1 + e^{\mathbf{w}^T \mathbf{x}_i}} \right)^{1-y_i} \right],$$

whose gradient is given by:

$$\ell'(\mathbf{w}) = - \sum_{i=1}^N \left( y_i - \frac{1}{1 + e^{-\mathbf{w}^T \mathbf{x}_i}} \right) \mathbf{x}_i.$$

Deliver python code to compute this gradient at a given  $\mathbf{w}$ .

**Problem 3.** Recall that gradient descent can be summarized as an iterative process that repeats until convergence the following update:

$$\mathbf{w}_{t+1} = \mathbf{w}_t - \eta \ell'(\mathbf{w}_t),$$

where  $\mathbf{w}_0$  is an initial *guess*, and  $\eta$  is the step size parameter. Deliver python code to implement gradient descent.

**Problem 4.** Use your implementations above to learn the optimal parameter  $\mathbf{w}^*$ . What value do you obtain for  $\mathbf{w}^*$ ?

**Problem 5.** What value or strategy did you use for  $\eta$ ?

**Problem 6.** Change  $\eta$  to another value and repeat the training process. What value did you obtain for  $\mathbf{w}^*$ ?

**Problem 7.** Explain which choice of  $\eta$  is better, and why.

**Problem 8.** Build your own feature vector  $\mathbf{x}$ . According to your AI system, would you have survived the titanic sinking? Briefly explain your conclusion.