

Optimization for Machine Learning (Spring 2026): Homework 2

- LO4ML stands for the textbook *Linear Algebra and Optimization for Machine Learning*.
 - You **must email your submission** as a **PDF file** to kbala@wsu.edu. You are welcome to write answers by hand and scan the pages. Put all the images on a PDF file, though.
 - Your file name should identify you in the following manner. If you are Napoleon Dynamite, you should **name your submission NapoleonDynamite_Hw2.pdf**. If you want to add more bits to the title, e.g., Math565, you could name it NapoleonDynamite_Math565_Hw2.pdf, for instance. But you should **start the file name with NapoleonDynamite. And please avoid white spaces in the file name**.
 - Begin the SUBJECT of your email submission with the same **FirstnameLastname**, expression, e.g., “**NapoleonDynamite Hw2 submission**”.
 - This homework is due by **10:00 PM on Tuesday, February 17**.
1. In this problem, you will explore the basic gradient descent (GD) algorithm and how its convergence could depend on the learning rate (step size) α . Recall the basic GD update step: $\mathbf{w} \leftarrow \mathbf{w} - \alpha \nabla J$. Consider the loss function $J(\mathbf{w}) = \frac{1}{2} \mathbf{w}^T A \mathbf{w}$ for $\mathbf{w} \in \mathbb{R}^2$ and a symmetric positive semidefinite matrix $A \in \mathbb{R}^{2 \times 2}$.
- (15) For a given starting point $\mathbf{w}_0 = [1 \ 1]^T$ and learning rate $\alpha = 0.1$, specify two different instances A_s and A_f of A on which GD runs *slow* and *fast*, respectively. Define on your own the exact characteristics of slow and fast modes of convergence. Also create visualizations of the two loss function surfaces J_s and J_f using A_s and A_f , respectively, for the A matrix.
 - (15) Computationally demonstrate the convergence behavior of GD for both the slow and fast instances of J . Include any code you want to share in the your homework PDF file itself (no need to send them separately).
 - (10) How does the rate of convergence of GD depend on the eigenvalues of A , or of the related matrix in the update step of GD?
 - (10) Explain why the same learning rate α that works well for A_f causes issues for A_s .
2. (20) LO4ML Problem 4.8.2 from Page 187. This is the same result presented as Lemma 8 in Lecture 8. For this as well as other problems, you are encouraged to use the notation we have seen in class: \mathbf{w} for the variable (parameter) vector and \mathbf{x}_i for the data vector of the i -th instance, etc., rather than the notation used by LO4ML, e.g., W, X_i , etc.
3. (20) LO4ML Problem 6 from Page 200.
4. (25) LO4ML Problem 12 from Page 201.