

# Assignment: Gradient Descent for Unconstrained Minimization

## Instructions

You will use **Python** or **R** to implement gradient descent to minimize the following unconstrained functions. For each problem:

- Implement gradient descent using a **fixed step size**.
- Stop when the gradient norm satisfies  $\|\nabla f(x)\| < \varepsilon$ , or after a maximum number of iterations.
- Plot the loss vs. iterations and, where applicable, a 2D or 3D visualization of the descent path.

## General Parameters

For each problem, you will be provided:

- The objective function  $f(x)$
- An initial starting point  $x_0$
- Step size  $\eta$
- Threshold  $\varepsilon$

## Problem Set

### Problem 1: Linear Regression Loss

**Function:**

$$f(\beta) = \frac{1}{2n} \sum_{i=1}^n (y_i - \mathbf{x}_i^\top \beta)^2$$

**Details:**

- Generate  $n = 100$  paired  $(x, y)$  where  $x \sim N(1, 2)$  and  $y \sim N(2 + 3x, 5)$  data ( $n = 100$ ,  $p = 2$ )
- Initial point:  $\beta_0 = [0, 0]$
- Step size:  $\eta = 0.01$
- Threshold:  $\varepsilon = 10^{-6}$

## Problem 2: Logistic Regression Negative Log-Likelihood

**Function:**

$$f(\beta) = \sum_{i=1}^n \log(1 + \exp(-y_i \mathbf{x}_i^\top \beta))$$

**Gradient:**

$$\nabla f(\beta) = - \sum_{i=1}^n y_i \mathbf{x}_i (1 - \sigma(y_i \mathbf{x}_i^\top \beta)) \quad \text{where } \sigma(z) = \frac{1}{1 + e^{-z}}$$

**Details:**

- Binary classification problem (data provided here in the link [https://docs.google.com/spreadsheets/d/13CmIStaYtiQqR\\_dhBPrkJHJINvVln9cepHypNinVQT3c/edit?gid=0#gid=0](https://docs.google.com/spreadsheets/d/13CmIStaYtiQqR_dhBPrkJHJINvVln9cepHypNinVQT3c/edit?gid=0#gid=0))
- Initial point:  $\beta_0 = [0, 0]$
- Step size:  $\eta = 0.05$
- Threshold:  $\varepsilon = 10^{-5}$

## Problem 3: Quadratic Convex Function

**Function:**

$$f(x) = x^\top A x + b^\top x \quad \text{with } A = \begin{bmatrix} 2 & 0 \\ 0 & 4 \end{bmatrix}, \quad b = \begin{bmatrix} -4 \\ -8 \end{bmatrix}$$

**Details:**

- Initial point:  $x_0 = [1, 1]$
- Step size:  $\eta = 0.1$
- Threshold:  $\varepsilon = 10^{-6}$

## Problem 4: Normal Distribution MLE (Negative Log-Likelihood)

**Function:**

$$f(\mu, \sigma) = \sum_{i=1}^n \log(\sigma) + \frac{(x_i - \mu)^2}{2\sigma^2}$$

**Note:** Treat  $\mu, \sigma > 0$  as unconstrained variables for gradient descent.

**Details:**

- Data:  $x_1, \dots, x_n$  provided here: [https://docs.google.com/spreadsheets/d/13CmIStaYtiQqR\\_dhBPrkHJINvVln9cepHypNinVQT3c/edit?gid=2023320122#gid=2023320122](https://docs.google.com/spreadsheets/d/13CmIStaYtiQqR_dhBPrkHJINvVln9cepHypNinVQT3c/edit?gid=2023320122#gid=2023320122)
- Initial point:  $\mu_0 = 0, \sigma_0 = 1$
- Step size:  $\eta = 0.01$
- Threshold:  $\varepsilon = 10^{-5}$

## Problem 5: Rosenbrock Function (Non-Convex)

**Function:**

$$f(x, y) = (1 - x)^2 + 100(y - x^2)^2$$

**Gradient:**

$$\nabla f(x, y) = \begin{bmatrix} -2(1 - x) - 400x(y - x^2) \\ 200(y - x^2) \end{bmatrix}$$

**Details:**

- Initial point:  $[x_0, y_0] = [-1, 1]$
- Step size:  $\eta = 0.001$
- Threshold:  $\varepsilon = 10^{-6}$

**Note:** This is a challenging non-convex function with a narrow curved valley.

## Submission

Please prepare and submit

- Your code (in Python or R)
- A short report with:
  - Plots of convergence
  - Final solution and number of iterations
  - Any observations or difficulties encountered