

Assignment: Build a Single Layer Perceptron Using Python Basics

Objective

Build a **Single Layer Perceptron from scratch** using **only Python basics** (no ML libraries) to solve a **binary classification problem** of your own design.

Problem Statement

Create a system that predicts whether a student will **Pass (1)** or **Fail (0)** based on:

- Study hours per day
- Class attendance percentage

 You must **create your own unique dataset**. Using online datasets or ChatGPT-generated data is not allowed.

Input Format

Each data point should be in the form:

[study_hours, attendance_percentage]

Example (format only, do not copy):

[4, 65]

Output

- 1 → Pass
 - 0 → Fail
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Mandatory Rules

Not Allowed

- numpy
- sklearn
- tensorflow / keras/ pytorch
- Copying full code from ChatGPT or classmates

✓ Must Use

- `for` loops
 - `if-else` conditions
 - `lists`
 - `functions`
 - user input (`input()`)
 - meaningful `print()` output
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⚙️ Task Breakdown

Task 1 Dataset Creation

- Create **10–15 data points**
- Store features in a list of lists
- Store labels in a separate list

💡 Labels must be assigned using **your own rule** (clearly explained in the code comment).

Task 2 Perceptron Initialization

Manually define:

- Two weights
- One bias
- Learning rate

You may choose random or fixed values, but you must justify your choice.

Task 3 Activation Function

Write a function that:

- Calculates the weighted sum
 - Applies a threshold of your own choice (step function)
 - Returns `0` or `1`
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Task 4 Training Loop

- Train the perceptron for multiple epochs

- For each data point:
 - Predict the output
 - Calculate the error
 - Update weights and bias

 The **weight update rule must be written and explained by you** (in comments).

Task 5: User Input Testing

After training:

- Take study hours and attendance as user input
- Predict pass/fail using the trained model
- Print a meaningful message
 - Example: *"The student is likely to Pass"*

Short Report (Required)

Students must submit a brief explanation covering:

1. How the dataset was created
2. Why the learning rate was chosen
3. How they verified the model is learning

 Identical datasets or explanations will be treated as plagiarism.

Bonus (Optional)

- Calculate accuracy
- Print training loss per epoch
- Show how weights change over time