# **Deep Learning Course**

Assignment 5

## **MLP Project**

In this assignment, you will be working with two different datasets. Your task is to implement Multilayer Perceptrons (MLPs) using PyTorch to solve a regression problem and a classification problem.

# **Dataset 1: Student Study Performance**

The first dataset is the "Student Study Performance" dataset. This dataset contains information about students' study habits and their performance in school. You can download the dataset from this <u>link</u>.

Your task is to predict students' performance based on their study habits. This is a regression problem.

### **Features**

There are five different features in this dataset:

- **gender**: sex of students -> (Male/female)
- race/ethnicity: ethnicity of students -> (Group A, B, C, D, E)
- parental level of education: parents' final education ->(bachelor's degree,some college,master's degree,associate's degree,- high school)
- lunch: having lunch before the test (standard or free/reduced)
- test preparation course: complete or not complete before the test

There are three target variables your model should predict:

- math score
- reading score
- writing score

#### Tasks

Here are some steps to guide you:

1. **Data Preprocessing**: Load the dataset and perform any necessary preprocessing steps. This may include handling missing values, encoding categorical variables, etc.

- 2. **Model Building**: Implement an MLP using PyTorch. Your network should have at least one hidden layer. You can choose the number of neurons in the hidden layer(s).
- 3. **Training**: Train your model using an appropriate loss function for regression. Monitor the loss function to ensure your model is learning.
- 4. **Evaluation**: Evaluate your model's performance on a separate test set.

### **Dataset 2: Obesity Levels**

The second dataset is the "Obesity Levels" dataset. This dataset contains information about individuals' eating habits and physical condition. You can download the dataset from this <u>link</u>.

Your task is to predict the level of obesity based on the given features. This is a classification problem.

#### **Features**

The dataset contains 16 features, and here are them:

- **Gender**: The gender of the individual.
- Age: The age of the individual.
- **Height**: The height of the individual in meters.
- **Weight**: The weight of the individual in kilograms.
- Family History with Overweight: Whether the individual has a family history of being overweight.
- FAVC (Frequent Consumption of High-Caloric Food): Whether the individual frequently consumes high-caloric food.
- FCVC (Frequency of Consumption of Vegetables): How frequently the individual consumes vegetables.
- NCP (Number of Main Meals): The number of meals the individual has daily.
- CAEC (Consumption of Food Between Meals): How often the individual eats between meals.
- **SMOKE**: Whether the individual smokes.
- CH2O (Consumption of Water Daily): The amount of water the individual consumes daily.
- SCC (Calories Consumption Monitoring): Whether the individual monitors their calorie consumption.
- FAF (Physical Activity Frequency): The frequency of physical activity of the individual.
- TUE (Time Using Technology Devices): The time the individual spends using technology devices.
- CALC (Consumption of Alcohol): The frequency of alcohol consumption by the individual.
- MTRANS (Transportation Used): The mode of transportation the individual uses.

The target variable is **NObeyesdad**, which represents the level of obesity in the individual. It takes seven values: Insufficient Weight, Normal Weight, Overweight Level I, Overweight Level II, Obesity Type I, Obesity Type II, and Obesity Type III.

### **Tasks**

Here are some steps to guide you:

- 1. **Data Preprocessing**: Load the dataset and perform any necessary preprocessing steps. This may include handling missing values, encoding categorical variables, etc.
- 2. **Model Building**: Implement an MLP using PyTorch. Your network should have at least one hidden layer. You can choose the number of neurons in the hidden layer(s).
- 3. **Training**: Train your model using an appropriate loss function for classification. Monitor the loss function to ensure your model is learning.
- 4. **Evaluation**: Evaluate your model's performance on a separate test set. You can use accuracy, precision, recall, F1-score, etc., as your evaluation metrics.

Please submit your Python notebooks with models and a short report describing your approach and results.