# Machine Learning Basics - Set 3a: Implementing a Single Neuron with NumPy

**Dataset:** A, B from previous sets and C: a tricker dataset with two features and points belonging either to class 0 or 1.

#### 1. Understanding Gradient Descent:

- a) Explain the concept of gradient descent in your own words.
- b) Implement a simple cost function for a single neuron:
  - def cost\_function(y\_true, y\_pred):
    # Your implementation here
    - # Hint: Use mean squared error
- c) Derive the analytical formula for the gradient of the cost function with respect to the weights for a single neuron, assuming a sigmoid activation function.

#### 2. Implementing a Single Neuron:

- a) Write a Python function to initialize weights for a single neuron:
  - def initialize\_weights(input\_dim):
    # Your implementation here
    - # Hint: Use np.random.randn()
- b) Implement the forward pass for a single neuron:

```
def forward_pass(X, weights, bias):
 # Your implementation here
 # Hint: Use np.dot() for matrix multiplication
```

c) Implement a simple activation function (the sigmoid as seen during the lectures):

```
def sigmoid(x):
 # Your implementation here
```

### 3. Gradient Descent for a Single Neuron:

- a) Implement a function to calculate the gradients:
  - def calculate\_gradients(X, y, y\_pred, weights):
    # Your implementation here
    # Hint: Use the formula derived in 1c
- b) Write a function to update the weights:

```
def update_weights(weights, gradients, learning_rate):
 # Your implementation here
```

c) Implement a simple training loop:

```
def train_neuron(X, y, learning_rate, num_epochs):
```

- # Your implementation here
- # Use functions from 3a and 3b
- # Return trained weights and cost history

## 4. Training and Evaluation:

- a) Train your single neuron on the provided dataset using the functions you've implemented.
- b) Plot the cost versus the number of epochs to visualize the training progress. Use matplotlib or any other plotting library.
- c) Evaluate the final predictions of your trained neuron on the training data:

def evaluate\_neuron(X, y, trained\_weights, trained\_bias):
 # Your implementation here
 # Calculate and return accuracy

Note: Use NumPy (*import numpy as np*) for all numerical operations. Remember to handle potential numerical instabilities, such as division by zero or log of zero.