

# Sigmoid Perceptron From Scratch Documentation

## Introduction

This document explains the implementation of a Sigmoid Perceptron from scratch using Python and NumPy. The purpose of this project is to understand how a single neuron based classifier works internally without relying on external machine learning libraries.

## Libraries Used

NumPy is used for numerical computations including vector operations and weight updates.

## Concept Overview

A perceptron is a fundamental neural unit that performs binary classification. The sigmoid activation function converts outputs into probability values between zero and one, making it suitable for classification tasks.

## Class Design

The SigmoidPerceptron class is responsible for initializing parameters, performing predictions, and training the model using gradient descent.

## Initialization

Weights are initialized randomly based on the number of input features. A bias term is added to shift the decision boundary.

## Sigmoid Function

The sigmoid function maps the weighted sum of inputs to a probability value which enables smooth learning during training.

## Prediction Process

Prediction is performed by computing a weighted sum of inputs, adding bias, and applying the sigmoid activation function.

## Training Process

The model is trained using supervised learning. For each epoch the model predicts outputs, calculates errors, computes gradients, and updates weights and bias accordingly.

## Error and Gradient

Error is calculated as the difference between target and predicted value. Gradients are derived from the sigmoid function to update parameters efficiently.

## Observed Issue

An error occurs when `input_size` is not passed during model initialization. This value must match the number of features in the dataset.

## Learning Outcomes

This project explains how neurons learn, how sigmoid activation enables probability outputs, and how gradient descent optimizes model parameters.

## Limitations

This implementation supports only linearly separable data and serves as a foundational learning model for deeper neural networks.

## Conclusion

Implementing a sigmoid perceptron from scratch provides a strong understanding of machine learning fundamentals and prepares learners for advanced models.

## Author

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