# BUILDING NEURAL NETWORKS AND CNN'S

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#### 1 Dataset Overview

The dataset consists of 766 samples and 8 features. A brief summary of the features is as follows:

• Number of samples: 766

• Number of features: 8

• Feature types:

- f1, f2, f4, f5, f6, f7: object

- f3, target: integer

• No missing values in the dataset.

#### **Key Statistics:**

f1	float64
f2	float64
f3	int64
f4	float64
f5	float64
f6	float64
f7	float64
target	int64

#### 2 Data Preprocessing

The following preprocessing steps were applied to clean and prepare the data:

- Replaced invalid characters in f1, f2, f4, f5, f6, f7 with None.
- Converted categorical features to numeric using pd.to\_numeric.
- Handled missing values by filling them with the column mean.
- Detected and handled outliers using Interquartile Range (IQR). In particular, outliers in f3 and f4 were handled by replacing them with the mean.

## 3 Data Visualizations

#### 3.1 Box Plot of f3

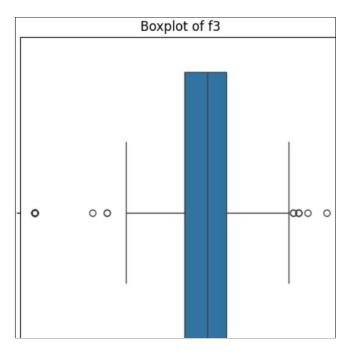


Figure 1: Box plot of feature f3.

This plot shows the distribution and presence of outliers in the f3 feature.

# 3.2 Correlation Heatmap

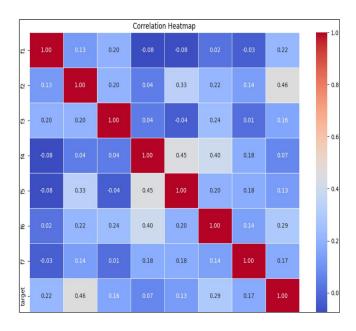


Figure 2: Correlation Heatmap between features.

The heatmap highlights the correlation between different features, with darker colors indicating stronger relationships.

#### 3.3 Histogram of Target Variable

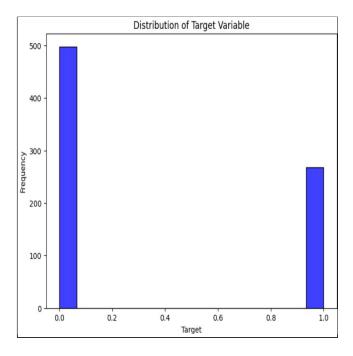


Figure 3: Distribution of the target variable (0/1).

The histogram shows the distribution of the target variable, with values of 0 and 1.

#### 4 Neural Network Architecture

The neural network implemented is a simple feedforward network with the following architecture:

- Input layer: 6 features.
- Two hidden layers, each with 64 neurons, followed by batch normalization and dropout (50%).
- Output layer: Single neuron with sigmoid activation for binary classification.

# 5 Performance Metrics and Analysis

The model was trained for 200 epochs. The final performance metrics on the test set are as follows:

• Test Accuracy: 79.22%

Test Loss: 0.4564Precision: 0.6786

• Recall: 0.7308

• F1 Score: 0.7037

## 5.1 Training, Validation, and Test Loss Over Epochs

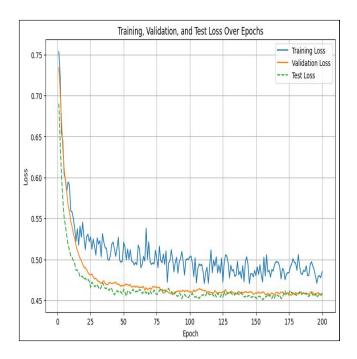


Figure 4: Training, Validation, and Test Loss Over Epochs.

## 5.2 Training, Validation, and Test Accuracy Over Epochs

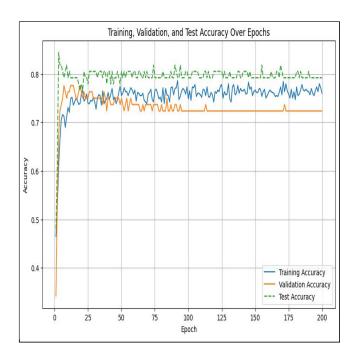


Figure 5: Training, Validation, and Test Accuracy Over Epochs.

## 5.3 Confusion Matrix

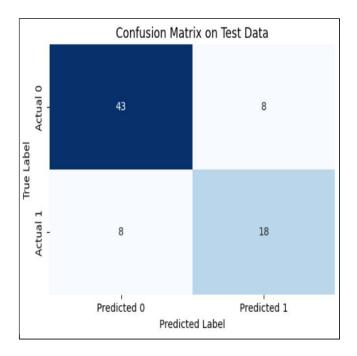


Figure 6: Confusion Matrix for Test Set.

### 5.4 ROC Curve

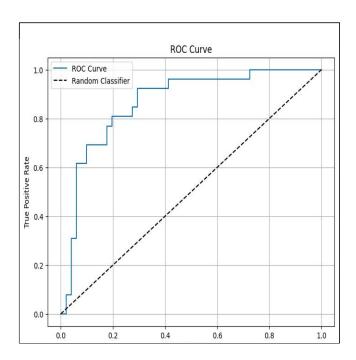


Figure 7: ROC Curve for the model.

The model achieved a balanced performance across precision, recall, and accuracy, making it suitable for binary classification.

# PART 2 REPORT

Table 1: Learning Rate Tuning

LEARNING RATE	TEST ACCURACY
0.0001	80.52
0.001	81.82
0.01	81.82

Table 2: Batch Size Tuning

LEARNING RATE	TEST ACCURACY
16	79.22
32	81.82
64	83.12

Table 3: Hidden Layer Configuration Tuning

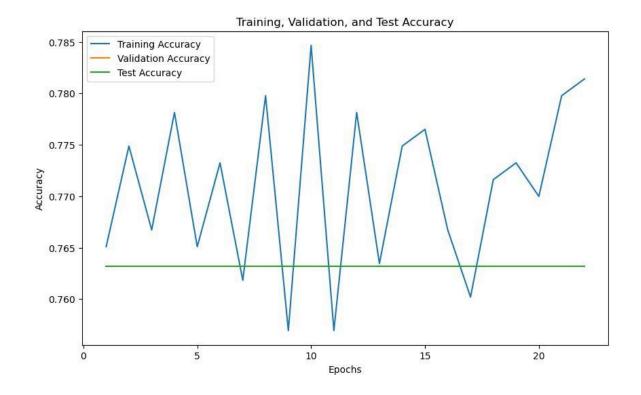
LEARNING RATE	TEST ACCURACY
[64]	85.71
[64, 64]	81.82
[128,64]	83.12

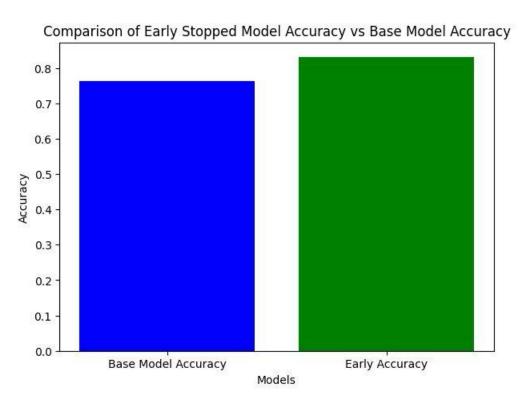
# **Analysis:**

**Learning Rate**: A rate of 0.001 performed best, indicating a balanced update step.

**Batch Size**: A size of 64 provided the best accuracy, possibly due to more stable gradient estimates.

**Hidden Layers**: Configurations with a single layer of size or two layers [64] performed similarly.





# **Methods for Improvement:**

In this I used four different models

- 1. Early Stopping: Prevents overfitting by halting training when validation loss ceases to improve.
- 2. Learning Rate Scheduler: Adjusts the learning rate to enhance convergence.
- 3. Batch Normalization: Stabilizes learning by normalizing inputs to each layer.
- 4. Gradient Accumulation: Simulates larger batch sizes without increasing memory usage.

## **Best Model Description**

Best Model Configuration:

Learning Rate: 0.001

Batch Size: 64

Hidden Layers: [128, 64]

Dropout Rate: Adjusted based on best performance

#### Performance and Analysis:

- The model with a learning rate of 0.001 achieved the highest accuracy of 0.8312.
- Visualizations can include accuracy over epochs to demonstrate improvements with the chosen methods.

#### **CITATIONS:**

Pandas: https://pandas.pydata.org/

Sckit-learn: <a href="https://scikit-learn.org/stable/">https://scikit-learn.org/stable/</a>

Numpy: <a href="https://numpy.org/">https://numpy.org/</a>

# Code of Assignment 1

Team Member	Assignment Part	Contribution
Samyak Shah	Part 1, Part 2	50%
Nischal	Part 1, Part 2	50%
Seemantula		