Alphabet Soup Deep Learning Model Report

Overview

The goal of this project is to develop a deep learning model to predict the success of Alphabet Soup-funded organizations based on historical application data. Using a neural network model (TensorFlow/Keras), the dataset was preprocessed, trained, optimized, and evaluated to achieve the best possible accuracy.

Results

Data Preprocessing

Target Variable:

 \bullet IS_SUCCESSFUL \to Indicates whether an organization was successfully funded.

Feature Variables:

• Application Type, Classification, Affiliation, Organization Type, Ask Amount, and other categorical features converted via **one-hot encoding**.

Removed Columns:

 \bullet EIN and NAME \to These are **identifiers** that do not contribute to predictive power.

Compiling, Training, and Evaluating the Model

Neural Network Structure:

• Input Layer: n input features (based on preprocessed dataset)

Hidden Layers:

- Layer 1: 80 neurons, **ReLU** activation
- Layer 2: 30 neurons, ReLU activation

Output Layer: 1 neuron, **Sigmoid** activation (binary classification)

Training Results:

- Initial Model Accuracy: ~72.99%
- Optimized Model Accuracy: ~72.96%
- Despite multiple adjustments, the model did not exceed the 75% accuracy goal.

Optimization Attempts:

- Added more neurons and layers to extract better features.
- Used Leaky ReLU activation in one model.
- Implemented **dropout layers** to reduce overfitting.
- Adjusted learning rates and batch sizes.
- Feature Selection (RFE) was applied to select the 10 most important features.

Summary

Did the model achieve the required 75% accuracy?

No, despite multiple optimization attempts, the highest accuracy achieved was ~72.99%.

Potential reasons for the accuracy limit:

- The dataset might not contain enough distinct information to separate successful vs. unsuccessful applicants.
- Some features **may not be strong predictors**, or better feature engineering is needed.
 - Deep learning may not be the best model for this classification problem.

Recommendation

To solve this classification problem, a different machine learning model might work better than a deep learning model:

Random Forest or XGBoost

• These models performed **similarly to the neural network** and require less tuning.

While deep learning provided **moderate accuracy**, trying **Random Forest or XGBoost** in a real-world scenario might lead to a better result **with less computational cost**.