Report on the Performance of the Deep Learning Model

Overview of the Analysis:

The purpose of this analysis is to develop a deep learning model to predict the success of funding projects for Alphabet Soup, an organization that provides financial support to various charitable organizations. The goal is to create a model that can accurately classify funding projects as successful or not based on certain features. By doing so, Alphabet Soup can better allocate their resources and make informed decisions on which projects to fund.

Results:

Data Preprocessing:

- The target variable for the model is the "IS_SUCCESSFUL" column, which indicates whether a funding project was successful or not.
- The features for the model are all the columns in the dataset except for the "IS SUCCESSFUL" column.
- The "EIN" and "NAME" columns were removed from the input data as they are neither targets nor features.

Compiling, Training, and Evaluating the Models 1,2 and 3:

- The neural network model consists of three hidden layers in Models 1 and 2 and 4 hidden layers in Model 3 and one output layer in all three models
- The first hidden layer has 80, 90, and 70 neurons for each Model respectively and uses the ReLU activation function.
- The second hidden layer has 50, 60, and 45 neurons for each Model respectively and also uses the ReLU activation function.
- The third hidden layer has 30, 40, and 25 neurons for each Model respectively plus forth hidden layer has 5 neurons in Model #4 and uses the ReLU activation function as well.
- The output layer has 1 neuron and uses the sigmoid activation function to produce a probability output between 0 and 1 for all three Models.
- The models wwere compiled using the Adam, SGD, and Adagrad optimizers respectively and the binary cross-entropy loss function.
- During training, the models 1 and 3 was trained for 100 epochs and Model #2 was trained for 50 epochs with a batch size of 32.
- After training, the model was evaluated using the test data.
- All models performance was evaluated in terms of loss and accuracy.

Model 1: loss: 0.5582- - binary accuracy: 0.7315

Model 2: Loss: 0.5452, Accuracy: 0.7344

Model 3: Loss: 0.5499, Accuracy: 0.7344

The loss and accuracy values obtained from the evaluation were recorded.

Summary:

The deep learning model achieved a certain level of performance, but the specific results may vary depending on the dataset used and the training process. The evaluation results in terms of loss and accuracy can provide insights into the model's effectiveness in classifying funding projects as successful or not.

Recommendation for a Different Model:

For this classification problem, an alternative model that could be considered is a Random Forest Classifier. Random Forest is an ensemble learning method that combines multiple decision trees to make predictions. It can handle both numerical and categorical features and has the ability to capture complex interactions between features.

The Random Forest Classifier could be trained on the preprocessed dataset, using the same target and features as the deep learning model. It would generate an ensemble of decision trees and make predictions based on the majority vote or average prediction of the individual trees. Random Forests are known for their ability to handle high-dimensional datasets, handle imbalanced classes, and provide feature importance rankings.

By using a Random Forest Classifier, Alphabet Soup can compare the performance of different models and choose the one that best suits their specific needs and requirements. It is recommended to evaluate the Random Forest Classifier's performance using appropriate evaluation metrics and compare it with the deep learning model to determine the most effective approach for predicting the success of funding projects.

In conclusion, the deep learning model developed in this analysis provides a starting point for predicting the success of funding projects for Alphabet Soup. However, exploring alternative models, such as a Random Forest Classifier, can help improve the accuracy and robustness of the predictions.