

Anderson Igben

Rice Data Analytics and Visualization

## Charity Funding Predictor

In the first model EIN and NAME columns were dropped from the dataset as a preprocessing step. Application Type and Classification was then analyzed for binning before the model was created. These were the targets while the other columns were the features. Once a split, training and testing dataset were created I went ahead and compiled train and evaluate the model. There were two layers and one output layer. The picture below visualizes what was done in code

### Compile, Train and Evaluate the Model

```
In [25]: # Define the model - deep neural net, i.e., the number of input features and
number_input_features = len( X_train_scaled[0])
hidden_nodes_layer1=90
hidden_nodes_layer2=50

nn = tf.keras.models.Sequential()

# First hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer1, input_dim=number_input_features, activation='relu'))

# Second hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer2, activation='relu'))

# Output layer
nn.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))

# Check the structure of the model
nn.summary()
```

Model: "sequential\_5"

Layer (type)	Output Shape	Param #
dense_13 (Dense)	(None, 90)	4050
dense_14 (Dense)	(None, 50)	4550
dense_15 (Dense)	(None, 1)	51

Total params: 8,651  
Trainable params: 8,651  
Non-trainable params: 0

Once completed the epochs was set to 100 and I went ahead to train the model. My final evaluation was set to an accuracy of Accuracy: 0.7269970774650574. On the other hand, to better improve this model and increase the accuracy score, in the preprocessing steps I dropped only the EIN columns and then evaluated the NAME columns. Columns for binning remained the same however I then added another layer to the model and kept the number of nodes the same. The picture below depicts changes made.

```
In [19]: # Define the model - deep neural net, i.e., the number of input features and
number_input_features = len( X_train_scaled[0])
hidden_nodes_layer1=90
hidden_nodes_layer2=50
hidden_nodes_layer3=25

nn = tf.keras.models.Sequential()

# First hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer1, input_dim=number_input_features, activation='relu'))

# Second hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer2, activation='relu'))

# Added Third hidden layer
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer3, activation='relu'))

# Output layer
nn.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))

# Check the structure of the model
nn.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 90)	40230
dense_1 (Dense)	(None, 50)	4550
dense_2 (Dense)	(None, 25)	1275
dense_3 (Dense)	(None, 1)	26

=====  
Total params: 46,081  
Trainable params: 46,081  
Non-trainable params: 0  
=====

After proceeding to compile the model and train it as well, I increased the accuracy to an impressive 0.7905539274215698 which was better than the initial model and even more than the required limit of 75%.