

## Alphabet Soup Report

### *Purpose*

This project helps the non-profit company Alphabet soup determine which of the loan applications are most likely to succeed at their ventures. With these results, Alphabet soup can be sure that its efforts have the biggest impact.

### Results

#### Variables

**IS\_SUCCESSFUL**—Was the money used effectively

The following parameters are the targets for the model:

The following parameters are the features of the model:

- **APPLICATION\_TYPE**—Alphabet Soup application type
- **AFFILIATION**—Affiliated sector of industry
- **CLASSIFICATION**—Government organization classification
- **USE\_CASE**—Use case for funding
- **ORGANIZATION**—Organization type
- **STATUS**—Active status
- **INCOME\_AMT**—Income classification
- **SPECIAL\_CONSIDERATIONS**—Special considerations for application
- **ASK\_AMT**—Funding amount requested
- 

The following parameters were removed because they are neither targets nor features  
**EIN** and **NAME**—Identification columns

#### Training

In creating the optimized model I decided to go with the following parameters

- 5 hidden layers (1 with 100 neurons, 3 with 50 neurons, and 1 with 30 neurons)
- Tanh activation functions throughout

I decided to go with this structure because I assumed that more hidden layers, and more neurons per layer would lead to more accurate models. I picked the tanh activation function before and have worked with sigmoid and relu before, so I wanted to try out the tanh function.

My assumptions about how to build a better model were incorrect because the optimized model had an accuracy of 78.5% while the original model had an accuracy of 72.9%. This means the optimized model performed worse than the original.

```
Epoch 16/17: 464/step - accuracy: 0.7358 - loss: 0.5426
Epoch 17/17: 464/step - accuracy: 0.7359 - loss: 0.5393
Epoch 18/17: 464/step - accuracy: 0.7452 - loss: 0.5364
Epoch 19/17: 464/step - accuracy: 0.7454 - loss: 0.5370
Epoch 20/17: 464/step - accuracy: 0.7429 - loss: 0.5385
Epoch 21/17: 464/step - accuracy: 0.7399 - loss: 0.5400
Epoch 22/17: 464/step - accuracy: 0.7402 - loss: 0.5380
Epoch 23/17: 464/step - accuracy: 0.7420 - loss: 0.5359
Epoch 24/17: 464/step - accuracy: 0.7390 - loss: 0.5403
Epoch 25/17: 464/step - accuracy: 0.7453 - loss: 0.5354
Epoch 26/17: 464/step - accuracy: 0.7408 - loss: 0.5369
Epoch 27/17: 464/step - accuracy: 0.7415 - loss: 0.5375
Epoch 28/17: 464/step - accuracy: 0.7399 - loss: 0.5389
Epoch 29/17: 464/step - accuracy: 0.7416 - loss: 0.5361
Epoch 30/17: 464/step - accuracy: 0.7395 - loss: 0.5275
```

```
Epoch 30/30: 464/step - accuracy: 0.7417 - loss: 0.5364
Epoch 31/30: 464/step - accuracy: 0.7462 - loss: 0.5323

[1]: Evaluate the model using the test data
model_loss, model_accuracy = m.evaluate(x_test_scaled, y_test, verbose=2)
print('Test: model_loss, accuracy: (model_accuracy)')

Test: 0.5381277333677909, Accuracy: 0.73787751548016

[2]: Export our model to H5 file
m.save('AlphabetSoupCharity_Optimization_Turner.h5')

WARNING:absl:You are saving your model as an H5 file ('model_save1'). We recommend saving your model as a Keras model ('model').
```

## Alterations

I also tried universal sigmoid activation functions throughout the model (with the same number of layers and neurons), but the accuracy was significantly worse, so the details are not reported.

## Summary

The potential for neural networks to help in this situation is evident. With enough tuning, Alphabet Soup can be confident that their loans will impact the world. However, I would recommend further refinements to the network before rolling it out to the field. Since neither model met the accuracy threshold of 75%, Alphabet Soup may find the current model ineffective. Therefore, I would wait until we clear the accuracy threshold before using the program.