

▼ Deliverable 1: Preprocessing the Data for a Neural Network

```
# Import our dependencies
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, OneHotEncoder
import pandas as pd
import tensorflow as tf
```

```
# Import and read the charity_data.csv.
import pandas as pd
application_df = pd.read_csv("/content/charity_data.csv")
application_df.head()
```



```
-----
FileNotFoundError                                Traceback (most recent call last)
<ipython-input-1-1d51bf91914b> in <module>
      7 # Import and read the charity_data.csv.
      8 import pandas as pd
----> 9 application_df = pd.read_csv("/content/charity_data.csv")
     10 application_df.head()
```

7 frames

```
/usr/local/lib/python3.7/dist-packages/pandas/io/common.py in
get_handle(path_or_buf, mode, encoding, compression, memory_map, is_text,
errors, storage_options)
    705         encoding=ioargs.encoding,
    706         errors=errors,
--> 707         newline="",
    708     )
    709     else:
```

```
FileNotFoundError: [Errno 2] No such file or directory:
'/content/charity_data.csv'
```

```
# from google.colab import drive
# drive.mount('/content/drive')
```

```
application_df.columns
```

```
Index(['EIN', 'NAME', 'APPLICATION_TYPE', 'AFFILIATION', 'CLASSIFICATION',
      'USE_CASE', 'ORGANIZATION', 'STATUS', 'INCOME_AMT',
      'SPECIAL_CONSIDERATIONS', 'ASK_AMT', 'IS_SUCCESSFUL'],
      dtype='object')
```

```
application_df.dtypes
```

```
EIN
```

```
int64
```

```
NAME object
APPLICATION_TYPE object
AFFILIATION object
CLASSIFICATION object
USE_CASE object
ORGANIZATION object
STATUS int64
INCOME_AMT object
SPECIAL_CONSIDERATIONS object
ASK_AMT int64
IS_SUCCESSFUL int64
dtype: object
```

```
# Drop the non-beneficial ID columns, 'EIN' and 'NAME'.
# YOUR CODE GOES HERE
application_df = application_df.drop(["EIN"], axis=1)

# Print out the Country value counts
application_df.head()
```

	NAME	APPLICATION_TYPE	AFFILIATION	CLASSIFICATION	USE_CASE	C
0	BLUE KNIGHTS MOTORCYCLE CLUB	T10	Independent	C1000	ProductDev	
1	AMERICAN CHESAPEAKE CLUB CHARITABLE TR	T3	Independent	C2000	Preservation	
2	ST CLOUD PROFESSIONAL FIREFIGHTERS	T5	CompanySponsored	C3000	ProductDev	
3	SOUTHSIDE ATHLETIC ASSOCIATION	T3	CompanySponsored	C2000	Preservation	
4	GENETIC RESEARCH INSTITUTE OF THE DESERT	T3	Independent	C1000	Heathcare	

```
application_counts = application_df['APPLICATION_TYPE'].value_counts()
application_counts
```

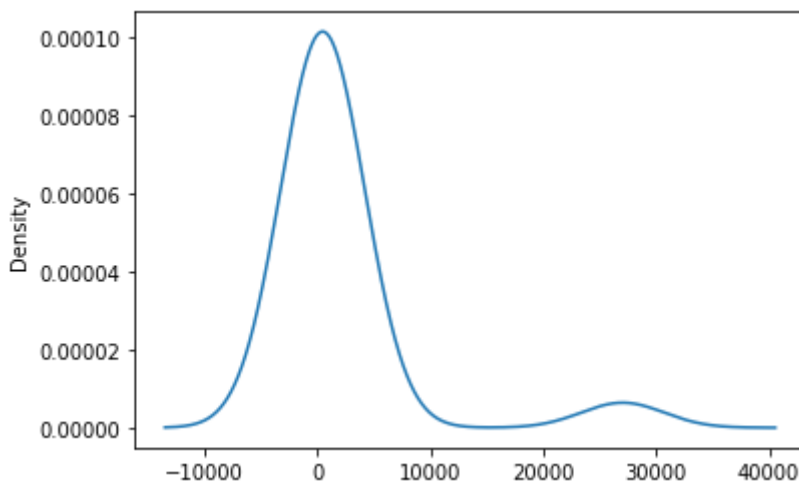
```
T3    27037
T4     1542
T6     1216
T5     1173
```

T19	1065
T8	737
T7	725
T10	528
T9	156
T13	66
T12	27
T2	16
T25	3
T14	3
T29	2
T15	2
T17	1

Name: APPLICATION_TYPE, dtype: int64

```
application_counts.plot.density()
```

<matplotlib.axes._subplots.AxesSubplot at 0x7fde48054890>



```
# Determine which values to replace if counts are less than ...?
```

```
replace_application = list(application_counts[application_counts<500].index)
```

```
# Replace in dataframe
```

```
for app in replace_application:
```

```
    application_df.APPLICATION_TYPE = application_df.APPLICATION_TYPE.replace(app, "Other")
```

```
# Check to make sure binning was successful
```

```
application_df.APPLICATION_TYPE.value_counts()
```

T3	27037
T4	1542
T6	1216
T5	1173
T19	1065
T8	737
T7	725
T10	528
Other	276

Name: APPLICATION_TYPE, dtype: int64

```
# Look at CLASSIFICATION value counts for binning
# YOUR CODE GOES HERE
class_counts = application_df.CLASSIFICATION.value_counts()
class_counts
```

```
C1000    17326
C2000     6074
C1200     4837
C3000     1918
C2100     1883
```

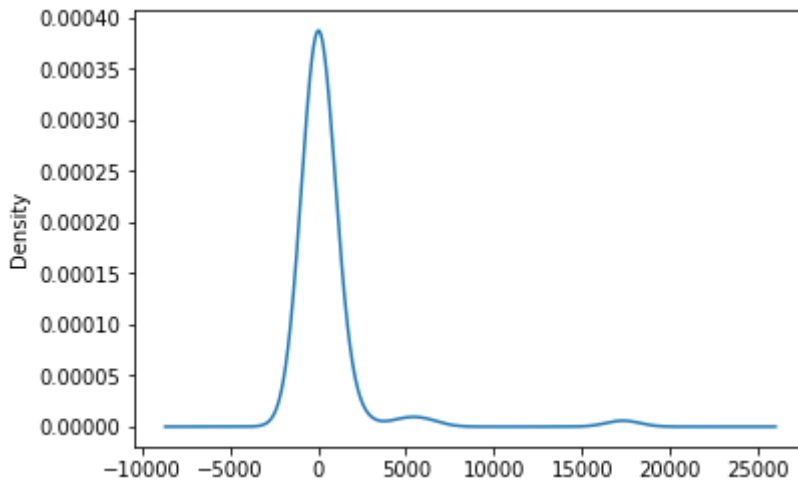
```
...
```

```
C4120         1
C8210         1
C2561         1
C4500         1
C2150         1
```

```
Name: CLASSIFICATION, Length: 71, dtype: int64
```

```
# Visualize the value counts of CLASSIFICATION
# YOUR CODE GOES HERE
class_counts.plot.density()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7fde47f7a4d0>
```



```
# Determine which values to replace if counts are less than ..?
# YOUR CODE GOES HERE
replace_class = list(class_counts[class_counts < 1000].index)
```

```
# Replace in dataframe
for cls in replace_class:
    application_df.CLASSIFICATION = application_df.CLASSIFICATION.replace(cls, "Other")
```

```
# Check to make sure binning was successful
application_df.CLASSIFICATION.value_counts()
```

```

C1000    17326
C2000     6074
C1200     4837
Other     2261
C3000     1918
C2100     1883
Name: CLASSIFICATION, dtype: int64

```

```
application_df.dtypes
```

```

NAME                object
APPLICATION_TYPE     object
AFFILIATION         object
CLASSIFICATION       object
USE_CASE            object
ORGANIZATION        object
STATUS              int64
INCOME_AMT          object
SPECIAL_CONSIDERATIONS object
ASK_AMT             int64
IS_SUCCESSFUL       int64
dtype: object

```

```

application_cat=application_df.dtypes[application_df.dtypes == "object"].index.tolist()
application_cat

```

```

['NAME',
 'APPLICATION_TYPE',
 'AFFILIATION',
 'CLASSIFICATION',
 'USE_CASE',
 'ORGANIZATION',
 'INCOME_AMT',
 'SPECIAL_CONSIDERATIONS']

```

```

# Create a OneHotEncoder instance
enc = OneHotEncoder(sparse=False)

```

```

# Fit and transform the OneHotEncoder using the categorical variable list
# YOUR CODE GOES HERE
encode_df = pd.DataFrame(enc.fit_transform(application_df[application_cat]))
# Add the encoded variable names to the dataframe
encode_df.columns = enc.get_feature_names(application_cat)
encode_df.head()

```

```
/usr/local/lib/python3.7/dist-packages/sklearn/utils/deprecation.py:87: FutureWarning:
warnings.warn(msg, category=FutureWarning)
```

	NAME_1 DAY RANCH RESCUE AND RURAL OKLAHOMA ANIMAL RESOURCE INC	NAME_100 BLACK MEN OF AMERICA	NAME_100 BLACK MEN OF MEMPHIS INC	NAME_100 BLACK MEN OF WEST GEORGIA INC	NAME_1150 WEBSTER STREET INC	NAME_116TH CAVALRY REGIMENT CHAPTER OF THE US CAVALRY & ARMOR ASSOCIATION	NAME_13TH BOMB SQUADRON ASSOCIATION	
0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
2	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
-	- -	- -	- -	- -	- -	- -	- -	

```
# Merge one-hot encoded features and drop the originals
application_df=application_df.merge(encode_df,left_index=True,right_index=True)
application_df=application_df.drop(application_cat,1)
application_df.head()
```

NameError

Traceback (most recent call last)

<ipython-input-1-acb515938c55> in <module>

1 # Merge one-hot encoded features and drop the originals

----> 2

application_df=application_df.merge(encode_df,left_index=True,right_index=True)

3 application_df=application_df.drop(application_cat,1)

4 application_df.head()

NameError: name 'application_df' is not defined

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```
# Split our preprocessed data into our features and target arrays
y=application_df['IS_SUCCESSFUL'].values
X=application_df.drop(['IS_SUCCESSFUL'], 1).values

# Split the preprocessed data into a training and testing dataset
X_train, X_test, y_train, y_test=train_test_split(X,y, random_state=42)
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:3: FutureWarning: In
This is separate from the ipykernel package so we can avoid doing imports until
```

```
# Create a StandardScaler instances
scaler = StandardScaler()
```

```
# Fit the StandardScaler
X_scaler = scaler.fit(X_train)

# Scale the data
X_train_scaled = X_scaler.transform(X_train)
X_test_scaled = X_scaler.transform(X_test)
```

▼ Deliverable 2: Compile, Train and Evaluate the Model

```
# Define the model - deep neural net, i.e., the number of input features and hidden nodes
# YOUR CODE GOES HERE

number_input_features = len(X_train[0])
hidden_nodes_layer1 = 8
hidden_nodes_layer2 = 5
nn = tf.keras.models.Sequential()

# First hidden layer
# YOUR CODE GOES HERE
nn.add(
    tf.keras.layers.Dense(units=hidden_nodes_layer1, input_dim=number_input_features,
    )
# Second hidden layer
# YOUR CODE GOES HERE
nn.add(tf.keras.layers.Dense(units=hidden_nodes_layer2, activation="relu"))

# Output layer
# YOUR CODE GOES HERE
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, 8)	352
dense_1 (Dense)	(None, 5)	45
dense_2 (Dense)	(None, 1)	6

```
=====
Total params: 403
Trainable params: 403
Non-trainable params: 0
=====
```

```

# Compile the model
# YOUR CODE GOES HERE
nn.compile(loss="binary_crossentropy", optimizer="adam", metrics=["accuracy"])

# Train the model
# YOUR CODE GOES HERE
fit_model = nn.fit(X_train,y_train,epochs=100)

Epoch 1/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 2/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 3/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 4/200
804/804 [=====] - 1s 2ms/step - loss: 0.6912 - accur
Epoch 5/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 6/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 7/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 8/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 9/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 10/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 11/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 12/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 13/200
804/804 [=====] - 1s 2ms/step - loss: 0.6912 - accur
Epoch 14/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 15/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 16/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 17/200
804/804 [=====] - 1s 2ms/step - loss: 0.6912 - accur
Epoch 18/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 19/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 20/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 21/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 22/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 23/200
804/804 [=====] - 2s 2ms/step - loss: 0.6911 - accur
Epoch 24/200

```



```

804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 25/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 26/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 27/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur
Epoch 28/200
804/804 [=====] - 2s 2ms/step - loss: 0.6912 - accur

```

```
# Evaluate the model using the test data
```

```

model_loss, model_accuracy = nn.evaluate(X_test_scaled,y_test,verbose=2)
print(f"Loss: {model_loss}, Accuracy: {model_accuracy}")

```

```

268/268 - 0s - loss: 1.2187 - accuracy: 0.5345 - 500ms/epoch - 2ms/step
Loss: 1.2186620235443115, Accuracy: 0.5344606637954712

```

```
nn.save("AlphabetSoupCharity.h5")
```

```

-----
NameError                                Traceback (most recent call last)
<ipython-input-1-3f97cdb2be7e> in <module>
----> 1 nn.save("AlphabetSoupCharity.h5")

```

```
NameError: name 'nn' is not defined
```

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