# **Baseline Datasets**

#### **Load Datasets**

```
import random
import numpy as np
import pandas as pd
from tensorflow.random import set_seed
random.seed(2024)
np.random.seed(2024)
set_seed(2024)

df_train = pd.read_csv("train.csv")
df_test = pd.read_csv("test.csv")
X_train = df_train.drop(columns = ["converter"])
Y_train = df_train.converter
X_test = df_test.drop(columns = ["converter"])
Y_test = df_test["converter"]
```

WARNING:tensorflow:From C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages \keras\src\losses.py:2976: The name tf.losses.sparse\_softmax\_cross\_entropy is deprecated. Please use tf.compat.v1.losses.sparse\_softmax\_cross\_entropy instead.

## **Normalization**

- 1. scale numerical columns
- 2. apply one-hot encoding to categorical columns

```
In [2]: from sklearn.preprocessing import StandardScaler
    from sklearn.preprocessing import OneHotEncoder

numerical_cols = X_train.select_dtypes(include=['float64', 'int64']).columns.tolist
    categorical_cols = X_train.select_dtypes(include=['object']).columns.tolist()

# Initialize the StandardScaler and OneHotEncoder
scaler = StandardScaler()
encoder = OneHotEncoder(sparse=False)

# scale
X_train_scaled = scaler.fit_transform(X_train[numerical_cols])
X_test_scaled = scaler.transform(X_test[numerical_cols])

# encode
X_train_encoded = encoder.fit_transform(X_train[categorical_cols])
X_test_encoded = encoder.transform(X_test[categorical_cols])
# concatenate
```

```
X_train_preprocessed = np.concatenate([X_train_scaled, X_train_encoded], axis=1)
X_test_preprocessed = np.concatenate([X_test_scaled, X_test_encoded], axis=1)
```

C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages\sklearn\preprocessing\\_
encoders.py:868: FutureWarning: `sparse` was renamed to `sparse\_output` in version
1.2 and will be removed in 1.4. `sparse\_output` is ignored unless you leave `spars
e` to its default value.
 warnings.warn(

#### **Neural Network**

```
In [3]: from tensorflow.keras.models import Sequential
    from tensorflow.keras.layers import Dense

NNmodel = Sequential()
    NNmodel.add(Dense(8, input_dim = len(X_train_preprocessed[0,:]), activation = "relu
    NNmodel.add(Dense(1, activation = "sigmoid")) # binary
    NNmodel.summary()
```

WARNING:tensorflow:From C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages \keras\src\backend.py:873: The name tf.get\_default\_graph is deprecated. Please use tf.compat.v1.get\_default\_graph instead.

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 8)	184
dense_1 (Dense)	(None, 1)	9

Total params: 193 (772.00 Byte)
Trainable params: 193 (772.00 Byte)
Non-trainable params: 0 (0.00 Byte)

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```
In [4]: NNmodel.compile(loss = "binary_crossentropy", optimizer = "adam", metrics = ["accur
history = NNmodel.fit(x = X_train_preprocessed, y = Y_train, epochs = 200, validati
```

WARNING:tensorflow:From C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages \keras\src\optimizers\\_\_init\_\_.py:309: The name tf.train.Optimizer is deprecated. Please use tf.compat.v1.train.Optimizer instead.

#### Epoch 1/200

WARNING:tensorflow:From C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages \keras\src\utils\tf\_utils.py:492: The name tf.ragged.RaggedTensorValue is deprecat ed. Please use tf.compat.v1.ragged.RaggedTensorValue instead.

WARNING:tensorflow:From C:\Users\43115\anaconda3\envs\python\_310\lib\site-packages \keras\src\engine\base\_layer\_utils.py:384: The name tf.executing\_eagerly\_outside\_f unctions is deprecated. Please use tf.compat.v1.executing\_eagerly\_outside\_function s instead.

```
20/20 [============== ] - 1s 9ms/step - loss: 0.8490 - accuracy: 0.
3997 - val_loss: 0.8206 - val_accuracy: 0.4551
Epoch 2/200
20/20 [============ ] - 0s 2ms/step - loss: 0.7783 - accuracy: 0.
4543 - val_loss: 0.7597 - val_accuracy: 0.4808
Epoch 3/200
20/20 [================ ] - 0s 3ms/step - loss: 0.7223 - accuracy: 0.
5329 - val_loss: 0.7137 - val_accuracy: 0.5321
Epoch 4/200
20/20 [============] - 0s 3ms/step - loss: 0.6791 - accuracy: 0.
6228 - val_loss: 0.6750 - val_accuracy: 0.6154
Epoch 5/200
20/20 [================ ] - 0s 3ms/step - loss: 0.6423 - accuracy: 0.
6838 - val_loss: 0.6435 - val_accuracy: 0.6603
Epoch 6/200
20/20 [=============== ] - 0s 3ms/step - loss: 0.6072 - accuracy: 0.
7319 - val loss: 0.6161 - val accuracy: 0.7115
20/20 [================ ] - 0s 3ms/step - loss: 0.5752 - accuracy: 0.
7464 - val_loss: 0.5893 - val_accuracy: 0.7244
Epoch 8/200
20/20 [================ ] - 0s 2ms/step - loss: 0.5457 - accuracy: 0.
7673 - val_loss: 0.5672 - val_accuracy: 0.7500
Epoch 9/200
20/20 [================ ] - 0s 2ms/step - loss: 0.5208 - accuracy: 0.
7705 - val_loss: 0.5489 - val_accuracy: 0.7372
Epoch 10/200
20/20 [============] - 0s 2ms/step - loss: 0.5006 - accuracy: 0.
7753 - val_loss: 0.5335 - val_accuracy: 0.7500
Epoch 11/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4843 - accuracy: 0.
7785 - val_loss: 0.5219 - val_accuracy: 0.7436
Epoch 12/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4718 - accuracy: 0.
7769 - val_loss: 0.5128 - val_accuracy: 0.7436
20/20 [================ ] - 0s 3ms/step - loss: 0.4623 - accuracy: 0.
7817 - val_loss: 0.5066 - val_accuracy: 0.7500
Epoch 14/200
20/20 [================= ] - 0s 2ms/step - loss: 0.4559 - accuracy: 0.
7817 - val_loss: 0.5006 - val_accuracy: 0.7500
```

Epoch 15/200

```
20/20 [================ ] - 0s 2ms/step - loss: 0.4497 - accuracy: 0.
7881 - val_loss: 0.4970 - val_accuracy: 0.7564
Epoch 16/200
20/20 [============] - 0s 2ms/step - loss: 0.4454 - accuracy: 0.
7913 - val_loss: 0.4933 - val_accuracy: 0.7564
Epoch 17/200
20/20 [================ ] - 0s 3ms/step - loss: 0.4416 - accuracy: 0.
7913 - val_loss: 0.4914 - val_accuracy: 0.7628
Epoch 18/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4390 - accuracy: 0.
7929 - val_loss: 0.4898 - val_accuracy: 0.7500
Epoch 19/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4363 - accuracy: 0.
7945 - val_loss: 0.4883 - val_accuracy: 0.7564
Epoch 20/200
20/20 [============== ] - 0s 2ms/step - loss: 0.4340 - accuracy: 0.
7961 - val_loss: 0.4866 - val_accuracy: 0.7500
Epoch 21/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4326 - accuracy: 0.
7945 - val_loss: 0.4855 - val_accuracy: 0.7564
Epoch 22/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4309 - accuracy: 0.
7978 - val_loss: 0.4842 - val_accuracy: 0.7564
Epoch 23/200
20/20 [============== ] - 0s 3ms/step - loss: 0.4296 - accuracy: 0.
7978 - val_loss: 0.4834 - val_accuracy: 0.7564
Epoch 24/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.4280 - accuracy: 0.
7994 - val_loss: 0.4826 - val_accuracy: 0.7564
Epoch 25/200
20/20 [============== ] - 0s 3ms/step - loss: 0.4267 - accuracy: 0.
7994 - val_loss: 0.4816 - val_accuracy: 0.7564
Epoch 26/200
20/20 [=========== ] - 0s 3ms/step - loss: 0.4252 - accuracy: 0.
8010 - val_loss: 0.4808 - val_accuracy: 0.7564
Epoch 27/200
20/20 [==============] - 0s 3ms/step - loss: 0.4253 - accuracy: 0.
8058 - val_loss: 0.4801 - val_accuracy: 0.7564
Epoch 28/200
20/20 [================ ] - 0s 3ms/step - loss: 0.4233 - accuracy: 0.
7994 - val_loss: 0.4801 - val_accuracy: 0.7500
Epoch 29/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4218 - accuracy: 0.
8042 - val_loss: 0.4792 - val_accuracy: 0.7564
Epoch 30/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4207 - accuracy: 0.
8074 - val_loss: 0.4790 - val_accuracy: 0.7628
Epoch 31/200
20/20 [================= ] - 0s 2ms/step - loss: 0.4198 - accuracy: 0.
8090 - val loss: 0.4791 - val accuracy: 0.7564
Epoch 32/200
20/20 [===========] - 0s 2ms/step - loss: 0.4190 - accuracy: 0.
8106 - val_loss: 0.4790 - val_accuracy: 0.7564
Epoch 33/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4185 - accuracy: 0.
8074 - val_loss: 0.4785 - val_accuracy: 0.7628
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Epoch 34/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4170 - accuracy: 0.
8074 - val_loss: 0.4782 - val_accuracy: 0.7628
Epoch 35/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4163 - accuracy: 0.
8122 - val_loss: 0.4779 - val_accuracy: 0.7628
Epoch 36/200
20/20 [================== ] - 0s 2ms/step - loss: 0.4152 - accuracy: 0.
8138 - val loss: 0.4777 - val accuracy: 0.7628
Epoch 37/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4143 - accuracy: 0.
8154 - val_loss: 0.4774 - val_accuracy: 0.7628
Epoch 38/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4135 - accuracy: 0.
8218 - val_loss: 0.4777 - val_accuracy: 0.7628
Epoch 39/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4127 - accuracy: 0.
8202 - val_loss: 0.4769 - val_accuracy: 0.7628
Epoch 40/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4119 - accuracy: 0.
8186 - val_loss: 0.4766 - val_accuracy: 0.7628
Epoch 41/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4111 - accuracy: 0.
8186 - val_loss: 0.4760 - val_accuracy: 0.7628
Epoch 42/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4101 - accuracy: 0.
8218 - val_loss: 0.4765 - val_accuracy: 0.7628
Epoch 43/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4096 - accuracy: 0.
8186 - val_loss: 0.4767 - val_accuracy: 0.7628
Epoch 44/200
20/20 [============== ] - 0s 2ms/step - loss: 0.4087 - accuracy: 0.
8186 - val_loss: 0.4760 - val_accuracy: 0.7692
Epoch 45/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4081 - accuracy: 0.
8202 - val_loss: 0.4763 - val_accuracy: 0.7692
Epoch 46/200
20/20 [============== ] - 0s 2ms/step - loss: 0.4072 - accuracy: 0.
8218 - val_loss: 0.4763 - val_accuracy: 0.7628
Epoch 47/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4062 - accuracy: 0.
8202 - val_loss: 0.4763 - val_accuracy: 0.7692
Epoch 48/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4057 - accuracy: 0.
8170 - val_loss: 0.4759 - val_accuracy: 0.7756
Epoch 49/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4052 - accuracy: 0.
8186 - val_loss: 0.4755 - val_accuracy: 0.7692
Epoch 50/200
20/20 [=================== ] - 0s 2ms/step - loss: 0.4046 - accuracy: 0.
8170 - val_loss: 0.4757 - val_accuracy: 0.7692
Epoch 51/200
20/20 [============ ] - 0s 2ms/step - loss: 0.4040 - accuracy: 0.
8186 - val_loss: 0.4766 - val_accuracy: 0.7692
20/20 [================== ] - 0s 2ms/step - loss: 0.4031 - accuracy: 0.
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8202 - val_loss: 0.4762 - val_accuracy: 0.7692
Epoch 53/200
20/20 [================ ] - 0s 2ms/step - loss: 0.4023 - accuracy: 0.
8186 - val_loss: 0.4760 - val_accuracy: 0.7628
Epoch 54/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.4017 - accuracy: 0.
8186 - val_loss: 0.4764 - val_accuracy: 0.7628
Epoch 55/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.4014 - accuracy: 0.
8202 - val_loss: 0.4757 - val_accuracy: 0.7628
Epoch 56/200
20/20 [============] - 0s 2ms/step - loss: 0.4004 - accuracy: 0.
8202 - val_loss: 0.4758 - val_accuracy: 0.7628
20/20 [============== ] - 0s 2ms/step - loss: 0.3999 - accuracy: 0.
8186 - val_loss: 0.4755 - val_accuracy: 0.7628
Epoch 58/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3996 - accuracy: 0.
8170 - val_loss: 0.4760 - val_accuracy: 0.7628
Epoch 59/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3992 - accuracy: 0.
8186 - val_loss: 0.4749 - val_accuracy: 0.7628
Epoch 60/200
20/20 [============] - 0s 2ms/step - loss: 0.3986 - accuracy: 0.
8186 - val_loss: 0.4757 - val_accuracy: 0.7628
Epoch 61/200
20/20 [==============] - 0s 2ms/step - loss: 0.3979 - accuracy: 0.
8218 - val_loss: 0.4759 - val_accuracy: 0.7628
Epoch 62/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3973 - accuracy: 0.
8170 - val loss: 0.4759 - val accuracy: 0.7692
20/20 [============ ] - 0s 2ms/step - loss: 0.3965 - accuracy: 0.
8154 - val_loss: 0.4758 - val_accuracy: 0.7692
Epoch 64/200
20/20 [=========== ] - 0s 3ms/step - loss: 0.3962 - accuracy: 0.
8186 - val_loss: 0.4756 - val_accuracy: 0.7692
Epoch 65/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3957 - accuracy: 0.
8186 - val_loss: 0.4764 - val_accuracy: 0.7756
Epoch 66/200
20/20 [===========] - 0s 2ms/step - loss: 0.3950 - accuracy: 0.
8202 - val_loss: 0.4764 - val_accuracy: 0.7692
Epoch 67/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3944 - accuracy: 0.
8202 - val_loss: 0.4769 - val_accuracy: 0.7692
Epoch 68/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3941 - accuracy: 0.
8202 - val_loss: 0.4773 - val_accuracy: 0.7692
Epoch 69/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3940 - accuracy: 0.
8202 - val_loss: 0.4772 - val_accuracy: 0.7756
Epoch 70/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3930 - accuracy: 0.
8202 - val_loss: 0.4770 - val_accuracy: 0.7692
Epoch 71/200
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20/20 [================ ] - 0s 2ms/step - loss: 0.3926 - accuracy: 0.
8234 - val_loss: 0.4766 - val_accuracy: 0.7692
Epoch 72/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3918 - accuracy: 0.
8218 - val_loss: 0.4765 - val_accuracy: 0.7692
Epoch 73/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3912 - accuracy: 0.
8218 - val_loss: 0.4766 - val_accuracy: 0.7692
20/20 [================ ] - 0s 3ms/step - loss: 0.3907 - accuracy: 0.
8250 - val_loss: 0.4769 - val_accuracy: 0.7692
Epoch 75/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3904 - accuracy: 0.
8250 - val_loss: 0.4772 - val_accuracy: 0.7756
Epoch 76/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3898 - accuracy: 0.
8218 - val_loss: 0.4766 - val_accuracy: 0.7756
Epoch 77/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3895 - accuracy: 0.
8250 - val_loss: 0.4764 - val_accuracy: 0.7692
Epoch 78/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3890 - accuracy: 0.
8234 - val_loss: 0.4767 - val_accuracy: 0.7628
Epoch 79/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3883 - accuracy: 0.
8250 - val_loss: 0.4767 - val_accuracy: 0.7756
Epoch 80/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3879 - accuracy: 0.
8250 - val_loss: 0.4776 - val_accuracy: 0.7756
Epoch 81/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3873 - accuracy: 0.
8234 - val_loss: 0.4772 - val_accuracy: 0.7692
Epoch 82/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3870 - accuracy: 0.
8234 - val_loss: 0.4777 - val_accuracy: 0.7692
Epoch 83/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3866 - accuracy: 0.
8234 - val_loss: 0.4774 - val_accuracy: 0.7692
Epoch 84/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3862 - accuracy: 0.
8234 - val_loss: 0.4775 - val_accuracy: 0.7692
Epoch 85/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3857 - accuracy: 0.
8250 - val_loss: 0.4775 - val_accuracy: 0.7692
Epoch 86/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3852 - accuracy: 0.
8234 - val_loss: 0.4780 - val_accuracy: 0.7692
Epoch 87/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3847 - accuracy: 0.
8218 - val loss: 0.4779 - val accuracy: 0.7692
Epoch 88/200
20/20 [===========] - 0s 2ms/step - loss: 0.3851 - accuracy: 0.
8218 - val_loss: 0.4779 - val_accuracy: 0.7692
Epoch 89/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3838 - accuracy: 0.
8250 - val_loss: 0.4779 - val_accuracy: 0.7692
```

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Epoch 90/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3837 - accuracy: 0.
8234 - val_loss: 0.4784 - val_accuracy: 0.7692
Epoch 91/200
20/20 [================ ] - 0s 3ms/step - loss: 0.3828 - accuracy: 0.
8266 - val_loss: 0.4784 - val_accuracy: 0.7692
Epoch 92/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3824 - accuracy: 0.
8266 - val loss: 0.4785 - val accuracy: 0.7692
Epoch 93/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3823 - accuracy: 0.
8266 - val_loss: 0.4785 - val_accuracy: 0.7692
Epoch 94/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3819 - accuracy: 0.
8250 - val_loss: 0.4790 - val_accuracy: 0.7692
Epoch 95/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3815 - accuracy: 0.
8250 - val_loss: 0.4786 - val_accuracy: 0.7692
Epoch 96/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3811 - accuracy: 0.
8266 - val_loss: 0.4791 - val_accuracy: 0.7692
Epoch 97/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3807 - accuracy: 0.
8266 - val_loss: 0.4798 - val_accuracy: 0.7692
Epoch 98/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3805 - accuracy: 0.
8266 - val_loss: 0.4799 - val_accuracy: 0.7692
Epoch 99/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3800 - accuracy: 0.
8250 - val_loss: 0.4799 - val_accuracy: 0.7692
Epoch 100/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3798 - accuracy: 0.
8234 - val_loss: 0.4798 - val_accuracy: 0.7692
Epoch 101/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3799 - accuracy: 0.
8234 - val_loss: 0.4798 - val_accuracy: 0.7692
Epoch 102/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3791 - accuracy: 0.
8250 - val_loss: 0.4800 - val_accuracy: 0.7692
Epoch 103/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3786 - accuracy: 0.
8283 - val_loss: 0.4800 - val_accuracy: 0.7628
Epoch 104/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3785 - accuracy: 0.
8299 - val_loss: 0.4802 - val_accuracy: 0.7692
Epoch 105/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3784 - accuracy: 0.
8299 - val_loss: 0.4800 - val_accuracy: 0.7692
Epoch 106/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3778 - accuracy: 0.
8315 - val_loss: 0.4804 - val_accuracy: 0.7692
Epoch 107/200
20/20 [============ ] - 0s 3ms/step - loss: 0.3775 - accuracy: 0.
8315 - val_loss: 0.4799 - val_accuracy: 0.7692
20/20 [============== ] - 0s 2ms/step - loss: 0.3772 - accuracy: 0.
```

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8283 - val_loss: 0.4807 - val_accuracy: 0.7692
Epoch 109/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3768 - accuracy: 0.
8299 - val_loss: 0.4805 - val_accuracy: 0.7692
Epoch 110/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3765 - accuracy: 0.
8315 - val_loss: 0.4813 - val_accuracy: 0.7756
Epoch 111/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3763 - accuracy: 0.
8315 - val_loss: 0.4806 - val_accuracy: 0.7692
Epoch 112/200
20/20 [============] - 0s 2ms/step - loss: 0.3759 - accuracy: 0.
8315 - val_loss: 0.4818 - val_accuracy: 0.7692
20/20 [============== ] - 0s 2ms/step - loss: 0.3760 - accuracy: 0.
8331 - val_loss: 0.4824 - val_accuracy: 0.7756
Epoch 114/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3754 - accuracy: 0.
8299 - val_loss: 0.4821 - val_accuracy: 0.7756
Epoch 115/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3752 - accuracy: 0.
8331 - val_loss: 0.4821 - val_accuracy: 0.7692
Epoch 116/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3751 - accuracy: 0.
8299 - val_loss: 0.4824 - val_accuracy: 0.7692
Epoch 117/200
20/20 [==============] - 0s 2ms/step - loss: 0.3749 - accuracy: 0.
8331 - val_loss: 0.4831 - val_accuracy: 0.7692
Epoch 118/200
20/20 [=============== ] - 0s 2ms/step - loss: 0.3745 - accuracy: 0.
8299 - val loss: 0.4828 - val accuracy: 0.7628
20/20 [============ ] - 0s 2ms/step - loss: 0.3744 - accuracy: 0.
8299 - val_loss: 0.4832 - val_accuracy: 0.7500
Epoch 120/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3738 - accuracy: 0.
8299 - val_loss: 0.4833 - val_accuracy: 0.7692
Epoch 121/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3736 - accuracy: 0.
8347 - val_loss: 0.4829 - val_accuracy: 0.7692
Epoch 122/200
20/20 [===========] - 0s 2ms/step - loss: 0.3733 - accuracy: 0.
8331 - val_loss: 0.4835 - val_accuracy: 0.7692
Epoch 123/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3730 - accuracy: 0.
8363 - val_loss: 0.4836 - val_accuracy: 0.7692
Epoch 124/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3730 - accuracy: 0.
8315 - val_loss: 0.4835 - val_accuracy: 0.7692
Epoch 125/200
20/20 [================ ] - 0s 3ms/step - loss: 0.3724 - accuracy: 0.
8315 - val_loss: 0.4841 - val_accuracy: 0.7692
Epoch 126/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3726 - accuracy: 0.
8363 - val_loss: 0.4851 - val_accuracy: 0.7692
Epoch 127/200
```

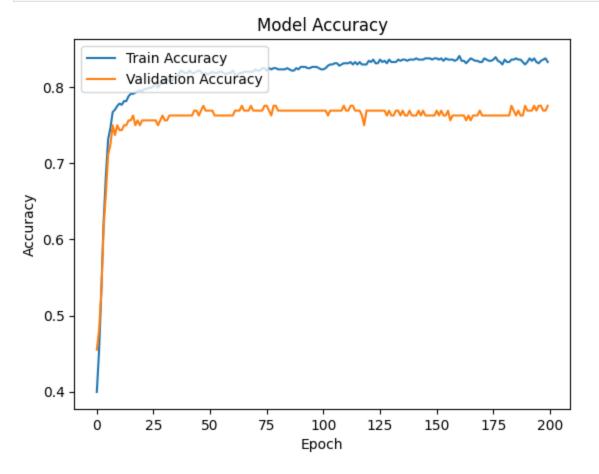
```
20/20 [================ ] - 0s 2ms/step - loss: 0.3719 - accuracy: 0.
8331 - val_loss: 0.4849 - val_accuracy: 0.7692
Epoch 128/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3716 - accuracy: 0.
8347 - val_loss: 0.4853 - val_accuracy: 0.7692
Epoch 129/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3717 - accuracy: 0.
8315 - val_loss: 0.4852 - val_accuracy: 0.7628
Epoch 130/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3715 - accuracy: 0.
8363 - val_loss: 0.4852 - val_accuracy: 0.7692
Epoch 131/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3709 - accuracy: 0.
8331 - val_loss: 0.4844 - val_accuracy: 0.7628
Epoch 132/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3709 - accuracy: 0.
8331 - val_loss: 0.4858 - val_accuracy: 0.7628
Epoch 133/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3705 - accuracy: 0.
8331 - val_loss: 0.4856 - val_accuracy: 0.7692
Epoch 134/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3704 - accuracy: 0.
8363 - val_loss: 0.4849 - val_accuracy: 0.7692
Epoch 135/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3697 - accuracy: 0.
8347 - val_loss: 0.4861 - val_accuracy: 0.7628
Epoch 136/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3696 - accuracy: 0.
8363 - val_loss: 0.4864 - val_accuracy: 0.7692
Epoch 137/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3691 - accuracy: 0.
8363 - val_loss: 0.4863 - val_accuracy: 0.7628
Epoch 138/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3691 - accuracy: 0.
8347 - val_loss: 0.4873 - val_accuracy: 0.7628
Epoch 139/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3687 - accuracy: 0.
8363 - val_loss: 0.4867 - val_accuracy: 0.7692
Epoch 140/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3684 - accuracy: 0.
8363 - val_loss: 0.4869 - val_accuracy: 0.7628
Epoch 141/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3687 - accuracy: 0.
8363 - val_loss: 0.4866 - val_accuracy: 0.7628
Epoch 142/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3687 - accuracy: 0.
8379 - val_loss: 0.4876 - val_accuracy: 0.7628
Epoch 143/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3677 - accuracy: 0.
8363 - val loss: 0.4878 - val accuracy: 0.7692
Epoch 144/200
20/20 [===========] - 0s 2ms/step - loss: 0.3674 - accuracy: 0.
8363 - val_loss: 0.4877 - val_accuracy: 0.7628
Epoch 145/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3672 - accuracy: 0.
8363 - val_loss: 0.4882 - val_accuracy: 0.7692
```

```
Epoch 146/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3669 - accuracy: 0.
8379 - val_loss: 0.4882 - val_accuracy: 0.7628
Epoch 147/200
20/20 [=============== ] - 0s 2ms/step - loss: 0.3671 - accuracy: 0.
8379 - val_loss: 0.4893 - val_accuracy: 0.7628
Epoch 148/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3665 - accuracy: 0.
8379 - val loss: 0.4897 - val accuracy: 0.7628
Epoch 149/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3663 - accuracy: 0.
8363 - val_loss: 0.4889 - val_accuracy: 0.7628
Epoch 150/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3660 - accuracy: 0.
8379 - val_loss: 0.4897 - val_accuracy: 0.7628
Epoch 151/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3659 - accuracy: 0.
8379 - val_loss: 0.4903 - val_accuracy: 0.7692
Epoch 152/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3654 - accuracy: 0.
8363 - val_loss: 0.4908 - val_accuracy: 0.7628
Epoch 153/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3651 - accuracy: 0.
8379 - val_loss: 0.4915 - val_accuracy: 0.7692
Epoch 154/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3650 - accuracy: 0.
8347 - val_loss: 0.4906 - val_accuracy: 0.7628
Epoch 155/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3646 - accuracy: 0.
8379 - val_loss: 0.4906 - val_accuracy: 0.7628
Epoch 156/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3645 - accuracy: 0.
8363 - val_loss: 0.4920 - val_accuracy: 0.7692
Epoch 157/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3642 - accuracy: 0.
8363 - val_loss: 0.4915 - val_accuracy: 0.7564
Epoch 158/200
8363 - val_loss: 0.4919 - val_accuracy: 0.7628
Epoch 159/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3642 - accuracy: 0.
8347 - val_loss: 0.4924 - val_accuracy: 0.7628
Epoch 160/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3636 - accuracy: 0.
8363 - val_loss: 0.4926 - val_accuracy: 0.7628
Epoch 161/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3642 - accuracy: 0.
8411 - val_loss: 0.4925 - val_accuracy: 0.7628
Epoch 162/200
8347 - val_loss: 0.4929 - val_accuracy: 0.7628
Epoch 163/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3628 - accuracy: 0.
8347 - val_loss: 0.4929 - val_accuracy: 0.7628
20/20 [================== ] - 0s 2ms/step - loss: 0.3630 - accuracy: 0.
```

```
8315 - val_loss: 0.4933 - val_accuracy: 0.7564
Epoch 165/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3621 - accuracy: 0.
8347 - val_loss: 0.4948 - val_accuracy: 0.7628
Epoch 166/200
20/20 [============== ] - 0s 2ms/step - loss: 0.3622 - accuracy: 0.
8379 - val_loss: 0.4951 - val_accuracy: 0.7564
Epoch 167/200
20/20 [=========== ] - 0s 2ms/step - loss: 0.3622 - accuracy: 0.
8363 - val_loss: 0.4952 - val_accuracy: 0.7628
Epoch 168/200
20/20 [============] - 0s 2ms/step - loss: 0.3617 - accuracy: 0.
8347 - val_loss: 0.4952 - val_accuracy: 0.7628
20/20 [============== ] - 0s 2ms/step - loss: 0.3614 - accuracy: 0.
8347 - val_loss: 0.4966 - val_accuracy: 0.7628
Epoch 170/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3616 - accuracy: 0.
8331 - val_loss: 0.4969 - val_accuracy: 0.7692
Epoch 171/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3608 - accuracy: 0.
8331 - val_loss: 0.4976 - val_accuracy: 0.7628
Epoch 172/200
20/20 [============ ] - 0s 2ms/step - loss: 0.3608 - accuracy: 0.
8379 - val_loss: 0.4975 - val_accuracy: 0.7628
Epoch 173/200
20/20 [==============] - 0s 2ms/step - loss: 0.3605 - accuracy: 0.
8363 - val_loss: 0.4985 - val_accuracy: 0.7628
Epoch 174/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3602 - accuracy: 0.
8347 - val loss: 0.4973 - val accuracy: 0.7628
Epoch 175/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3603 - accuracy: 0.
8347 - val_loss: 0.4974 - val_accuracy: 0.7628
Epoch 176/200
20/20 [=============== ] - 0s 2ms/step - loss: 0.3602 - accuracy: 0.
8363 - val_loss: 0.4987 - val_accuracy: 0.7628
Epoch 177/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3595 - accuracy: 0.
8395 - val_loss: 0.4989 - val_accuracy: 0.7628
Epoch 178/200
20/20 [============] - 0s 2ms/step - loss: 0.3596 - accuracy: 0.
8347 - val_loss: 0.4990 - val_accuracy: 0.7628
Epoch 179/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3594 - accuracy: 0.
8331 - val_loss: 0.5004 - val_accuracy: 0.7628
Epoch 180/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3596 - accuracy: 0.
8299 - val_loss: 0.5005 - val_accuracy: 0.7628
Epoch 181/200
20/20 [================ ] - 0s 2ms/step - loss: 0.3591 - accuracy: 0.
8363 - val_loss: 0.5016 - val_accuracy: 0.7628
Epoch 182/200
20/20 [================= ] - 0s 2ms/step - loss: 0.3588 - accuracy: 0.
8331 - val_loss: 0.5009 - val_accuracy: 0.7628
Epoch 183/200
```

```
20/20 [================ ] - 0s 2ms/step - loss: 0.3586 - accuracy: 0.
       8331 - val_loss: 0.5014 - val_accuracy: 0.7628
       Epoch 184/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3584 - accuracy: 0.
       8363 - val_loss: 0.5016 - val_accuracy: 0.7756
       Epoch 185/200
       20/20 [================ ] - 0s 2ms/step - loss: 0.3584 - accuracy: 0.
       8363 - val_loss: 0.5025 - val_accuracy: 0.7692
       Epoch 186/200
       20/20 [================ ] - 0s 2ms/step - loss: 0.3577 - accuracy: 0.
       8379 - val_loss: 0.5025 - val_accuracy: 0.7628
       Epoch 187/200
       20/20 [================ ] - 0s 2ms/step - loss: 0.3578 - accuracy: 0.
       8363 - val_loss: 0.5026 - val_accuracy: 0.7692
       Epoch 188/200
       20/20 [=========== ] - 0s 2ms/step - loss: 0.3576 - accuracy: 0.
       8363 - val_loss: 0.5034 - val_accuracy: 0.7628
       Epoch 189/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3574 - accuracy: 0.
       8331 - val_loss: 0.5041 - val_accuracy: 0.7628
       Epoch 190/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3575 - accuracy: 0.
       8299 - val_loss: 0.5039 - val_accuracy: 0.7756
       Epoch 191/200
       20/20 [============== ] - 0s 2ms/step - loss: 0.3571 - accuracy: 0.
       8331 - val_loss: 0.5044 - val_accuracy: 0.7692
       Epoch 192/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3568 - accuracy: 0.
       8379 - val_loss: 0.5051 - val_accuracy: 0.7692
       Epoch 193/200
       20/20 [============== ] - 0s 2ms/step - loss: 0.3570 - accuracy: 0.
       8347 - val_loss: 0.5049 - val_accuracy: 0.7692
       Epoch 194/200
       20/20 [=========== ] - 0s 2ms/step - loss: 0.3568 - accuracy: 0.
       8379 - val_loss: 0.5048 - val_accuracy: 0.7756
       Epoch 195/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3566 - accuracy: 0.
       8331 - val_loss: 0.5050 - val_accuracy: 0.7692
       Epoch 196/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3567 - accuracy: 0.
       8315 - val_loss: 0.5058 - val_accuracy: 0.7756
       Epoch 197/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3561 - accuracy: 0.
       8347 - val_loss: 0.5058 - val_accuracy: 0.7756
       Epoch 198/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3563 - accuracy: 0.
       8363 - val_loss: 0.5055 - val_accuracy: 0.7692
       Epoch 199/200
       20/20 [================ ] - 0s 2ms/step - loss: 0.3558 - accuracy: 0.
       8379 - val loss: 0.5064 - val accuracy: 0.7692
       Epoch 200/200
       20/20 [============ ] - 0s 2ms/step - loss: 0.3557 - accuracy: 0.
       8331 - val_loss: 0.5068 - val_accuracy: 0.7756
In [5]: import matplotlib.pyplot as plt
       plt.plot(history.history['accuracy'], label='Train Accuracy')
```

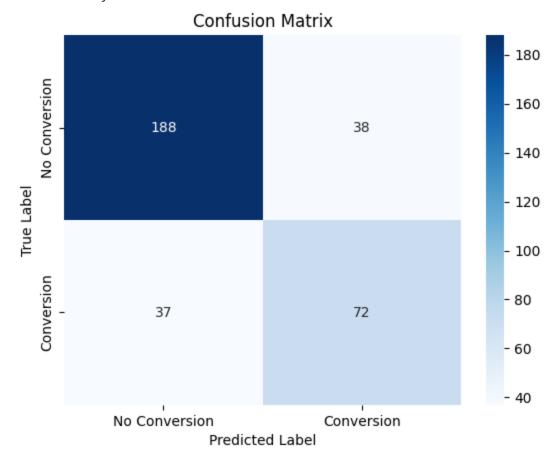
```
plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epoch')
plt.legend(loc='upper left')
plt.savefig('Graphs/NN1_accuracy.png')
plt.show()
plt.close()
```



# **Test dataset**

```
In [6]:
        from sklearn.metrics import confusion_matrix
        from sklearn.metrics import accuracy_score
        import seaborn as sns
        test_pred_prob = NNmodel.predict(X_test_preprocessed)
        test_preds = (test_pred_prob > 0.5).astype(int)
        accuracy = accuracy_score(Y_test, test_preds)
        print(f'Test Accuracy: {accuracy * 100:.2f}%')
        cm = confusion_matrix(Y_test, test_preds)
        sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['No Conversion', 'C
        plt.ylabel('True Label')
        plt.xlabel('Predicted Label')
        plt.title('Confusion Matrix')
        plt.savefig('Graphs/NN1_cm1.png')
        plt.show()
        plt.close()
```

11/11 [==========] - Os 800us/step Test Accuracy: 77.61%



# **Completer Dataset**

# **Datasets**

```
In [7]: df_train = pd.read_csv("train_completer.csv")
    df_test = pd.read_csv("test_completer.csv")
    X_train = df_train.drop(columns = ["converter"])
    Y_train = df_train.converter
    X_test = df_test.drop(columns = ["converter"])
    Y_test = df_test["converter"]
```

# Normalization

```
In [8]: numerical_cols = X_train.select_dtypes(include=['float64', 'int64']).columns.tolist
    categorical_cols = X_train.select_dtypes(include=['object']).columns.tolist()

# Initialize the StandardScaler and OneHotEncoder
scaler = StandardScaler()
encoder = OneHotEncoder(sparse=False)

# scale
```

```
X_train_scaled = scaler.fit_transform(X_train[numerical_cols])
X_test_scaled = scaler.transform(X_test[numerical_cols])
# encode
X_train_encoded = encoder.fit_transform(X_train[categorical_cols])
X_test_encoded = encoder.transform(X_test[categorical_cols])
# concatenate
X_train_preprocessed = np.concatenate([X_train_scaled, X_train_encoded], axis=1)
X_test_preprocessed = np.concatenate([X_test_scaled, X_test_encoded], axis=1)
C:\Users\43115\anaconda3\envs\python_310\lib\site-packages\sklearn\preprocessing\_encoders.py:868: FutureWarning: `sparse` was renamed to `sparse_output` in version
1.2 and will be removed in 1.4. `sparse_output` is ignored unless you leave `sparse` to its default value.
    warnings.warn(
```

## NN model

```
In [9]: NNmodel = Sequential()
    NNmodel.add(Dense(8, input_dim = len(X_train_preprocessed[0,:]), activation = "relu
    NNmodel.add(Dense(1, activation = "sigmoid")) # binary
    NNmodel.summary()
```

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_2 (Dense)	(None, 8)	184
dense_3 (Dense)	(None, 1)	9

Total params: 193 (772.00 Byte) Trainable params: 193 (772.00 Byte) Non-trainable params: 0 (0.00 Byte)

```
In [10]: NNmodel.compile(loss = "binary_crossentropy", optimizer = "adam", metrics = ["accur
history = NNmodel.fit(x = X_train_preprocessed, y = Y_train, epochs = 200, validati
```

```
Epoch 1/200
232 - val_loss: 0.8398 - val_accuracy: 0.3684
Epoch 2/200
30 - val_loss: 0.8209 - val_accuracy: 0.3684
Epoch 3/200
5/5 [================== ] - 0s 8ms/step - loss: 0.6576 - accuracy: 0.56
95 - val_loss: 0.8035 - val_accuracy: 0.3684
Epoch 4/200
94 - val_loss: 0.7867 - val_accuracy: 0.4211
Epoch 5/200
93 - val_loss: 0.7714 - val_accuracy: 0.4211
Epoch 6/200
24 - val_loss: 0.7581 - val_accuracy: 0.4737
Epoch 7/200
5/5 [================== ] - 0s 7ms/step - loss: 0.6120 - accuracy: 0.66
89 - val_loss: 0.7458 - val_accuracy: 0.5263
Epoch 8/200
55 - val_loss: 0.7337 - val_accuracy: 0.5263
Epoch 9/200
21 - val_loss: 0.7221 - val_accuracy: 0.5263
Epoch 10/200
21 - val_loss: 0.7117 - val_accuracy: 0.5526
Epoch 11/200
5/5 [==========] - 0s 7ms/step - loss: 0.5794 - accuracy: 0.68
87 - val_loss: 0.7024 - val_accuracy: 0.5789
Epoch 12/200
85 - val_loss: 0.6928 - val_accuracy: 0.6316
Epoch 13/200
5/5 [==========] - 0s 7ms/step - loss: 0.5649 - accuracy: 0.74
17 - val_loss: 0.6841 - val_accuracy: 0.6316
Epoch 14/200
50 - val_loss: 0.6760 - val_accuracy: 0.6579
Epoch 15/200
16 - val_loss: 0.6680 - val_accuracy: 0.6842
Epoch 16/200
82 - val_loss: 0.6601 - val_accuracy: 0.7105
Epoch 17/200
15 - val_loss: 0.6520 - val_accuracy: 0.7105
Epoch 18/200
47 - val_loss: 0.6444 - val_accuracy: 0.7105
5/5 [==================== ] - 0s 7ms/step - loss: 0.5232 - accuracy: 0.80
```

```
79 - val_loss: 0.6362 - val_accuracy: 0.7105
Epoch 20/200
46 - val_loss: 0.6274 - val_accuracy: 0.7368
Epoch 21/200
79 - val_loss: 0.6189 - val_accuracy: 0.7368
Epoch 22/200
46 - val_loss: 0.6111 - val_accuracy: 0.7632
Epoch 23/200
5/5 [=================] - 0s 7ms/step - loss: 0.4967 - accuracy: 0.80
79 - val_loss: 0.6030 - val_accuracy: 0.7632
Epoch 24/200
5/5 [=============== ] - 0s 7ms/step - loss: 0.4902 - accuracy: 0.80
79 - val_loss: 0.5959 - val_accuracy: 0.7632
Epoch 25/200
5/5 [===========] - 0s 7ms/step - loss: 0.4849 - accuracy: 0.82
12 - val_loss: 0.5886 - val_accuracy: 0.7895
Epoch 26/200
12 - val_loss: 0.5818 - val_accuracy: 0.7895
Epoch 27/200
12 - val_loss: 0.5741 - val_accuracy: 0.7895
Epoch 28/200
5/5 [==========] - 0s 7ms/step - loss: 0.4659 - accuracy: 0.82
12 - val_loss: 0.5672 - val_accuracy: 0.7895
Epoch 29/200
5/5 [=================] - 0s 8ms/step - loss: 0.4603 - accuracy: 0.82
78 - val_loss: 0.5606 - val_accuracy: 0.7895
Epoch 30/200
78 - val_loss: 0.5547 - val_accuracy: 0.7895
Epoch 31/200
5/5 [=================] - 0s 7ms/step - loss: 0.4499 - accuracy: 0.82
78 - val_loss: 0.5493 - val_accuracy: 0.7895
Epoch 32/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4449 - accuracy: 0.82
12 - val_loss: 0.5437 - val_accuracy: 0.7895
Epoch 33/200
12 - val_loss: 0.5382 - val_accuracy: 0.7895
Epoch 34/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4365 - accuracy: 0.82
12 - val_loss: 0.5335 - val_accuracy: 0.7895
Epoch 35/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4326 - accuracy: 0.82
12 - val_loss: 0.5292 - val_accuracy: 0.7895
5/5 [================== ] - 0s 7ms/step - loss: 0.4289 - accuracy: 0.82
12 - val_loss: 0.5252 - val_accuracy: 0.7895
Epoch 37/200
78 - val_loss: 0.5219 - val_accuracy: 0.7895
Epoch 38/200
```

```
5/5 [================== ] - 0s 8ms/step - loss: 0.4221 - accuracy: 0.82
12 - val_loss: 0.5183 - val_accuracy: 0.7895
Epoch 39/200
12 - val_loss: 0.5144 - val_accuracy: 0.7895
Epoch 40/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4160 - accuracy: 0.82
78 - val_loss: 0.5109 - val_accuracy: 0.7895
Epoch 41/200
5/5 [============] - 0s 7ms/step - loss: 0.4132 - accuracy: 0.82
78 - val_loss: 0.5084 - val_accuracy: 0.8158
Epoch 42/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4103 - accuracy: 0.82
78 - val_loss: 0.5054 - val_accuracy: 0.8158
Epoch 43/200
5/5 [================ ] - 0s 7ms/step - loss: 0.4077 - accuracy: 0.82
78 - val_loss: 0.5032 - val_accuracy: 0.8158
Epoch 44/200
78 - val_loss: 0.4999 - val_accuracy: 0.8158
Epoch 45/200
5/5 [==========] - 0s 7ms/step - loss: 0.4027 - accuracy: 0.82
78 - val_loss: 0.4973 - val_accuracy: 0.8158
Epoch 46/200
5/5 [================== ] - 0s 7ms/step - loss: 0.4003 - accuracy: 0.82
12 - val_loss: 0.4947 - val_accuracy: 0.8158
Epoch 47/200
12 - val_loss: 0.4921 - val_accuracy: 0.8158
Epoch 48/200
5/5 [================ ] - 0s 7ms/step - loss: 0.3956 - accuracy: 0.82
78 - val_loss: 0.4901 - val_accuracy: 0.8158
Epoch 49/200
11 - val_loss: 0.4884 - val_accuracy: 0.8158
Epoch 50/200
5/5 [==========] - 0s 7ms/step - loss: 0.3914 - accuracy: 0.84
11 - val_loss: 0.4871 - val_accuracy: 0.8158
Epoch 51/200
11 - val_loss: 0.4857 - val_accuracy: 0.8158
Epoch 52/200
11 - val_loss: 0.4840 - val_accuracy: 0.8158
Epoch 53/200
11 - val_loss: 0.4828 - val_accuracy: 0.8158
Epoch 54/200
5/5 [================== ] - 0s 8ms/step - loss: 0.3835 - accuracy: 0.84
11 - val loss: 0.4815 - val accuracy: 0.8158
Epoch 55/200
11 - val_loss: 0.4807 - val_accuracy: 0.8158
Epoch 56/200
11 - val loss: 0.4789 - val accuracy: 0.8158
```

```
Epoch 57/200
5/5 [================= ] - 0s 8ms/step - loss: 0.3781 - accuracy: 0.84
11 - val_loss: 0.4778 - val_accuracy: 0.8158
Epoch 58/200
5/5 [=================] - 0s 7ms/step - loss: 0.3762 - accuracy: 0.84
11 - val_loss: 0.4771 - val_accuracy: 0.8158
Epoch 59/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3744 - accuracy: 0.84
11 - val loss: 0.4764 - val accuracy: 0.8158
Epoch 60/200
11 - val_loss: 0.4765 - val_accuracy: 0.8158
Epoch 61/200
11 - val_loss: 0.4753 - val_accuracy: 0.8158
Epoch 62/200
77 - val_loss: 0.4743 - val_accuracy: 0.8158
Epoch 63/200
5/5 [================== ] - 0s 8ms/step - loss: 0.3676 - accuracy: 0.84
77 - val_loss: 0.4737 - val_accuracy: 0.8158
Epoch 64/200
77 - val_loss: 0.4727 - val_accuracy: 0.8158
Epoch 65/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3642 - accuracy: 0.84
77 - val_loss: 0.4728 - val_accuracy: 0.8158
Epoch 66/200
43 - val_loss: 0.4721 - val_accuracy: 0.8158
Epoch 67/200
5/5 [==========] - 0s 7ms/step - loss: 0.3609 - accuracy: 0.85
43 - val_loss: 0.4718 - val_accuracy: 0.8158
Epoch 68/200
43 - val_loss: 0.4706 - val_accuracy: 0.8158
Epoch 69/200
43 - val_loss: 0.4703 - val_accuracy: 0.8158
Epoch 70/200
43 - val_loss: 0.4690 - val_accuracy: 0.8158
Epoch 71/200
43 - val_loss: 0.4691 - val_accuracy: 0.8158
Epoch 72/200
09 - val_loss: 0.4699 - val_accuracy: 0.8158
Epoch 73/200
09 - val_loss: 0.4696 - val_accuracy: 0.8158
Epoch 74/200
09 - val_loss: 0.4680 - val_accuracy: 0.8158
5/5 [=================== ] - 0s 7ms/step - loss: 0.3484 - accuracy: 0.86
```

```
09 - val_loss: 0.4679 - val_accuracy: 0.8158
Epoch 76/200
09 - val_loss: 0.4681 - val_accuracy: 0.8158
Epoch 77/200
09 - val_loss: 0.4678 - val_accuracy: 0.8158
Epoch 78/200
09 - val_loss: 0.4680 - val_accuracy: 0.8158
Epoch 79/200
5/5 [=================] - 0s 8ms/step - loss: 0.3427 - accuracy: 0.86
09 - val_loss: 0.4679 - val_accuracy: 0.8158
5/5 [=============== ] - 0s 7ms/step - loss: 0.3413 - accuracy: 0.86
09 - val_loss: 0.4676 - val_accuracy: 0.8158
Epoch 81/200
5/5 [===========] - 0s 7ms/step - loss: 0.3401 - accuracy: 0.86
09 - val_loss: 0.4667 - val_accuracy: 0.8158
Epoch 82/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3386 - accuracy: 0.86
09 - val_loss: 0.4654 - val_accuracy: 0.8158
Epoch 83/200
09 - val_loss: 0.4644 - val_accuracy: 0.8158
Epoch 84/200
5/5 [==========] - 0s 7ms/step - loss: 0.3361 - accuracy: 0.86
09 - val_loss: 0.4636 - val_accuracy: 0.8158
Epoch 85/200
5/5 [=================] - 0s 8ms/step - loss: 0.3349 - accuracy: 0.86
09 - val loss: 0.4627 - val accuracy: 0.8158
Epoch 86/200
09 - val_loss: 0.4624 - val_accuracy: 0.8158
Epoch 87/200
5/5 [=================] - 0s 7ms/step - loss: 0.3324 - accuracy: 0.86
09 - val_loss: 0.4620 - val_accuracy: 0.8158
Epoch 88/200
09 - val_loss: 0.4621 - val_accuracy: 0.8158
Epoch 89/200
5/5 [================= ] - 0s 8ms/step - loss: 0.3301 - accuracy: 0.86
09 - val_loss: 0.4610 - val_accuracy: 0.8158
Epoch 90/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3290 - accuracy: 0.86
09 - val_loss: 0.4603 - val_accuracy: 0.8158
Epoch 91/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3279 - accuracy: 0.86
75 - val_loss: 0.4604 - val_accuracy: 0.8158
Epoch 92/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3266 - accuracy: 0.86
75 - val_loss: 0.4599 - val_accuracy: 0.8158
Epoch 93/200
5/5 [================= ] - 0s 8ms/step - loss: 0.3254 - accuracy: 0.87
42 - val_loss: 0.4593 - val_accuracy: 0.8158
Epoch 94/200
```

```
5/5 [================== ] - 0s 7ms/step - loss: 0.3245 - accuracy: 0.87
42 - val_loss: 0.4592 - val_accuracy: 0.8158
Epoch 95/200
08 - val_loss: 0.4587 - val_accuracy: 0.8158
Epoch 96/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3223 - accuracy: 0.88
08 - val_loss: 0.4575 - val_accuracy: 0.8158
Epoch 97/200
5/5 [============] - 0s 7ms/step - loss: 0.3212 - accuracy: 0.88
08 - val_loss: 0.4571 - val_accuracy: 0.8158
Epoch 98/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3201 - accuracy: 0.88
74 - val_loss: 0.4560 - val_accuracy: 0.8158
Epoch 99/200
5/5 [=============== ] - 0s 8ms/step - loss: 0.3193 - accuracy: 0.88
08 - val_loss: 0.4547 - val_accuracy: 0.8158
Epoch 100/200
08 - val_loss: 0.4544 - val_accuracy: 0.8158
Epoch 101/200
5/5 [==========] - 0s 7ms/step - loss: 0.3170 - accuracy: 0.88
08 - val_loss: 0.4539 - val_accuracy: 0.8158
Epoch 102/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3160 - accuracy: 0.88
08 - val_loss: 0.4531 - val_accuracy: 0.8158
Epoch 103/200
74 - val_loss: 0.4526 - val_accuracy: 0.8158
Epoch 104/200
5/5 [=============== ] - 0s 7ms/step - loss: 0.3137 - accuracy: 0.88
74 - val_loss: 0.4530 - val_accuracy: 0.8158
Epoch 105/200
40 - val_loss: 0.4523 - val_accuracy: 0.8158
Epoch 106/200
5/5 [==========] - 0s 7ms/step - loss: 0.3118 - accuracy: 0.89
40 - val_loss: 0.4517 - val_accuracy: 0.8158
Epoch 107/200
5/5 [================] - 0s 7ms/step - loss: 0.3107 - accuracy: 0.89
40 - val_loss: 0.4511 - val_accuracy: 0.8158
Epoch 108/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3098 - accuracy: 0.89
40 - val_loss: 0.4509 - val_accuracy: 0.8158
Epoch 109/200
5/5 [=================] - 0s 8ms/step - loss: 0.3090 - accuracy: 0.89
40 - val_loss: 0.4492 - val_accuracy: 0.8158
Epoch 110/200
40 - val loss: 0.4486 - val accuracy: 0.8158
Epoch 111/200
40 - val_loss: 0.4472 - val_accuracy: 0.8158
Epoch 112/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3061 - accuracy: 0.89
40 - val loss: 0.4462 - val accuracy: 0.8158
```

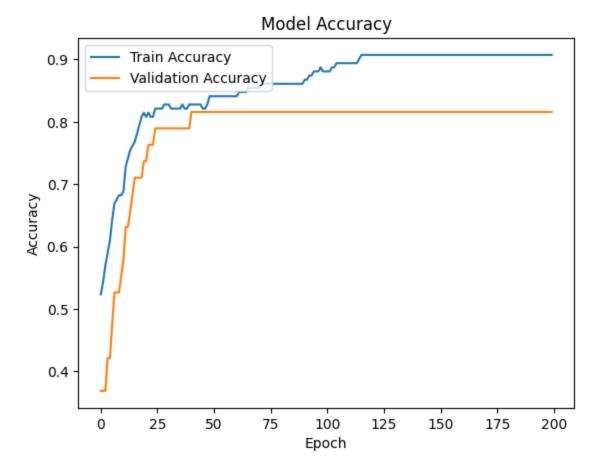
```
Epoch 113/200
5/5 [================== ] - 0s 7ms/step - loss: 0.3050 - accuracy: 0.89
40 - val_loss: 0.4462 - val_accuracy: 0.8158
Epoch 114/200
5/5 [================] - 0s 8ms/step - loss: 0.3042 - accuracy: 0.89
40 - val_loss: 0.4463 - val_accuracy: 0.8158
Epoch 115/200
5/5 [================== ] - 0s 8ms/step - loss: 0.3033 - accuracy: 0.90
07 - val_loss: 0.4457 - val_accuracy: 0.8158
Epoch 116/200
73 - val_loss: 0.4454 - val_accuracy: 0.8158
Epoch 117/200
73 - val_loss: 0.4447 - val_accuracy: 0.8158
Epoch 118/200
73 - val_loss: 0.4440 - val_accuracy: 0.8158
Epoch 119/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2997 - accuracy: 0.90
73 - val_loss: 0.4440 - val_accuracy: 0.8158
Epoch 120/200
73 - val_loss: 0.4428 - val_accuracy: 0.8158
Epoch 121/200
73 - val_loss: 0.4422 - val_accuracy: 0.8158
Epoch 122/200
73 - val_loss: 0.4418 - val_accuracy: 0.8158
Epoch 123/200
5/5 [==========] - 0s 7ms/step - loss: 0.2963 - accuracy: 0.90
73 - val_loss: 0.4420 - val_accuracy: 0.8158
Epoch 124/200
73 - val_loss: 0.4420 - val_accuracy: 0.8158
Epoch 125/200
73 - val_loss: 0.4412 - val_accuracy: 0.8158
Epoch 126/200
73 - val_loss: 0.4406 - val_accuracy: 0.8158
Epoch 127/200
73 - val_loss: 0.4399 - val_accuracy: 0.8158
Epoch 128/200
73 - val_loss: 0.4391 - val_accuracy: 0.8158
Epoch 129/200
73 - val_loss: 0.4383 - val_accuracy: 0.8158
Epoch 130/200
73 - val_loss: 0.4383 - val_accuracy: 0.8158
5/5 [=================== ] - 0s 7ms/step - loss: 0.2896 - accuracy: 0.90
```

```
73 - val_loss: 0.4372 - val_accuracy: 0.8158
Epoch 132/200
73 - val_loss: 0.4375 - val_accuracy: 0.8158
Epoch 133/200
73 - val_loss: 0.4365 - val_accuracy: 0.8158
Epoch 134/200
73 - val_loss: 0.4358 - val_accuracy: 0.8158
Epoch 135/200
5/5 [=================] - 0s 7ms/step - loss: 0.2865 - accuracy: 0.90
73 - val_loss: 0.4360 - val_accuracy: 0.8158
5/5 [=============== ] - 0s 7ms/step - loss: 0.2857 - accuracy: 0.90
73 - val_loss: 0.4355 - val_accuracy: 0.8158
Epoch 137/200
5/5 [==========] - 0s 8ms/step - loss: 0.2850 - accuracy: 0.90
73 - val_loss: 0.4351 - val_accuracy: 0.8158
Epoch 138/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2844 - accuracy: 0.90
73 - val_loss: 0.4333 - val_accuracy: 0.8158
Epoch 139/200
73 - val_loss: 0.4328 - val_accuracy: 0.8158
Epoch 140/200
5/5 [==========] - 0s 7ms/step - loss: 0.2828 - accuracy: 0.90
73 - val_loss: 0.4317 - val_accuracy: 0.8158
Epoch 141/200
5/5 [=================] - 0s 7ms/step - loss: 0.2819 - accuracy: 0.90
73 - val loss: 0.4314 - val accuracy: 0.8158
Epoch 142/200
73 - val_loss: 0.4317 - val_accuracy: 0.8158
Epoch 143/200
5/5 [=================] - 0s 7ms/step - loss: 0.2806 - accuracy: 0.90
73 - val_loss: 0.4316 - val_accuracy: 0.8158
Epoch 144/200
73 - val_loss: 0.4311 - val_accuracy: 0.8158
Epoch 145/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2791 - accuracy: 0.90
73 - val_loss: 0.4304 - val_accuracy: 0.8158
Epoch 146/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2783 - accuracy: 0.90
73 - val_loss: 0.4300 - val_accuracy: 0.8158
Epoch 147/200
5/5 [================== ] - 0s 8ms/step - loss: 0.2777 - accuracy: 0.90
73 - val_loss: 0.4302 - val_accuracy: 0.8158
Epoch 148/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2770 - accuracy: 0.90
73 - val_loss: 0.4300 - val_accuracy: 0.8158
Epoch 149/200
73 - val_loss: 0.4294 - val_accuracy: 0.8158
Epoch 150/200
```

```
5/5 [================== ] - 0s 7ms/step - loss: 0.2756 - accuracy: 0.90
73 - val_loss: 0.4286 - val_accuracy: 0.8158
Epoch 151/200
73 - val_loss: 0.4279 - val_accuracy: 0.8158
Epoch 152/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2743 - accuracy: 0.90
73 - val_loss: 0.4274 - val_accuracy: 0.8158
Epoch 153/200
5/5 [============] - 0s 7ms/step - loss: 0.2736 - accuracy: 0.90
73 - val_loss: 0.4274 - val_accuracy: 0.8158
Epoch 154/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2730 - accuracy: 0.90
73 - val_loss: 0.4270 - val_accuracy: 0.8158
Epoch 155/200
73 - val_loss: 0.4278 - val_accuracy: 0.8158
Epoch 156/200
73 - val_loss: 0.4278 - val_accuracy: 0.8158
Epoch 157/200
5/5 [==========] - 0s 7ms/step - loss: 0.2710 - accuracy: 0.90
73 - val_loss: 0.4274 - val_accuracy: 0.8158
Epoch 158/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2704 - accuracy: 0.90
73 - val_loss: 0.4273 - val_accuracy: 0.8158
Epoch 159/200
73 - val_loss: 0.4271 - val_accuracy: 0.8158
Epoch 160/200
5/5 [================ ] - 0s 8ms/step - loss: 0.2691 - accuracy: 0.90
73 - val_loss: 0.4263 - val_accuracy: 0.8158
Epoch 161/200
73 - val_loss: 0.4265 - val_accuracy: 0.8158
Epoch 162/200
5/5 [==========] - 0s 7ms/step - loss: 0.2677 - accuracy: 0.90
73 - val_loss: 0.4266 - val_accuracy: 0.8158
Epoch 163/200
73 - val_loss: 0.4271 - val_accuracy: 0.8158
Epoch 164/200
73 - val_loss: 0.4266 - val_accuracy: 0.8158
Epoch 165/200
73 - val_loss: 0.4262 - val_accuracy: 0.8158
Epoch 166/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2653 - accuracy: 0.90
73 - val loss: 0.4266 - val accuracy: 0.8158
Epoch 167/200
73 - val_loss: 0.4266 - val_accuracy: 0.8158
Epoch 168/200
73 - val_loss: 0.4267 - val_accuracy: 0.8158
```

```
Epoch 169/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2634 - accuracy: 0.90
73 - val_loss: 0.4259 - val_accuracy: 0.8158
Epoch 170/200
5/5 [=================] - 0s 7ms/step - loss: 0.2627 - accuracy: 0.90
73 - val_loss: 0.4259 - val_accuracy: 0.8158
Epoch 171/200
5/5 [================== ] - 0s 7ms/step - loss: 0.2621 - accuracy: 0.90
73 - val_loss: 0.4261 - val_accuracy: 0.8158
Epoch 172/200
73 - val_loss: 0.4258 - val_accuracy: 0.8158
Epoch 173/200
73 - val_loss: 0.4255 - val_accuracy: 0.8158
Epoch 174/200
73 - val_loss: 0.4251 - val_accuracy: 0.8158
Epoch 175/200
5/5 [================== ] - 0s 8ms/step - loss: 0.2597 - accuracy: 0.90
73 - val_loss: 0.4247 - val_accuracy: 0.8158
Epoch 176/200
73 - val_loss: 0.4251 - val_accuracy: 0.8158
Epoch 177/200
73 - val_loss: 0.4253 - val_accuracy: 0.8158
Epoch 178/200
73 - val_loss: 0.4249 - val_accuracy: 0.8158
Epoch 179/200
5/5 [==========] - 0s 7ms/step - loss: 0.2573 - accuracy: 0.90
73 - val_loss: 0.4253 - val_accuracy: 0.8158
Epoch 180/200
73 - val_loss: 0.4249 - val_accuracy: 0.8158
Epoch 181/200
73 - val_loss: 0.4254 - val_accuracy: 0.8158
Epoch 182/200
73 - val_loss: 0.4249 - val_accuracy: 0.8158
Epoch 183/200
73 - val_loss: 0.4248 - val_accuracy: 0.8158
Epoch 184/200
73 - val_loss: 0.4238 - val_accuracy: 0.8158
Epoch 185/200
73 - val_loss: 0.4230 - val_accuracy: 0.8158
Epoch 186/200
73 - val_loss: 0.4228 - val_accuracy: 0.8158
5/5 [==================== ] - 0s 7ms/step - loss: 0.2529 - accuracy: 0.90
```

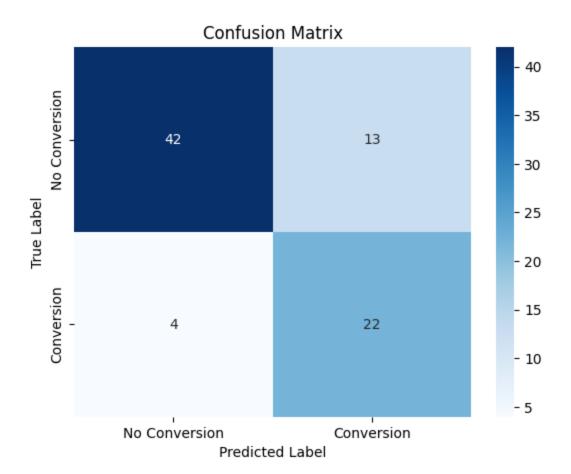
```
73 - val_loss: 0.4235 - val_accuracy: 0.8158
     Epoch 188/200
     73 - val_loss: 0.4231 - val_accuracy: 0.8158
     Epoch 189/200
     73 - val_loss: 0.4239 - val_accuracy: 0.8158
     Epoch 190/200
     73 - val_loss: 0.4252 - val_accuracy: 0.8158
     Epoch 191/200
     73 - val_loss: 0.4247 - val_accuracy: 0.8158
     Epoch 192/200
     73 - val_loss: 0.4247 - val_accuracy: 0.8158
     Epoch 193/200
     73 - val_loss: 0.4256 - val_accuracy: 0.8158
     Epoch 194/200
     73 - val_loss: 0.4256 - val_accuracy: 0.8158
     Epoch 195/200
     73 - val_loss: 0.4258 - val_accuracy: 0.8158
     Epoch 196/200
     5/5 [==========] - 0s 7ms/step - loss: 0.2481 - accuracy: 0.90
     73 - val_loss: 0.4249 - val_accuracy: 0.8158
     Epoch 197/200
     5/5 [=================] - 0s 8ms/step - loss: 0.2475 - accuracy: 0.90
     73 - val loss: 0.4255 - val accuracy: 0.8158
     Epoch 198/200
     73 - val loss: 0.4258 - val accuracy: 0.8158
     Epoch 199/200
     5/5 [=================] - 0s 8ms/step - loss: 0.2464 - accuracy: 0.90
     73 - val_loss: 0.4257 - val_accuracy: 0.8158
     Epoch 200/200
     73 - val_loss: 0.4259 - val_accuracy: 0.8158
In [11]: plt.plot(history.history['accuracy'], label='Train Accuracy')
     plt.plot(history.history['val_accuracy'], label='Validation Accuracy')
     plt.title('Model Accuracy')
     plt.ylabel('Accuracy')
     plt.xlabel('Epoch')
     plt.legend(loc='upper left')
     plt.savefig('Graphs/NN2_accuracy.png')
     plt.show()
     plt.close()
```



```
In [12]: test_pred_prob = NNmodel.predict(X_test_preprocessed)
    test_preds = (test_pred_prob > 0.5).astype(int)
    accuracy = accuracy_score(Y_test, test_preds)
    print(f'Test Accuracy: {accuracy * 100:.2f}%')
    cm = confusion_matrix(Y_test, test_preds)
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['No Conversion', 'C plt.ylabel('True Label')
    plt.xlabel('Predicted Label')
    plt.title('Confusion Matrix')
    plt.savefig('Graphs/NN2_cm2.png')
    plt.show()
    plt.close()
```

3/3 [======] - 0s 1ms/step Test Accuracy: 79.01%

file:///C:/Users/43115/Downloads/Neural Network (4).html



In [ ]: