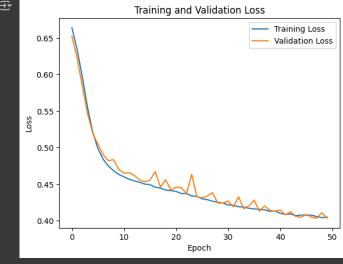
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from \ sklearn.model\_selection \ import \ train\_test\_split
{\it from} \ {\it keras.models import Sequential}
from keras.layers import Dense
from keras.losses import BinaryCrossentropy
data = pd.read_csv('spotify_preprocessed.csv')
# Shuffle the data
data = data.sample(frac=1).reset_index(drop=True)
# Split the data into features and labels
X = data.drop(columns=['target'])
y = data['target']
# Split the data into training and test set
X_train_full, X_test, y_train_full, y_test = train_test_split(X, y, test_size=0.1, random_state=42)
# Split the training set into training and validation sets
X_train, X_val, y_train, y_val = train_test_split(X_train_full, y_train_full, test_size=0.2, random_state=42)
# Build the neural network model
def build_model():
    model = Sequential()
    model.add(Dense(32, input_dim=X.shape[1], activation='relu'))
model.add(Dense(32, activation='relu'))
    model.add(Dense(1, activation='sigmoid'))
    return model
model = build_model()
# Compile the model
model.compile(loss=BinaryCrossentropy(),
              optimizer=SGD(learning_rate=0.01),
               metrics=['accuracy'])
# Train the model
history = model.fit(X_train, y_train, epochs=50,
                     batch size=16.
                     validation_data=(X_val, y_val))
<del>____</del>
```

```
7/14/24, 11:53 PM
                                                                                     Belii_Assignment#10 - Colab
         288/288 [=
                                                      - 1s 3ms/step - loss: 0.4094 - accuracy: 0.8150 - val_loss: 0.4125 - val_accuracy: 0.8203
                                                        1s 4ms/step - loss: 0.4070 - accuracy: 0.8146 - val_loss: 0.4059 - val_accuracy: 0.8273
         Epoch 45/50
                                                        1s 4ms/step - loss: 0.4079 - accuracy: 0.8111 - val_loss: 0.4049 - val_accuracy: 0.8281
         288/288 [=
         Epoch 46/50
                                                        1s 4ms/step - loss: 0.4078 - accuracy: 0.8152 - val_loss: 0.4089 - val_accuracy: 0.8177
         288/288 [=
         Epoch 47/50
                                                        1s 4ms/step - loss: 0.4077 - accuracy: 0.8133 - val_loss: 0.4051 - val_accuracy: 0.8264
         288/288 [==
         Epoch 48/50
         288/288 [=
                                                        1s 4ms/step - loss: 0.4059 - accuracy: 0.8131 - val_loss: 0.4036 - val_accuracy: 0.8255
         Epoch 49/50
         288/288 [===
Epoch 50/50
                                                        1s 4ms/step - loss: 0.4045 - accuracy: 0.8124 - val_loss: 0.4113 - val_accuracy: 0.8151
         288/288 [===
                                                        2s 5ms/step - loss: 0.4052 - accuracy: 0.8174 - val_loss: 0.4035 - val_accuracy: 0.8325
    # Plot the training and validation loss
    plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
    plt.title('Training and Validation Loss')
    plt.xlabel('Epoch')
    plt.ylabel('Loss')
    plt.legend()
    plt.show()
     <del>____</del>
                                   Training and Validation Loss
                                                                   Training Loss
             0.65
                                                                   Validation Loss
```



Evaluate the model on the training and validation sets train_loss, train_accuracy = model.evaluate(X_train, y_train, verbose=0) val_loss, val_accuracy = model.evaluate(X_val, y_val, verbose=0)

print(f'Training Loss: {train_loss}, Training Accuracy: {train_accuracy}')
print(f'Validation Loss: {val_loss}, Validation Accuracy: {val_accuracy}')

Training Loss: 0.3991946578025818, Training Accuracy: 0.8174120783805847
Validation Loss: 0.40354645252227783, Validation Accuracy: 0.8324652910232544

```
Test different model designs
def experiment_with_models():
    configurations =
          Igurations = {
{"layers": [32, 32], "epochs": 50, "batch_size": 16, "learning_rate": 0.01},
{"layers": [64, 32], "epochs": 50, "batch_size": 16, "learning_rate": 0.01},
{"layers": [32, 32, 16], "epochs": 50, "batch_size": 16, "learning_rate": 0.01},
{"layers": [64, 64], "epochs": 50, "batch_size": 16, "learning_rate": 0.01},
{"layers": [32, 16], "epochs": 50, "batch_size": 16, "learning_rate": 0.01},
     for config in configurations:
          model = Sequential()
          for units in config["layers"]:
               if model.layers:
                    model.add(Dense(units, activation='relu'))
         model.add(Dense(units, input_dim=X.shape[1], activation='relu'))
model.add(Dense(1, activation='sigmoid'))
         metrics=['accuracy'])
          history = model.fit(X_train, y_train,
                                   epochs=config["epochs"],
                                   batch_size=config["batch_size"],
                                   validation_data=(X_val, y_val),
                                    verbose=0)
          train_loss, train_accuracy = model.evaluate(X_train, y_train, verbose=0)
          val_loss, val_accuracy = model.evaluate(X_val, y_val, verbose=0)
          results.append({
               "Configuration": config,
               "Training Loss": train_loss,
               "Training Accuracy": train_accuracy,
               "Validation Loss": val_loss,
               "Validation Accuracy": val_accuracy
          # Plot the training and validation loss for each configuration
          plt.figure()
         plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
          plt.title(f'Training and Validation Loss for config {config}')
          plt.xlabel('Epoch')
          plt.ylabel('Loss')
          plt.legend()
          plt.show()
     return pd.DataFrame(results)
results_df = experiment_with_models()
```

```
print(results_df)
```

