

```
#Jack Baxter
#Final Project: AAI 511
#11 August 2025
```

```
#import necessary packages
import pretty_midi
import pandas as pd
import numpy as np
from google.colab import drive
import os
from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import to_categorical
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense, Dropout
from tensorflow.keras.layers import LSTM, TimeDistributed
from sklearn.metrics import accuracy_score, precision_score, recall_score
```

```
#initialize connection to google drive data
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True)

```
#create main dataframe for piano rolls and composer id
main_music = pd.DataFrame(columns = ['song', 'composer_id'])
```

```
#initialize standard midi & piano roll parameters
hz = 100
timesteps = 4000
pitchrange = 128
```

```
#define function to get piano roll from midi file
#function created by LLM (xAI, 2025)
def get_piano_roll(midi_path, fs=100, fixed_length=4000):
    try:
        midi_data = pretty_midi.PrettyMIDI(midi_path)
        piano_roll = midi_data.get_piano_roll(fs=fs)
        piano_roll = piano_roll / 127.0
        current_length = piano_roll.shape[1]
        if current_length < fixed_length:
            padded = np.zeros((pitchrange, fixed_length))
            padded[:, :current_length] = piano_roll
            return padded
        else:
            return piano_roll[:, :fixed_length]
    except Exception as e:
        print(f"Error processing {midi_path}: {e}")
        return None
```

```
#previous dataset append function - not used in final
piano_roll = get_piano_roll(midi_path, hz, pitchrange)
if piano_roll is not None:
    new_row = pd.DataFrame({
        'song': [piano_roll],
        'composer_id': [composer_id]
    })
    main_music = pd.concat([main_music, new_row], ignore_index=True)
print(main_music)
print("\nDataFrame Info:")
print(main_music.dtypes)
print("\nShape of first piano roll:", main_music['song'][0].shape if not main_music.empty else "No data")
print(main_music.head())
```

song composer_id
0 [[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,... 1

```
DataFrame Info:
song          object
composer_id   object
dtype: object
```

Shape of first piano roll: (128, 128)

song composer_id

0 [[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...

1

```
#define function to evaluate final model
#function created by LLM (xAI, 2025)
def evaluate_model(model, X_test, y_test, model_name):
    y_pred = model.predict(X_test)
    y_pred_classes = np.argmax(y_pred, axis=1)
    y_test_classes = np.argmax(y_test, axis=1)
    accuracy = accuracy_score(y_test_classes, y_pred_classes)
    precision = precision_score(y_test_classes, y_pred_classes, average='weighted')
    recall = recall_score(y_test_classes, y_pred_classes, average='weighted')
    print(f"{model_name} Performance:")
    print(f"Accuracy: {accuracy:.4f}")
    print(f"Precision: {precision:.4f}")
    print(f"Recall: {recall:.4f}")
    print()
    return accuracy, precision, recall

#establish data file path, specifiy piano roll hyper param, initialize list for
#data, iterate through directory and convert midi files to piano rolls
#bach data
data_dir = '/content/drive/My Drive/Bach'
data = []
fs = 100
fixed_length = 4000
compid = 1
for file_name in os.listdir(data_dir):
    midi_path = os.path.join(data_dir, file_name)
    piano_roll = get_piano_roll(midi_path, fs, fixed_length)
    if piano_roll is not None:
        composer_id = compid
        data.append({'song': piano_roll, 'composer_id': composer_id})

#establish data file path, specifiy piano roll hyper param, initialize list for
#data, iterate through directory and convert midi files to piano rolls
#beethoven data
data_dir = '/content/drive/My Drive/Beethoven'
fs = 100
fixed_length = 4000
compid = 2
max_entries = 20
for file_name in os.listdir(data_dir):
    midi_path = os.path.join(data_dir, file_name)
    piano_roll = get_piano_roll(midi_path, fs, fixed_length)
    if piano_roll is not None:
        composer_id = compid
        data.append({'song': piano_roll, 'composer_id': composer_id})
    if len(data) >= max_entries: # Stop after collecting 10 valid entries
        break

#establish data file path, specifiy piano roll hyper param, initialize list for
#data, iterate through directory and convert midi files to piano rolls
#chopin data
data_dir = '/content/drive/My Drive/Chopin'
fs = 100
fixed_length = 4000
compid = 3
max_entries = 30
for file_name in os.listdir(data_dir):
    midi_path = os.path.join(data_dir, file_name)
    piano_roll = get_piano_roll(midi_path, fs, fixed_length)
    if piano_roll is not None:
        composer_id = compid
        data.append({'song': piano_roll, 'composer_id': composer_id})
    if len(data) >= max_entries: # Stop after collecting 10 valid entries
        break

→ /usr/local/lib/python3.11/dist-packages/pretty_midi/pretty_midi.py:100: RuntimeWarning: Tempo, Key or Time signature change
warnings.warn(
```

#establish data file path, specifiy piano roll hyper param, initialize list for
#data, iterate through directory and convert midi files to piano rolls
#mozart data

```

data_dir = '/content/drive/My Drive/Mozart'
fs = 100
fixed_length = 4000
compid = 4
max_entries = 40
for file_name in os.listdir(data_dir):
    midi_path = os.path.join(data_dir, file_name)
    piano_roll = get_piano_roll(midi_path, fs, fixed_length)
    if piano_roll is not None:
        composer_id = compid
        data.append({'song': piano_roll, 'composer_id': composer_id})
    if len(data) >= max_entries: # Stop after collecting 10 valid entries
        break

```

```

#inspect dataframe
main_music = pd.DataFrame(data)

```

```

#inspect/evaluate dataframe
print(main_music)
print(main_music.dtypes)

```

		song	composer_id
0	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
1	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
2	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
3	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
4	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
5	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
6	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
7	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
8	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
9	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		1
10	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
11	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
12	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
13	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
14	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
15	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
16	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
17	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
18	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
19	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		2
20	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
21	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
22	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
23	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
24	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
25	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
26	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
27	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
28	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
29	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		3
30	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
31	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
32	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
33	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
34	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
35	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
36	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
37	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
38	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4
39	[[0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0, 0.0,...		4

```

song          object
composer_id   int64
dtype: object

```

```

#define X and Y vars for training
#one-hot encode y vars for training clarity
x = np.array(main_music['song'].tolist())
y = main_music['composer_id'].values - 1
y = to_categorical(y, num_classes=4)
#investigate format
print(y)
print(x.shape)

→ [[1. 0. 0. 0.]
 [1. 0. 0. 0.]
 [1. 0. 0. 0.]
 [1. 0. 0. 0.]]

```

```
#split train/test data
xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.2, random_state=42)

#format data for CNN expansion (additional channel for CNN format)
#both training and test data
xtraincnn = np.expand_dims(xtrain, axis=-1)
xtestcnn = np.expand_dims(xtest, axis=-1)

#initialize CNN model, sequential with multiple layers and activation functions
cnn_model = Sequential([
    Conv2D(32, (3, 3), activation='relu', input_shape=(xtrain.shape[1], xtrain.shape[2], 1)),
    MaxPooling2D((2, 2)),
    Conv2D(64, (3, 3), activation='relu'),
    MaxPooling2D((2, 2)),
    Flatten(),
    Dense(128, activation='relu'),
    Dropout(0.5),
    Dense(4, activation='softmax')
])
→ /usr/local/lib/python3.11/dist-packages/keras/src/layers/convolutional/base_conv.py:113: UserWarning: Do not pass an `input_
    super().__init__(activity_regularizer=activity_regularizer, **kwargs)

#initialize LSTM model, sequential with multiple layers and activation functions
lstm_model = Sequential([
    LSTM(128, input_shape=(xtrain.shape[1], xtrain.shape[2]), return_sequences=True),
    LSTM(64),
    Dense(64, activation='relu'),
    Dropout(0.5),
    Dense(4, activation='softmax')
])
→ /usr/local/lib/python3.11/dist-packages/keras/src/layers/rnn/rnn.py:199: UserWarning: Do not pass an `input_shape`/`input_di
    super().__init__(**kwargs)

#compile models with adam optimizer and CCentropy
#include epoch accuracy metric for real time data monitoring
cnn_model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

```
lstm_model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
cnn_model.fit(xtraincnn, ytrain, epochs=10, batch_size=32, validation_split=0.2)
lstm_model.fit(xtrain, ytrain, epochs=10, batch_size=32, validation_split=0.2)
```

→ Epoch 1/10

```
#evaluation metrics CNN
cnn_accuracy, cnn_precision, cnn_recall = evaluate_model(cnn_model, xtestcnn, ytest, "CNN Model")
```

→ 1/1 ————— 1s 529ms/step
 CNN Model Performance:
 Accuracy: 1.0000
 Precision: 1.0000
 Recall: 1.0000

```
#evaluation metrics LSTM
lstm_accuracy, lstm_precision, lstm_recall = evaluate_model(lstm_model, xtest, ytest, "LSTM Model")
```

→ 1/1 ————— 1s 800ms/step
 LSTM Model Performance:
 Accuracy: 0.0000
 Precision: 0.0000
 Recall: 0.0000

```
/usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565: UndefinedMetricWarning: Precision is ill-defined when all predictions are zero - warn_prf(average, modifier, f"{{metric.capitalize()}} is", len(result))
/usr/local/lib/python3.11/dist-packages/sklearn/metrics/_classification.py:1565: UndefinedMetricWarning: Recall is ill-defined when all labels in a class are zero - warn_prf(average, modifier, f"{{metric.capitalize()}} is", len(result))
```

```
#hyperparam adjustments and retraining for CNN:
from tensorflow.keras.layers import BatchNormalization, Activation, AveragePooling2D
```

```
cnn_model2 = Sequential([
    Conv2D(32, (3, 3), input_shape=(xtrain.shape[1], xtrain.shape[2], 1)),
    BatchNormalization(),
    Activation('relu'),
    AveragePooling2D((2, 2)),
    Conv2D(64, (3, 3)),
    BatchNormalization(),
    Activation('relu'),
    AveragePooling2D((2, 2)),
    Conv2D(128, (3, 3)),
    BatchNormalization(),
    Activation('relu'),
    AveragePooling2D((2, 2)),
    Flatten(),
    Dense(128, activation='relu'),
    Dropout(0.5),
    Dense(4, activation='softmax')
])
```

```
#adjusted epochs 10->20
cnn_model2.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
cnn_model2.fit(xtraincnn, ytrain, epochs=20, batch_size=32, validation_split=0.2)
cnn_accuracy, cnn_precision, cnn_recall = evaluate_model(cnn_model2, xtestcnn, ytest, "CNN Model 2")
```

```
#hyperparam adjustments and retraining for LSTM:
```

```
lstm_model2 = Sequential([
    LSTM(128, input_shape=(xtrain.shape[1], xtrain.shape[2]),
         return_sequences=True, dropout=0.3),
    LSTM(128, dropout=0.3),
    Dense(64, activation='relu'),
    Dropout(0.5),
    Dense(4, activation='softmax')
])
```

```
#adjusted epochs 10->20
lstm_model2.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
lstm_model2.fit(xtrain, ytrain, epochs=20, batch_size=32, validation_split=0.2)
lstm_accuracy, lstm_precision, lstm_recall = evaluate_model(lstm_model2, xtest, ytest, "LSTM Model 2")
```

