

task 3

May 12, 2025

```
[3]: # Necessary libraries
import h5py
import numpy as np
import matplotlib.pyplot as plt
import math
from collections import Counter
import tensorflow as tf
from sklearn.model_selection import train_test_split
from sklearn.utils import shuffle
from sklearn.preprocessing import LabelEncoder
from sklearn.metrics import confusion_matrix, ConfusionMatrixDisplay

from tensorflow.keras.utils import to_categorical

import librosa, librosa.display, numpy as np, matplotlib.pyplot as plt, pandas_
↪as pd
from tensorflow.keras.models import load_model
import pathlib
```

```
[6]: # loading the file
data = h5py.File('bird_spectrograms.hdf5', 'r')

bird_names = list(data.keys())
print(bird_names)

#labeling them for readability
names = {
    'amecro': 'American Crow',
    'amerob': 'American Robin',
    'bewwre': 'Bewicks Wren',
    'bkcchi': 'Black capped Chickadee',
    'daejun': 'Dark eyed Junco',
    'houfin': 'House Finch',
    'houspa': 'House Sparrow',
    'norfli': 'Northern Flicker',
    'rewbla': 'Red winged Blackbird',
    'sonspa': 'Song Sparrow',
```

```

        'spotow': 'Spotted Towhee',
        'whcspa': 'White crowned Sparrow'
    }
    print(names)

    print("All bird species:")
    all_birds = [names[code] for code in bird_names]
    for bird in all_birds:
        print(bird)

    for code in bird_names:
        shape = data[code].shape
        print(f"{code:7s} {names.get(code, 'Unknown')}:25s} {shape}")

    n_cols, n_rows = 3, math.ceil(len(bird_names) / 3)
    fig, axes = plt.subplots(n_rows, n_cols, figsize=(12, 4 * n_rows))
    axes = axes.flatten()

    for i, code in enumerate(bird_names):
        spectro = data[code][:, :, 0]
        axes[i].imshow(spectro, aspect="auto", origin="lower", cmap="gray")
        axes[i].set_title(f"{code} | {names.get(code, '')}", fontsize=9)
        axes[i].set_xlabel("Time bins")
        axes[i].set_ylabel("Frequency bins")

    for j in range(i + 1, len(axes)):
        axes[j].axis("off")

    plt.tight_layout()
    plt.show()

```

```

['amecro', 'amerob', 'bewwre', 'bkcchi', 'daejun', 'houfin', 'houspa', 'norfli',
'rewbla', 'sonspa', 'spotow', 'whcspa']
{'amecro': 'American Crow', 'amerob': 'American Robin', 'bewwre': 'Bewicks
Wren', 'bkcchi': 'Black capped Chickadee', 'daejun': 'Dark eyed Junco',
'houfin': 'House Finch', 'houspa': 'House Sparrow', 'norfli': 'Northern
Flicker', 'rewbla': 'Red winged Blackbird', 'sonspa': 'Song Sparrow', 'spotow':
'Spotted Towhee', 'whcspa': 'White crowned Sparrow'}

```

All bird species:

American Crow

American Robin

Bewicks Wren

Black capped Chickadee

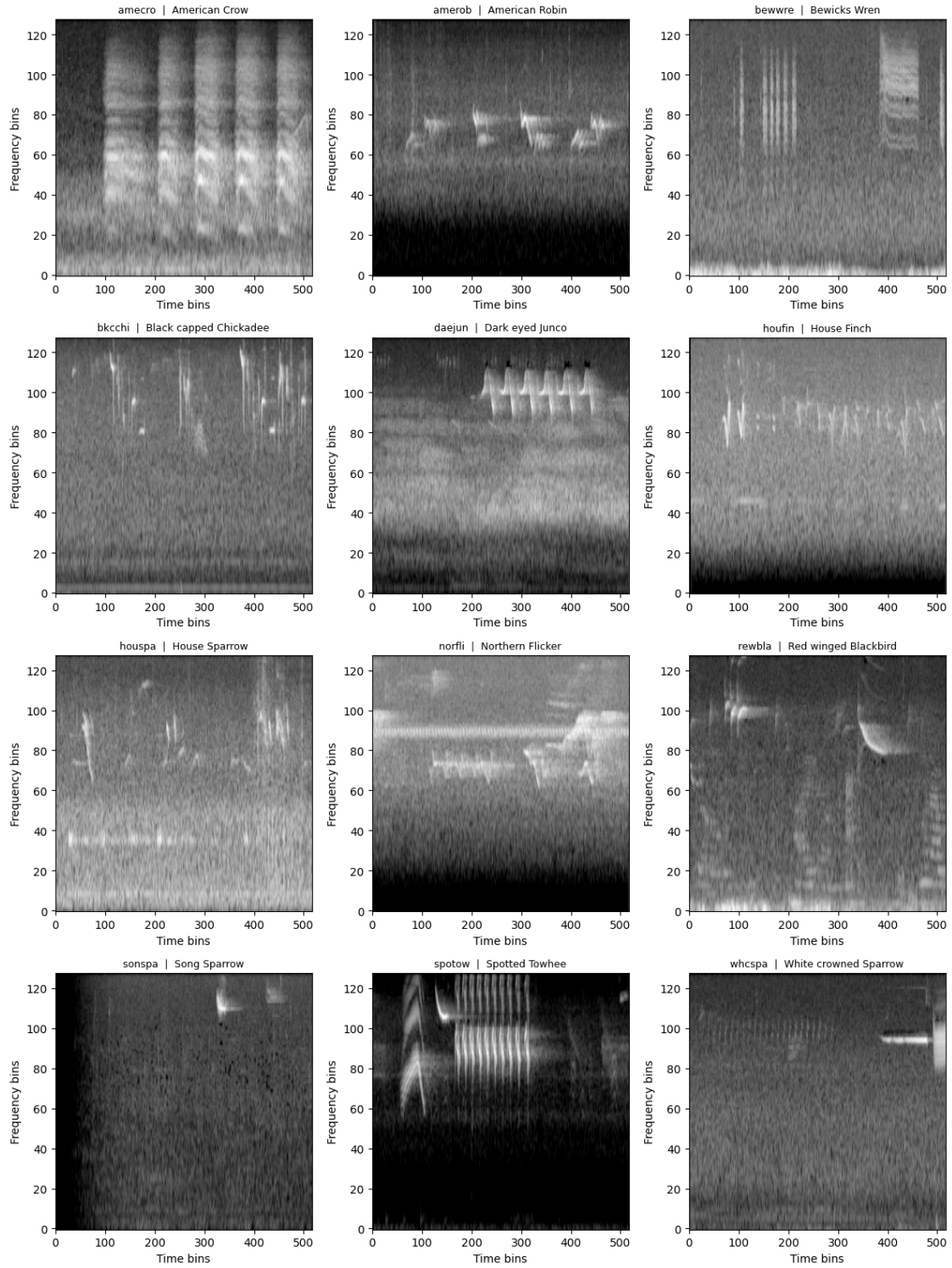
Dark eyed Junco

House Finch

House Sparrow

Northern Flicker

Red winged Blackbird
 Song Sparrow
 Spotted Towhee
 White crowned Sparrow
 amecro American Crow (128, 517, 66)
 amerob American Robin (128, 517, 172)
 bewwre Bewicks Wren (128, 517, 144)
 bkcchi Black capped Chickadee (128, 517, 45)
 daejun Dark eyed Junco (128, 517, 125)
 houfin House Finch (128, 517, 84)
 houspa House Sparrow (128, 517, 630)
 norfli Northern Flicker (128, 517, 37)
 rewbla Red winged Blackbird (128, 517, 187)
 sonspa Song Sparrow (128, 517, 263)
 spotow Spotted Towhee (128, 517, 137)
 whcspa White crowned Sparrow (128, 517, 91)



```
[8]: import os
import h5py
```

```

import numpy as np
import librosa
import matplotlib.pyplot as plt
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense,
↳ Dropout
from tensorflow.keras.utils import to_categorical
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.models import load_model

spectro_path = '/Users/alekh/Desktop/birds/bird_spectrograms.hdf5'
audios = {
    "test1": "/Users/alekh/Desktop/birds/test1.wav",
    "test2": "/Users/alekh/Desktop/birds/test2.wav",
    "test3": "/Users/alekh/Desktop/birds/test3.wav"
}
time_windows = {
    "test1": (15, 17),
    "test2": (0, 2),
    "test3": (2, 4)
}

spectrogram_list = []
labels = []
with h5py.File(spectro_path, 'r') as f:
    species = list(f.keys())
    for key in species:
        data = f[key][...]
        if data.ndim == 3:
            for i in range(data.shape[2]):
                spectrogram_list.append(data[:, :, i])
                labels.append(key)
        elif data.ndim == 2:
            spectrogram_list.append(data)
            labels.append(key)
        else:
            raise ValueError(f"ndim={data.ndim}")

spectrogram_list = np.array(spectrogram_list)
labels = np.array(labels)
label_encoder = LabelEncoder().fit(labels)
labels_encoded = label_encoder.transform(labels)
labels_onehot = to_categorical(labels_encoded)
mel_bins, time_frames = spectrogram_list.shape[1], spectrogram_list.shape[2]
spectrogram_list = spectrogram_list.reshape(-1, mel_bins, time_frames, 1).
↳ astype(np.float32) / 255.0

```

```

model_three = Sequential([
    Conv2D(32, (3,3), activation='relu', input_shape=(mel_bins, time_frames, 1)),
    MaxPooling2D((2,2)),
    Conv2D(64, (3,3), activation='relu'),
    MaxPooling2D((2,2)),
    Conv2D(128, (3,3), activation='relu'),
    MaxPooling2D((2,2)),
    Flatten(),
    Dense(128, activation='relu'),
    Dropout(0.5),
    Dense(len(species), activation='softmax')
])
model_three.compile(optimizer='adam', loss='categorical_crossentropy',
    metrics=['accuracy'])
model_three.fit(spectrogram_list, labels_onehot, epochs=5, batch_size=32)

test_data = []
for name, path in audios.items():
    audio, sr = librosa.load(path, sr=22050)
    start, end = time_windows[name]
    clip = audio[int(start*sr):int(end*sr)]
    S = librosa.feature.melspectrogram(y=clip, sr=sr, n_fft=2048,
    hop_length=512, n_mels=mel_bins)
    S_db = librosa.power_to_db(S, ref=np.max)
    if S_db.shape[1] < time_frames:
        S_db = np.pad(S_db, ((0,0),(0, time_frames - S_db.shape[1])),
    'constant')
    else:
        S_db = S_db[:, :time_frames]
    test_data.append(S_db)

test_data = np.array(test_data).reshape(-1, mel_bins, time_frames, 1).astype(np.
    float32) / 255.0
test_predictions = model_three.predict(test_data, verbose=0)
test_one = np.argmax(test_predictions, axis=1)
test_class = label_encoder.inverse_transform(test_one)
for i, pred in enumerate(test_class, 1):
    print(f"Test Spectrogram {i} predicted as {pred}")

```

Epoch 1/5

62/62 [=====] - 186s 3s/step - loss: 2.3170 - accuracy: 0.2988

Epoch 2/5

62/62 [=====] - 176s 3s/step - loss: 2.2268 - accuracy: 0.3175

Epoch 3/5

```
62/62 [=====] - 185s 3s/step - loss: 2.2217 - accuracy:
0.3170
Epoch 4/5
62/62 [=====] - 178s 3s/step - loss: 2.1683 - accuracy:
0.3155
Epoch 5/5
62/62 [=====] - 185s 3s/step - loss: 2.1283 - accuracy:
0.3155
Test Spectrogram 1 predicted as houspa
Test Spectrogram 2 predicted as houspa
Test Spectrogram 3 predicted as houspa
```

```
[22]: import matplotlib.pyplot as plt

for idx, name in enumerate(audios):
    probs = test_predictions[idx]
    plt.figure(figsize=(6,4))
    plt.bar(label_encoder.classes_, probs)
    plt.xticks(rotation=45, ha='right')
    plt.title(f"{name} softmax probabilities")
    plt.tight_layout()
    plt.show()
```

