

task 3

May 12, 2025

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[1]: # 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
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[2]: # 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',
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    '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()

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['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'}

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All bird species:

American Crow

American Robin

Bewicks Wren

Black capped Chickadee

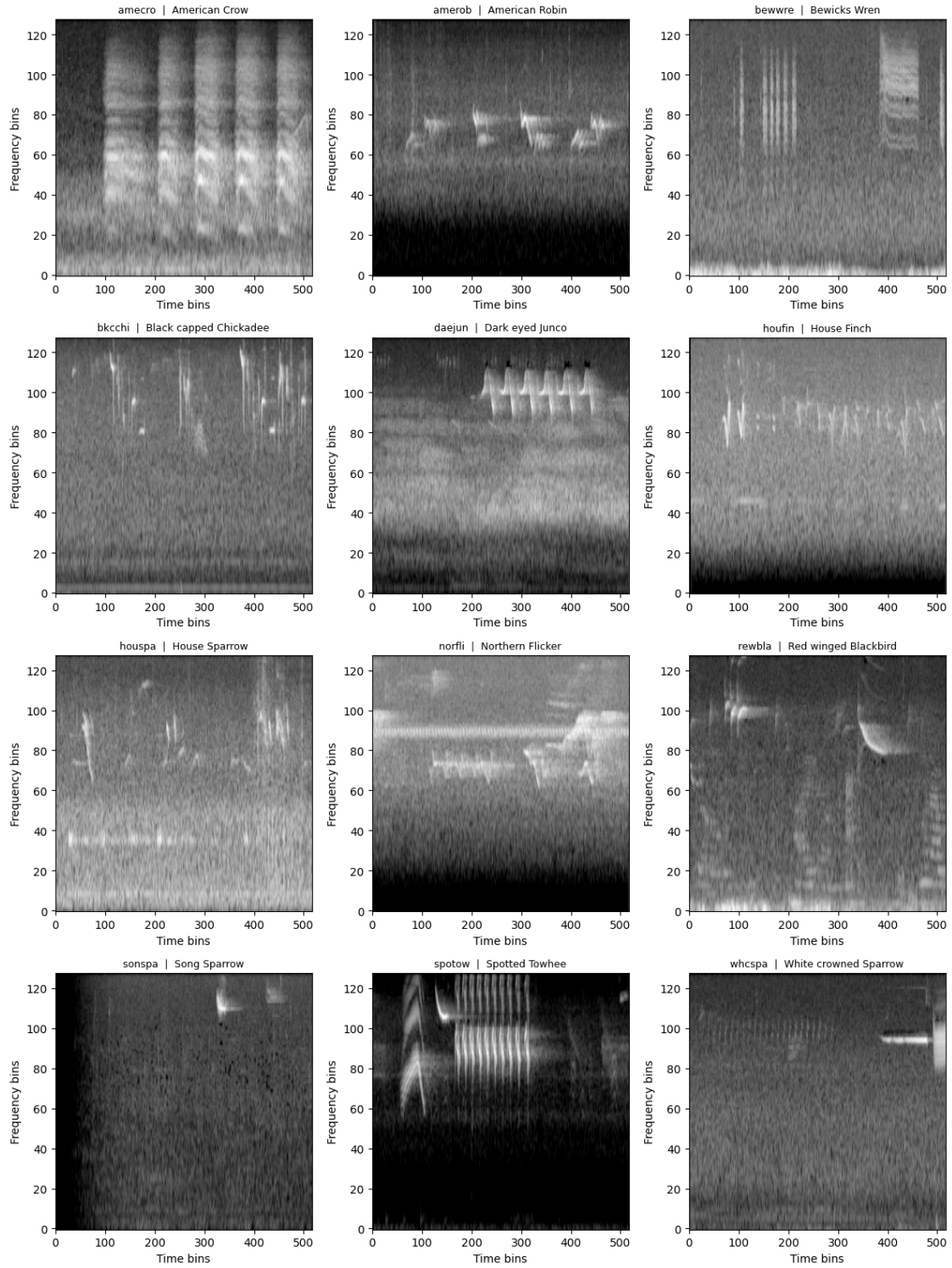
Dark eyed Junco

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)



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[3]: import os
import h5py
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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

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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}")

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Epoch 1/5

62/62 [=====] - 97s 2s/step - loss: 2.2666 - accuracy: 0.3029

Epoch 2/5

62/62 [=====] - 88s 1s/step - loss: 2.2182 - accuracy: 0.3165

Epoch 3/5

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62/62 [=====] - 95s 2s/step - loss: 2.1717 - accuracy:
0.3150
Epoch 4/5
62/62 [=====] - 89s 1s/step - loss: 2.1132 - accuracy:
0.3246
Epoch 5/5
62/62 [=====] - 89s 1s/step - loss: 2.0288 - accuracy:
0.3523
Test Spectrogram 1 predicted as sonspa
Test Spectrogram 2 predicted as sonspa
Test Spectrogram 3 predicted as sonspa

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[4]: import numpy as np
import pandas as pd
import librosa

bird_codes = label_encoder.classes_
bird_names = [names[c] for c in bird_codes]

hop = int(time_frames * 0.5)
rows = []

for clip_name, path in audios.items():
    y, sr = librosa.load(path, sr=22050)
    s, e = time_windows[clip_name]
    seg = y[int(s*sr):int(e*sr)]

    S = librosa.feature.melspectrogram(
        y=seg, sr=sr,
        n_fft=2048, hop_length=512, n_mels=mel_bins
    )
    S_db = librosa.power_to_db(S, ref=np.max)
    S_norm = np.clip((S_db + 80) / 80, 0, 1)

    n_cols = S_norm.shape[1]
    if n_cols < time_frames:
        pad = time_frames - n_cols
        S_norm = np.pad(S_norm, ((0,0),(0,pad)), mode='constant')
        n_cols = time_frames

    starts = range(0, n_cols - time_frames + 1, hop)
    patches = np.stack([S_norm[:, i:i+time_frames] for i in starts])[..., None]

    P = model_three.predict(patches, verbose=0)
    average = P.mean(axis=0)
    top_three = np.argsort(average)[-3:][::-1]

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rows.append({
    "clip": clip_name,
    "top1_species": bird_names[top_three[0]],
    "top1_probability": float(average[top_three[0]]),
    "top2_species": bird_names[top_three[1]],
    "top2_probabaility": float(average[top_three[1]]),
    "top3_species": bird_names[top_three[2]],
    "top3_probabaility": float(average[top_three[2]]),
})

df_top3 = pd.DataFrame(rows)
print("The top 3 external clip predictions")
display(df_top3)

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The top 3 external clip predictions

	clip	top1_species	top1_probability	top2_species	top2_probabaility \
0	test1	House Sparrow	0.805277	American Robin	0.126353
1	test2	House Sparrow	0.643778	American Robin	0.204123
2	test3	House Sparrow	0.853333	American Robin	0.108637

	top3_species	top3_probabaility
0	Song Sparrow	0.023167
1	Song Sparrow	0.053612
2	Song Sparrow	0.013637