# Music\_Genre\_Analysis

#### December 7, 2023

```
[1]: # Mount Google Drive to the Colab VM.
     from google.colab import drive
     drive.mount('/content/drive')
     FOLDERNAME = "COMPSCI 682/compsci-682-project"
     assert FOLDERNAME is not None, "[!] Enter the foldername."
     import sys
     sys.path.append('/content/drive/My Drive/{}'.format(FOLDERNAME))
     # %cd /content/drive/My\ Drive/$FOLDERNAME/datasets
     # !wget https://raw.githubusercontent.com/coreyker/dnn-mgr/master/gtzan/
      \hookrightarrow train\_filtered.txt
     # !wget https://raw.githubusercontent.com/coreyker/dnn-mgr/master/gtzan/
      \hookrightarrow valid\_filtered.txt
     # !wget https://raw.githubusercontent.com/coreyker/dnn-mgr/master/gtzan/
      \rightarrow test\ filtered.txt
     %cd /content/drive/My\ Drive/$FOLDERNAME
    Mounted at /content/drive
    /content/drive/My Drive/COMPSCI 682/compsci-682-project
```

```
[2]: # Setting up locale
import locale
locale.getpreferredencoding = lambda: "UTF-8"
```

```
[3]: %load_ext autoreload %autoreload 2

from src.configuration import *
from src.dataset import *
from src.networks import *
from src.training import *
from src.evaluation import *
from src.utils import *
import src.preprocess as pp
```

```
[4]: # Setting up environment for using GPU or CPU as per availability dtype = torch.float32
```

Using device: cuda

```
[5]: # Preprocess the dataset to extract and save features
if pp.are_features_extracted("datasets/features"):
    print("Features already extracted. Proceed further...")
else:
    print("Features absent. Processing...")
    pp.generate_spectrograms(
        data_path="datasets/genres",
        save_path="datasets/features"
    )
    print("Features extracted. Proceed further...")

# Note that one song in GTZAN dataset contains data in unknown/corrupt format:
    www will delete if this exists.
if os.path.exists("datasets/genres/jazz/jazz.00054.wav"):
    os.remove("datasets/genres/jazz/jazz.00054.wav")
```

Features already extracted. Proceed further...

```
[6]: # Log details about the datasets (total 929 instances of well-split manually.
     →annotated data from coreyker/dnn-mgr/)
     train_dataset = GTZANFeatureDataset(split="train")
     valid_dataset = GTZANFeatureDataset(split="valid")
     test_dataset = GTZANFeatureDataset(split="test")
     print(f"Number of train instances: {len(train_dataset)} ({len(train_dataset) / ___
      4929 * 100:.2f}%)") # Expect 442 (47.58%)
     print(f"Number of valid instances: {len(valid_dataset)} ({len(valid_dataset) / ___
     929 * 100:.2f%)") # Expect 197 (21.21%)
     print(f"Number of test instances: {len(test_dataset )} ({len(test_dataset ) / ___
      929 * 100:.2f}%)") # Expect 290 (31.22%)
     print("Number of class labels: ", len(CLASS_LABELS)) # Expect 10
     print("Values of Class labels: ", CLASS_LABELS) # Expect ['blues', 'classical', L
      -'country', 'disco', 'hiphop', 'jazz', 'metal', 'pop', 'reggae', 'rock']
    Number of train instances: 442 (47.58%)
    Number of valid instances: 197 (21.21%)
```

```
Number of train instances: 442 (47.58%)

Number of valid instances: 197 (21.21%)

Number of test instances: 290 (31.22%)

Number of class labels: 10

Values of Class labels: ['blues', 'classical', 'country', 'disco', 'hiphop', 'jazz', 'metal', 'pop', 'reggae', 'rock']
```

```
[9]: # Initialize the data loaders for mel spectrograms
train_loader = get_dataloader(split='train')
```

```
valid_loader = get_dataloader(split='valid')
    test_loader = get_dataloader(split='test')
    train_melspectrogram, train_genre = next(iter(train_loader))
    test_melspectrogram, test_genre = next(iter(test_loader))
    print('Training Mini-Batch Tensor Shape: %s' % str(train_melspectrogram.shape))
    print('Test Mini-Batch Tensor Shape: %s' % str(test_melspectrogram.shape))
    Training Mini-Batch Tensor Shape: torch.Size([32, 3, 369, 496])
    Test Mini-Batch Tensor Shape: torch.Size([32, 3, 369, 496])
[14]: speccnn_model = SpectrogramCNN() # Custom model built by us
    train(speccnn_model, "best_speccnn_model_melspectrogram.ckpt", train_loader,__
      ⇔valid_loader, device, num_epochs=50, lr=1e-3, debug = True)
    Epoch: [1/50], Train loss: 2.3869
    Epoch: [1/50], Valid loss: 2.3214, Valid accuracy: 0.1168
    _____
    Saving the best model at 0 epochs!
    _____
    Epoch: [2/50], Train loss: 2.1533
    Epoch: [2/50], Valid loss: 2.2430, Valid accuracy: 0.1827
    _____
    Saving the best model at 1 epochs!
    _____
    Epoch: [3/50], Train loss: 2.0085
    Epoch: [3/50], Valid loss: 2.0722, Valid accuracy: 0.2589
    _____
    Saving the best model at 2 epochs!
    _____
    Epoch: [4/50], Train loss: 1.8515
    Epoch: [4/50], Valid loss: 1.9482, Valid accuracy: 0.2792
    _____
    Saving the best model at 3 epochs!
    -----
    Epoch: [5/50], Train loss: 1.7510
    Epoch: [5/50], Valid loss: 1.8721, Valid accuracy: 0.2995
    ______
    Saving the best model at 4 epochs!
    Epoch: [6/50], Train loss: 1.7109
    Epoch: [6/50], Valid loss: 1.8246, Valid accuracy: 0.3503
    Saving the best model at 5 epochs!
    _____
    Epoch: [7/50], Train loss: 1.5847
```

Epoch: [7/50], Valid loss: 1.8030, Valid accuracy: 0.3503 \_\_\_\_\_ Saving the best model at 6 epochs! Epoch: [8/50], Train loss: 1.5075 Epoch: [8/50], Valid loss: 1.8001, Valid accuracy: 0.3503 Saving the best model at 7 epochs! -----Epoch: [9/50], Train loss: 1.4672 Epoch: [9/50], Valid loss: 1.7914, Valid accuracy: 0.4162 \_\_\_\_\_ Saving the best model at 8 epochs! \_\_\_\_\_ Epoch: [10/50], Train loss: 1.3864 Epoch: [10/50], Valid loss: 1.7558, Valid accuracy: 0.4213 \_\_\_\_\_ Saving the best model at 9 epochs! \_\_\_\_\_ Epoch: [11/50], Train loss: 1.3149 Epoch: [11/50], Valid loss: 1.7544, Valid accuracy: 0.4365 -----Saving the best model at 10 epochs! Epoch: [12/50], Train loss: 1.2878 Epoch: [12/50], Valid loss: 1.7501, Valid accuracy: 0.3909 \_\_\_\_\_ Saving the best model at 11 epochs! \_\_\_\_\_ Epoch: [13/50], Train loss: 1.2474 Epoch: [13/50], Valid loss: 1.7368, Valid accuracy: 0.4518 \_\_\_\_\_ Saving the best model at 12 epochs! \_\_\_\_\_ Epoch: [14/50], Train loss: 1.2404 Epoch: [14/50], Valid loss: 1.7093, Valid accuracy: 0.4112 \_\_\_\_\_ Saving the best model at 13 epochs! Epoch: [15/50], Train loss: 1.1439 Epoch: [15/50], Valid loss: 1.6229, Valid accuracy: 0.4772 \_\_\_\_\_ Saving the best model at 14 epochs! \_\_\_\_\_\_ Epoch: [16/50], Train loss: 1.1385 Epoch: [16/50], Valid loss: 1.6777, Valid accuracy: 0.4112 -----

```
Epoch: [17/50], Train loss: 1.1260
Epoch: [17/50], Valid loss: 1.7568, Valid accuracy: 0.4365
_____
Epoch: [18/50], Train loss: 1.1500
Epoch: [18/50], Valid loss: 1.7880, Valid accuracy: 0.3807
_____
Epoch: [19/50], Train loss: 1.0458
Epoch: [19/50], Valid loss: 1.5949, Valid accuracy: 0.4619
_____
Saving the best model at 18 epochs!
-----
Epoch: [20/50], Train loss: 0.9913
Epoch: [20/50], Valid loss: 1.6537, Valid accuracy: 0.4873
-----
_____
Epoch: [21/50], Train loss: 0.9270
Epoch: [21/50], Valid loss: 1.5872, Valid accuracy: 0.4569
_____
Saving the best model at 20 epochs!
_____
Epoch: [22/50], Train loss: 0.9402
Epoch: [22/50], Valid loss: 1.5329, Valid accuracy: 0.5127
_____
Saving the best model at 21 epochs!
_____
Epoch: [23/50], Train loss: 0.8958
Epoch: [23/50], Valid loss: 1.4871, Valid accuracy: 0.5330
______
Saving the best model at 22 epochs!
_____
Epoch: [24/50], Train loss: 0.9156
Epoch: [24/50], Valid loss: 1.5683, Valid accuracy: 0.4975
_____
_____
Epoch: [25/50], Train loss: 0.9326
Epoch: [25/50], Valid loss: 1.5265, Valid accuracy: 0.5025
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_____
Epoch: [26/50], Train loss: 0.8228
Epoch: [26/50], Valid loss: 1.5199, Valid accuracy: 0.5127
_____
_____
Epoch: [27/50], Train loss: 0.8579
Epoch: [27/50], Valid loss: 1.5397, Valid accuracy: 0.4822
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Epoch: [28/50], Train loss: 0.8008
Epoch: [28/50], Valid loss: 1.7213, Valid accuracy: 0.4061
_____
Epoch: [29/50], Train loss: 0.7935
Epoch: [29/50], Valid loss: 1.5113, Valid accuracy: 0.5127
_____
Epoch: [30/50], Train loss: 0.7439
Epoch: [30/50], Valid loss: 1.6411, Valid accuracy: 0.4772
_____
_____
Epoch: [31/50], Train loss: 0.7535
Epoch: [31/50], Valid loss: 1.5133, Valid accuracy: 0.5025
_____
_____
Epoch: [32/50], Train loss: 0.7146
Epoch: [32/50], Valid loss: 1.4982, Valid accuracy: 0.5127
_____
_____
Epoch: [33/50], Train loss: 0.7296
Epoch: [33/50], Valid loss: 1.5069, Valid accuracy: 0.5279
_____
_____
Epoch: [34/50], Train loss: 0.6943
Epoch: [34/50], Valid loss: 1.5344, Valid accuracy: 0.5228
_____
_____
Epoch: [35/50], Train loss: 0.6830
Epoch: [35/50], Valid loss: 1.5097, Valid accuracy: 0.5127
______
-----
Epoch: [36/50], Train loss: 0.6574
Epoch: [36/50], Valid loss: 1.5780, Valid accuracy: 0.5330
_____
_____
Epoch: [37/50], Train loss: 0.7001
Epoch: [37/50], Valid loss: 1.6931, Valid accuracy: 0.4315
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Epoch: [38/50], Train loss: 0.7511
Epoch: [38/50], Valid loss: 1.7850, Valid accuracy: 0.4213
_____
_____
Epoch: [39/50], Train loss: 0.6465
Epoch: [39/50], Valid loss: 1.5255, Valid accuracy: 0.4873
```

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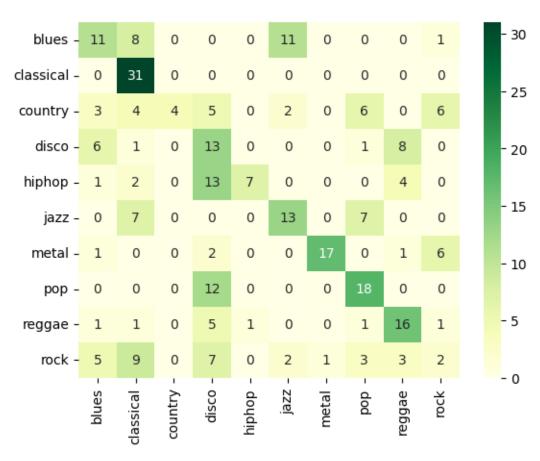
```
Epoch: [40/50], Train loss: 0.6158
   Epoch: [40/50], Valid loss: 1.5328, Valid accuracy: 0.5178
   ______
   Epoch: [41/50], Train loss: 0.5935
   Epoch: [41/50], Valid loss: 1.4699, Valid accuracy: 0.5178
   Saving the best model at 40 epochs!
   -----
   Epoch: [42/50], Train loss: 0.5237
   Epoch: [42/50], Valid loss: 1.5705, Valid accuracy: 0.4873
   _____
   _____
   Epoch: [43/50], Train loss: 0.5755
   Epoch: [43/50], Valid loss: 1.7177, Valid accuracy: 0.4822
   ______
   _____
   Epoch: [44/50], Train loss: 0.4974
   Epoch: [44/50], Valid loss: 1.6607, Valid accuracy: 0.5178
   _____
   _____
   Epoch: [45/50], Train loss: 0.5514
   Epoch: [45/50], Valid loss: 1.6658, Valid accuracy: 0.5127
   _____
   _____
   Epoch: [46/50], Train loss: 0.5526
   Epoch: [46/50], Valid loss: 1.7019, Valid accuracy: 0.4670
   _____
   _____
   Epoch: [47/50], Train loss: 0.5831
   Epoch: [47/50], Valid loss: 1.7592, Valid accuracy: 0.4924
   _____
   Epoch: [48/50], Train loss: 0.5194
   Epoch: [48/50], Valid loss: 1.5291, Valid accuracy: 0.5076
   _____
   _____
   Epoch: [49/50], Train loss: 0.4708
   Epoch: [49/50], Valid loss: 1.8334, Valid accuracy: 0.4467
   -----
   Epoch: [50/50], Train loss: 0.4783
   Epoch: [50/50], Valid loss: 1.6373, Valid accuracy: 0.5381
   _____
[15]: evaluate(speccnn_model, "best_speccnn_model_melspectrogram.ckpt", test_loader,__
     ⊶device)
```

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#### Model at best\_speccnn\_model\_melspectrogram.ckpt:

\_\_\_\_\_

Accuracy: 0.4552 Precision: 0.5522 Recall: 0.4585 F1 Score: 0.4379



[16]: resnet\_model = ResNet18() # ResNet-18
train(resnet\_model, "best\_resnet\_model\_melspectrogram.ckpt", train\_loader,

→valid\_loader, device, num\_epochs=50, lr=1e-3, debug = True)

Downloading: "https://download.pytorch.org/models/resnet18-f37072fd.pth" to /root/.cache/torch/hub/checkpoints/resnet18-f37072fd.pth 100%| | 44.7M/44.7M [00:00<00:00, 145MB/s]

Epoch: [1/50], Train loss: 2.3955

Epoch: [1/50], Valid loss: 2.3073, Valid accuracy: 0.0812

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Saving the best model at 0 epochs!

Epoch: [2/50], Train loss: 2.1335 Epoch: [2/50], Valid loss: 2.1240, Valid accuracy: 0.2843 Saving the best model at 1 epochs! \_\_\_\_\_ Epoch: [3/50], Train loss: 1.9365 Epoch: [3/50], Valid loss: 1.9931, Valid accuracy: 0.2843 -----Saving the best model at 2 epochs! \_\_\_\_\_ Epoch: [4/50], Train loss: 1.7967 Epoch: [4/50], Valid loss: 1.9101, Valid accuracy: 0.3350 \_\_\_\_\_ Saving the best model at 3 epochs! -----Epoch: [5/50], Train loss: 1.6978 Epoch: [5/50], Valid loss: 1.8816, Valid accuracy: 0.3198 -----Saving the best model at 4 epochs! \_\_\_\_\_ Epoch: [6/50], Train loss: 1.6063 Epoch: [6/50], Valid loss: 1.7941, Valid accuracy: 0.3756 Saving the best model at 5 epochs! \_\_\_\_\_ Epoch: [7/50], Train loss: 1.5040 Epoch: [7/50], Valid loss: 1.7717, Valid accuracy: 0.4112 -----Saving the best model at 6 epochs! -----Epoch: [8/50], Train loss: 1.4448 Epoch: [8/50], Valid loss: 1.7084, Valid accuracy: 0.4162 -----Saving the best model at 7 epochs! \_\_\_\_\_ Epoch: [9/50], Train loss: 1.3726 Epoch: [9/50], Valid loss: 1.6656, Valid accuracy: 0.4213 -----Saving the best model at 8 epochs! \_\_\_\_\_ Epoch: [10/50], Train loss: 1.3262 Epoch: [10/50], Valid loss: 1.6611, Valid accuracy: 0.4315 \_\_\_\_\_\_ Saving the best model at 9 epochs! \_\_\_\_\_ Epoch: [11/50], Train loss: 1.2689 Epoch: [11/50], Valid loss: 1.6475, Valid accuracy: 0.3858

```
Saving the best model at 10 epochs!
_____
Epoch: [12/50], Train loss: 1.2254
Epoch: [12/50], Valid loss: 1.6055, Valid accuracy: 0.4365
_____
Saving the best model at 11 epochs!
_____
Epoch: [13/50], Train loss: 1.1951
Epoch: [13/50], Valid loss: 1.5923, Valid accuracy: 0.4569
_____
Saving the best model at 12 epochs!
_____
Epoch: [14/50], Train loss: 1.1423
Epoch: [14/50], Valid loss: 1.5881, Valid accuracy: 0.4467
-----
Saving the best model at 13 epochs!
Epoch: [15/50], Train loss: 1.1189
Epoch: [15/50], Valid loss: 1.5715, Valid accuracy: 0.4975
_____
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.0586
Epoch: [16/50], Valid loss: 1.5510, Valid accuracy: 0.4721
_____
Saving the best model at 15 epochs!
_____
Epoch: [17/50], Train loss: 1.0420
Epoch: [17/50], Valid loss: 1.5534, Valid accuracy: 0.4315
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Epoch: [18/50], Train loss: 1.0187
Epoch: [18/50], Valid loss: 1.5501, Valid accuracy: 0.4619
   -----
Saving the best model at 17 epochs!
  -----
Epoch: [19/50], Train loss: 0.9717
Epoch: [19/50], Valid loss: 1.5272, Valid accuracy: 0.4721
 -----
Saving the best model at 18 epochs!
_____
Epoch: [20/50], Train loss: 0.9559
Epoch: [20/50], Valid loss: 1.5191, Valid accuracy: 0.5127
_____
Saving the best model at 19 epochs!
-----
Epoch: [21/50], Train loss: 0.9529
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Epoch: [21/50], Valid loss: 1.5105, Valid accuracy: 0.4670
_____
Saving the best model at 20 epochs!
Epoch: [22/50], Train loss: 0.9311
Epoch: [22/50], Valid loss: 1.5138, Valid accuracy: 0.5127
_____
Epoch: [23/50], Train loss: 0.9178
Epoch: [23/50], Valid loss: 1.5460, Valid accuracy: 0.4772
_____
_____
Epoch: [24/50], Train loss: 0.9017
Epoch: [24/50], Valid loss: 1.5230, Valid accuracy: 0.4822
_____
_____
Epoch: [25/50], Train loss: 0.8748
Epoch: [25/50], Valid loss: 1.4834, Valid accuracy: 0.4975
_____
Saving the best model at 24 epochs!
_____
Epoch: [26/50], Train loss: 0.8338
Epoch: [26/50], Valid loss: 1.5083, Valid accuracy: 0.4975
_____
Epoch: [27/50], Train loss: 0.8477
Epoch: [27/50], Valid loss: 1.4804, Valid accuracy: 0.4975
_____
Saving the best model at 26 epochs!
______
Epoch: [28/50], Train loss: 0.8290
Epoch: [28/50], Valid loss: 1.5165, Valid accuracy: 0.4772
_____
Epoch: [29/50], Train loss: 0.7977
Epoch: [29/50], Valid loss: 1.4653, Valid accuracy: 0.5076
  -----
Saving the best model at 28 epochs!
_____
Epoch: [30/50], Train loss: 0.7973
Epoch: [30/50], Valid loss: 1.5164, Valid accuracy: 0.4975
_____
_____
Epoch: [31/50], Train loss: 0.7552
Epoch: [31/50], Valid loss: 1.4579, Valid accuracy: 0.4924
  -----
Saving the best model at 30 epochs!
```

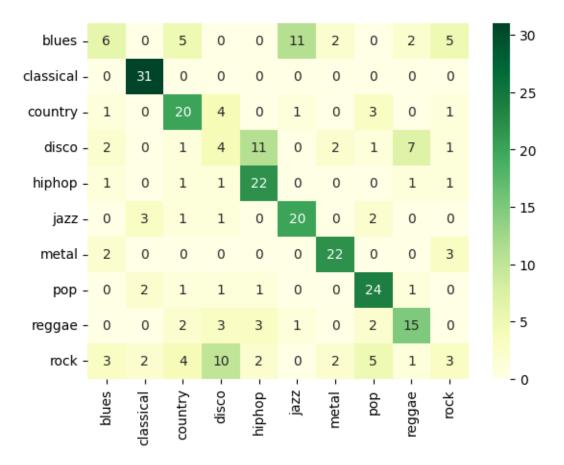
\_\_\_\_\_

```
Epoch: [32/50], Train loss: 0.7621
Epoch: [32/50], Valid loss: 1.5094, Valid accuracy: 0.5076
_____
Epoch: [33/50], Train loss: 0.7266
Epoch: [33/50], Valid loss: 1.4865, Valid accuracy: 0.4822
_____
Epoch: [34/50], Train loss: 0.7114
Epoch: [34/50], Valid loss: 1.4782, Valid accuracy: 0.5279
_____
_____
Epoch: [35/50], Train loss: 0.7387
Epoch: [35/50], Valid loss: 1.4780, Valid accuracy: 0.4975
_____
_____
Epoch: [36/50], Train loss: 0.7001
Epoch: [36/50], Valid loss: 1.4801, Valid accuracy: 0.5076
_____
_____
Epoch: [37/50], Train loss: 0.6755
Epoch: [37/50], Valid loss: 1.4662, Valid accuracy: 0.5076
_____
_____
Epoch: [38/50], Train loss: 0.6840
Epoch: [38/50], Valid loss: 1.4776, Valid accuracy: 0.5279
_____
_____
Epoch: [39/50], Train loss: 0.6859
Epoch: [39/50], Valid loss: 1.5038, Valid accuracy: 0.4873
______
-----
Epoch: [40/50], Train loss: 0.6749
Epoch: [40/50], Valid loss: 1.4728, Valid accuracy: 0.5076
_____
_____
Epoch: [41/50], Train loss: 0.6814
Epoch: [41/50], Valid loss: 1.4639, Valid accuracy: 0.5127
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_____
Epoch: [42/50], Train loss: 0.6635
Epoch: [42/50], Valid loss: 1.5012, Valid accuracy: 0.4975
_____
_____
Epoch: [43/50], Train loss: 0.6325
Epoch: [43/50], Valid loss: 1.4611, Valid accuracy: 0.5279
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```

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Epoch: [44/50], Train loss: 0.6191
   Epoch: [44/50], Valid loss: 1.5016, Valid accuracy: 0.5025
   ______
   Epoch: [45/50], Train loss: 0.6006
   Epoch: [45/50], Valid loss: 1.4706, Valid accuracy: 0.4924
   _____
   Epoch: [46/50], Train loss: 0.5869
   Epoch: [46/50], Valid loss: 1.4670, Valid accuracy: 0.5127
   _____
   _____
   Epoch: [47/50], Train loss: 0.5818
   Epoch: [47/50], Valid loss: 1.4780, Valid accuracy: 0.5076
   _____
   _____
   Epoch: [48/50], Train loss: 0.5846
   Epoch: [48/50], Valid loss: 1.4876, Valid accuracy: 0.4975
   _____
   _____
   Epoch: [49/50], Train loss: 0.5866
   Epoch: [49/50], Valid loss: 1.4719, Valid accuracy: 0.4975
   _____
   _____
   Epoch: [50/50], Train loss: 0.5664
   Epoch: [50/50], Valid loss: 1.5047, Valid accuracy: 0.5228
   _____
[17]: evaluate(resnet_model, "best_resnet_model_melspectrogram.ckpt", test_loader,
     ⊶device)
   Model at best_resnet_model_melspectrogram.ckpt:
```

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Accuracy: 0.5759 Precision: 0.5328 Recall: 0.5839 F1 Score: 0.5472



```
[18]: effnet_model = EfficientNetV2S() # EfficientNetv2-S
train(effnet_model, "best_effnet_model_melspectrogram.ckpt", train_loader,
valid_loader, device, num_epochs=50, lr=1e-3, debug = True)
```

Downloading: "https://download.pytorch.org/models/efficientnet\_v2\_s-dd5fe13b.pth" to /root/.cache/torch/hub/checkpoints/efficientnet\_v2\_s-dd5fe13b.pth

100%| | 82.7M/82.7M [00:00<00:00, 163MB/s]

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Epoch: [1/50], Train loss: 2.2671

Epoch: [1/50], Valid loss: 2.2570, Valid accuracy: 0.1675

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Saving the best model at 0 epochs!

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Epoch: [2/50], Train loss: 2.0468

Epoch: [2/50], Valid loss: 1.9998, Valid accuracy: 0.3858

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Saving the best model at 1 epochs!

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Epoch: [3/50], Train loss: 1.8906

```
Epoch: [3/50], Valid loss: 1.8688, Valid accuracy: 0.4162
_____
Saving the best model at 2 epochs!
Epoch: [4/50], Train loss: 1.7604
Epoch: [4/50], Valid loss: 1.7926, Valid accuracy: 0.4670
Saving the best model at 3 epochs!
-----
Epoch: [5/50], Train loss: 1.6472
Epoch: [5/50], Valid loss: 1.7379, Valid accuracy: 0.4518
_____
Saving the best model at 4 epochs!
-----
Epoch: [6/50], Train loss: 1.5641
Epoch: [6/50], Valid loss: 1.6771, Valid accuracy: 0.4772
_____
Saving the best model at 5 epochs!
_____
Epoch: [7/50], Train loss: 1.4886
Epoch: [7/50], Valid loss: 1.6492, Valid accuracy: 0.4721
  ._____
Saving the best model at 6 epochs!
Epoch: [8/50], Train loss: 1.4251
Epoch: [8/50], Valid loss: 1.6155, Valid accuracy: 0.4873
_____
Saving the best model at 7 epochs!
     -----
Epoch: [9/50], Train loss: 1.3717
Epoch: [9/50], Valid loss: 1.6029, Valid accuracy: 0.4772
_____
Saving the best model at 8 epochs!
_____
Epoch: [10/50], Train loss: 1.3139
Epoch: [10/50], Valid loss: 1.5703, Valid accuracy: 0.4822
-----
Saving the best model at 9 epochs!
Epoch: [11/50], Train loss: 1.2841
Epoch: [11/50], Valid loss: 1.5493, Valid accuracy: 0.4822
_____
Saving the best model at 10 epochs!
______
Epoch: [12/50], Train loss: 1.2887
Epoch: [12/50], Valid loss: 1.5374, Valid accuracy: 0.4670
-----
```

Saving the best model at 11 epochs!

Epoch: [13/50], Train loss: 1.2071 Epoch: [13/50], Valid loss: 1.5322, Valid accuracy: 0.5127 Saving the best model at 12 epochs! \_\_\_\_\_ Epoch: [14/50], Train loss: 1.1962 Epoch: [14/50], Valid loss: 1.5248, Valid accuracy: 0.5025 -----Saving the best model at 13 epochs! \_\_\_\_\_ Epoch: [15/50], Train loss: 1.1229 Epoch: [15/50], Valid loss: 1.4869, Valid accuracy: 0.5127 -----Saving the best model at 14 epochs! -----Epoch: [16/50], Train loss: 1.1373 Epoch: [16/50], Valid loss: 1.4883, Valid accuracy: 0.5127 \_\_\_\_\_ \_\_\_\_\_ Epoch: [17/50], Train loss: 1.1243 Epoch: [17/50], Valid loss: 1.4837, Valid accuracy: 0.4975 -----Saving the best model at 16 epochs! \_\_\_\_\_ Epoch: [18/50], Train loss: 1.0452 Epoch: [18/50], Valid loss: 1.4629, Valid accuracy: 0.5127 \_\_\_\_\_ Saving the best model at 17 epochs! \_\_\_\_\_\_ Epoch: [19/50], Train loss: 1.0333 Epoch: [19/50], Valid loss: 1.4650, Valid accuracy: 0.5381 \_\_\_\_\_ Epoch: [20/50], Train loss: 1.0108 Epoch: [20/50], Valid loss: 1.4612, Valid accuracy: 0.5381 -----Saving the best model at 19 epochs! Epoch: [21/50], Train loss: 1.0000 Epoch: [21/50], Valid loss: 1.4652, Valid accuracy: 0.5381 \_\_\_\_\_ \_\_\_\_\_ Epoch: [22/50], Train loss: 0.9891 Epoch: [22/50], Valid loss: 1.4588, Valid accuracy: 0.5228 -----Saving the best model at 21 epochs!

\_\_\_\_\_

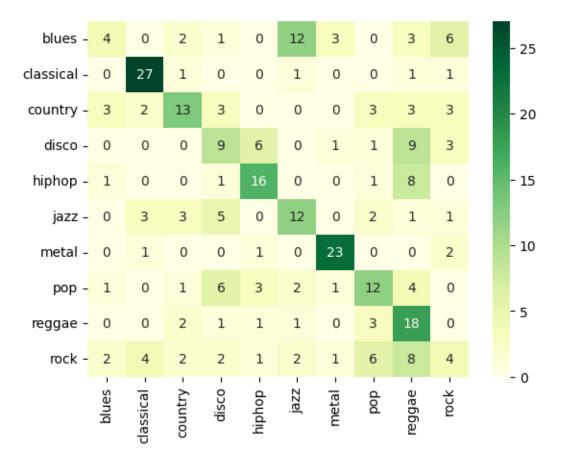
```
Epoch: [23/50], Train loss: 0.9882
Epoch: [23/50], Valid loss: 1.4621, Valid accuracy: 0.5076
_____
Epoch: [24/50], Train loss: 0.9314
Epoch: [24/50], Valid loss: 1.4397, Valid accuracy: 0.5533
Saving the best model at 23 epochs!
-----
Epoch: [25/50], Train loss: 0.9457
Epoch: [25/50], Valid loss: 1.4431, Valid accuracy: 0.5279
_____
_____
Epoch: [26/50], Train loss: 0.9337
Epoch: [26/50], Valid loss: 1.4345, Valid accuracy: 0.5381
-----
Saving the best model at 25 epochs!
Epoch: [27/50], Train loss: 0.8747
Epoch: [27/50], Valid loss: 1.4420, Valid accuracy: 0.5279
_____
-----
Epoch: [28/50], Train loss: 0.8849
Epoch: [28/50], Valid loss: 1.4333, Valid accuracy: 0.5381
_____
Saving the best model at 27 epochs!
_____
Epoch: [29/50], Train loss: 0.8956
Epoch: [29/50], Valid loss: 1.4272, Valid accuracy: 0.5330
______
Saving the best model at 28 epochs!
-----
Epoch: [30/50], Train loss: 0.8845
Epoch: [30/50], Valid loss: 1.4422, Valid accuracy: 0.5431
_____
_____
Epoch: [31/50], Train loss: 0.8332
Epoch: [31/50], Valid loss: 1.4071, Valid accuracy: 0.5533
-----
Saving the best model at 30 epochs!
_____
Epoch: [32/50], Train loss: 0.8530
Epoch: [32/50], Valid loss: 1.4228, Valid accuracy: 0.5381
_____
_____
Epoch: [33/50], Train loss: 0.8252
Epoch: [33/50], Valid loss: 1.4358, Valid accuracy: 0.5279
```

Epoch: [34/50], Train loss: 0.8246 Epoch: [34/50], Valid loss: 1.4159, Valid accuracy: 0.5381 \_\_\_\_\_ Epoch: [35/50], Train loss: 0.8065 Epoch: [35/50], Valid loss: 1.4288, Valid accuracy: 0.5076 \_\_\_\_\_ Epoch: [36/50], Train loss: 0.8108 Epoch: [36/50], Valid loss: 1.4071, Valid accuracy: 0.5330 \_\_\_\_\_ Saving the best model at 35 epochs! \_\_\_\_\_ Epoch: [37/50], Train loss: 0.7804 Epoch: [37/50], Valid loss: 1.4186, Valid accuracy: 0.5381 \_\_\_\_\_ Epoch: [38/50], Train loss: 0.7810 Epoch: [38/50], Valid loss: 1.4064, Valid accuracy: 0.5431 \_\_\_\_\_ Saving the best model at 37 epochs! \_\_\_\_\_ Epoch: [39/50], Train loss: 0.7888 Epoch: [39/50], Valid loss: 1.4061, Valid accuracy: 0.5431 \_\_\_\_\_ Saving the best model at 38 epochs! \_\_\_\_\_ Epoch: [40/50], Train loss: 0.8049 Epoch: [40/50], Valid loss: 1.3956, Valid accuracy: 0.5228 -----Saving the best model at 39 epochs! \_\_\_\_\_\_ Epoch: [41/50], Train loss: 0.7687 Epoch: [41/50], Valid loss: 1.4054, Valid accuracy: 0.5330 \_\_\_\_\_ Epoch: [42/50], Train loss: 0.7261 Epoch: [42/50], Valid loss: 1.4183, Valid accuracy: 0.5279 -----Epoch: [43/50], Train loss: 0.7542 Epoch: [43/50], Valid loss: 1.3982, Valid accuracy: 0.5533 \_\_\_\_\_\_ \_\_\_\_\_ Epoch: [44/50], Train loss: 0.7369 Epoch: [44/50], Valid loss: 1.4138, Valid accuracy: 0.5431

\_\_\_\_\_

```
Epoch: [45/50], Train loss: 0.7196
   Epoch: [45/50], Valid loss: 1.3965, Valid accuracy: 0.5584
   Epoch: [46/50], Train loss: 0.7206
   Epoch: [46/50], Valid loss: 1.4135, Valid accuracy: 0.5482
   _____
   Epoch: [47/50], Train loss: 0.7074
   Epoch: [47/50], Valid loss: 1.4243, Valid accuracy: 0.5381
   _____
   _____
   Epoch: [48/50], Train loss: 0.6868
   Epoch: [48/50], Valid loss: 1.4037, Valid accuracy: 0.5330
   ______
   _____
   Epoch: [49/50], Train loss: 0.7314
   Epoch: [49/50], Valid loss: 1.3936, Valid accuracy: 0.5482
   _____
   Saving the best model at 48 epochs!
   _____
   Epoch: [50/50], Train loss: 0.7123
   Epoch: [50/50], Valid loss: 1.4491, Valid accuracy: 0.5076
   _____
[19]: evaluate(effnet_model, "best_effnet_model_melspectrogram.ckpt", test_loader,__
     ⊶device)
   _____
   Model at best_effnet_model_melspectrogram.ckpt:
   _____
```

Accuracy: 0.4759 Precision: 0.4677 Recall: 0.4850 F1 Score: 0.4618



[21]: speccnn\_model = SpectrogramCNN() # Custom model built by us train(speccnn\_model, "best\_speccnn\_model\_spectrogram.ckpt", train\_loader, ovalid\_loader, device, num\_epochs=50, lr=1e-3, debug = True)

-----

Epoch: [1/50], Train loss: 2.3476

Epoch: [1/50], Valid loss: 2.3532, Valid accuracy: 0.1015

-----

```
Saving the best model at 0 epochs!
_____
Epoch: [2/50], Train loss: 2.1226
Epoch: [2/50], Valid loss: 2.2427, Valid accuracy: 0.2030
-----
Saving the best model at 1 epochs!
Epoch: [3/50], Train loss: 1.9534
Epoch: [3/50], Valid loss: 2.1090, Valid accuracy: 0.2234
_____
Saving the best model at 2 epochs!
_____
Epoch: [4/50], Train loss: 1.9066
Epoch: [4/50], Valid loss: 2.0617, Valid accuracy: 0.2538
  -----
Saving the best model at 3 epochs!
_____
Epoch: [5/50], Train loss: 1.7968
Epoch: [5/50], Valid loss: 2.0515, Valid accuracy: 0.2386
_____
Saving the best model at 4 epochs!
_____
Epoch: [6/50], Train loss: 1.7070
Epoch: [6/50], Valid loss: 2.0640, Valid accuracy: 0.2386
_____
_____
Epoch: [7/50], Train loss: 1.6809
Epoch: [7/50], Valid loss: 2.0097, Valid accuracy: 0.2589
  -----
Saving the best model at 6 epochs!
-----
Epoch: [8/50], Train loss: 1.6461
Epoch: [8/50], Valid loss: 1.9424, Valid accuracy: 0.2741
_____
Saving the best model at 7 epochs!
_____
Epoch: [9/50], Train loss: 1.5796
Epoch: [9/50], Valid loss: 1.8627, Valid accuracy: 0.3503
._____
Saving the best model at 8 epochs!
_____
Epoch: [10/50], Train loss: 1.5517
Epoch: [10/50], Valid loss: 1.9404, Valid accuracy: 0.2792
_____
_____
Epoch: [11/50], Train loss: 1.4858
Epoch: [11/50], Valid loss: 1.8595, Valid accuracy: 0.3452
```

```
Saving the best model at 10 epochs!
_____
Epoch: [12/50], Train loss: 1.4501
Epoch: [12/50], Valid loss: 1.7558, Valid accuracy: 0.3503
-----
Saving the best model at 11 epochs!
Epoch: [13/50], Train loss: 1.4462
Epoch: [13/50], Valid loss: 1.7753, Valid accuracy: 0.3959
_____
_____
Epoch: [14/50], Train loss: 1.4252
Epoch: [14/50], Valid loss: 1.8635, Valid accuracy: 0.3452
-----
_____
Epoch: [15/50], Train loss: 1.3634
Epoch: [15/50], Valid loss: 1.7090, Valid accuracy: 0.3909
  -----
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.2687
Epoch: [16/50], Valid loss: 1.6937, Valid accuracy: 0.3858
_____
Saving the best model at 15 epochs!
_____
Epoch: [17/50], Train loss: 1.2803
Epoch: [17/50], Valid loss: 1.6481, Valid accuracy: 0.4518
_____
Saving the best model at 16 epochs!
______
Epoch: [18/50], Train loss: 1.2441
Epoch: [18/50], Valid loss: 1.6717, Valid accuracy: 0.3909
_____
Epoch: [19/50], Train loss: 1.2097
Epoch: [19/50], Valid loss: 1.8737, Valid accuracy: 0.3350
_____
_____
Epoch: [20/50], Train loss: 1.1819
Epoch: [20/50], Valid loss: 1.6173, Valid accuracy: 0.4112
-----
Saving the best model at 19 epochs!
     _____
Epoch: [21/50], Train loss: 1.0820
Epoch: [21/50], Valid loss: 1.5776, Valid accuracy: 0.4162
-----
Saving the best model at 20 epochs!
```

\_\_\_\_\_

```
Epoch: [22/50], Train loss: 1.0732
Epoch: [22/50], Valid loss: 1.5224, Valid accuracy: 0.4772
-----
Saving the best model at 21 epochs!
-----
Epoch: [23/50], Train loss: 1.0836
Epoch: [23/50], Valid loss: 1.6139, Valid accuracy: 0.4112
_____
Epoch: [24/50], Train loss: 1.0650
Epoch: [24/50], Valid loss: 1.5281, Valid accuracy: 0.4264
_____
_____
Epoch: [25/50], Train loss: 1.0410
Epoch: [25/50], Valid loss: 1.5797, Valid accuracy: 0.4569
______
_____
Epoch: [26/50], Train loss: 0.9533
Epoch: [26/50], Valid loss: 1.5648, Valid accuracy: 0.4061
_____
_____
Epoch: [27/50], Train loss: 0.9818
Epoch: [27/50], Valid loss: 1.6942, Valid accuracy: 0.4112
_____
_____
Epoch: [28/50], Train loss: 0.9679
Epoch: [28/50], Valid loss: 1.7874, Valid accuracy: 0.3909
_____
_____
Epoch: [29/50], Train loss: 0.9290
Epoch: [29/50], Valid loss: 1.4163, Valid accuracy: 0.4822
_____
Saving the best model at 28 epochs!
_____
Epoch: [30/50], Train loss: 1.0066
Epoch: [30/50], Valid loss: 1.6017, Valid accuracy: 0.4416
_____
_____
Epoch: [31/50], Train loss: 0.9105
Epoch: [31/50], Valid loss: 1.3102, Valid accuracy: 0.5736
-----
Saving the best model at 30 epochs!
    ._____
Epoch: [32/50], Train loss: 0.8618
Epoch: [32/50], Valid loss: 1.5078, Valid accuracy: 0.5381
-----
_____
```

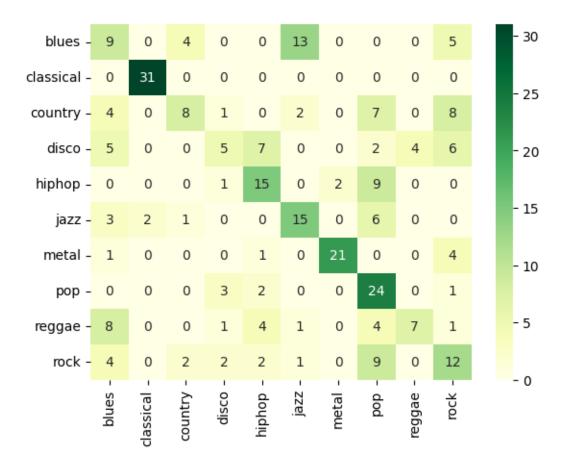
Epoch: [33/50], Train loss: 0.8695

```
Epoch: [33/50], Valid loss: 1.4798, Valid accuracy: 0.4619
_____
Epoch: [34/50], Train loss: 0.8235
Epoch: [34/50], Valid loss: 1.9080, Valid accuracy: 0.3756
_____
Epoch: [35/50], Train loss: 0.7966
Epoch: [35/50], Valid loss: 1.5019, Valid accuracy: 0.4365
_____
_____
Epoch: [36/50], Train loss: 0.8384
Epoch: [36/50], Valid loss: 1.8467, Valid accuracy: 0.3756
_____
_____
Epoch: [37/50], Train loss: 0.7751
Epoch: [37/50], Valid loss: 1.5109, Valid accuracy: 0.4873
_____
Epoch: [38/50], Train loss: 0.7567
Epoch: [38/50], Valid loss: 1.5357, Valid accuracy: 0.4822
______
_____
Epoch: [39/50], Train loss: 0.7465
Epoch: [39/50], Valid loss: 1.3781, Valid accuracy: 0.5279
_____
_____
Epoch: [40/50], Train loss: 0.7674
Epoch: [40/50], Valid loss: 1.6174, Valid accuracy: 0.4924
_____
_____
Epoch: [41/50], Train loss: 0.7267
Epoch: [41/50], Valid loss: 1.3523, Valid accuracy: 0.6142
_____
Epoch: [42/50], Train loss: 0.7179
Epoch: [42/50], Valid loss: 1.4985, Valid accuracy: 0.4975
_____
Epoch: [43/50], Train loss: 0.6909
Epoch: [43/50], Valid loss: 1.6161, Valid accuracy: 0.5076
_____
_____
Epoch: [44/50], Train loss: 0.6369
Epoch: [44/50], Valid loss: 1.4548, Valid accuracy: 0.5228
-----
_____
```

Epoch: [45/50], Train loss: 0.6227

```
Epoch: [45/50], Valid loss: 1.3285, Valid accuracy: 0.5635
   _____
   Epoch: [46/50], Train loss: 0.6566
   Epoch: [46/50], Valid loss: 1.5960, Valid accuracy: 0.4924
   _____
   Epoch: [47/50], Train loss: 0.6413
   Epoch: [47/50], Valid loss: 1.8024, Valid accuracy: 0.4416
   _____
   Epoch: [48/50], Train loss: 0.6780
   Epoch: [48/50], Valid loss: 1.4749, Valid accuracy: 0.5076
   -----
   _____
   Epoch: [49/50], Train loss: 0.6462
   Epoch: [49/50], Valid loss: 1.4643, Valid accuracy: 0.4924
   _____
   Epoch: [50/50], Train loss: 0.5690
   Epoch: [50/50], Valid loss: 1.6031, Valid accuracy: 0.4822
   _____
[22]: evaluate(speccnn_model, "best_speccnn_model_spectrogram.ckpt", test_loader,__
     ⊸device)
   _____
   Model at best_speccnn_model_spectrogram.ckpt:
   -----
```

Accuracy: 0.5069 Precision: 0.5342 Recall: 0.5063 F1 Score: 0.4959



```
[23]: resnet_model = ResNet18() # ResNet-18
train(resnet_model, "best_resnet_model_spectrogram.ckpt", train_loader,
valid_loader, device, num_epochs=50, lr=1e-3, debug = True)
```

\_\_\_\_\_

Epoch: [1/50], Train loss: 2.3556

Epoch: [1/50], Valid loss: 2.2668, Valid accuracy: 0.1472

\_\_\_\_\_

Saving the best model at 0 epochs!

-----

Epoch: [2/50], Train loss: 2.0943

Epoch: [2/50], Valid loss: 2.0968, Valid accuracy: 0.2843

\_\_\_\_\_

Saving the best model at 1 epochs!

-----

Epoch: [3/50], Train loss: 1.8901

Epoch: [3/50], Valid loss: 1.9381, Valid accuracy: 0.3807

-----

Saving the best model at 2 epochs!

-----

```
Epoch: [4/50], Train loss: 1.7322
Epoch: [4/50], Valid loss: 1.8843, Valid accuracy: 0.3147
-----
Saving the best model at 3 epochs!
-----
Epoch: [5/50], Train loss: 1.6034
Epoch: [5/50], Valid loss: 1.8213, Valid accuracy: 0.3959
-----
Saving the best model at 4 epochs!
_____
Epoch: [6/50], Train loss: 1.5131
Epoch: [6/50], Valid loss: 1.7716, Valid accuracy: 0.3807
_____
Saving the best model at 5 epochs!
_____
Epoch: [7/50], Train loss: 1.4220
Epoch: [7/50], Valid loss: 1.7174, Valid accuracy: 0.4518
-----
Saving the best model at 6 epochs!
_____
Epoch: [8/50], Train loss: 1.3687
Epoch: [8/50], Valid loss: 1.7472, Valid accuracy: 0.3706
_____
Epoch: [9/50], Train loss: 1.3314
Epoch: [9/50], Valid loss: 1.6606, Valid accuracy: 0.3959
_____
Saving the best model at 8 epochs!
    _____
Epoch: [10/50], Train loss: 1.3040
Epoch: [10/50], Valid loss: 1.6960, Valid accuracy: 0.4264
_____
Epoch: [11/50], Train loss: 1.2089
Epoch: [11/50], Valid loss: 1.6314, Valid accuracy: 0.4061
_____
Saving the best model at 10 epochs!
_____
Epoch: [12/50], Train loss: 1.1964
Epoch: [12/50], Valid loss: 1.6418, Valid accuracy: 0.4467
-----
_____
Epoch: [13/50], Train loss: 1.1388
Epoch: [13/50], Valid loss: 1.6168, Valid accuracy: 0.4213
_____
Saving the best model at 12 epochs!
-----
```

Epoch: [14/50], Train loss: 1.1167

```
Epoch: [14/50], Valid loss: 1.6181, Valid accuracy: 0.4315
_____
Epoch: [15/50], Train loss: 1.1033
Epoch: [15/50], Valid loss: 1.5978, Valid accuracy: 0.4619
_____
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.0406
Epoch: [16/50], Valid loss: 1.5682, Valid accuracy: 0.4467
_____
Saving the best model at 15 epochs!
-----
Epoch: [17/50], Train loss: 1.0167
Epoch: [17/50], Valid loss: 1.5730, Valid accuracy: 0.4518
-----
_____
Epoch: [18/50], Train loss: 1.0046
Epoch: [18/50], Valid loss: 1.5607, Valid accuracy: 0.4822
_____
Saving the best model at 17 epochs!
______
Epoch: [19/50], Train loss: 0.9678
Epoch: [19/50], Valid loss: 1.5302, Valid accuracy: 0.4315
_____
Saving the best model at 18 epochs!
_____
Epoch: [20/50], Train loss: 0.9245
Epoch: [20/50], Valid loss: 1.5484, Valid accuracy: 0.4670
______
_____
Epoch: [21/50], Train loss: 0.9133
Epoch: [21/50], Valid loss: 1.5159, Valid accuracy: 0.4670
_____
Saving the best model at 20 epochs!
_____
Epoch: [22/50], Train loss: 0.8821
Epoch: [22/50], Valid loss: 1.5298, Valid accuracy: 0.4619
-----
_____
Epoch: [23/50], Train loss: 0.8804
Epoch: [23/50], Valid loss: 1.5278, Valid accuracy: 0.4315
_____
_____
Epoch: [24/50], Train loss: 0.8817
Epoch: [24/50], Valid loss: 1.5081, Valid accuracy: 0.4721
-----
```

Saving the best model at 23 epochs!

Epoch: [25/50], Train loss: 0.8346 Epoch: [25/50], Valid loss: 1.5110, Valid accuracy: 0.4518 \_\_\_\_\_ Epoch: [26/50], Train loss: 0.8368 Epoch: [26/50], Valid loss: 1.5064, Valid accuracy: 0.4772 \_\_\_\_\_ Saving the best model at 25 epochs! \_\_\_\_\_ Epoch: [27/50], Train loss: 0.8280 Epoch: [27/50], Valid loss: 1.5385, Valid accuracy: 0.4721 \_\_\_\_\_ \_\_\_\_\_\_ Epoch: [28/50], Train loss: 0.8102 Epoch: [28/50], Valid loss: 1.4871, Valid accuracy: 0.4772 \_\_\_\_\_ Saving the best model at 27 epochs! \_\_\_\_\_ Epoch: [29/50], Train loss: 0.7801 Epoch: [29/50], Valid loss: 1.5004, Valid accuracy: 0.4365 \_\_\_\_\_ \_\_\_\_\_ Epoch: [30/50], Train loss: 0.7781 Epoch: [30/50], Valid loss: 1.4831, Valid accuracy: 0.4822 \_\_\_\_\_ Saving the best model at 29 epochs! \_\_\_\_\_ Epoch: [31/50], Train loss: 0.7546 Epoch: [31/50], Valid loss: 1.4890, Valid accuracy: 0.4619 \_\_\_\_\_ -----Epoch: [32/50], Train loss: 0.7577 Epoch: [32/50], Valid loss: 1.4938, Valid accuracy: 0.4670 \_\_\_\_\_ \_\_\_\_\_ Epoch: [33/50], Train loss: 0.7270 Epoch: [33/50], Valid loss: 1.4773, Valid accuracy: 0.4822 -----Saving the best model at 32 epochs! \_\_\_\_\_ Epoch: [34/50], Train loss: 0.7112 Epoch: [34/50], Valid loss: 1.4922, Valid accuracy: 0.4518 \_\_\_\_\_ \_\_\_\_\_ Epoch: [35/50], Train loss: 0.7358 Epoch: [35/50], Valid loss: 1.5126, Valid accuracy: 0.4822

Epoch: [36/50], Train loss: 0.6861 Epoch: [36/50], Valid loss: 1.5050, Valid accuracy: 0.4619 \_\_\_\_\_ Epoch: [37/50], Train loss: 0.7093 Epoch: [37/50], Valid loss: 1.5082, Valid accuracy: 0.4670 \_\_\_\_\_ \_\_\_\_\_ Epoch: [38/50], Train loss: 0.7083 Epoch: [38/50], Valid loss: 1.4669, Valid accuracy: 0.4721 \_\_\_\_\_ Saving the best model at 37 epochs! \_\_\_\_\_ Epoch: [39/50], Train loss: 0.6526 Epoch: [39/50], Valid loss: 1.5025, Valid accuracy: 0.4670 \_\_\_\_\_ Epoch: [40/50], Train loss: 0.6965 Epoch: [40/50], Valid loss: 1.4639, Valid accuracy: 0.4822 \_\_\_\_\_ Saving the best model at 39 epochs! \_\_\_\_\_ Epoch: [41/50], Train loss: 0.6440 Epoch: [41/50], Valid loss: 1.4729, Valid accuracy: 0.4873 \_\_\_\_\_ \_\_\_\_\_ Epoch: [42/50], Train loss: 0.6379 Epoch: [42/50], Valid loss: 1.4661, Valid accuracy: 0.4670 \_\_\_\_\_ \_\_\_\_\_ Epoch: [43/50], Train loss: 0.6107 Epoch: [43/50], Valid loss: 1.4670, Valid accuracy: 0.4822 \_\_\_\_\_ Epoch: [44/50], Train loss: 0.6100 Epoch: [44/50], Valid loss: 1.4760, Valid accuracy: 0.4721 \_\_\_\_\_ Epoch: [45/50], Train loss: 0.6327 Epoch: [45/50], Valid loss: 1.4822, Valid accuracy: 0.4822 \_\_\_\_\_ \_\_\_\_\_ Epoch: [46/50], Train loss: 0.6170 Epoch: [46/50], Valid loss: 1.4656, Valid accuracy: 0.4721 -----\_\_\_\_\_

Epoch: [47/50], Train loss: 0.5987

```
Epoch: [47/50], Valid loss: 1.4594, Valid accuracy: 0.5025
```

-----

Saving the best model at 46 epochs!

-----

Epoch: [48/50], Train loss: 0.5978

Epoch: [48/50], Valid loss: 1.5000, Valid accuracy: 0.4619

\_\_\_\_\_

Epoch: [49/50], Train loss: 0.5713

Epoch: [49/50], Valid loss: 1.4498, Valid accuracy: 0.4924

\_\_\_\_\_

Saving the best model at 48 epochs!

-----

Epoch: [50/50], Train loss: 0.6043

Epoch: [50/50], Valid loss: 1.5022, Valid accuracy: 0.4873

-----

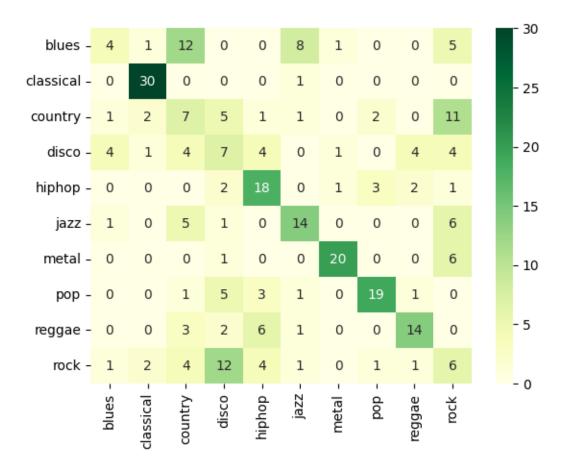
## 

-----

### Model at best\_resnet\_model\_spectrogram.ckpt:

-----

Accuracy: 0.4793 Precision: 0.5030 Recall: 0.4857 F1 Score: 0.4850



```
[25]: effnet_model = EfficientNetV2S() # EfficientNetv2-S
train(effnet_model, "best_effnet_model_spectrogram.ckpt", train_loader,
valid_loader, device, num_epochs=50, lr=1e-3, debug = True)
```

\_\_\_\_\_

Epoch: [1/50], Train loss: 2.2556

Epoch: [1/50], Valid loss: 2.2398, Valid accuracy: 0.2437

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Saving the best model at 0 epochs!

\_\_\_\_\_

Epoch: [2/50], Train loss: 2.0613

Epoch: [2/50], Valid loss: 2.0607, Valid accuracy: 0.3096

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Saving the best model at 1 epochs!

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Epoch: [3/50], Train loss: 1.9198

Epoch: [3/50], Valid loss: 1.9549, Valid accuracy: 0.3553

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Saving the best model at 2 epochs!

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```
Epoch: [4/50], Train loss: 1.8167
Epoch: [4/50], Valid loss: 1.9096, Valid accuracy: 0.3909
  _____
Saving the best model at 3 epochs!
_____
Epoch: [5/50], Train loss: 1.7358
Epoch: [5/50], Valid loss: 1.8527, Valid accuracy: 0.3959
-----
Saving the best model at 4 epochs!
_____
Epoch: [6/50], Train loss: 1.6660
Epoch: [6/50], Valid loss: 1.8249, Valid accuracy: 0.3959
_____
Saving the best model at 5 epochs!
_____
Epoch: [7/50], Train loss: 1.6021
Epoch: [7/50], Valid loss: 1.7840, Valid accuracy: 0.4010
  -----
Saving the best model at 6 epochs!
_____
Epoch: [8/50], Train loss: 1.6030
Epoch: [8/50], Valid loss: 1.7812, Valid accuracy: 0.3909
-----
Saving the best model at 7 epochs!
_____
Epoch: [9/50], Train loss: 1.5356
Epoch: [9/50], Valid loss: 1.7459, Valid accuracy: 0.3959
_____
Saving the best model at 8 epochs!
_____
Epoch: [10/50], Train loss: 1.4268
Epoch: [10/50], Valid loss: 1.6983, Valid accuracy: 0.4162
_____
Saving the best model at 9 epochs!
______
Epoch: [11/50], Train loss: 1.4666
Epoch: [11/50], Valid loss: 1.7271, Valid accuracy: 0.4213
_____
Epoch: [12/50], Train loss: 1.3943
Epoch: [12/50], Valid loss: 1.6651, Valid accuracy: 0.4264
_____
Saving the best model at 11 epochs!
_____
Epoch: [13/50], Train loss: 1.3845
Epoch: [13/50], Valid loss: 1.6752, Valid accuracy: 0.4162
-----
```

```
Epoch: [14/50], Train loss: 1.2636
Epoch: [14/50], Valid loss: 1.6517, Valid accuracy: 0.4315
_____
Saving the best model at 13 epochs!
-----
Epoch: [15/50], Train loss: 1.3411
Epoch: [15/50], Valid loss: 1.6512, Valid accuracy: 0.4213
_____
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.2617
Epoch: [16/50], Valid loss: 1.6058, Valid accuracy: 0.4365
_____
Saving the best model at 15 epochs!
_____
Epoch: [17/50], Train loss: 1.2531
Epoch: [17/50], Valid loss: 1.6221, Valid accuracy: 0.4315
_____
Epoch: [18/50], Train loss: 1.2752
Epoch: [18/50], Valid loss: 1.6522, Valid accuracy: 0.4061
  _____
_____
Epoch: [19/50], Train loss: 1.1442
Epoch: [19/50], Valid loss: 1.5943, Valid accuracy: 0.4264
  -----
Saving the best model at 18 epochs!
_____
Epoch: [20/50], Train loss: 1.1869
Epoch: [20/50], Valid loss: 1.5913, Valid accuracy: 0.4365
-----
Saving the best model at 19 epochs!
_____
Epoch: [21/50], Train loss: 1.1411
Epoch: [21/50], Valid loss: 1.5752, Valid accuracy: 0.4416
_____
Saving the best model at 20 epochs!
_____
Epoch: [22/50], Train loss: 1.2226
Epoch: [22/50], Valid loss: 1.6182, Valid accuracy: 0.4213
_____
_____
Epoch: [23/50], Train loss: 1.1000
Epoch: [23/50], Valid loss: 1.5776, Valid accuracy: 0.4416
_____
Epoch: [24/50], Train loss: 1.1617
Epoch: [24/50], Valid loss: 1.5776, Valid accuracy: 0.4619
```

\_\_\_\_\_ -----Epoch: [25/50], Train loss: 1.1352 Epoch: [25/50], Valid loss: 1.5665, Valid accuracy: 0.4670 -----Saving the best model at 24 epochs! Epoch: [26/50], Train loss: 1.0710 Epoch: [26/50], Valid loss: 1.5595, Valid accuracy: 0.4569 \_\_\_\_\_ Saving the best model at 25 epochs! \_\_\_\_\_ Epoch: [27/50], Train loss: 1.0420 Epoch: [27/50], Valid loss: 1.5652, Valid accuracy: 0.4518 \_\_\_\_\_ \_\_\_\_\_ Epoch: [28/50], Train loss: 1.0898 Epoch: [28/50], Valid loss: 1.5556, Valid accuracy: 0.4721 \_\_\_\_\_ Saving the best model at 27 epochs! \_\_\_\_\_ Epoch: [29/50], Train loss: 1.0172 Epoch: [29/50], Valid loss: 1.5368, Valid accuracy: 0.4721 \_\_\_\_\_ Saving the best model at 28 epochs! \_\_\_\_\_ Epoch: [30/50], Train loss: 1.0606 Epoch: [30/50], Valid loss: 1.5399, Valid accuracy: 0.4873 \_\_\_\_\_ \_\_\_\_\_ Epoch: [31/50], Train loss: 1.0445 Epoch: [31/50], Valid loss: 1.5475, Valid accuracy: 0.4518 \_\_\_\_\_ -----Epoch: [32/50], Train loss: 1.0019 Epoch: [32/50], Valid loss: 1.5356, Valid accuracy: 0.4619 \_\_\_\_\_ Saving the best model at 31 epochs! Epoch: [33/50], Train loss: 1.0381 Epoch: [33/50], Valid loss: 1.5316, Valid accuracy: 0.4721 \_\_\_\_\_ Saving the best model at 32 epochs! \_\_\_\_\_ Epoch: [34/50], Train loss: 0.9833 Epoch: [34/50], Valid loss: 1.5333, Valid accuracy: 0.4416 -----

```
Epoch: [35/50], Train loss: 0.9886
Epoch: [35/50], Valid loss: 1.5436, Valid accuracy: 0.4569
_____
Epoch: [36/50], Train loss: 0.9556
Epoch: [36/50], Valid loss: 1.5581, Valid accuracy: 0.4365
_____
Epoch: [37/50], Train loss: 1.0276
Epoch: [37/50], Valid loss: 1.5556, Valid accuracy: 0.4670
_____
_____
Epoch: [38/50], Train loss: 0.9976
Epoch: [38/50], Valid loss: 1.5541, Valid accuracy: 0.4518
_____
_____
Epoch: [39/50], Train loss: 0.9778
Epoch: [39/50], Valid loss: 1.5508, Valid accuracy: 0.4518
  .____
_____
Epoch: [40/50], Train loss: 0.9837
Epoch: [40/50], Valid loss: 1.5559, Valid accuracy: 0.4619
_____
_____
Epoch: [41/50], Train loss: 0.9578
Epoch: [41/50], Valid loss: 1.5279, Valid accuracy: 0.4670
_____
Saving the best model at 40 epochs!
    ._____
Epoch: [42/50], Train loss: 0.9555
Epoch: [42/50], Valid loss: 1.5313, Valid accuracy: 0.4772
_____
Epoch: [43/50], Train loss: 0.9198
Epoch: [43/50], Valid loss: 1.5212, Valid accuracy: 0.4822
_____
Saving the best model at 42 epochs!
_____
Epoch: [44/50], Train loss: 0.9405
Epoch: [44/50], Valid loss: 1.5098, Valid accuracy: 0.4772
_____
Saving the best model at 43 epochs!
    _____
Epoch: [45/50], Train loss: 0.9729
Epoch: [45/50], Valid loss: 1.5352, Valid accuracy: 0.4721
-----
_____
```

Epoch: [46/50], Train loss: 0.8716

Epoch: [46/50], Valid loss: 1.5265, Valid accuracy: 0.4721

\_\_\_\_\_

Epoch: [47/50], Train loss: 0.9491

Epoch: [47/50], Valid loss: 1.5464, Valid accuracy: 0.4670

\_\_\_\_\_

Epoch: [48/50], Train loss: 0.8708

Epoch: [48/50], Valid loss: 1.5380, Valid accuracy: 0.4670

\_\_\_\_\_ \_\_\_\_\_

Epoch: [49/50], Train loss: 0.8430

Epoch: [49/50], Valid loss: 1.5330, Valid accuracy: 0.4670

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Epoch: [50/50], Train loss: 0.8536

Epoch: [50/50], Valid loss: 1.5187, Valid accuracy: 0.4721

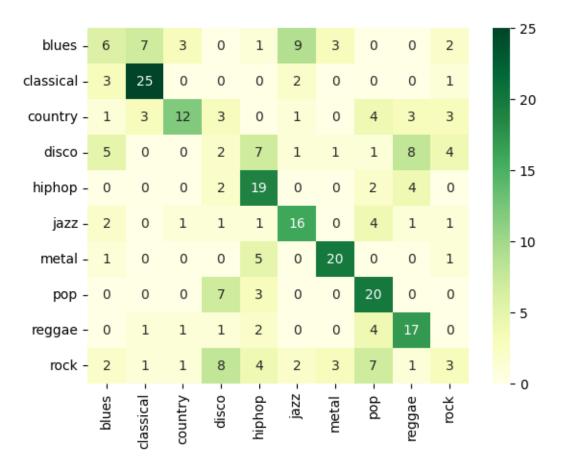
\_\_\_\_\_

[26]: evaluate(effnet\_model, "best\_effnet\_model\_spectrogram.ckpt", test\_loader, \_\_ ⊶device)

.\_\_\_\_\_

Model at best\_effnet\_model\_spectrogram.ckpt:

Accuracy: 0.4828 Precision: 0.4611 Recall: 0.4920 F1 Score: 0.4639



[28]: speccnn\_model = SpectrogramCNN() # Custom model built by us train(speccnn\_model, "best\_speccnn\_model\_mfcc.ckpt", train\_loader,\_u \( \text{-valid\_loader, device, num\_epochs=50, lr=1e-3, debug = True} \)

\_\_\_\_\_

Epoch: [1/50], Train loss: 2.4066

Epoch: [1/50], Valid loss: 2.3220, Valid accuracy: 0.0863

-----

Saving the best model at 0 epochs! \_\_\_\_\_ Epoch: [2/50], Train loss: 2.2881 Epoch: [2/50], Valid loss: 2.2668, Valid accuracy: 0.1878 -----Saving the best model at 1 epochs! Epoch: [3/50], Train loss: 2.1422 Epoch: [3/50], Valid loss: 2.1297, Valid accuracy: 0.2589 \_\_\_\_\_ Saving the best model at 2 epochs! \_\_\_\_\_ Epoch: [4/50], Train loss: 2.0765 Epoch: [4/50], Valid loss: 2.0243, Valid accuracy: 0.2487 .\_\_\_\_\_ Saving the best model at 3 epochs! \_\_\_\_\_ Epoch: [5/50], Train loss: 1.9848 Epoch: [5/50], Valid loss: 2.1202, Valid accuracy: 0.1523 \_\_\_\_\_ \_\_\_\_\_ Epoch: [6/50], Train loss: 1.8935 Epoch: [6/50], Valid loss: 2.0617, Valid accuracy: 0.2234 .\_\_\_\_\_ Epoch: [7/50], Train loss: 1.8449 Epoch: [7/50], Valid loss: 2.0534, Valid accuracy: 0.2183 \_\_\_\_\_ -----Epoch: [8/50], Train loss: 1.7546 Epoch: [8/50], Valid loss: 2.0162, Valid accuracy: 0.2437 \_\_\_\_\_ Saving the best model at 7 epochs! \_\_\_\_\_ Epoch: [9/50], Train loss: 1.7436 Epoch: [9/50], Valid loss: 1.9246, Valid accuracy: 0.2640 -----Saving the best model at 8 epochs! Epoch: [10/50], Train loss: 1.6648 Epoch: [10/50], Valid loss: 1.8775, Valid accuracy: 0.2893 \_\_\_\_\_ Saving the best model at 9 epochs! \_\_\_\_\_ Epoch: [11/50], Train loss: 1.6407 Epoch: [11/50], Valid loss: 1.9037, Valid accuracy: 0.2843 -----

```
Epoch: [12/50], Train loss: 1.5746
Epoch: [12/50], Valid loss: 1.9437, Valid accuracy: 0.2081
_____
Epoch: [13/50], Train loss: 1.5703
Epoch: [13/50], Valid loss: 1.8332, Valid accuracy: 0.3147
Saving the best model at 12 epochs!
-----
Epoch: [14/50], Train loss: 1.5182
Epoch: [14/50], Valid loss: 1.9521, Valid accuracy: 0.2386
_____
_____
Epoch: [15/50], Train loss: 1.4718
Epoch: [15/50], Valid loss: 1.8141, Valid accuracy: 0.2944
-----
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.4213
Epoch: [16/50], Valid loss: 1.7709, Valid accuracy: 0.3096
_____
Saving the best model at 15 epochs!
_____
Epoch: [17/50], Train loss: 1.3495
Epoch: [17/50], Valid loss: 1.8753, Valid accuracy: 0.3096
_____
_____
Epoch: [18/50], Train loss: 1.3635
Epoch: [18/50], Valid loss: 1.7349, Valid accuracy: 0.3249
_____
Saving the best model at 17 epochs!
-----
Epoch: [19/50], Train loss: 1.2684
Epoch: [19/50], Valid loss: 1.6701, Valid accuracy: 0.4010
  _____
Saving the best model at 18 epochs!
_____
Epoch: [20/50], Train loss: 1.2454
Epoch: [20/50], Valid loss: 1.8039, Valid accuracy: 0.3401
-----
Epoch: [21/50], Train loss: 1.2232
Epoch: [21/50], Valid loss: 2.0306, Valid accuracy: 0.3147
_____
_____
Epoch: [22/50], Train loss: 1.1614
Epoch: [22/50], Valid loss: 1.6857, Valid accuracy: 0.3655
```

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-----Epoch: [23/50], Train loss: 1.1442 Epoch: [23/50], Valid loss: 1.7334, Valid accuracy: 0.3807 \_\_\_\_\_ Epoch: [24/50], Train loss: 1.1353 Epoch: [24/50], Valid loss: 1.6638, Valid accuracy: 0.4112 \_\_\_\_\_ Saving the best model at 23 epochs! \_\_\_\_\_ Epoch: [25/50], Train loss: 1.1123 Epoch: [25/50], Valid loss: 1.7463, Valid accuracy: 0.3807 \_\_\_\_\_ \_\_\_\_\_\_ Epoch: [26/50], Train loss: 1.0956 Epoch: [26/50], Valid loss: 1.7404, Valid accuracy: 0.3909 \_\_\_\_\_ Epoch: [27/50], Train loss: 1.0641 Epoch: [27/50], Valid loss: 1.6413, Valid accuracy: 0.4112 \_\_\_\_\_ Saving the best model at 26 epochs! \_\_\_\_\_ Epoch: [28/50], Train loss: 1.0649 Epoch: [28/50], Valid loss: 1.7941, Valid accuracy: 0.3858 \_\_\_\_\_ \_\_\_\_\_ Epoch: [29/50], Train loss: 1.0329 Epoch: [29/50], Valid loss: 1.9574, Valid accuracy: 0.3350 \_\_\_\_\_ \_\_\_\_\_ Epoch: [30/50], Train loss: 1.0030 Epoch: [30/50], Valid loss: 1.7554, Valid accuracy: 0.3706 \_\_\_\_\_ \_\_\_\_\_ Epoch: [31/50], Train loss: 1.0165 Epoch: [31/50], Valid loss: 1.8816, Valid accuracy: 0.3807 \_\_\_\_\_ Epoch: [32/50], Train loss: 0.9762 Epoch: [32/50], Valid loss: 1.6346, Valid accuracy: 0.4467 \_\_\_\_\_ Saving the best model at 31 epochs! \_\_\_\_\_ Epoch: [33/50], Train loss: 0.9408 Epoch: [33/50], Valid loss: 1.8096, Valid accuracy: 0.4213 -----

```
Epoch: [34/50], Train loss: 0.9181
Epoch: [34/50], Valid loss: 1.6966, Valid accuracy: 0.4213
_____
Epoch: [35/50], Train loss: 0.8816
Epoch: [35/50], Valid loss: 1.7079, Valid accuracy: 0.4365
_____
Epoch: [36/50], Train loss: 0.8841
Epoch: [36/50], Valid loss: 1.9066, Valid accuracy: 0.3706
_____
_____
Epoch: [37/50], Train loss: 0.8595
Epoch: [37/50], Valid loss: 1.7505, Valid accuracy: 0.4467
_____
_____
Epoch: [38/50], Train loss: 0.8731
Epoch: [38/50], Valid loss: 1.8575, Valid accuracy: 0.4061
_____
_____
Epoch: [39/50], Train loss: 0.8284
Epoch: [39/50], Valid loss: 1.6176, Valid accuracy: 0.4467
_____
Saving the best model at 38 epochs!
_____
Epoch: [40/50], Train loss: 0.8243
Epoch: [40/50], Valid loss: 1.9219, Valid accuracy: 0.3807
_____
_____
Epoch: [41/50], Train loss: 0.7945
Epoch: [41/50], Valid loss: 1.9255, Valid accuracy: 0.4061
-----
Epoch: [42/50], Train loss: 0.7697
Epoch: [42/50], Valid loss: 1.8254, Valid accuracy: 0.4416
_____
_____
```

Epoch: [43/50], Train loss: 0.7389

Epoch: [43/50], Valid loss: 1.8405, Valid accuracy: 0.4416

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Epoch: [44/50], Train loss: 0.7271

Epoch: [44/50], Valid loss: 1.8929, Valid accuracy: 0.4112

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Epoch: [45/50], Train loss: 0.7190

Epoch: [45/50], Valid loss: 2.1704, Valid accuracy: 0.3858

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-----

Epoch: [46/50], Train loss: 0.7764

Epoch: [46/50], Valid loss: 1.8365, Valid accuracy: 0.4365

\_\_\_\_\_\_

Epoch: [47/50], Train loss: 0.7362

Epoch: [47/50], Valid loss: 1.9890, Valid accuracy: 0.4264

\_\_\_\_\_

Epoch: [48/50], Train loss: 0.6971

Epoch: [48/50], Valid loss: 2.0569, Valid accuracy: 0.4061

\_\_\_\_\_

Epoch: [49/50], Train loss: 0.6852

Epoch: [49/50], Valid loss: 2.0974, Valid accuracy: 0.3858

-----

Epoch: [50/50], Train loss: 0.6999

Epoch: [50/50], Valid loss: 1.7070, Valid accuracy: 0.4873

\_\_\_\_\_

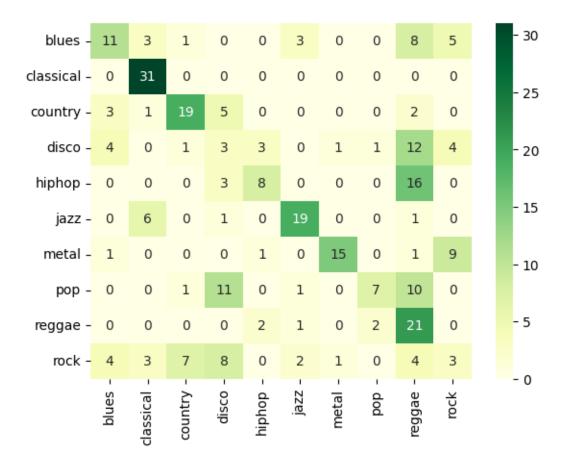
[29]: evaluate(speccnn\_model, "best\_speccnn\_model\_mfcc.ckpt", test\_loader, device)

-----

Model at best\_speccnn\_model\_mfcc.ckpt:

\_\_\_\_\_

Accuracy: 0.4724 Precision: 0.5242 Recall: 0.4782 F1 Score: 0.4646



```
[30]: resnet_model = ResNet18() # ResNet-18
train(resnet_model, "best_resnet_model_mfcc.ckpt", train_loader, valid_loader, udevice, num_epochs=50, lr=1e-3, debug = True)
```

\_\_\_\_\_

Epoch: [1/50], Train loss: 2.3303

Epoch: [1/50], Valid loss: 2.2937, Valid accuracy: 0.1371

\_\_\_\_\_

Saving the best model at 0 epochs!

-----

Epoch: [2/50], Train loss: 2.0969

Epoch: [2/50], Valid loss: 2.1279, Valid accuracy: 0.2234

\_\_\_\_\_

Saving the best model at 1 epochs!

-----

Epoch: [3/50], Train loss: 1.9504

Epoch: [3/50], Valid loss: 2.0178, Valid accuracy: 0.2843

-----

Saving the best model at 2 epochs!

-----

```
Epoch: [4/50], Train loss: 1.8661
Epoch: [4/50], Valid loss: 1.9508, Valid accuracy: 0.3046
Saving the best model at 3 epochs!
-----
Epoch: [5/50], Train loss: 1.7613
Epoch: [5/50], Valid loss: 1.9537, Valid accuracy: 0.2843
_____
Epoch: [6/50], Train loss: 1.6955
Epoch: [6/50], Valid loss: 1.8949, Valid accuracy: 0.3147
_____
Saving the best model at 5 epochs!
______
Epoch: [7/50], Train loss: 1.5960
Epoch: [7/50], Valid loss: 1.8734, Valid accuracy: 0.3096
_____
Saving the best model at 6 epochs!
_____
Epoch: [8/50], Train loss: 1.5350
Epoch: [8/50], Valid loss: 1.8466, Valid accuracy: 0.3147
  ._____
Saving the best model at 7 epochs!
Epoch: [9/50], Train loss: 1.4929
Epoch: [9/50], Valid loss: 1.8484, Valid accuracy: 0.3350
_____
_____
Epoch: [10/50], Train loss: 1.4284
Epoch: [10/50], Valid loss: 1.8275, Valid accuracy: 0.3096
-----
Saving the best model at 9 epochs!
_____
Epoch: [11/50], Train loss: 1.3980
Epoch: [11/50], Valid loss: 1.8222, Valid accuracy: 0.3299
_____
Saving the best model at 10 epochs!
_____
Epoch: [12/50], Train loss: 1.3516
Epoch: [12/50], Valid loss: 1.8074, Valid accuracy: 0.3553
-----
Saving the best model at 11 epochs!
     _____
Epoch: [13/50], Train loss: 1.3117
Epoch: [13/50], Valid loss: 1.7867, Valid accuracy: 0.3452
  -----
Saving the best model at 12 epochs!
```

\_\_\_\_\_

```
Epoch: [14/50], Train loss: 1.2827
Epoch: [14/50], Valid loss: 1.7758, Valid accuracy: 0.3655
-----
Saving the best model at 13 epochs!
-----
Epoch: [15/50], Train loss: 1.2339
Epoch: [15/50], Valid loss: 1.7608, Valid accuracy: 0.3858
_____
Saving the best model at 14 epochs!
_____
Epoch: [16/50], Train loss: 1.2272
Epoch: [16/50], Valid loss: 1.7592, Valid accuracy: 0.3655
_____
Saving the best model at 15 epochs!
_____
Epoch: [17/50], Train loss: 1.2362
Epoch: [17/50], Valid loss: 1.7835, Valid accuracy: 0.3503
_____
Epoch: [18/50], Train loss: 1.1599
Epoch: [18/50], Valid loss: 1.7596, Valid accuracy: 0.3604
  _____
_____
Epoch: [19/50], Train loss: 1.1356
Epoch: [19/50], Valid loss: 1.7818, Valid accuracy: 0.3553
_____
_____
Epoch: [20/50], Train loss: 1.1088
Epoch: [20/50], Valid loss: 1.7554, Valid accuracy: 0.3756
_____
Saving the best model at 19 epochs!
-----
Epoch: [21/50], Train loss: 1.1069
Epoch: [21/50], Valid loss: 1.7550, Valid accuracy: 0.3807
  ._____
Saving the best model at 20 epochs!
 ._____
Epoch: [22/50], Train loss: 1.0799
Epoch: [22/50], Valid loss: 1.7458, Valid accuracy: 0.3807
-----
Saving the best model at 21 epochs!
_____
Epoch: [23/50], Train loss: 1.0416
Epoch: [23/50], Valid loss: 1.7841, Valid accuracy: 0.3249
_____
Epoch: [24/50], Train loss: 1.0296
```

Epoch: [24/50], Valid loss: 1.7532, Valid accuracy: 0.3655

.\_\_\_\_\_ -----Epoch: [25/50], Train loss: 1.0126 Epoch: [25/50], Valid loss: 1.7406, Valid accuracy: 0.3909 -----Saving the best model at 24 epochs! Epoch: [26/50], Train loss: 0.9980 Epoch: [26/50], Valid loss: 1.7632, Valid accuracy: 0.3350 \_\_\_\_\_ \_\_\_\_\_ Epoch: [27/50], Train loss: 0.9789 Epoch: [27/50], Valid loss: 1.7424, Valid accuracy: 0.4010 \_\_\_\_\_ \_\_\_\_\_ Epoch: [28/50], Train loss: 0.9580 Epoch: [28/50], Valid loss: 1.7521, Valid accuracy: 0.3604 \_\_\_\_\_ Epoch: [29/50], Train loss: 0.9471 Epoch: [29/50], Valid loss: 1.7739, Valid accuracy: 0.3756 \_\_\_\_\_ \_\_\_\_\_ Epoch: [30/50], Train loss: 0.9155 Epoch: [30/50], Valid loss: 1.7438, Valid accuracy: 0.3858 \_\_\_\_\_ \_\_\_\_\_ Epoch: [31/50], Train loss: 0.8895 Epoch: [31/50], Valid loss: 1.7621, Valid accuracy: 0.3706 \_\_\_\_\_ \_\_\_\_\_ Epoch: [32/50], Train loss: 0.9272 Epoch: [32/50], Valid loss: 1.7424, Valid accuracy: 0.3604 \_\_\_\_\_ Epoch: [33/50], Train loss: 0.8784 Epoch: [33/50], Valid loss: 1.7762, Valid accuracy: 0.3604 \_\_\_\_\_ Epoch: [34/50], Train loss: 0.8781 Epoch: [34/50], Valid loss: 1.7604, Valid accuracy: 0.3604 \_\_\_\_\_ \_\_\_\_\_ Epoch: [35/50], Train loss: 0.8557 Epoch: [35/50], Valid loss: 1.7757, Valid accuracy: 0.3655 -----\_\_\_\_\_

Epoch: [36/50], Train loss: 0.8471

```
Epoch: [36/50], Valid loss: 1.7682, Valid accuracy: 0.3706
_____
Epoch: [37/50], Train loss: 0.8482
Epoch: [37/50], Valid loss: 1.7820, Valid accuracy: 0.3807
_____
Epoch: [38/50], Train loss: 0.8194
Epoch: [38/50], Valid loss: 1.7455, Valid accuracy: 0.3807
_____
_____
Epoch: [39/50], Train loss: 0.8016
Epoch: [39/50], Valid loss: 1.7538, Valid accuracy: 0.3959
_____
_____
Epoch: [40/50], Train loss: 0.8435
Epoch: [40/50], Valid loss: 1.7915, Valid accuracy: 0.3655
_____
Epoch: [41/50], Train loss: 0.8026
Epoch: [41/50], Valid loss: 1.7543, Valid accuracy: 0.3858
_____
_____
Epoch: [42/50], Train loss: 0.7649
Epoch: [42/50], Valid loss: 1.7919, Valid accuracy: 0.3401
_____
_____
Epoch: [43/50], Train loss: 0.7859
Epoch: [43/50], Valid loss: 1.7398, Valid accuracy: 0.4162
_____
Saving the best model at 42 epochs!
-----
Epoch: [44/50], Train loss: 0.7705
Epoch: [44/50], Valid loss: 1.7622, Valid accuracy: 0.3858
_____
_____
Epoch: [45/50], Train loss: 0.7555
Epoch: [45/50], Valid loss: 1.7631, Valid accuracy: 0.3706
-----
_____
Epoch: [46/50], Train loss: 0.7608
Epoch: [46/50], Valid loss: 1.7705, Valid accuracy: 0.3756
_____
_____
Epoch: [47/50], Train loss: 0.7281
Epoch: [47/50], Valid loss: 1.7782, Valid accuracy: 0.3655
-----
```

Epoch: [48/50], Train loss: 0.7034

Epoch: [48/50], Valid loss: 1.7695, Valid accuracy: 0.4112

\_\_\_\_\_\_

Epoch: [49/50], Train loss: 0.6871

Epoch: [49/50], Valid loss: 1.7685, Valid accuracy: 0.3807

\_\_\_\_\_

Epoch: [50/50], Train loss: 0.7057

Epoch: [50/50], Valid loss: 1.7879, Valid accuracy: 0.3959

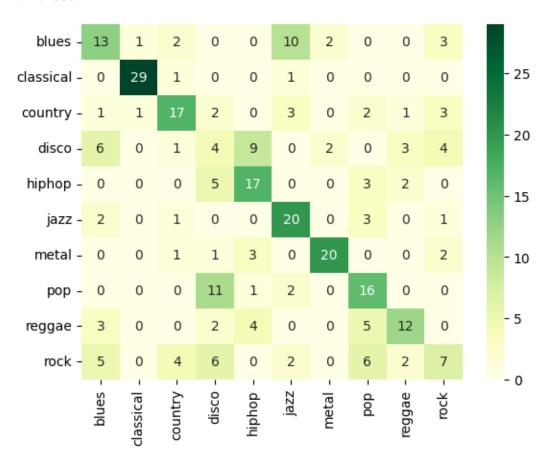
\_\_\_\_\_

## [31]: evaluate(resnet\_model, "best\_resnet\_model\_mfcc.ckpt", test\_loader, device)

-----

## Model at best\_resnet\_model\_mfcc.ckpt:

Accuracy: 0.5345 Precision: 0.5394 Recall: 0.5384 F1 Score: 0.5332



```
[32]: effnet_model = EfficientNetV2S() # EfficientNetv2-S
    train(effnet_model, "best_effnet_model_mfcc.ckpt", train_loader, valid_loader, u
     ⇒device, num_epochs=50, lr=1e-3, debug = True)
    _____
    Epoch: [1/50], Train loss: 2.2662
    Epoch: [1/50], Valid loss: 2.2680, Valid accuracy: 0.1777
    Saving the best model at 0 epochs!
    _____
    Epoch: [2/50], Train loss: 2.1213
    Epoch: [2/50], Valid loss: 2.1392, Valid accuracy: 0.2487
    -----
    Saving the best model at 1 epochs!
    Epoch: [3/50], Train loss: 2.0286
    Epoch: [3/50], Valid loss: 2.0662, Valid accuracy: 0.2640
      _____
    Saving the best model at 2 epochs!
    _____
    Epoch: [4/50], Train loss: 1.9275
    Epoch: [4/50], Valid loss: 2.0285, Valid accuracy: 0.2437
    Saving the best model at 3 epochs!
    _____
    Epoch: [5/50], Train loss: 1.8642
    Epoch: [5/50], Valid loss: 1.9819, Valid accuracy: 0.2893
    _____
    Saving the best model at 4 epochs!
    _____
    Epoch: [6/50], Train loss: 1.8015
    Epoch: [6/50], Valid loss: 1.9661, Valid accuracy: 0.2690
    -----
    Saving the best model at 5 epochs!
    _____
    Epoch: [7/50], Train loss: 1.7581
    Epoch: [7/50], Valid loss: 1.9456, Valid accuracy: 0.2741
    _____
    Saving the best model at 6 epochs!
    _____
    Epoch: [8/50], Train loss: 1.6695
    Epoch: [8/50], Valid loss: 1.9221, Valid accuracy: 0.3147
    -----
    Saving the best model at 7 epochs!
    _____
    Epoch: [9/50], Train loss: 1.6417
```

```
Epoch: [9/50], Valid loss: 1.9186, Valid accuracy: 0.3046
_____
Saving the best model at 8 epochs!
Epoch: [10/50], Train loss: 1.5905
Epoch: [10/50], Valid loss: 1.8966, Valid accuracy: 0.2995
Saving the best model at 9 epochs!
-----
Epoch: [11/50], Train loss: 1.5796
Epoch: [11/50], Valid loss: 1.8983, Valid accuracy: 0.2944
_____
_____
Epoch: [12/50], Train loss: 1.5499
Epoch: [12/50], Valid loss: 1.9101, Valid accuracy: 0.2893
-----
_____
Epoch: [13/50], Train loss: 1.5156
Epoch: [13/50], Valid loss: 1.8906, Valid accuracy: 0.2995
_____
Saving the best model at 12 epochs!
_____
Epoch: [14/50], Train loss: 1.4607
Epoch: [14/50], Valid loss: 1.8904, Valid accuracy: 0.3046
_____
Saving the best model at 13 epochs!
_____
Epoch: [15/50], Train loss: 1.4291
Epoch: [15/50], Valid loss: 1.8690, Valid accuracy: 0.3198
_____
Saving the best model at 14 epochs!
-----
Epoch: [16/50], Train loss: 1.4380
Epoch: [16/50], Valid loss: 1.8656, Valid accuracy: 0.3147
   -----
Saving the best model at 15 epochs!
  _____
Epoch: [17/50], Train loss: 1.4205
Epoch: [17/50], Valid loss: 1.8658, Valid accuracy: 0.2944
-----
Epoch: [18/50], Train loss: 1.3972
Epoch: [18/50], Valid loss: 1.8771, Valid accuracy: 0.3096
_____
_____
Epoch: [19/50], Train loss: 1.4016
Epoch: [19/50], Valid loss: 1.8794, Valid accuracy: 0.3096
```

Epoch: [20/50], Train loss: 1.3475 Epoch: [20/50], Valid loss: 1.8805, Valid accuracy: 0.2944 \_\_\_\_\_ Epoch: [21/50], Train loss: 1.3811 Epoch: [21/50], Valid loss: 1.8785, Valid accuracy: 0.3046 \_\_\_\_\_ Epoch: [22/50], Train loss: 1.3168 Epoch: [22/50], Valid loss: 1.8864, Valid accuracy: 0.3299 \_\_\_\_\_ \_\_\_\_\_ Epoch: [23/50], Train loss: 1.2649 Epoch: [23/50], Valid loss: 1.8764, Valid accuracy: 0.3096 \_\_\_\_\_ \_\_\_\_\_ Epoch: [24/50], Train loss: 1.2791 Epoch: [24/50], Valid loss: 1.8570, Valid accuracy: 0.3198 \_\_\_\_\_ Saving the best model at 23 epochs! \_\_\_\_\_ Epoch: [25/50], Train loss: 1.2991 Epoch: [25/50], Valid loss: 1.8498, Valid accuracy: 0.3096 \_\_\_\_\_ Saving the best model at 24 epochs! \_\_\_\_\_ Epoch: [26/50], Train loss: 1.2633 Epoch: [26/50], Valid loss: 1.8578, Valid accuracy: 0.2944 \_\_\_\_\_ \_\_\_\_\_ Epoch: [27/50], Train loss: 1.2350 Epoch: [27/50], Valid loss: 1.8617, Valid accuracy: 0.2995 \_\_\_\_\_ Epoch: [28/50], Train loss: 1.2652 Epoch: [28/50], Valid loss: 1.8549, Valid accuracy: 0.3046 \_\_\_\_\_ Epoch: [29/50], Train loss: 1.3000 Epoch: [29/50], Valid loss: 1.8547, Valid accuracy: 0.2944 \_\_\_\_\_ \_\_\_\_\_ Epoch: [30/50], Train loss: 1.2627 Epoch: [30/50], Valid loss: 1.8539, Valid accuracy: 0.2995 -----\_\_\_\_\_

Epoch: [31/50], Train loss: 1.2106

```
Epoch: [31/50], Valid loss: 1.8615, Valid accuracy: 0.3147
_____
Epoch: [32/50], Train loss: 1.1706
Epoch: [32/50], Valid loss: 1.8727, Valid accuracy: 0.3147
_____
Epoch: [33/50], Train loss: 1.2238
Epoch: [33/50], Valid loss: 1.8635, Valid accuracy: 0.3096
_____
_____
Epoch: [34/50], Train loss: 1.1770
Epoch: [34/50], Valid loss: 1.8645, Valid accuracy: 0.3147
_____
_____
Epoch: [35/50], Train loss: 1.1690
Epoch: [35/50], Valid loss: 1.8637, Valid accuracy: 0.3198
_____
Epoch: [36/50], Train loss: 1.2060
Epoch: [36/50], Valid loss: 1.8528, Valid accuracy: 0.2995
_____
_____
Epoch: [37/50], Train loss: 1.1201
Epoch: [37/50], Valid loss: 1.8568, Valid accuracy: 0.3096
_____
_____
Epoch: [38/50], Train loss: 1.1145
Epoch: [38/50], Valid loss: 1.8276, Valid accuracy: 0.3299
_____
Saving the best model at 37 epochs!
-----
Epoch: [39/50], Train loss: 1.0926
Epoch: [39/50], Valid loss: 1.8506, Valid accuracy: 0.3096
_____
_____
Epoch: [40/50], Train loss: 1.1463
Epoch: [40/50], Valid loss: 1.8553, Valid accuracy: 0.2944
-----
_____
Epoch: [41/50], Train loss: 1.0948
Epoch: [41/50], Valid loss: 1.8542, Valid accuracy: 0.3046
_____
_____
Epoch: [42/50], Train loss: 1.1167
Epoch: [42/50], Valid loss: 1.8505, Valid accuracy: 0.3249
-----
```

```
Epoch: [43/50], Train loss: 1.1420
```

Epoch: [43/50], Valid loss: 1.8444, Valid accuracy: 0.3147

\_\_\_\_\_\_

Epoch: [44/50], Train loss: 1.1404

Epoch: [44/50], Valid loss: 1.8737, Valid accuracy: 0.3096

\_\_\_\_\_\_

Epoch: [45/50], Train loss: 1.0893

Epoch: [45/50], Valid loss: 1.8588, Valid accuracy: 0.3046

\_\_\_\_\_\_

Epoch: [46/50], Train loss: 1.0606

Epoch: [46/50], Valid loss: 1.8799, Valid accuracy: 0.2995

\_\_\_\_\_

Epoch: [47/50], Train loss: 1.0471

Epoch: [47/50], Valid loss: 1.8722, Valid accuracy: 0.3249

\_\_\_\_\_\_

Epoch: [48/50], Train loss: 1.0454

Epoch: [48/50], Valid loss: 1.8639, Valid accuracy: 0.3350

\_\_\_\_\_\_

Epoch: [49/50], Train loss: 1.0374

Epoch: [49/50], Valid loss: 1.9092, Valid accuracy: 0.3350

\_\_\_\_\_\_

Epoch: [50/50], Train loss: 1.0123

Epoch: [50/50], Valid loss: 1.8795, Valid accuracy: 0.3299

-----

## [33]: evaluate(effnet\_model, "best\_effnet\_model\_mfcc.ckpt", test\_loader, device)

-----

Model at best\_effnet\_model\_mfcc.ckpt:

-----

Accuracy: 0.4483 Precision: 0.4390 Recall: 0.4485 F1 Score: 0.4315

blues -	10	2	5	1	1	3	3	2	1	3	
classical -	0	28	2	0	0	0	0	0	1	0	- 25
country -	1	1	14	5	0	0	1	4	0	4	- 20
disco -	3	0	2	4	8	0	4	1	6	1	
hiphop -	0	0	1	5	14	0	1	5	1	0	- 15
jazz -	5	5	2	0	0	10	0	4	0	1	
metal -	2	0	1	1	0	1	20	0	0	2	- 10
pop -	0	1	1	3	1	0	0	21	3	0	
reggae -	1	0	0	6	3	1	0	8	6	1	- 5
rock -	3	2	1	8	5	0	1	6	3	3	- 0
	- sənlq	classical -	country -	disco -	- doydiy	- jazz	metal -	- dod	reggae -	rock -	- <b>U</b>