model.mfcc

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1 CS-72: Accelerated Computational Linguistics

1.1 Final Project Code

1.1.1 Emotion Detection in Audio

1.1.2 Team Members:

- Aiwei Zhang
- Amittai Siavava
- Carlos Guerrero Alvarez

This is a variant of our model that uses MFCC (Mel-frequency cepstral coefficients) features extracted from audio.

```
[]: # %pip install pytorch_lightning
# %pip install transformers
# %pip install torchmetrics
# %pip install soundfile
# %pip install librosa
# %pip install ipywidgets
```

```
[]: # imports
import torch
import torch.nn as nn
import torch.optim as optim
# import math
import pytorch_lightning as pl
# import torchaudio
import torchaudio
import torchmetrics
from torch.optim.lr_scheduler import ReduceLROnPlateau
```

```
/usr/lib/python3/dist-packages/pkg_resources/__init__.py:116:
PkgResourcesDeprecationWarning: 1.12.1-git20200711.33e2d80-dfsg1-0.6 is an
invalid version and will not be supported in a future release
   warnings.warn(
/usr/lib/python3/dist-packages/pkg_resources/__init__.py:116:
PkgResourcesDeprecationWarning: 1.12.1-git20200711.33e2d80-dfsg1-0.6 is an
invalid version and will not be supported in a future release
   warnings.warn(
```

```
[]: torch.cuda.is_available()
[]: True
[]: from local_dataset import AudioEmotionsDataset
     # import TQDMProgressBar
     from pytorch_lightning.callbacks import TQDMProgressBar
[ ]: BATCH_SIZE = 32
     dataset = AudioEmotionsDataset("/home/ubuntu/siavava-west-1/test/data/
      →audio-emotions", batch_size=BATCH_SIZE, max_size=200, feature_type="mfcc")
     train = dataset.train_dataloader
     test = dataset.test_dataloader
    Some weights of Wav2Vec2ForCTC were not initialized from the model checkpoint at
    facebook/wav2vec2-base-960h and are newly initialized:
    ['wav2vec2.masked_spec_embed']
    You should probably TRAIN this model on a down-stream task to be able to use it
    for predictions and inference.
    /usr/lib/python3/dist-packages/scipy/__init__.py:146: UserWarning: A NumPy
    version >=1.17.3 and <1.25.0 is required for this version of SciPy (detected
    version 1.25.2
      warnings.warn(f"A NumPy version >={np minversion} and <{np maxversion}"
    LOADED: 2400
[]: class Reshape(nn.Module):
        def __init__(self):
            super(Reshape, self).__init__()
             # self.shape = shape
        def forward(self, x):
             # return x.view((x.shape[0], *self.shape))
            return x.view(x.shape[0], 1, x.shape[1])
        def call (self, x):
             return self.forward(x)
     class SpeechEmotionRecognitionModel(pl.LightningModule):
        def __init__(self, input_size, num_classes, dim_feedforward=2048,_
      dim_model=1024, nhead=8, num_encoder_layers=6, num_decoder_layers=6, lr=0.5,__
      ⇒dropout=0.1):
             super(SpeechEmotionRecognitionModel, self).__init__()
             self.lr = lr
             self.layers = nn.Sequential(
```

```
Reshape(),
          nn.Conv1d(in_channels=1, kernel_size=40, out_channels=7),
          nn.ReLU(),
          nn.Dropout(p=0.2),
          nn.Flatten(start_dim=1, end_dim=2),
          nn.Linear(7, 7),
          nn.Softmax(dim=1),
          # nn.Sigmoid()
      )
      # initialize the metrics
      self.loss_function = nn.CrossEntropyLoss()
      self.precision = torchmetrics.Precision(task='multiclass',__
→num_classes=num_classes, average="macro")
      self.recall = torchmetrics.Recall(task='multiclass',__
→num_classes=num_classes, average="macro")
      self.F1 = torchmetrics.F1Score(task='multiclass', __

¬num_classes=num_classes, average="macro")
  def forward(self, src):
      output = self.layers(src)
      return output
  def training_step(self, batch, batch_idx):
      src, tgt = batch
      output = self(src)
      loss = self.loss_function(output, torch.argmax(tgt, axis=1))
      self.log('cross entropy loss_step', loss, on_step=True, on_epoch=True, u
→prog_bar=True)
      return loss
  def validation step(self, batch, batch idx):
      src, tgt = batch
      output = self(src)
      loss = self.loss_function(output, torch.argmax(tgt, axis=1))
      # print(f"VALIDATION LOSS: {loss}")
  def configure_optimizers(self):
      optimizer = torch.optim.Adam(self.parameters(), lr=self.lr)
      scheduler = {
           'scheduler': ReduceLROnPlateau(optimizer, mode='min', factor=0.1, ____
→patience=2, verbose=True),
           'monitor': 'cross entropy loss_step', # Name of the metric to_
\hookrightarrow monitor
```

```
'interval': 'epoch',
                 'frequency': 1,
             }
             return {'optimizer': optimizer, 'lr_scheduler': scheduler}
         # function for evaluating the quality of output and target
         def evaluation(self, output, target, loss):
            precision = self.precision(output, target)
            recall = self.recall(output, target)
             f1 = self.F1(output, target)
                               {loss}")
            print(f"CE:
            print(f"PRECISION: {precision}")
             print(f"RECALL: {recall}")
                                {f1}")
            print(f"F1:
[]: model = SpeechEmotionRecognitionModel(input_size=dataset.feature_count,__
      num_classes=dataset.class_count)
     # for p in model.parameters():
     # if p.dim() > 1:
              nn.init.xavier_uniform_(p)
     # callbacks=[TQDMProgressBar(refresh_rate=10)]
     trainer = pl.Trainer(default_root_dir='checkpoints', accelerator="auto", __

→max_epochs=50, min_epochs=2, log_every_n_steps=1)
    GPU available: True (cuda), used: True
    TPU available: False, using: 0 TPU cores
    IPU available: False, using: 0 IPUs
    HPU available: False, using: 0 HPUs
[]: for batch in test:
        X, y = batch
         X = X.cuda(0)
         y = y.cuda(0)
         print(f"X: {X.device}")
         print(f"y: {y.device}")
         print(f"model: {model.device}")
         model = model.cuda(0)
         output = model(X)
         # print out put device
         print(f"output: {X.device}")
         model.evaluation(output, y, model.loss_function(output, y))
         # for i in range(32):
```

```
print(f"\{torch.argmax(output[i]):2d\} \ / \ \{torch.argmax(y[i]):2d\} >>>_{\sqcup}
      \hookrightarrow \{torch.max(output[i]):.2f\}")
         break
    X: cuda:0
    y: cuda:0
    model: cpu
    output: cuda:0
               2.068202495574951
    PRECISION: 0.42825111746788025
    RECALL: 0.4973958432674408
    F1:
              0.4602409601211548
[]: torch.set_float32_matmul_precision('medium')
     trainer.fit(model, train_dataloaders=train, val_dataloaders=test)
    LOCAL_RANK: O - CUDA_VISIBLE_DEVICES: [0]
      | Name
                      | Type
                                            | Params
    0 | lavers
                     | Sequential
                                            | 343
    1 | loss_function | CrossEntropyLoss
                                           1 0
    2 | precision
                    | MulticlassPrecision | 0
    3 | recall
                      | MulticlassRecall
                      | MulticlassF1Score
    4 | F1
                                           1 0
    343
              Trainable params
    0
              Non-trainable params
              Total params
    343
    0.001
              Total estimated model params size (MB)
                                | 0/? [00:00<?, ?it/s]
    Sanity Checking: |
    Training: |
                    | 0/? [00:00<?, ?it/s]
    Validation: |
                          | 0/? [00:00<?, ?it/s]
    Validation: |
                          | 0/? [00:00<?, ?it/s]
    Validation: |
                          | 0/? [00:00<?, ?it/s]
    Validation: |
                          | 0/? [00:00<?, ?it/s]
                       | 0/? [00:00<?, ?it/s]
    Validation: |
    Validation: |
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    Validation: |
                        | 0/? [00:00<?, ?it/s]
                           | 0/? [00:00<?, ?it/s]
    Validation: |
    Epoch 00008: reducing learning rate of group 0 to 5.0000e-02.
    Validation: |
                          | 0/? [00:00<?, ?it/s]
```

```
| 0/? [00:00<?, ?it/s]
Validation: |
Validation: |
                       | 0/? [00:00<?, ?it/s]
Epoch 00011: reducing learning rate of group 0 to 5.0000e-03.
Validation: |
                       | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Epoch 00014: reducing learning rate of group 0 to 5.0000e-04.
Validation: |
                      | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Epoch 00017: reducing learning rate of group 0 to 5.0000e-05.
Validation: |
                       | 0/? [00:00<?, ?it/s]
Validation: |
                      | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Epoch 00020: reducing learning rate of group 0 to 5.0000e-06.
Validation: |
                      | 0/? [00:00<?, ?it/s]
Validation: |
                      | 0/? [00:00<?, ?it/s]
                       | 0/? [00:00<?, ?it/s]
Validation: |
Epoch 00023: reducing learning rate of group 0 to 5.0000e-07.
                       | 0/? [00:00<?, ?it/s]
Validation: |
Validation: |
                       | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
Epoch 00026: reducing learning rate of group 0 to 5.0000e-08.
Validation: |
                      | 0/? [00:00<?, ?it/s]
Validation: |
                       | 0/? [00:00<?, ?it/s]
                       | 0/? [00:00<?, ?it/s]
Validation: |
Validation: |
                       | 0/? [00:00<?, ?it/s]
                       | 0/? [00:00<?, ?it/s]
Validation: |
Epoch 00031: reducing learning rate of group 0 to 5.0000e-09.
                       | 0/? [00:00<?, ?it/s]
Validation: |
                      | 0/? [00:00<?, ?it/s]
Validation: |
Validation: |
                       | 0/? [00:00<?, ?it/s]
```

```
Validation: |
                            | 0/? [00:00<?, ?it/s]
    Validation: |
                            | 0/? [00:00<?, ?it/s]
    Validation: |
                            | 0/? [00:00<?, ?it/s]
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    Validation: |
    Validation: |
                            | 0/? [00:00<?, ?it/s]
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    Validation: |
    Validation: |
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    Validation: |
                            | 0/? [00:00<?, ?it/s]
                            | 0/? [00:00<?, ?it/s]
    Validation: |
    Validation: |
                            | 0/? [00:00<?, ?it/s]
    Validation: |
                            | 0/? [00:00<?, ?it/s]
    `Trainer.fit` stopped: `max_epochs=50` reached.
[]: for batch in test:
         X, y = batch
         X = X.cuda(0)
         y = y.cuda(0)
         print(f"X: {X.device}")
         print(f"y: {y.device}")
         print(f"model: {model.device}")
         model = model.cuda(0)
         output = model(X)
         # print out put device
         print(f"output: {X.device}")
         model.evaluation(output, y, model.loss_function(output, y))
         # for i in range(32):
         # print(f''\{torch.argmax(output[i]):2d\} | \{torch.argmax(y[i]):2d\} >>>_{\sqcup}
      \hookrightarrow {torch.max(output[i]):.2f}")
         break
    X: cuda:0
```

| 0/? [00:00<?, ?it/s]

Validation: |

y: cuda:0
model: cuda:0

output: cuda:0

CE: 1.8528821468353271
PRECISION: 0.6099656224250793
RECALL: 0.6041666269302368
F1: 0.6068861484527588

```
[]: # create 'models' directory if nonexistent
    # import os
    # if not os.path.exists("models"):
    # os.makedirs("models")
    # save model weights
    # torch.save(model.state_dict(), "./models/model_weights.pth")
```