DLHLP 2021 Fall HW1 E2E ASR

2021.10.03

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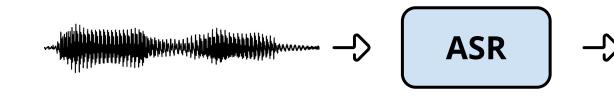
Outline

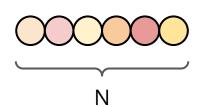
- Intro to E2E ASR (CTC)
- MiniASR Toolkit
 - Preprocess
 - Modify Config
 - Training
 - Testing (evaluation metric)
- Hints for improving performance
- Reference

Preliminaries

- Strongly recommended:
 - Speech Recognition 1
 - o Speech Recognition 2
 - Speech Recognition 3
 - o Speech Recognition 4
 - o Speech Recognition 5
 - Speech Recognition 6
- Optional:
 - Language Modeling

End-to-end Automatic Speech Recognition

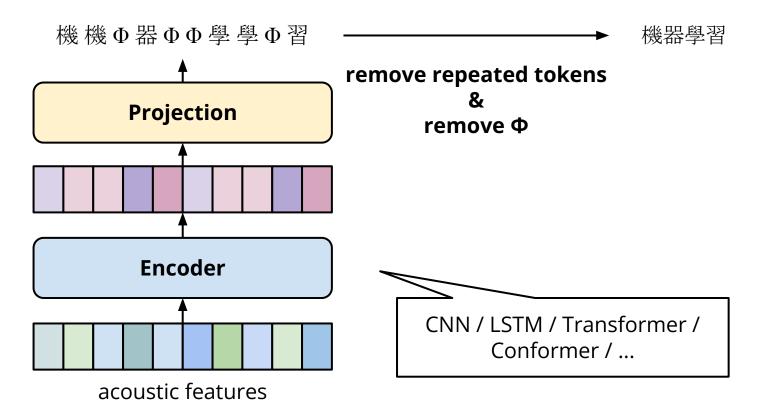




T acoustic features (continuous)

text tokens (discrete)

Connectionist Temporal Classification (CTC)



Connectionist Temporal Classification (CTC)

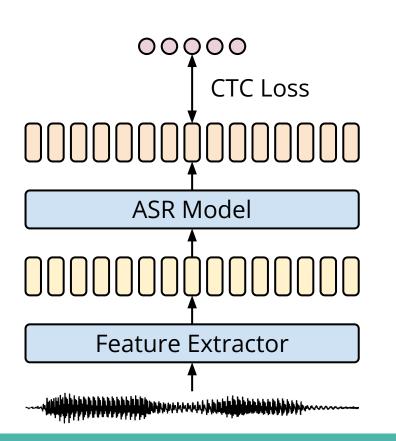
Target: Maximize P(機器學習 | X)

$$P(機器學習 \mid X) = P(機機\Phi器\Phi\Phi學學\Phi習 \mid X)$$

- + P(Φ機Φ器ΦΦΦ學習Φ | X)
- $+ P(機\Phi\Phi 器器器學\Phi習習 | X)$
- + ...

Too many possibilities! **Dynamic Programming**

End-to-end Automatic Speech Recognition



text token sequence: $L \times V$ ($L \leq T'$)

probability distributions over all possible vocabularies: $T' \times V$ ($T' \leq T$)

acoustic features: $T \times D$ (e.g. $T = T_{raw} / 160$)

raw waveform: $T_{raw} \times 1$ (sample rate = 16kHz)

Toolkit

- https://github.com/vectominist/MiniASR
- A mini, simple, and fast E2E ASR toolkit. (still in development)
- Core: <u>PyTorch</u>, <u>PyTorch Lightning</u>, <u>S3PRL</u>
- https://github.com/vectominist/MiniASR/blob/main/example/example libr ispeech training.ipynb



Installation

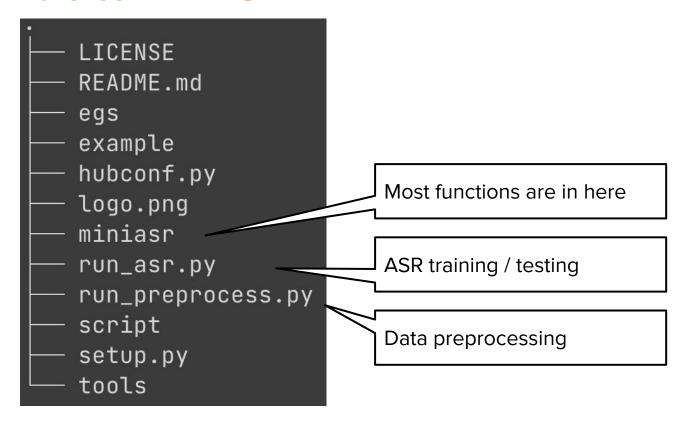
```
git clone https://github.com/vectominist/MiniASR.git
cd MiniASR
pip3 install -e ./
```

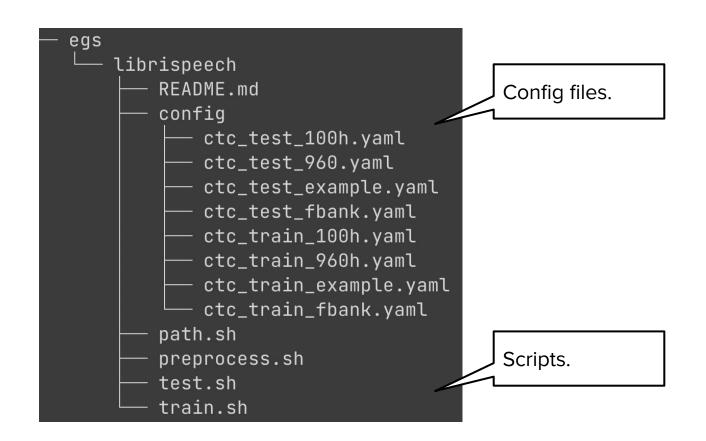
Available commands:

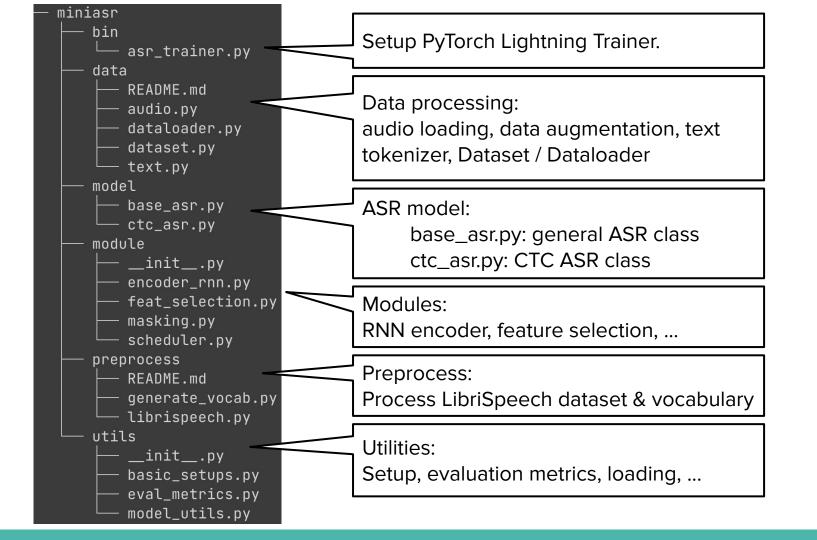
miniasr-asr

miniasr-preprocess

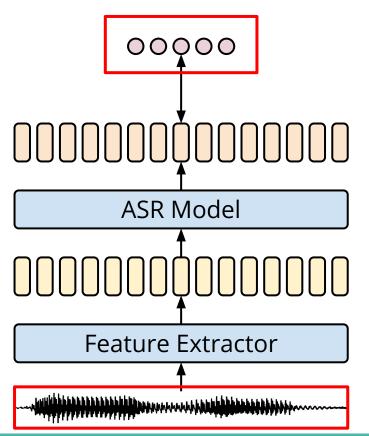
Brief Intro to MiniASR







Step 1: Data Preprocessing



- Training set: Libri-light fine-tuning set
 - https://github.com/facebookresearch/libri-light
 - o 10 hours
 - English audio books
- Development set: LibriSpeech dev-clean
 - https://www.openslr.org/12
- Testing set: LibriSpeech test-clean
 - https://www.openslr.org/12

Step 1: Data Preprocessing

miniasr-preprocess

- --corpus Corpus name.
- --path Path to dataset.
- --set Which subsets to be processed.
- --out Output directory.
- --gen-vocab Specify whether to generate
 vocabulary files.
- --char-vocab-size Character vocabulary size.

- ▼ MiniASR
 - → data
 - LibriSpeech
 - libri_dev
 - libri_test
 - ▼ libri_train_1h
 - data_dict.json
 - data_list_sorted.json
 - text.txt
 - vocab_char.txt
 - libri_train_9h
 - librispeech_finetuning

Step 1: Data Preprocessing

data/libri_train_1h/data_list_sorted.json

```
data_dict.json
data_list_sorted.json
text.txt
vocab_char.txt
```

```
BOWER SAID SILAS WHEN THE BARGAIN
  SOCIABLE GLASS OF OLD JAMAIKEY WARM SHOULD MEET YOUR VIEWS I AM NOT THE MAN TO BEGRUDGE IT YOU ARE AWARE OF MY
BEING POOR COMPANY SIR REPLIED MISTER VENUS BUT BE IT SO".
        "file": "/work/harry87122/dataset/librispeech_finetuning/1h/4/other/978/125137/978-125137-0011.flac"
                                         THE WHOLE A VEXATIOUS NECESSITY HER OBSERVATION
INCLINED HER TO THINK IT RATHER A DREARY STATE",
        "file": "/work/harry87122/dataset/librispeech_finetuning/1h/4/clean/248/130644/248-130644-0008.flac"
                                            TO A SINGLE LIFE BUT
      HERBS A PEERAGE WILL NOT QUITE DO INSTEAD OF LEADERSHIP TO THE MAN WHO MEANT TO LEAD AND THIS DELICATE
LIMBED SYLPH OF TWENTY MEANT TO LEAD",
        "file": "/work/harry87122/dataset/librispeech_finetuning/1h/4/clean/248/130644/248-130644-0010.flac"
```

Step 2-1: Modify Config Files

egs/librispeech/config/ctc_train_example.yaml

Do not modify this part unless you want to use other training data.

Step 2-1: Modify Config Files

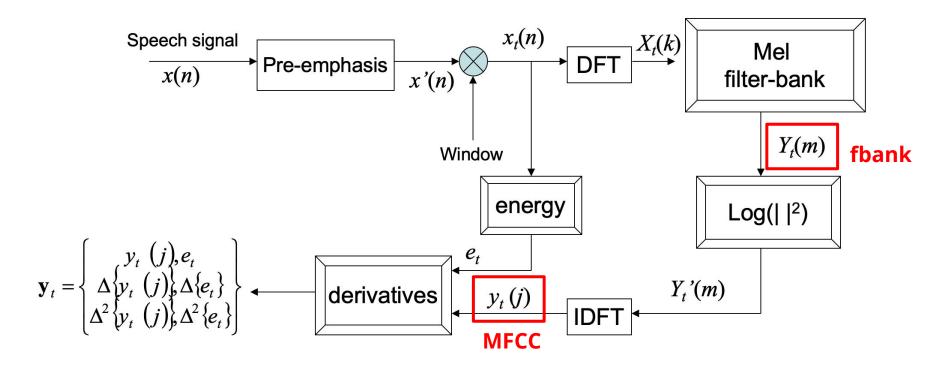
egs/librispeech/config/ctc_train_example.yaml

```
model:
  name: ctc_asr
  extractor:
                   feature
    name: fbank
                   extractor
    train: false
    feature: hidden_states
  encoder:
    hid_dim: 256
    n_layers: 3
                           model
    module: GRU
                           architecture
    bidirectional: true
    dropout: 0.2
```

```
optim:
  algo: Adam
  kwargs:
    lr: 0.0001
specaugment:
  freq_mask_range: [0, 20]
  freq_mask_num: 2
  time_mask_range: [0, 40]
  time_mask_num: 2
  time_mask_max: 1.0
  time_warp_w: 80
```

comment this part to disable SpecAugment

Feature Extractors



Feature Extractors

Source: https://github.com/s3prl/s3prl

Feature	Name	Default Dim	Stride	Window	Backend
Spectrogram	spectrogram	257	10ms	25ms	torchaudio-kaldi
FBANK	fbank	80 + delta1 + delta2	10ms	25ms	torchaudio-kaldi
MFCC	mfcc	13 + delta1 + delta2	10ms	25ms	torchaudio-kaldi
Mel	mel	80	10ms	25ms	torchaudio
Linear	linear	201	10ms	25ms	torchaudio

Step 2-1: Modify Config Files

egs/librispeech/config/ctc_train_example.yaml

```
hparam:
train_batch_size: 32
val_batch_size: 32
accum_grad: 1
grad_clip: 5
njobs: 4
pin_memory: true
```

PyTorch Lightning Trainer:

https://pytorch-lightning.readthedo cs.io/en/latest/common/trainer.htm l#trainer-class-api

```
trainer:
  max_epochs: 500
  max_steps: 100000
  check_val_every_n_epoch: 5
  gpus: 1
  precision: 16
  logger: true
  log_every_n_steps: 5
                                 model name
  flush_logs_every_n_steps: 5
 default_root_dir: model/ctc_libri-10h_char
  deterministic: true
```

Step 2-2: Modify Config Files via Command Line

Step 3: Training

```
minasr-asr --config egs/librispeech/config/ctc_train_example.yaml

or

python3 run_asr.py \
    --config egs/librispeech/config/ctc_train_example.yaml
```

The results will be saved in model/ctc_libri-10h_char/

Step 3: Training (Resuming from a Checkpoint)

minasr-asr --ckpt model/ctc_libri-10h_char/???.ckpt

Step 4: Testing

```
minasr-asr \
     --config egs/librispeech/config/ctc_test_example.yaml \
     --test \
     --ckpt model/ctc_libri-10h_char/???.ckpt
```

The results will be saved in model/ctc_libri-10h_char/

Step 4: Testing

- → MiniASR
 - data
 - egs
 - lightning_logs
 - miniasr
 - miniasr.egg-info
 - model
 - ctc_libri-10h_char
 - lightning_logs
 - epoch=19-step=1719.ckpt
 - epoch=29-step=2579.ckpt
 - epoch=34-step=3009.ckpt
 - epoch=39-step=3439.ckpt
 - epoch=44-step=3869.ckpt
 - hyps.txt
 - model_dev_config.yaml
 - model_train_config.yaml
 - refs.txt

Model Checkpoints

Hypothesis

Character Error Rate (CER)

Reference

```
I like math until they added the alphabet in it.
```

Hypothesis

```
I like mat until they addded the alphabat in it.

deletion insertion substitution
```

Word Error Rate (WER)

Reference

```
I like math until they added the alphabet in it.
```

Hypothesis

```
I like until they a added the alphapad in it.

deletion insertion substitution
```

Step 4: Testing character error rate (report this) substitution deletion insertion Character errors #Tok Sub Del Err SErr #Snt Ins 2620 281530 14.9 11.8 2.7 29.5 100.0 Word errors #Snt Del Err SErr #Tok Sub Ins 2620 52576 62.3 6.3 4.1 72.7 100.0 word error rate

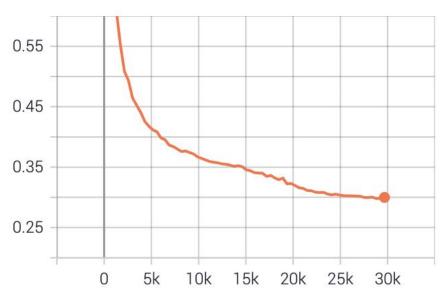
Baseline -- FBANK (CER%)

Method	dev-clean	test-clean
FBANK	29.8	29.5

Learning Curve

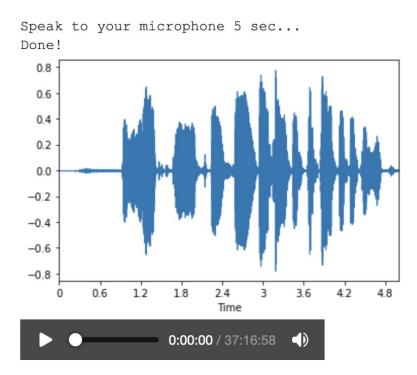
% tensorboard --logdir model

val_cer



Play with Your Model

https://github.com/vectominis
 t/MiniASR/blob/main/example
 /recognition.ipynb



```
[ ] waves = [load_waveform('audio_ds.wav').to('cuda')]
  hyps = model.recognize(waves)
  print(hyps[0])
```

I LIKED MATH UNTIL THEY ADDED THE ALPHABET IN IT

What you can do to improve performance?

- Modify configs
 - egs/librispeech/config/ctc_train_example.yaml
- Data augmentation
 - o w/ or w/o SpecAugment?
 - Speed perturbation
- Implement other model architectures
 - o Encoder's module (CNN, GRU, LSTM, Transformer, Conformer)
 - Framework (LAS / RNN-T)
- Input features
 - MFCC / fbank / spectrogram?

- CTC
 - https://www.cs.toronto.edu/~graves/icml_2006.pdf
 - http://proceedings.mlr.press/v32/graves14.pdf
 - https://distill.pub/2017/ctc/
- Data Augmentation
 - SpecAugment: https://arxiv.org/abs/1904.08779
 - Adaptive SpecAugment: https://arxiv.org/abs/1912.05533
 - Speed Perturbation:
 https://www.danielpovey.com/files/2015 interspeech augmentation.pdf
- ASR Frameworks
 - o RNN-T: https://arxiv.org/abs/1211.3711
 - LAS: https://arxiv.org/abs/1508.01211

- ASR Architectures
 - Transformer: https://arxiv.org/abs/1706.03762
 - Conformer: https://arxiv.org/abs/2005.08100
- Conventional ASR & Other Speech Processing Topics
 - http://ocw.aca.ntu.edu.tw/ntu-ocw/ocw/cou/104S204
- ASR Toolkits
 - Kaldi: https://github.com/kaldi-asr/kaldi (conventional ASR)
 - ESPnet: https://github.com/espnet/espnet
 - SpeechBrain: https://github.com/speechbrain/speechbrain
 - fairseq: https://github.com/pytorch/fairseq
 - flashlight: https://github.com/flashlight/flashlight

- Libraries
 - PyTorch: https://pytorch.org/docs/stable/index.html
 - PyTorch Ligtning: https://pytorch-lightning.readthedocs.io/en/latest/
 - torchaudio: https://pytorch.org/audio/stable/index.html
- Colab
 - Saving files to Google Drive:
 https://www.wongwonggoods.com/python/python-colab-mount-google-drive/