```
gpu info = !nvidia-smi
gpu_info = '\n'.join(gpu_info)
if gpu_info.find('failed') >= 0:
 print('Not connected to a GPU')
else:
 print(gpu_info)
   Thu Dec 7 17:43:56 2023
    NVIDIA-SMI 525.105.17 Driver Version: 525.105.17 CUDA Version: 12.0
   GPU Name Persistence-M Bus-Id Disp.A | Volatile Uncorr. ECC |
   | Fan Temp Perf Pwr:Usage/Cap| Memory-Usage | GPU-Util Compute M. | MTG M. |
   +-----
   Processes:
    GPU GI CI PID Type Process name
                                           GPU Memory |
                                            Usage
   No running processes found
```

We need the librosa package to load audio files and the jiwer to evaluate our fine-tuned model using the <u>word error rate (WER)</u> metric ¹.

```
%%capture
!pip install datasets==1.18.3
!pip install transformers==4.17.0
!pip install jiwer
```

Prepare Data, Tokenizer, Feature Extractor

- New Section
- Create Wav2Vec2CTCTokenizer

Let's start by loading the dataset and taking a look at its structure.

```
from datasets import load dataset, load metric
timit = load dataset("timit asr")
     Downloading:
                                                          7.06k/? [00:00<00:00, 442kB/s]
                                                          2.64k/? [00:00<00:00, 142kB/s]
     Downloading:
     Downloading and preparing dataset timit asr/clean (download: 828.75 MiB, generated: 7.96
     Downloading: 100%
                                                              869M/869M [02:45<00:00, 6.90MB/s]
     Dataset timit_asr downloaded and prepared to /root/.cache/huggingface/datasets/timit_asr
timit
     DatasetDict({
         train: Dataset({
             features: ['file', 'audio', 'text', 'phonetic_detail', 'word_detail',
     'dialect_region', 'sentence_type', 'speaker_id', 'id'],
             num rows: 4620
         })
         test: Dataset({
             features: ['file', 'audio', 'text', 'phonetic_detail', 'word_detail',
     'dialect_region', 'sentence_type', 'speaker_id', 'id'],
             num rows: 1680
         })
     })
timit = timit.remove_columns(["phonetic_detail", "word_detail", "dialect_region", "id", "ser
```

Let's write a short function to display some random samples of the dataset and run it a couple of times to get a feeling for the transcriptions.

```
from datasets import ClassLabel
import random
import pandas as pd
from IPython.display import display, HTML
def show random elements(dataset, num examples=10):
    assert num_examples <= len(dataset), "Can't pick more elements than there are in the dat
    picks = []
    for in range(num examples):
        pick = random.randint(0, len(dataset)-1)
        while pick in picks:
             pick = random.randint(0, len(dataset)-1)
        picks.append(pick)
    df = pd.DataFrame(dataset[picks])
    display(HTML(df.to html()))
show random elements(timit["train"].remove columns(["audio", "file"]), num examples=10)
                                                                    text
      0
                              What targets have we successfully knocked out?
      1
                                   Both figures would go higher in later years.
      2
                                             Jeff's toy go-cart never worked!
                                           Does Hindu ideology honor cows?
      3
      4
                                       Should giraffes be kept in small zoos?
      5
         And their chroniclers are not the dramatic poets but the prose novelists.
      6
                                     It is a big project, not to be taken lightly.
      7
                   We apply auditory modeling to computer speech recognition.
      8
                                    Don't ask me to carry an oily rag like that.
      9
           Those who are not purists use canned vegetables when making stew.
import re
chars_to_ignore_regex = '[\,\?\.\!\-\;\:\"]'
def remove special characters(batch):
    batch["text"] = re.sub(chars_to_ignore_regex, '', batch["text"]).lower() + " "
    return batch
timit = timit.map(remove_special_characters)
```

```
WARNING:datasets.fingerprint:Parameter 'function'=<function remove_special_characters at
    4620/? [00:00<00:00, 10758.98ex/s]
    1680/? [00:00<00:00, 7869.35ex/s]
```

```
show_random_elements(timit["train"].remove_columns(["audio", "file"]))
```

100%

	text		
0	often you'll get back more than you put in		
1	to be passive to be girlishly shy was palpably absurd		
2	don't ask me to carry an oily rag like that		
3	the hallway opens into a huge chamber		
4	irish youngsters eat fresh kippers for breakfast		
5	don't ask me to carry an oily rag like that		
6	there are canoes ideal for fishing in protected waters or for camping trips		
7	don't ask me to carry an oily rag like that		
8	turbulent tides rose as much as fifty feet		
9	trespassing is forbidden and subject to penalty		
all_te vocab	ract_all_chars(batch): ext = " ".join(batch["text"]) = list(set(all_text)) ext = " ".yocab": [vocab], "all_text": [all_text]}		
cabs =	timit.map(extract_all_chars, batched=True, batch_size=-1, keep_in_memory=True,		
100	1/1 [00:00<00:00, 15.26ba/s]		

Now, we create the union of all distinct letters in the training dataset and test dataset and convert the resulting list into an enumerated dictionary.

1/1 [00:00<00:00, 23.64ba/s]

```
vocab_list = list(set(vocabs["train"]["vocab"][0]) | set(vocabs["test"]["vocab"][0]))
vocab_dict = {v: k for k, v in enumerate(vocab_list)}
vocab_dict
```

```
{'n': 0,
       'f': 1,
      'w': 2,
      "'": 3,
      's': 4,
      'k': 5,
       'c': 6,
       '1': 7,
       'e': 8,
       'z': 9,
       'u': 10,
       'v': 11,
       'y': 12,
       'a': 13,
      'x': 14,
       'm': 15,
       'q': 16,
      'j': 17,
       'd': 18,
      'h': 19,
       't': 20,
       'i': 21,
       'o': 22,
       'b': 23,
       'g': 24,
       'r': 25,
      ' ': 26,
       'p': 27}
vocab_dict["|"] = vocab_dict[" "]
del vocab_dict[" "]
vocab_dict["[UNK]"] = len(vocab_dict)
vocab_dict["[PAD]"] = len(vocab_dict)
len(vocab_dict)
     30
```

Let's now save the vocabulary as a json file.

```
import json
with open('vocab.json', 'w') as vocab_file:
    json.dump(vocab_dict, vocab_file)
```

In a final step, we use the json file to instantiate an object of the Wav2Vec2CTCTokenizer class.

```
from transformers import Wav2Vec2CTCTokenizer
```

```
tokenizer = Wav2Vec2CTCTokenizer("./vocab.json", unk_token="[UNK]", pad_token="[PAD]", word_
```

Create Wav2Vec2 Feature Extractor

New Section

```
from transformers import Wav2Vec2FeatureExtractor

feature_extractor = Wav2Vec2FeatureExtractor(feature_size=1, sampling_rate=16000, padding_va

from transformers import Wav2Vec2Processor

processor = Wav2Vec2Processor(feature_extractor=feature_extractor, tokenizer=tokenizer)
```

Next, we can prepare the dataset.

```
timit["train"][0]["file"]
```

'/root/.cache/huggingface/datasets/downloads/extracted/404950a46da14eac65eb4e2a8317b1372fh3971d980d91d5d5h221275h1fd7e0/data/TRATN/DR4/MMDM0/ST681.WAV'

Wav2Vec2 expects the input in the format of a 1-dimensional array of 16 kHz. This means that the audio file has to be loaded and resampled.

Thankfully, datasets does this automatically when calling the column audio. Let try it out.

```
import IPython.display as ipd
import numpy as np
import random
rand int = random.randint(0, len(timit["train"]))
print(timit["train"][rand_int]["text"])
ipd.Audio(data=np.asarray(timit["train"][rand_int]["audio"]["array"]), autoplay=True, rate=1
     the proof that you are seeking is not available in books
           0:00 / 0:03
rand_int = random.randint(0, len(timit["train"]))
print("Target text:", timit["train"][rand_int]["text"])
print("Input array shape:", np.asarray(timit["train"][rand int]["audio"]["array"]).shape)
print("Sampling rate:", timit["train"][rand_int]["audio"]["sampling_rate"])
     Target text: first add milk to the shredded cheese
     Input array shape: (38605,)
     Sampling rate: 16000
def prepare dataset(batch):
    audio = batch["audio"]
    # batched output is "un-batched" to ensure mapping is correct
    batch["input_values"] = processor(audio["array"], sampling_rate=audio["sampling_rate"]).
    batch["input length"] = len(batch["input values"])
    with processor.as_target_processor():
        batch["labels"] = processor(batch["text"]).input_ids
    return batch
Let's apply the data preparation function to all examples.
timit = timit.map(prepare dataset, remove columns=timit.column names["train"], num proc=4)
max_input_length_in_sec = 4.0
timit["train"] = timit["train"].filter(lambda x: x < max_input_length_in_sec * processor.fea</pre>
     100%
                                                  5/5 [00:00<00:00, 154.87ba/s]
```

Awesome, now we are ready to start training!

```
import torch
from dataclasses import dataclass, field
from typing import Any, Dict, List, Optional, Union
@dataclass
class DataCollatorCTCWithPadding:
   Data collator that will dynamically pad the inputs received.
        processor (:class:`~transformers.Wav2Vec2Processor`)
            The processor used for processing the data.
        padding (:obj:`bool`, :obj:`str` or :class:`~transformers.tokenization_utils_base.Pa
            Select a strategy to pad the returned sequences (according to the model's paddir
            among:
            * :obj:`True` or :obj:`'longest'`: Pad to the longest sequence in the batch (or
              sequence if provided).
            * :obj:`'max length'`: Pad to a maximum length specified with the argument :obj:
              maximum acceptable input length for the model if that argument is not provided
            * :obj:`False` or :obj:`'do_not_pad'` (default): No padding (i.e., can output a
              different lengths).
    .....
    processor: Wav2Vec2Processor
    padding: Union[bool, str] = True
    def __call__(self, features: List[Dict[str, Union[List[int], torch.Tensor]]]) -> Dict[st
        # split inputs and labels since they have to be of different lenghts and need
        # different padding methods
        input features = [{"input values": feature["input values"]} for feature in features]
        label features = [{"input ids": feature["labels"]} for feature in features]
        batch = self.processor.pad(
            input_features,
            padding=self.padding,
            return tensors="pt",
        with self.processor.as_target_processor():
            labels_batch = self.processor.pad(
                label_features,
                padding=self.padding,
                return tensors="pt",
            )
        # replace padding with -100 to ignore loss correctly
        labels = labels_batch["input_ids"].masked_fill(labels_batch.attention_mask.ne(1), -1
        batch["labels"] = labels
        return batch
```

```
data_collator = DataCollatorCTCWithPadding(processor=processor, padding=True)
```

Next, the evaluation metric is defined. As mentioned earlier, the predominant metric in ASR is the word error rate (WER), hence we will use it in this notebook as well.

```
wer_metric = load_metric("wer")
     Downloading:
                                                          4.48k/? [00:00<00:00, 266kB/s]
from transformers import TrainingArguments
output_dir = '/content/repo_name/'
training args = TrainingArguments(
  output_dir=output_dir,
  group_by_length=True,
  per_device_train_batch_size=8,
  evaluation_strategy="steps",
  num_train_epochs=30,
  fp16=True,
  gradient_checkpointing=True,
  save steps=500,
  eval_steps=500,
  logging steps=500,
  learning_rate=1e-4,
  weight_decay=0.005,
  warmup steps=1000,
  save_total_limit=2,
)
```

Now, all instances can be passed to Trainer and we are ready to start training!

```
from transformers import Trainer

trainer = Trainer(
    model=model,
    data_collator=data_collator,
    args=training_args,
    compute_metrics=compute_metrics,
    train_dataset=timit["train"],
    eval_dataset=timit["test"],
    tokenizer=processor.feature_extractor,
)

Using amp half precision backend
```

```
function ConnectButton(){
    console.log("Connect pushed");
    document.querySelector("#top-toolbar > colab-connect-button").shadowRoot.querySelector("#connect")
}
setInterval(ConnectButton,60000);
```

trainer.train()

①

The following columns in the training set don't have a corresponding argument in `Wav2\ /usr/local/lib/python3.10/dist-packages/transformers/optimization.py:306: FutureWarning: warnings.warn(

***** Running training *****

Num examples = 3978

Num Epochs = 30

Instantaneous batch size per device = 8

Total train batch size (w. parallel, distributed & accumulation) = 8

Gradient Accumulation steps = 1

Total optimization steps = 14940

/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

[2501/14940 18:59 < 1:34:33, 2.19 it/s, Epoch 5.02/30]

Step Training Loss Validation Loss Wer 500 3.499400 1.680385 0.964579 1000 0.845700 0.563984 0.545310 1500 0.438600 0.462594 0.458687 2000 0.297700 0.422367 0.422162

___ [76/210 00:21 < 00:37, 3.53 it/s]

The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****

Num examples = 1680

Batch size = 8

Saving model checkpoint to /content/repo_name/checkpoint-500

Configuration saved in /content/repo name/checkpoint-500/config.json

Model weights saved in /content/repo name/checkpoint-500/pytorch model.bin

Feature extractor saved in /content/repo_name/checkpoint-500/preprocessor_config.json /usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****

Num examples = 1680

Batch size = 8

Saving model checkpoint to /content/repo_name/checkpoint-1000

Configuration saved in /content/repo_name/checkpoint-1000/config.json

Model weights saved in /content/repo name/checkpoint-1000/pytorch model.bin

Feature extractor saved in /content/repo_name/checkpoint-1000/preprocessor_config.json /usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****

Num examples = 1680

Batch size = 8

Saving model checkpoint to /content/repo name/checkpoint-1500

Configuration saved in /content/repo name/checkpoint-1500/config.json

Model weights saved in /content/repo_name/checkpoint-1500/pytorch_model.bin

Feature extractor saved in /content/repo_name/checkpoint-1500/preprocessor_config.json Deleting older checkpoint [/content/repo_name/checkpoint-500] due to args.save_total_lin/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

The following columns in the evaluation set don't have a corresponding argument in `Wav

***** Kunning Evaluation *****
Num examples = 1680

Batch size = 8

Saving model checkpoint to /content/repo_name/checkpoint-2000

Configuration saved in /content/repo_name/checkpoint-2000/config.json

Model weights saved in /content/repo name/checkpoint-2000/pytorch model.bin

Feature extractor saved in /content/repo_name/checkpoint-2000/preprocessor_config.json Deleting older checkpoint [/content/repo_name/checkpoint-1000] due to args.save_total_li/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****

Num examples = 1680 Batch size = 8

[9001/14940 1:09:59 < 46:11, 2.14 it/s, Epoch 18.07/30]

Step	Training Loss	Validation Loss	Wer
500	3.499400	1.680385	0.964579
1000	0.845700	0.563984	0.545310
1500	0.438600	0.462594	0.458687
2000	0.297700	0.422367	0.422162
2500	0.227100	0.435139	0.410861
3000	0.194400	0.429970	0.394666
3500	0.159300	0.498936	0.390600
4000	0.137700	0.412528	0.392737
4500	0.124900	0.434000	0.376886
5000	0.108900	0.429038	0.382951
5500	0.101700	0.418560	0.371718
6000	0.092500	0.530390	0.384122
6500	0.084200	0.504215	0.371718
7000	0.072900	0.509529	0.373716
7500	0.069600	0.473375	0.372407
8000	0.060100	0.441955	0.365447
8500	0.059300	0.507855	0.358142

[4/210 00:00 < 01:00, 3.40 it/s]

Saving model checkpoint to /content/repo_name/checkpoint-2500

Configuration saved in /content/repo_name/checkpoint-2500/config.json

Model weights saved in /content/repo name/checkpoint-2500/pytorch model.bin

Feature extractor saved in /content/repo_name/checkpoint-2500/preprocessor_config.json Deleting older checkpoint [/content/repo_name/checkpoint-1500] due to args.save_total_li/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc warnings.warn(

The following columns in the evaluation set don't have a corresponding argument in `Wax

```
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-3000
Configuration saved in /content/repo name/checkpoint-3000/config.json
Model weights saved in /content/repo name/checkpoint-3000/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-3000/preprocessor config.json
Deleting older checkpoint [/content/repo name/checkpoint-2000] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wav
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-3500
Configuration saved in /content/repo name/checkpoint-3500/config.json
Model weights saved in /content/repo name/checkpoint-3500/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-3500/preprocessor config.json
Deleting older checkpoint [/content/repo_name/checkpoint-2500] due to args.save_total_li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wav
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-4000
Configuration saved in /content/repo name/checkpoint-4000/config.json
Model weights saved in /content/repo name/checkpoint-4000/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-4000/preprocessor config.json
Deleting older checkpoint [/content/repo name/checkpoint-3000] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wax
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-4500
Configuration saved in /content/repo_name/checkpoint-4500/config.json
Model weights saved in /content/repo name/checkpoint-4500/pytorch model.bin
Feature extractor saved in /content/repo_name/checkpoint-4500/preprocessor_config.json
Deleting older checkpoint [/content/repo_name/checkpoint-3500] due to args.save_total_li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-5000
Configuration saved in /content/repo_name/checkpoint-5000/config.json
Model weights saved in /content/repo name/checkpoint-5000/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-5000/preprocessor config.json
Deleting older checkpoint [/content/repo_name/checkpoint-4000] due to args.save_total_li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wax
***** Running Evaluation *****
```

```
Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo_name/checkpoint-5500
Configuration saved in /content/repo_name/checkpoint-5500/config.json
Model weights saved in /content/repo name/checkpoint-5500/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-5500/preprocessor config.json
Deleting older checkpoint [/content/repo_name/checkpoint-4500] due to args.save_total_li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wav
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-6000
Configuration saved in /content/repo_name/checkpoint-6000/config.json
Model weights saved in /content/repo name/checkpoint-6000/pytorch model.bin
Feature extractor saved in /content/repo_name/checkpoint-6000/preprocessor_config.json
Deleting older checkpoint [/content/repo name/checkpoint-5000] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-6500
Configuration saved in /content/repo_name/checkpoint-6500/config.json
Model weights saved in /content/repo name/checkpoint-6500/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-6500/preprocessor config.json
Deleting older checkpoint [/content/repo name/checkpoint-5500] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wav
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo name/checkpoint-7000
Configuration saved in /content/repo name/checkpoint-7000/config.json
Model weights saved in /content/repo name/checkpoint-7000/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-7000/preprocessor config.json
Deleting older checkpoint [/content/repo name/checkpoint-6000] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
Saving model checkpoint to /content/repo_name/checkpoint-7500
Configuration saved in /content/repo name/checkpoint-7500/config.json
Model weights saved in /content/repo name/checkpoint-7500/pytorch model.bin
Feature extractor saved in /content/repo name/checkpoint-7500/preprocessor config.json
Deleting older checkpoint [/content/repo name/checkpoint-6500] due to args.save total li
/usr/local/lib/python3.10/dist-packages/torch/utils/checkpoint.py:429: UserWarning: torc
  warnings.warn(
The following columns in the evaluation set don't have a corresponding argument in `Wa\
***** Running Evaluation *****
  Num examples = 1680
  Batch size = 8
```