



+ Code + Text

✓ RAM
Disk

Editing



```
[2] from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
[4] # %%capture
!pip install -U torch==1.6.0 sentence-transformers==0.3.3 transformers==3.0.2 faiss-cpu==1.6.3
```

Custom Dataloader

```
[32] from torch.utils.data import Dataset
from sklearn.datasets import fetch_20newsgroups

from typing import List
import bisect
import torch
import logging
import numpy as np
from tqdm import tqdm
from sentence_transformers import SentenceTransformer
from sentence_transformers.readers import InputExample

class Dataset20news(Dataset):

    def __init__(self, model: SentenceTransformer, provide_positive: bool = True,
                 provide_negative: bool = True):
        """
        Converts input examples to a SentenceLabelDataset usable to train the model with
        SentenceTransformer.smart_batching_collate as the collate_fn for the DataLoader
        Assumes only one sentence per InputExample and labels as integers from 0 to max_num_labels
        and should be used in combination with dataset_reader.LabelSentenceReader.
        Labels with only one example are ignored.
        smart_batching_collate as collate_fn is required because it transforms the tokenized texts to the tensors.
        :param examples:
            the input examples for the training
        :param model:
            the Sentence BERT model for the conversion
        :param provide_positive:
            set this to False, if you don't need a positive example (e.g. for BATCH_HARD_TRIPLET_LOSS).
        :param provide_negative:
            set this to False, if you don't need a negative example (e.g. for BATCH_HARD_TRIPLET_LOSS
            or MULTIPLE_NEGATIVES_RANKING_LOSS).
        """
        self.model = model
        self.groups_right_border = []
        self.grouped_inputs = []
        self.grouped_labels = []

        self.data = self.download_dataset()
        self.convert_input_examples(self.data[0], self.model)

        self.idx = np.arange(len(self.grouped_inputs))

        self.provide_positive = provide_positive
        self.provide_negative = provide_negative

    def download_dataset(self):
        dataset_all = {}

        for subset in ['train', 'test']:
            dataset = fetch_20newsgroups(subset=subset, remove=('headers', 'footers', 'quotes'), shuffle=True)

            dataset_all[subset] = [InputExample(guid=id, texts=[text], label=label) for id, (text, label) in enumerate(zip(dataset.data, dataset.target))]

            # only 100 elements to quickly test the pipeline
            if subset == "train":
                dataset_all[subset] = dataset_all[subset][:100]

        return dataset_all['train'], dataset_all['test']

    def convert_input_examples(self, examples: List[InputExample], model: SentenceTransformer):
        """
        Converts input examples to a SentenceLabelDataset.
        """
```

```

Assumes only one sentence per InputExample and labels as integers from 0 to max_num_labels
and should be used in combination with dataset_reader.LabelSentenceReader.
Labels with only one example are ignored.
:param examples:
    the input examples for the training
:param model
    the Sentence Transformer model for the conversion
"""

inputs = []
labels = []

label_sent_mapping = {}
too_long = 0
label_type = None

# Group examples and labels
# Add examples with the same label to the same dict
for ex_index, example in enumerate(tqdm(examples, desc="Convert dataset")):
    if label_type is None:
        if isinstance(example.label, int):
            label_type = torch.long
        elif isinstance(example.label, float):
            label_type = torch.float

        # tokenized_text = model.tokenize(example.texts)['input_ids'][0]

        # tokenized_text = example.texts

        tokenized_text = model.tokenizer.encode(example.texts[0], padding="max_length", max_length=512, truncation=True,)

        if hasattr(model, 'max_seq_length') and model.max_seq_length is not None and model.max_seq_length > 0 and len(tokenized_text) > model.max_seq_length:
            too_long += 1

        if example.label in label_sent_mapping:
            label_sent_mapping[example.label].append(ex_index)
        else:
            label_sent_mapping[example.label] = [ex_index]

        inputs.append(tokenized_text)
        labels.append(example.label)

# Group sentences, such that sentences with the same label
# are besides each other. Only take labels with at least 2 examples
distinct_labels = list(label_sent_mapping.keys())
for i in range(len(distinct_labels)):
    label = distinct_labels[i]
    if len(label_sent_mapping[label]) >= 2:
        self.grouped_inputs.extend([inputs[j] for j in label_sent_mapping[label]])
        self.grouped_labels.extend([labels[j] for j in label_sent_mapping[label]])
        self.groups_right_border.append(len(self.grouped_inputs)) #At which position does this label group / bucket end?

self.grouped_labels = torch.tensor(self.grouped_labels, dtype=label_type)
logging.info("Num sentences: %d" % (len(self.grouped_inputs)))
logging.info("Sentences longer than max_sequence_length: {}".format(too_long))
logging.info("Number of labels with >1 examples: {}".format(len(distinct_labels)))

def __getitem__(self, item):
    if not self.provide_positive and not self.provide_negative:
        return [self.grouped_inputs[item]], self.grouped_labels[item]

    # Anchor element
    anchor = self.grouped_inputs[item]

    # Check start and end position for this label in our list of grouped sentences
    group_idx = bisect.bisect_right(self.groups_right_border, item)
    left_border = 0 if group_idx == 0 else self.groups_right_border[group_idx - 1]
    right_border = self.groups_right_border[group_idx]

    if self.provide_positive:
        positive_item_idx = np.random.choice(np.concatenate([self.idxes[left_border:item], self.idxes[item + 1:right_border]]))
        positive = self.grouped_inputs[positive_item_idx]
    else:
        positive = []

    if self.provide_negative:
        negative_item_idx = np.random.choice(np.concatenate([self.idxes[0:left_border], self.idxes[right_border:]]))
        negative = self.grouped_inputs[negative_item_idx]
    else:
        negative = []

    return [anchor, positive, negative], self.grouped_labels[item]

```

```
def __len__(self):
    return len(self.grouped_inputs)
```

➤ Loading model and dataloader

```
✓ [33] from sentence_transformers import SentenceTransformer
24s from torch.utils.data import DataLoader

# load model
model_name = 'distilbert-base-nli-mean-tokens'
model = SentenceTransformer(model_name)

# load dataset
train_batch_size = 8
train_dataset = Dataset20news(model=model)
train_dataloader = DataLoader(train_dataset,
                              batch_size=train_batch_size,
                              shuffle=True,
                              )

100%|██████████| 245M/245M [00:06<00:00, 35.2MB/s]
Downloading 20news dataset. This may take a few minutes.
INFO:sklearn.datasets.twenty_newsgroups:Downloading 20news dataset. This may take a few minutes.
Downloading dataset from https://ndownloader.figshare.com/files/5975967 (14 MB)
INFO:sklearn.datasets.twenty_newsgroups:Downloading dataset from https://ndownloader.figshare.com/files/5975967 (14 MB)
Convert dataset: 100%|██████████| 100/100 [00:00<00:00, 189.24it/s]
```

➤ Evaluation and test loader

```
✓ [34] import random
from sentence_transformers.evaluation import TripletEvaluator
from tqdm import tqdm

✓ [35] test_dataset = train_dataset.data[1]
val_dataset, test_dataset = test_dataset[:len(test_dataset)//2], test_dataset[len(test_dataset)//2:]
```

```
✓ [36] def examples_to_triplets(examples):
2s
    triplets = []
    grouped_examples = {}

    for example in examples:
        if example.label in grouped_examples.keys():
            grouped_examples[example.label].append(example)
        else:
            grouped_examples[example.label] = [example]

    for example in tqdm(examples) :
        pos = example
        while pos.guid == example.guid:
            pos = random.choice([sample for sample in grouped_examples[example.label]])

        neg_label = random.choice([i for i in grouped_examples if i != example.label])
        neg = random.choice(grouped_examples[neg_label])

        triplets.append(InputExample(texts=[example.texts[0], pos.texts[0], neg.texts[0]]))

    return triplets
```

```
✓ [37] examples = val_dataset
2s triplets = examples_to_triplets(examples)
triplets = triplets[:10]
dev_evaluator = TripletEvaluator.from_input_examples(triplets)
dev_evaluator(model)
```

```
100%|██████████| 3766/3766 [00:00<00:00, 17020.55it/s]
0.6
```

➤ Model fine-tuning

```
✓ [38] from sentence_transformers import losses
10s
train_loss = losses.TripletLoss(model)


#Tune the model
model.fit(train_objectives=[(train_dataloader, train_loss)],
          evaluator=dev_evaluator,
          epochs=1,
          evaluation_steps=1000,
          warmup_steps=100,
          output_path="/content/drive/MyDrive/vocads_challenge/model")
```

Epoch: 100%  1/1 [00:10<00:00, 10.12s/it]

Iteration: 100%  13/13 [00:07<00:00, 1.72it/s]

▼ Testing the model

```
✓ [39] examples = test_dataset
3s
triplets = examples_to_triplets(examples)
triplets = triplets[:100]
test_evaluator = TripletEvaluator.from_input_examples(triplets)
test_evaluator(model)
```

100%  | 3766/3766 [00:00<00:00, 48701.21it/s]
0.66

▼ Building embeddings

```
✓ [27] %cd /content/drive/MyDrive/vocads_challenge
0s
/content/drive/MyDrive/vocads_challenge
```

```
✓ [29] !pip freeze > requirements.txt
```

```
!python build_embeddings.py
```

Batches: 100% 1415/1415 [03:32<00:00, 6.65it/s]

▼ Testing app.py

```
[1] !pip install flask-ngrok
!pip install flask==0.12.2
```

```
✓ [15] !python app.py
```