

```
In [13]: from google.colab import drive
drive.mount('/content/gdrive')
```

Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount("/content/gdrive", force_remount=True).

```
In [14]: %cd /content/gdrive/My Drive/Colab Notebooks/Notebooks/Luodingo
```

```
import torch
import torch.optim as optim

import seq2seq_multilayer_gru
from sequence_model_trainer import TrainModel

from torchtext.data import Field, LabelField
from torchtext.data import TabularDataset
from torchtext.data import Iterator, BucketIterator

%load_ext autoreload
%autoreload 2
```

/content/gdrive/My Drive/Colab Notebooks/Notebooks/Luodingo
The autoreload extension is already loaded. To reload it, use:
%reload_ext autoreload

```
In [0]: MASKED_TEXT = Field(
        sequential=True,
        tokenize=lambda x: x.split(),
    )

    TARGET_TEXT = Field(
        sequential=True,
        tokenize=lambda x: x.split(),
    )

    # LABEL = LabelField(dtype=torch.Long)

    fields = [('id', None), ('keywords', MASKED_TEXT), ('target', TARGET_TEXT)]
```

```
In [16]: %cd /content/gdrive/My Drive/Colab Notebooks/Datasets
train, val, test = TabularDataset.splits(
    path='Masked Corpus',
    train='train.csv',
    validation='val.csv',
    test='test.csv',
    format='csv',
    skip_header=True,
    fields=fields
)
```

/content/gdrive/My Drive/Colab Notebooks/Datasets

```
In [0]: MASKED_TEXT.build_vocab(train)
TARGET_TEXT.build_vocab(train)
# LABEL.build_vocab(train)
```

```
In [0]: BATCH_SIZE = 32

device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')

train_iter, val_iter, test_iter = BucketIterator.splits(
    (train, val, test),
    batch_size=BATCH_SIZE,
    device=device,
    sort_key=lambda x: len(x.keywords))
```

```
In [0]: # EMB_DIM=bert.config.to_dict()['hidden_size']
EMB_DIM=256
ENC_INPUT_DIM=len(MASKED_TEXT.vocab)
DEC_INPUT_DIM=len(TARGET_TEXT.vocab)
OUTPUT_DIM=DEC_INPUT_DIM
N_LAYER=4
HID_DIM=1024
DROPOUT=0.3

model = seq2seq_multilayer_gru.Seq2Seq(
    enc_input_dim=ENC_INPUT_DIM,
    dec_input_dim=DEC_INPUT_DIM,
    emb_dim=EMB_DIM,
    enc_hid_dim=HID_DIM,
    dec_hid_dim=HID_DIM,
    n_layers=N_LAYER,
    output_dim=OUTPUT_DIM,
    device=device,
    dropout=DROPOUT
).to(device)
```

```
In [0]: LEARNING_RATE = 0.0001

adam = torch.optim.Adam(model.parameters(), lr=LEARNING_RATE)
cross_e = torch.nn.CrossEntropyLoss()
```

```
In [0]: trainer = TrainModel(
    model=model,
    train_iterator=train_iter,
    val_iterator=val_iter,
    optimizer=adam,
    criterion=cross_e,
    output_dim=OUTPUT_DIM
)
```

```
In [0]: N_EPOCHS = 200
        CLIP = 1
        trainer.epoch(n_epochs=N_EPOCHS, clip=CLIP, model_name='seq2seq-multilayer-gru.pt')
```

```
Epoch: 01 | Time: 0m 51s
        Train Loss: 3.570 | Train PPL: 35.501
        Val. Loss: 5.155 | Val. PPL: 173.218
Epoch: 02 | Time: 0m 51s
        Train Loss: 3.024 | Train PPL: 20.564
        Val. Loss: 5.115 | Val. PPL: 166.445
Epoch: 03 | Time: 0m 51s
        Train Loss: 2.951 | Train PPL: 19.126
        Val. Loss: 5.139 | Val. PPL: 170.565
Epoch: 04 | Time: 0m 51s
        Train Loss: 2.892 | Train PPL: 18.024
        Val. Loss: 5.233 | Val. PPL: 187.324
Epoch: 05 | Time: 0m 51s
        Train Loss: 2.821 | Train PPL: 16.801
        Val. Loss: 5.093 | Val. PPL: 162.935
Epoch: 06 | Time: 0m 51s
        Train Loss: 2.740 | Train PPL: 15.490
        Val. Loss: 4.900 | Val. PPL: 134.298
Epoch: 07 | Time: 0m 51s
        Train Loss: 2.694 | Train PPL: 14.797
        Val. Loss: 4.945 | Val. PPL: 140.430
Epoch: 08 | Time: 0m 51s
        Train Loss: 2.714 | Train PPL: 15.091
        Val. Loss: 4.972 | Val. PPL: 144.382
Epoch: 09 | Time: 0m 51s
        Train Loss: 2.646 | Train PPL: 14.092
        Val. Loss: 4.867 | Val. PPL: 129.919
Epoch: 10 | Time: 0m 51s
        Train Loss: 2.534 | Train PPL: 12.601
        Val. Loss: 4.685 | Val. PPL: 108.356
Epoch: 11 | Time: 0m 51s
        Train Loss: 2.411 | Train PPL: 11.142
        Val. Loss: 4.754 | Val. PPL: 116.044
Epoch: 12 | Time: 0m 51s
        Train Loss: 2.337 | Train PPL: 10.352
        Val. Loss: 4.677 | Val. PPL: 107.442
Epoch: 13 | Time: 0m 51s
        Train Loss: 2.292 | Train PPL: 9.896
        Val. Loss: 4.454 | Val. PPL: 85.932
Epoch: 14 | Time: 0m 51s
        Train Loss: 2.219 | Train PPL: 9.197
        Val. Loss: 4.563 | Val. PPL: 95.904
Epoch: 15 | Time: 0m 51s
        Train Loss: 2.149 | Train PPL: 8.572
        Val. Loss: 4.210 | Val. PPL: 67.335
Epoch: 16 | Time: 0m 51s
        Train Loss: 2.084 | Train PPL: 8.036
        Val. Loss: 4.283 | Val. PPL: 72.488
Epoch: 17 | Time: 0m 51s
        Train Loss: 2.009 | Train PPL: 7.457
        Val. Loss: 4.084 | Val. PPL: 59.386
Epoch: 18 | Time: 0m 51s
        Train Loss: 1.939 | Train PPL: 6.949
        Val. Loss: 3.955 | Val. PPL: 52.179
Epoch: 19 | Time: 0m 51s
        Train Loss: 1.852 | Train PPL: 6.376
        Val. Loss: 3.932 | Val. PPL: 51.030
Epoch: 20 | Time: 0m 51s
        Train Loss: 1.810 | Train PPL: 6.108
        Val. Loss: 3.806 | Val. PPL: 44.981
Epoch: 21 | Time: 0m 51s
        Train Loss: 1.746 | Train PPL: 5.732
        Val. Loss: 3.685 | Val. PPL: 39.843
Epoch: 22 | Time: 0m 51s
        Train Loss: 1.675 | Train PPL: 5.338
        Val. Loss: 3.451 | Val. PPL: 31.532
Epoch: 23 | Time: 0m 51s
        Train Loss: 1.609 | Train PPL: 5.000
        Val. Loss: 3.392 | Val. PPL: 29.721
Epoch: 24 | Time: 0m 51s
        Train Loss: 1.574 | Train PPL: 4.824
        Val. Loss: 3.287 | Val. PPL: 26.774
Epoch: 25 | Time: 0m 51s
        Train Loss: 1.515 | Train PPL: 4.549
        Val. Loss: 3.236 | Val. PPL: 25.425
Epoch: 26 | Time: 0m 51s
        Train Loss: 1.465 | Train PPL: 4.328
```

	Val. Loss: 3.191	Val. PPL: 24.320
Epoch: 27	Time: 0m 51s	
	Train Loss: 1.427	Train PPL: 4.166
	Val. Loss: 3.188	Val. PPL: 24.247
Epoch: 28	Time: 0m 51s	
	Train Loss: 1.394	Train PPL: 4.032
	Val. Loss: 3.037	Val. PPL: 20.853
Epoch: 29	Time: 0m 52s	
	Train Loss: 1.354	Train PPL: 3.874
	Val. Loss: 2.909	Val. PPL: 18.337
Epoch: 30	Time: 0m 51s	
	Train Loss: 1.310	Train PPL: 3.705
	Val. Loss: 3.124	Val. PPL: 22.730
Epoch: 31	Time: 0m 50s	
	Train Loss: 1.294	Train PPL: 3.647
	Val. Loss: 2.930	Val. PPL: 18.725
Epoch: 32	Time: 0m 50s	
	Train Loss: 1.249	Train PPL: 3.488
	Val. Loss: 2.871	Val. PPL: 17.653
Epoch: 33	Time: 0m 50s	
	Train Loss: 1.215	Train PPL: 3.372
	Val. Loss: 2.718	Val. PPL: 15.157
Epoch: 34	Time: 0m 51s	
	Train Loss: 1.176	Train PPL: 3.241
	Val. Loss: 2.758	Val. PPL: 15.765
Epoch: 35	Time: 0m 51s	
	Train Loss: 1.148	Train PPL: 3.151
	Val. Loss: 2.918	Val. PPL: 18.510
Epoch: 36	Time: 0m 51s	
	Train Loss: 1.116	Train PPL: 3.053
	Val. Loss: 2.692	Val. PPL: 14.763
Epoch: 37	Time: 0m 51s	
	Train Loss: 1.083	Train PPL: 2.954
	Val. Loss: 2.778	Val. PPL: 16.081
Epoch: 38	Time: 0m 51s	
	Train Loss: 1.046	Train PPL: 2.848
	Val. Loss: 2.640	Val. PPL: 14.015
Epoch: 39	Time: 0m 51s	
	Train Loss: 1.027	Train PPL: 2.793
	Val. Loss: 2.574	Val. PPL: 13.124
Epoch: 40	Time: 0m 51s	
	Train Loss: 1.006	Train PPL: 2.735
	Val. Loss: 2.615	Val. PPL: 13.669
Epoch: 41	Time: 0m 50s	
	Train Loss: 0.974	Train PPL: 2.648
	Val. Loss: 2.715	Val. PPL: 15.099
Epoch: 42	Time: 0m 50s	
	Train Loss: 0.949	Train PPL: 2.583
	Val. Loss: 2.490	Val. PPL: 12.058
Epoch: 43	Time: 0m 51s	
	Train Loss: 0.914	Train PPL: 2.494
	Val. Loss: 2.478	Val. PPL: 11.918
Epoch: 44	Time: 0m 51s	
	Train Loss: 0.891	Train PPL: 2.436
	Val. Loss: 2.383	Val. PPL: 10.836
Epoch: 45	Time: 0m 51s	
	Train Loss: 0.857	Train PPL: 2.356
	Val. Loss: 2.496	Val. PPL: 12.130
Epoch: 46	Time: 0m 51s	
	Train Loss: 0.853	Train PPL: 2.347
	Val. Loss: 2.511	Val. PPL: 12.312
Epoch: 47	Time: 0m 50s	
	Train Loss: 0.839	Train PPL: 2.314
	Val. Loss: 2.646	Val. PPL: 14.094
Epoch: 48	Time: 0m 51s	
	Train Loss: 0.796	Train PPL: 2.218
	Val. Loss: 2.451	Val. PPL: 11.601
Epoch: 49	Time: 0m 51s	
	Train Loss: 0.786	Train PPL: 2.195
	Val. Loss: 2.407	Val. PPL: 11.100
Epoch: 50	Time: 0m 51s	
	Train Loss: 0.750	Train PPL: 2.116
	Val. Loss: 2.439	Val. PPL: 11.466
Epoch: 51	Time: 0m 51s	
	Train Loss: 0.735	Train PPL: 2.085
	Val. Loss: 2.443	Val. PPL: 11.502
Epoch: 52	Time: 0m 51s	
	Train Loss: 0.703	Train PPL: 2.019
	Val. Loss: 2.514	Val. PPL: 12.352
Epoch: 53	Time: 0m 50s	
	Train Loss: 0.682	Train PPL: 1.978
	Val. Loss: 2.308	Val. PPL: 10.058

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Epoch: 54 | Time: 0m 51s
  Train Loss: 0.661 | Train PPL: 1.937
  Val. Loss: 2.243 | Val. PPL: 9.421
Epoch: 55 | Time: 0m 50s
  Train Loss: 0.637 | Train PPL: 1.891
  Val. Loss: 2.267 | Val. PPL: 9.655
Epoch: 56 | Time: 0m 51s
  Train Loss: 0.618 | Train PPL: 1.856
  Val. Loss: 2.298 | Val. PPL: 9.952
Epoch: 57 | Time: 0m 51s
  Train Loss: 0.592 | Train PPL: 1.807
  Val. Loss: 2.264 | Val. PPL: 9.621
Epoch: 58 | Time: 0m 51s
  Train Loss: 0.571 | Train PPL: 1.769
  Val. Loss: 2.258 | Val. PPL: 9.564
Epoch: 59 | Time: 0m 51s
  Train Loss: 0.559 | Train PPL: 1.750
  Val. Loss: 2.243 | Val. PPL: 9.426
Epoch: 60 | Time: 0m 51s
  Train Loss: 0.545 | Train PPL: 1.725
  Val. Loss: 2.116 | Val. PPL: 8.295
Epoch: 61 | Time: 0m 51s
  Train Loss: 0.522 | Train PPL: 1.685
  Val. Loss: 2.161 | Val. PPL: 8.681
Epoch: 62 | Time: 0m 51s
  Train Loss: 0.495 | Train PPL: 1.640
  Val. Loss: 2.145 | Val. PPL: 8.541
Epoch: 63 | Time: 0m 51s
  Train Loss: 0.480 | Train PPL: 1.616
  Val. Loss: 2.094 | Val. PPL: 8.115
Epoch: 64 | Time: 0m 51s
  Train Loss: 0.455 | Train PPL: 1.577
  Val. Loss: 2.115 | Val. PPL: 8.286
Epoch: 65 | Time: 0m 51s
  Train Loss: 0.441 | Train PPL: 1.555
  Val. Loss: 2.120 | Val. PPL: 8.331
Epoch: 66 | Time: 0m 51s
  Train Loss: 0.421 | Train PPL: 1.523
  Val. Loss: 2.104 | Val. PPL: 8.201
Epoch: 67 | Time: 0m 51s
  Train Loss: 0.403 | Train PPL: 1.497
  Val. Loss: 2.040 | Val. PPL: 7.691
```

In [0]: