BERT_step_by_step

June 25, 2020

PF06 - NSGC: Neural Spell & Grammar Checker (en/pt)

Author: Rafael Ito

e-mail: ito.rafael@gmail.com

0. Dataset and Description

Name: CoNLL-2014, JFLEG, BEA

Description: in this notebook we will use BERT and T5 to predict words in a sentence to perform a spell and grammar checker for Portuguese and English languages. For English, we will use the BERT and T5 models from transformers library (huggingface) and evaluate the performance in CoNLL-2014 and JFLEG datasets. For Portuguese, we will use the transformers/neuralmind BERT version and a custom dataset for evaluation.

1. Libraries and packages

1.1 Check device

```
[1]: import torch
  device = torch.device('cpu')
  if torch.cuda.is_available():
      device_model = torch.cuda.get_device_name(0)
  print('GPU model:', device_model)
```

GPU model: Tesla P100-PCIE-16GB

1.2 Install packages

1.3 Import libraries

```
[3]: #-----
     # general
    import torch
    import numpy as np
    import pandas as pd
    import sys
    import os
    import pdb
    import codecs
    import subprocess
    from multiprocessing import cpu_count
    # NLP
    from transformers import T5Tokenizer, BertTokenizer, BertForMaskedLM, __
     →T5ForConditionalGeneration
    import enchant
    import nltk
    nltk.download('words')
    from nltk.corpus import words
    #-----
    # Edit distance algorithms
    from strsimpy.levenshtein import Levenshtein
    from strsimpy.normalized_levenshtein import NormalizedLevenshtein
    from strsimpy.weighted_levenshtein import WeightedLevenshtein
    from strsimpy.weighted_levenshtein import CharacterSubstitutionInterface
    from strsimpy.damerau import Damerau
    from strsimpy.optimal_string_alignment import OptimalStringAlignment
    # random seed generator
    seed = 42
    np.random.seed(seed)
    torch.manual_seed(seed)
    torch.cuda.manual_seed(seed)
    #-----
     # Suppress some of the logging
    import logging
```

[nltk_data] Downloading package words to /root/nltk_data...
[nltk_data] Package words is already up-to-date!
Torch version: 1.5.1+cu101

1.4 Device info

```
[4]: import torch
    device = torch.device('cpu')
    if torch.cuda.is_available():
        device = torch.device('cuda')
        device_model = torch.cuda.get_device_name(0)
        device_memory = torch.cuda.get_device_properties(device).total_memory / 1e9
    #----
    print('Device:', device)
    print('GPU model:', device_model)
    print('GPU memory: {0:.2f} GB'.format(device_memory))
    print('#----')
    print('CPU cores:', cpu_count())
   Device: cuda
   GPU model: Tesla P100-PCIE-16GB
   GPU memory: 17.07 GB
   #-----
   CPU cores: 4
```

2. Custom functions and classes

2.1 Function to read file

```
[5]:
    function that reads a file and return its text
#-----
parameters:
    - path: path of the file to be read
    - encoding: encoding to be used
returns:
        file content as list of strings
'''

def read_file(path, encoding='utf-8'):
        with codecs.open(path, encoding=encoding) as f:
        return f.read().splitlines()
```

2.2 Function to write in file

2.3 Function to get tokenizer

```
[7]:
     function that returns the tokenizer associated to a string
     #-----
    parameters:
        tokenizer:
          BERT options:
            - 'bert-base-cased'
            - 'bert-large-cased'
            - 'bert-base-uncased'
            - 'bert-large-uncased'
          T5 options:
            - 't5-small'
            - 't5-base'
            - 't5-large'
            - 't5-3b'
            - 't5-11b'
          otherwise raise an error
    returns:
        Hugging Face's tokenizer
    def get_tokenizer(tokenizer):
        # BERT
        if ((tokenizer == 'bert-base-cased') or
            (tokenizer == 'bert-large-cased') or
            (tokenizer == 'bert-base-uncased') or
            (tokenizer == 'bert-large-uncased') or
            (tokenizer == 'neuralmind/bert-large-portuguese-cased') or
            (tokenizer == 'neuralmind/bert-base-portuguese-cased')):
            return BertTokenizer.from_pretrained(tokenizer)
         # T5
        elif ((tokenizer == 't5-small') or
              (tokenizer == 't5-base') or
```

```
(tokenizer == 't5-large') or
  (tokenizer == 't5-3b') or
   (tokenizer == 't5-11b')):
   return T5Tokenizer.from_pretrained(tokenizer)
#-----else:
   raise ValueError(f'Unsupported tokenizer: {tokenizer}')
```

2.4 Function to get model

```
[8]:
     function that returns the the network model associated to a string
     #_____
    parameters:
        model_name:
          BERT models:
                                                      # BERT base cased [en] (110 M_{\sqcup}
           - 'bert-base-cased'
      ⇔params)
           - 'bert-large-cased'
                                                       # BERT large cased [en] (340 M_{\square}
      ⇔params)
                                                       # BERT base uncased [en] (110 Mill
            - 'bert-base-uncased'
      ⇔params)
           - 'bert-large-uncased'
                                                       # BERT large uncased [en] (340 M
      ⇔params)
           - 'neuralmind/bert-base-portuguese-cased' # BERT base cased [pt] (110 M_{
m LI}
            - 'neuralmind/bert-large-portuguese-cased' # BERT large cased [pt] (340 M_{
m oldsymbol{\sqcup}}
      ⇔params)
          T5 models:
            - 't5-small' (60 M params)
            - 't5-base' (220 M params)
            - 't5-large' (770 M params)
            - 't5-3B' (2.8 B params)
- 't5-11B' (11 B params)
          otherwise raise an error
    returns:
       Hugging Face's model
    def get_model(model_name):
        # BERT
        if ((model_name == 'bert-base-cased') or
                                                                       # BERT base cased
     (model_name == 'bert-large-cased') or
                                                                       # BERT large cased
      (model_name == 'bert-base-uncased') or
                                                                       # BERT base _
      →uncased [en]
            (model_name == 'bert-large-uncased') or
                                                                      # BERT large_
     →uncased [en]
            (model_name == 'neuralmind/bert-base-portuguese-cased') or # BERT base cased_
```

2.5 Function to edit distance algorithm

```
[9]: '''
     function that returns the algorithm to calculate the edit distance
     parameters:
                                  | algorithm | metric? |
         algorithm:
                                   +----+
             - 'levenshtein' | Levenshtein | yes |
- 'normalized' | Normalized Levenshtein | no |
             - 'weighted' | Weighted Levenshtein | no

- 'damerau' | Damerau-Levenshtein | yes

- 'osa' | Optimal String Alignment | no
         otherwise raise an error +-----+
     returns:
         edit distance algorithm
     def get_distance_algorithm(algorithm):
         if (algorithm == 'levenshtein'):
             return Levenshtein()
         elif (algorithm == 'normalized'):
            return NormalizedLevenshtein()
         elif (algorithm == 'weighted'):
             return
         elif (algorithm == 'damerau'):
             return Damerau()
         elif (algorithm == 'osa'):
             return OptimalStringAlignment()
             raise ValueError(f'Unsupported algorithm: {algorithm}')
```

2.6 Function to calculate GLEU score

```
[10]: '''
      function that receives text files and calculate GLEU score
     #-----
     parameters:
         - src: source file
         - ref: reference file(s)
         - hyp: hypothesis file
         - n: n-gram order
          - num_iter: number of GLEU iterations
         - sent: sentence level scores
     returns:
         GLEU score (float)
     def calc_gleu(src, ref, hyp, n=4, num_iter=500, sent=False):
         gleu_calculator.load_sources(src)
         gleu_calculator.load_references(ref)
         if len(ref) == 1:
             print("There is one reference. NOTE: GLEU is not computing the confidence ⊔
       →interval.")
             gleu = [g for g in gleu_calculator.run_iterations(
                 num_iterations=num_iter,
                 source=src,
                 hypothesis=hyp,
                 per_sent=sent)][0][0]
         else:
             gleu = [g for g in gleu_calculator.run_iterations(
                 num_iterations=num_iter,
                 source=src,
                 hypothesis=hyp,
                 per_sent=sent)][0][0]
          #print(qleu)
         return float(gleu)*100
```

2.7 Function to calculate MaxMatch score

2.8 Function parse M2 file

3. Datasets

3.1 CoNLL-2013

3.1.1 Download

```
[13]: # test set
! wget -q -nc https://www.comp.nus.edu.sg/~nlp/conll13st/release2.3.1.tar.gz
! tar -xzf release2.3.1.tar.gz
! rm release2.3.1.tar.gz
```

3.1.2 Test set

3.1.3 Sample

```
[15]: print('original sentence:')
    print(conll_2013_test_src[0])
    #------
    print('\nannotation:')
    print(*conll_2013_test_ref[0:4], sep='\n')
```

original sentence:

In modern digital world , electronic products are widely used in daily lives

```
annotation:
S In modern digital world , electronic products are widely used in daily lives
such as Smart phones , computers and etc .
A 1 1|||ArtOrDet|||the|||REQUIRED|||-NONE-|||0
A 12 13|||Nn|||life|||REQUIRED|||-NONE-|||0
A 15 16|||Mec|||smart|||REQUIRED|||-NONE-|||0
```

3.2 CoNLL-2014

3.2.1 Download

```
[16]: ## training set
    #from google.colab import drive
    #drive.mount('/gdrive')
    #------
# test set
! wget -q -nc https://www.comp.nus.edu.sg/~nlp/conll14st/conll14st-test-data.tar.gz
! tar -xzf conll14st-test-data.tar.gz
! rm conll14st-test-data.tar.gz
```

3.2.2 Training set

3.2.3 Test set

```
[18]: # import test set
#------
# source
m2_file = '/content/conll14st-test-data/noalt/official-2014.1.m2'
output_file = '/content/conll14st-test-data/noalt/official-2014.1.src'
conll_2014_test_src = m2_parser(m2_file, output_file)
# reference
conll_2014_test_ref = read_file(m2_file)
```

3.2.4 Sample

```
[19]: print('original sentence:')
    print(conll_2014_test_src[3])
    #------
    print('\nannotation:')
```

```
print(*conll_2014_test_ref[7:9], sep='\n')

original sentence:
People get certain disease because of genetic changes .

annotation:
S People get certain disease because of genetic changes .
A 3 4|||Nn|||diseases|||REQUIRED|||-NONE-|||0
```

3.3 JFLEG

3.3.1 Download

```
[20]: # clone GitHub repo
! git clone --quiet https://github.com/keisks/jfleg.git 2> /dev/null
```

3.3.2 Training set

```
[21]: # import training set
#-------
# source
jfleg_train_src = read_file('jfleg/dev/dev.src')
# references
jfleg_train_ref0 = read_file('jfleg/dev/dev.ref0')
jfleg_train_ref1 = read_file('jfleg/dev/dev.ref1')
jfleg_train_ref2 = read_file('jfleg/dev/dev.ref2')
jfleg_train_ref3 = read_file('jfleg/dev/dev.ref3')
```

3.3.3 Test set

```
[22]: # import test set
#-------
# source
jfleg_test_src = read_file('jfleg/test/test.src')
# references
jfleg_test_ref0 = read_file('jfleg/test/test.ref0')
jfleg_test_ref1 = read_file('jfleg/test/test.ref1')
jfleg_test_ref2 = read_file('jfleg/test/test.ref2')
jfleg_test_ref3 = read_file('jfleg/test/test.ref3')
```

3.3.4 Sample

```
[23]: # print source and references example
print('source sentence:')
print(jfleg_test_src[0])
#------
print('\nreferences sentences:')
print(jfleg_test_ref0[0])
print(jfleg_test_ref1[0])
print(jfleg_test_ref2[0])
print(jfleg_test_ref3[0])
```

```
source sentence:
New and new technology has been introduced to the society .

references sentences:
New technology has been introduced to society .
New technology has been introduced into the society .
Newer and newer technology has been introduced into society .
Newer and newer technology has been introduced to the society .
```

3.4 BEA

3.4.1 Download

```
[24]: # download test data
! wget -q -nc https://www.cl.cam.ac.uk/research/nl/bea2019st/data/wi+locness_v2.1.

→bea19.tar.gz
! tar -xzf wi+locness_v2.1.bea19.tar.gz
! rm wi+locness_v2.1.bea19.tar.gz
```

3.4.2 Training set

```
[25]: # import test set
      # source
      # read A, B, C M2 file
      m2_file_A = '/content/wi+locness/m2/A.train.gold.bea19.m2'
      m2_file_B = '/content/wi+locness/m2/B.train.gold.bea19.m2'
      m2_file_C = '/content/wi+locness/m2/C.train.gold.bea19.m2'
      # read and concatenate all files
      m2_ABC_file = read_file(m2_file_A) + read_file(m2_file_B) + read_file(m2_file_C)
      # save to a file
      m2_file = '/content/wi+locness/m2/ABC.train.gold.bea19.m2'
      with open(m2_file, 'w') as f:
          for line in m2_ABC_file:
              f.write('%s\n' %line)
      output_file = '/content/wi+locness/m2/ABCN.train.gold.bea19.src'
      bea_train_src = m2_parser(m2_file, output_file)
      # reference
      bea_train_ref = read_file(m2_file)
```

3.4.3 Development set

```
[26]: # import test set
#------
# source
m2_file = '/content/wi+locness/m2/ABCN.dev.gold.bea19.m2'
output_file = '/content/wi+locness/m2/ABCN.dev.gold.bea19.src'
bea_test_src = m2_parser(m2_file, output_file)
# reference
bea_test_ref = read_file(m2_file)
```

3.4.4 Sample

```
[27]: print('original sentence:')
    print(bea_train_src[0])
#------
print('\nannotation:')
    print(*bea_train_ref[0:2], sep='\n')

original sentence:
    My town is a medium size city with eighty thousand inhabitants .

annotation:
    S My town is a medium size city with eighty thousand inhabitants .
    A 5 6||R:OTHER||- sized||REQUIRED||-NONE-||0
```

3.5 ReGRA

3.5.1 Import

```
[28]: # mount drive to access file with sentences
from google.colab import drive
drive.mount('/gdrive')
```

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

3.5.2 Test set

```
[29]: # source
    regra_src_file = '/gdrive/My Drive/Colab Notebooks/IA376E/Final Project/ReGRA/src.txt'
    #regra_src = read_file(regra_src_file, encoding='latin-1')
    regra_src = read_file(regra_src_file, encoding='utf-8')
#-------
# reference
    regra_ref_file = '/gdrive/My Drive/Colab Notebooks/IA376E/Final Project/ReGRA/ref.txt'
    #regra_ref = read_file(regra_ref_file, encoding='latin-1')
    regra_ref = read_file(regra_ref_file, encoding='utf-8')
```

3.5.4 Sample

```
[30]: print('original sentences:')
    print(*regra_src[1000:1003], sep='\n')
#------
print('\nreference sentences:')
    print(*regra_ref[1000:1003], sep='\n')
```

original sentences:

Uma delegação de padeiros vem prestar seu apoio as mulheres dos grevistas. Uma era ítala-brasileira.

Uma frota de navios norte-americanos se dirigiste ao Mar Mediterrâneo.

reference sentences:

Uma delegação de padeiros vem prestar seu apoio às mulheres dos grevistas.

Uma era ítalo-brasileira. Uma frota de navios norte-americanos se dirige ao Mar Mediterrâneo.

4. Evaluation Metrics

4.1 M^2 (MaxMatch) score

4.1.1 Getting the M^2 scorer

```
[31]: # get m2scorer
! wget -q -nc https://www.comp.nus.edu.sg/~nlp/sw/m2scorer.tar.gz
! tar -xzf m2scorer.tar.gz
! rm m2scorer.tar.gz
```

4.1.2 Testing the M^2 scorer

```
[32]: # getting examples
    src = '/content/m2scorer/example/system2'
    ref = '/content/m2scorer/example/source_gold'

[33]: # source
    print('source sentences:')
    print(*read_file(src), sep='\n')

    source sentences:
    A cat sat on mat .
    The dog .
    Giant otters are apex predator .
```

```
[34]: # reference
print('reference sentences:')
print(*read_file(ref), sep='\n')
```

```
reference sentences:

S The cat sat at mat .

A 3 4|||Prep|||on|||REQUIRED|||-NONE-|||0

A 4 4|||ArtOrDet|||the||a|||REQUIRED|||-NONE-|||0

S The dog .

A 1 2|||NN|||dogs|||REQUIRED|||-NONE-|||0

A -1 -1|||noop|||-NONE-|||-NONE-|||-NONE-|||1

S Giant otters is an apex predator .

A 2 3|||SVA|||are|||REQUIRED|||-NONE-|||0

A 3 4|||ArtOrDet|||-NONE-|||REQUIRED|||-NONE-|||0

A 5 6|||NN|||predators|||REQUIRED|||-NONE-|||0

A 1 2|||NN|||otter|||REQUIRED|||-NONE-|||1
```

```
[35]: # score
score = m2scorer(src, ref)
print(score)
```

Precision : 0.7500 Recall : 0.6000 F_0.5 : 0.7143

4.2 GLEU score

https://github.com/keisks/jfleg

4.2.1 Getting the GLEU scorer

```
[36]: # import gleu metric
sys.path.append('/content/jfleg/eval/')
from gleu import GLEU
gleu_calculator = GLEU()
```

4.2.2 Testing the GLEU scorer

There is one reference. NOTE: GLEU is not computing the confidence interval. GLEU = 100.00

GLEU = 40.47

```
[39]: # hyp = ref
#-------
# source file
src = 'jfleg/test/test.src'
#------
# ref0
hyp = 'jfleg/test/test.ref0'
ref = ['jfleg/test/test.ref1', 'jfleg/test/test.ref2', 'jfleg/test/test.ref3']
ref0 = calc_gleu(src, ref, hyp);
```

```
# ref1
hyp = 'jfleg/test/test.ref1'
ref = ['jfleg/test/test.ref0', 'jfleg/test/test.ref2', 'jfleg/test/test.ref3']
ref1 = calc_gleu(src, ref, hyp);
# ref2
hyp = 'jfleg/test/test.ref2'
ref = ['jfleg/test/test.ref0', 'jfleg/test/test.ref1', 'jfleg/test/test.ref3']
ref2 = calc_gleu(src, ref, hyp);
#-----
# ref3
hyp = 'jfleg/test/test.ref3'
ref = ['jfleg/test/test.ref0', 'jfleg/test/test.ref1', 'jfleg/test/test.ref2']
ref3 = calc_gleu(src, ref, hyp);
#-----
print(f'ref0 = {ref0:.2f}')
print(f'ref1 = {ref1:.2f}')
print(f'ref2 = {ref2:.2f}')
print(f'ref3 = {ref3:.2f}')
print('#----')
print(f'mean = {(ref0 + ref1 + ref2 + ref3) / 4:.2f}')
ref0 = 61.32
```

reference table:

system	GLEU (dev)	GLEU (test)
SOURCE	38.21	40.54
REFERENCE	55.26	62.37

4.3 Edit distance

4.3.1 Getting distances algorithms

https://github.com/luozhouyang/python-string-similarity#damerau-levenshtein

```
[40]: levenshtein = get_distance_algorithm('levenshtein')
  damerau = get_distance_algorithm('damerau')
  normalized = get_distance_algorithm('normalized')
  weighted = get_distance_algorithm('weighted')
  osa = get_distance_algorithm('osa')
```

4.3.2 Testing Damerau-Levenshtein distance algorithm

```
[41]: # distance = 1: character removed
    print('distance =', damerau.distance('Covid-19', 'Covid-9'))

distance = 1

[42]: # distance = 2: character removed & character inserted
    print('distance =', damerau.distance('Covid-19', 'Codiv-19'))

distance = 2

[43]: # distance = 1: transposition of two adjacent characters
    print('distance =', damerau.distance('Covid-19', 'Covid-91'))

distance = 1
```

5. Tokenizer

5.1 BERT

5.2 T5

```
[45]: #tokenizer = get_tokenizer('t5-small')

#tokenizer = get_tokenizer('t5-base')

#tokenizer = get_tokenizer('t5-large')

#tokenizer = get_tokenizer('t5-3b')

#tokenizer = get_tokenizer('t5-11b')
```

6. Model

6.1 BERT

```
#model = get_model('neuralmind/bert-base-portuguese-cased') # BERT base cased [pt] __ \( \times 438 \) MB

model = get_model('neuralmind/bert-large-portuguese-cased') # BERT large cased [pt] 1.
\( \times 34 \) GB
```

6.2 T5

```
[47]: #model = get_model('t5-small') # 242 MB

#model = get_model('t5-base') # 892 MB

#model = get_model('t5-large') # 2.95 GB

#model = get_model('t5-3b') # 11.4 GB

#model = get_model('t5-11b') # ??.? GB
```

7. Sentence Correction Suggestion

7.1 BERT-based function

Hyperparameters

```
[48]: # topk model output predictions used to compare
k = 10
# Damerau-Levenshtein
edit_distance = get_distance_algorithm('damerau')
# threshold distance to suggest correction
threshold = 5
```

Function

```
[49]: def suggest(sentences, tokenizer, model, distance, split=False, k=20, threshold=5, u
       →device='cpu'):
          model.to(device)
          sentences_suggested = []
          for sentence in sentences:
              if split:
                 tokenized = sentence.split()
                                                                               # dummy
       \rightarrow tokenizer
              else:
                  tokenized = tokenizer.tokenize(sentence)
                                                                               # tokenize
              tokenized_ids = tokenizer.encode(tokenized)
                                                                              # '[CLS]' +
       ⇒ qet word ids + '[SEP]'
              single_input_ids = torch.LongTensor(tokenized_ids).to(device) # convert list_
       \rightarrowto tensor
              input_ids = single_input_ids.repeat(len(single_input_ids)-2, 1) # repeat tensor
              #_____
              # mask tokens
              for i in range(len(input_ids)):
                  input_ids[i][i+1] = tokenizer.mask_token_id
              \# predict the top-k tokens for the masked ones
              topk_pred_pt = torch.zeros((len(tokenized), k))
              for i, masked_sentence in enumerate(input_ids):
                  model_output = model(input_ids = masked_sentence.unsqueeze(dim=0))
```

```
logits = model_output[0]
        _, predicted_ids = torch.topk(logits, k, sorted=True)
        topk_pred_pt[i] = predicted_ids.squeeze()[i+1]
    # convert ids back to words
    topk_pred_tokens = [] # list of lists
    for masked_sentence in topk_pred_pt:
        pred_list = []
        for predictions in masked_sentence:
            pred_list.append(tokenizer.decode([predictions.tolist()]))
        topk_pred_tokens.append(pred_list)
    # compare predictions and calculate edit distance
    suggestion = []
    for i, masked_token in enumerate(tokenized):
        # check if masked token is in predictions
        if masked_token in topk_pred_tokens[i]:
            # if it is, no correction is suggested
            suggestion.append(masked_token)
        else:
            # using distance?
            if (distance != None):
                # if masked token not in predictions, calculate distance
                dist = torch.zeros(k)
                for j, prediction in enumerate(topk_pred_tokens[i]):
                    dist[j] = edit_distance.distance(masked_token, prediction)
                # check if minimum distance is under a limiar
                if torch.min(dist).item() <= threshold:</pre>
                    # if it is, make suggestions
                    # argmin returns the last index --> workaround: flip the tensor
                    min_index = len(dist) - torch.argmin(dist.flip(0)).item() - 1
                    suggestion.append(topk_pred_tokens[i][min_index])
                else:
                    # if it is not, make no correction suggestion
                   suggestion.append(masked_token)
            # greedy suggestion
                suggestion.append(topk_pred_tokens[i][0])
    sentences_suggested.append(' '.join(suggestion))
return sentences_suggested
```

7.2 Step-by-Step

```
[50]: model.to(device);
```

7.2.0 Hyperparameters

```
[51]: # topk model output predictions used to compare
k = 2
# Damerau-Levenshtein
edit_distance = get_distance_algorithm('damerau')
# threshold distance to suggest correction
threshold = 2
```

7.2.1 Get sentence

```
[52]: # get sentence
sentence = regra_src[325]
sentence
```

[52]: 'Ele comprou este carro à prazo.'

```
[53]: # get reference
  reference = regra_ref[325]
  reference
```

[53]: 'Ele comprou este carro a prazo.'

7.2.2 Tokenize

```
[54]: # tokenize
#tokenized = sentence.split()
tokenized = tokenizer.tokenize(sentence)
tokenized
```

- [54]: ['Ele', 'comprou', 'este', 'carro', 'à', 'prazo', '.']
- [55]: # '[CLS]' + get word ids + '[SEP]'
 tokenized_ids = tokenizer.encode(tokenized)
 tokenized_ids
- [55]: [101, 787, 10107, 860, 3883, 353, 6620, 119, 102]
- [56]: # convert list to tensor
 single_input_ids = torch.LongTensor(tokenized_ids).to(device)
 single_input_ids
- [56]: tensor([101, 787, 10107, 860, 3883, 353, 6620, 119, 102], device='cuda:0')
- [57]: # repeat tensor
 input_ids = single_input_ids.repeat(len(single_input_ids)-2, 1)
 input_ids
- [57]: tensor([[101, 787, 10107, 860, 3883, 353, 6620, 119, 102], [101, 787, 10107, 3883, 353, 6620, 119, 102], 860, [101, 787, 10107, 3883, 353, 6620, 119, 102], 860,

```
[ 101,
          787, 10107,
                        860,
                              3883,
                                      353, 6620,
                                                    119,
                                                           102],
 [ 101,
          787, 10107,
                        860,
                              3883,
                                      353,
                                            6620,
                                                    119,
                                                           102],
 [ 101,
          787, 10107,
                              3883,
                                      353,
                                            6620,
                                                    119,
                                                           102],
                        860,
 [ 101,
          787, 10107,
                        860,
                              3883,
                                      353, 6620,
                                                    119,
                                                           102]],
device='cuda:0')
```

7.2.3 Mask tokens

```
[58]: # mask tokens
for i in range(len(input_ids)):
    input_ids[i][i+1] = tokenizer.mask_token_id
input_ids
```

```
103, 10107,
                                     860,
[58]: tensor([[ 101,
                                          3883,
                                                  353, 6620,
                                                                119,
                                                                       102],
             [ 101,
                       787, 103,
                                     860,
                                          3883,
                                                  353, 6620,
                                                                119,
                                                                       102],
             [ 101,
                       787, 10107,
                                           3883,
                                                  353, 6620,
                                                                119.
                                                                       1027.
                                     103,
             [ 101,
                       787, 10107,
                                     860,
                                           103,
                                                  353, 6620,
                                                                119,
                                                                       102],
             [ 101,
                       787, 10107,
                                          3883,
                                                  103,
                                                       6620,
                                                                119,
                                                                       102],
                                     860,
                       787, 10107,
                                     860, 3883,
             [ 101,
                                                  353,
                                                        103,
                                                                119,
                                                                       102],
             [ 101,
                       787, 10107,
                                     860, 3883,
                                                  353, 6620,
                                                                103,
                                                                       102]],
            device='cuda:0')
```

7.2.4 Top-k predictions

```
[59]: topk_pred_pt = torch.zeros((len(tokenized), k))
    for i, masked_sentence in enumerate(input_ids):
        model_output = model(input_ids = masked_sentence.unsqueeze(dim=0))
        logits = model_output[0]
        _, predicted_ids = torch.topk(logits, k, sorted=True)
        topk_pred_pt[i] = predicted_ids.squeeze()[i+1]
        topk_pred_pt
```

7.2.5 Convert IDs back to words

```
[60]: # convert ids back to words
topk_pred_tokens = [] # list of lists
for masked_sentence in topk_pred_pt:
    pred_list = []
    for predictions in masked_sentence:
        pred_list.append(tokenizer.decode([predictions.tolist()]))
    topk_pred_tokens.append(pred_list)
topk_pred_tokens
```

7.2.6 Compare predictions and calculate distance

```
[61]: # compare predictions and calculate edit distance
      suggestion = []
      for i, masked_token in enumerate(tokenized):
          # check if masked token is in predictions
          if masked_token in topk_pred_tokens[i]:
              # if it is, no correction is suggested
              suggestion.append(masked_token)
          else:
              # using distance?
              if (edit_distance != None):
                  # if masked token not in predictions, calculate distance
                  dist = torch.zeros(k)
                  for j, prediction in enumerate(topk_pred_tokens[i]):
                      dist[j] = edit_distance.distance(masked_token, prediction)
                  # check if minimum distance is under a limiar
                  if torch.min(dist).item() <= threshold:</pre>
                      # if it is, make suggestions
                      # argmin returns the last index --> workaround: flip the tensor
                      min_index = len(dist) - torch.argmin(dist.flip(0)).item() - 1
                      suggestion.append(topk_pred_tokens[i][min_index])
                  #----
                  else:
                      # if it is not, make no correction suggestion
                      suggestion.append(masked_token)
              # greedy suggestion
                  suggestion.append(topk_pred_tokens[i][0])
      ' '.join(suggestion)
```

[61]: 'Ele comprou este carro a prazo .'

End of the notebook