# T5\_step\_by\_step

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

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## 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

Requirement already satisfied: strsimpy in /usr/local/lib/python3.6/dist-packages (0.1.6)

## 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)
```

[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

```
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):
         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)
```

## 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'
                                                      # BERT base cased [en] (110 M_{11}
      →params)
                                                     # BERT large cased [en] (340 M_{\sqcup}
            - 'bert-large-cased'
      ⇔params)
           - 'bert-base-uncased'
                                                      # BERT base uncased [en] (110 M_{\square}
      ⇔params)
           - 'bert-large-uncased'
                                                      # BERT large uncased [en] (340 M_
      ⇔params)
           - 'neuralmind/bert-base-portuguese-cased' # BERT base cased [pt] (110 M_{
m lue}
      ⇔params)
            - 'neuralmind/bert-large-portuguese-cased' \# BERT large cased [pt] (340 M_{\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):
        if ((model_name == 'bert-base-cased') or
                                                                      # BERT base cased
     → [en]
            (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]
```

## 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 Levenshtein | no
| Damerau-Levenshtein | yes
| Optimal String Alignment | no
             - 'weighted'
                                                             | yes |
             - 'damerau'
         otherwise raise an error +-----+
        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')
```

#### 3.5.2 Test set

## 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')
```

## 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-|||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

```
[37]: # hyp = ref
#------
src = 'jfleg/test/test.src'
ref = ['jfleg/test/test.ref0']
hyp = 'jfleg/test/test.ref0'
print(f'GLEU = {calc_gleu(src, ref, hyp):.2f}')
```

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

GLEU = 40.47

```
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}')
```

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**

#### 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 T5-based function

Hyperparameters

```
[48]: # number of output predictions
k = 30
# beams used in beam search
b = 50
# Damerau-Levenshtein
edit_distance = get_distance_algorithm('damerau')
# threshold distance to suggest correction
threshold = 5
```

#### **Function**

```
[49]: def suggest_t5(sentences, tokenizer, model, distance, split=False, k=30, b=50, u
       →threshold=5, device='cpu'):
          model.to(device)
          sentences_suggested = []
          for sentence in sentences:
              # split and add mask
              # tokenize
             tokenized_raw = sentence.split()
              tokenized = tokenized_raw.copy()
              tokenized.append('</s>')
              # repeat tensor
             repeated = [tokenized*1 for _ in range(len(tokenized_raw))]
              #-----
              # mask tokens (insert '<extra_id_0>')
              for i, seq in enumerate(repeated):
                 seq[i] = '<extra_id_0>'
              # joing tokens back
              joined = []
              for seq in repeated:
                 joined.append(' '.join(seq))
              # encode sentences
              input_ids = []
              for masked_sentence in joined:
                  input_ids.append(tokenizer.encode(masked_sentence,_
       →add_special_tokens=True, return_tensors='pt'))
```

```
# top-k predictions
      topk_pred_pt = torch.zeros((len(repeated), k))
      for i, masked_sentence in enumerate(input_ids):
           # model predict
          model_output = model.generate(input_ids = masked_sentence.to(device),_
→num_beams=b, num_return_sequences=k, max_length=3)
          topk_pred_pt[i] = model_output[:,-1]
      topk_pred_pt.long()
       # 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
       # compare predictions and calculate edit distance
       suggestion = []
      for i, masked_token in enumerate(tokenized_raw):
           # 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]: # number of output predictions
k = 5
# beams used in beam search
b = 10
# Damerau-Levenshtein
edit_distance = get_distance_algorithm('damerau')
# threshold distance to suggest correction
threshold = 2
```

#### 7.2.1 Get sentence

```
[52]: # get sentence
sentence = jfleg_train_src[721]
sentence
```

[52]: 'People also do not do nothing . '

```
[53]: # get references
print(jfleg_train_ref0[721])
print(jfleg_train_ref1[721])
print(jfleg_train_ref2[721])
print(jfleg_train_ref3[721])
```

```
People also do not do anything . People also do not do nothing . People also do something . People also do not do nothing .
```

#### 7.2.2 Tokenize

```
[54]: # tokenize
  tokenized_raw = sentence.split()
  tokenized = tokenized_raw.copy()
  tokenized.append('</s>')
  tokenized
```

```
[54]: ['People', 'also', 'do', 'not', 'do', 'nothing', '.', '</s>']
```

```
[55]: # repeat tensor
repeated = [tokenized*1 for _ in range(len(tokenized_raw))]
repeated
```

```
['People', 'also', 'do', 'not', 'do', 'nothing', '.', '</s>'],
['People', 'also', 'do', 'not', 'do', 'nothing', '.', '</s>'],
['People', 'also', 'do', 'not', 'do', 'nothing', '.', '</s>']]
```

#### 7.2.3 Mask tokens

```
[56]: # insert '<extra_id_0>'
    for i, seq in enumerate(repeated):
        seq[i] = '<extra_id_0>'
    repeated

[56]: [['<extra_id_0>', 'also', 'do', 'not', 'do', 'nothing', '.', '</s>'],
        ['People', '<extra_id_0>', 'do', 'not', 'do', 'nothing', '.', '</s>'],
        ['People', 'also', '<extra_id_0>', 'not', 'do', 'nothing', '.', '</s>'],
```

```
['People', 'also', '<extra_id_0>', 'not', 'do', 'nothing', '.', '</s>'],
['People', 'also', 'do', '<extra_id_0>', 'do', 'nothing', '.', '</s>'],
['People', 'also', 'do', 'not', '<extra_id_0>', 'nothing', '.', '</s>'],
['People', 'also', 'do', 'not', 'do', '<extra_id_0>', '.', '</s>'],
['People', 'also', 'do', 'not', 'do', 'nothing', '<extra_id_0>', '.', '</s>']]
```

```
[57]: # joing tokens back
joined = []
for seq in repeated:
    joined.append(' '.join(seq))
joined
```

#### 7.2.4 Encoding sentences

```
[58]: input_ids = []
for masked_sentence in joined:
    input_ids.append(tokenizer.encode(masked_sentence, add_special_tokens=True,
    return_tensors='pt'))
input_ids
```

```
103, 1327,
                                                                             1]]),
[58]: [tensor([[32099,
                          92,
                                103,
                                        59,
                                                              3,
                                                                      5,
                                                                     5,
       tensor([[ 2449, 32099,
                                103,
                                        59,
                                              103, 1327,
                                                              3,
                                                                             1]]),
       tensor([[ 2449,
                          92, 32099,
                                        59,
                                              103, 1327,
                                                              3,
                                                                     5,
                                                                             1]]),
       tensor([[ 2449,
                          92,
                                103, 32099,
                                              103, 1327,
                                                              3,
                                                                      5,
                                                                             1]]),
                                        59, 32099, 1327,
       tensor([[ 2449,
                          92,
                                                                      5,
                                                                             1]]),
                                103,
                                                              3,
                                                                     5,
       tensor([[ 2449,
                          92,
                                103,
                                        59,
                                              103, 32099,
                                                              3,
                                                                             1]]),
       tensor([[ 2449,
                          92,
                                103,
                                        59,
                                              103, 1327, 32099,
                                                                      1]])]
```

#### 7.2.5 Top-k predictions

#### 7.2.6 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
```

## 7.2.7 Compare predictions and calculate distance

[61]: 'People also do not do anything .'

## End of the notebook