## In [1]: %load\_ext memory\_profiler

```
In [2]: import os
    import gensim
    import jieba
    import zhconv
    from gensim.corpora import WikiCorpus
    from datetime import datetime
    from typing import List

if (not os.path.isfile("dict.txt.big")):
    ! wget https://github.com/fxsjy/jieba/raw/master/extra_dict/dic
    jieba.set_dictionary("dict.txt.big")
```

```
In [3]: ZhWiki = "zhwiki-20230501-pages-articles-multistream.xml.bz2"
   !du -sh $ZhWiki
!md5 $ZhWiki
!file $ZhWiki
```

2.6G zhwiki-20230501-pages-articles-multistream.xml.bz2
MD5 (zhwiki-20230501-pages-articles-multistream.xml.bz2) = 27e78ff
901bcd3803955d9373537dd3f
zhwiki-20230501-pages-articles-multistream.xml.bz2: bzip2 compress
ed data, block size = 900k

```
In [4]: |import spacy
        spacy.cli.download("zh core web sm")
        spacy.cli.download("en_core_web_sm")
        nlp_zh = spacy.load("zh_core_web_sm")
        nlp en = spacy.load("en core web sm")
        Collecting zh-core-web-sm==3.5.0
          Downloading https://github.com/explosion/spacy-models/releases/d
        ownload/zh_core_web_sm-3.5.0/zh_core_web_sm-3.5.0-py3-none-any.whl
        (https://github.com/explosion/spacy-models/releases/download/zh co
        re web sm-3.5.0/zh core web sm-3.5.0-pv3-none-anv.whl) (48.5 MB)
                                                     - 48.5/48.5 MB 52.1 kB
        /s eta 0:00:00
        Requirement already satisfied: spacy<3.6.0,>=3.5.0 in /Users/hsiu/
        opt/anaconda3/lib/python3.9/site-packages (from zh-core-web-sm==3.
        5.0) (3.5.2)
        Collecting spacy-pkuseg<0.1.0,>=0.0.27
          Downloading spacy_pkuseg-0.0.32-cp39-cp39-macosx_11_0_arm64.whl
        (2.3 MB)
                                                     — 2.3/2.3 MB 56.0 kB/s
        eta 0:00:00
        Requirement already satisfied: jinja2 in /Users/hsiu/opt/anaconda3
        /lib/python3.9/site-packages (from spacy<3.6.0,>=3.5.0->zh-core-we
        b-sm==3.5.0) (2.11.3)
        Requirement already satisfied: spacy-legacy<3.1.0,>=3.0.11 in /Use
In [5]: STOPWORDS = nlp_zh.Defaults.stop_words | nlp_en.Defaults.stop_word
        for word in STOPWORDS.copy():
            STOPWORDS.add(zhconv.convert(word, "zh-tw"))
In [6]: def preprocess_and_tokenize(text, token_min_len = 1, token_max_len
            if (lower):
                text = text.lower()
            text = zhconv.convert(text, "zh-tw")
            return [
                token for token in jieba.cut(text, cut_all = False)
                if token min len <= len(token) <= token max len and token n</pre>
            1
```

```
In [7]: %%time
        %memit
        print(f"Parsing {ZhWiki}...")
        wiki_corpus = WikiCorpus(ZhWiki, token_min_len=1)
```

Parsing zhwiki-20230501-pages-articles-multistream.xml.bz2...

/Users/hsiu/opt/anaconda3/lib/python3.9/site-packages/gensim/utils .py:1332: UserWarning: detected OSX with python3.8+; aliasing chun kize to chunkize serial

warnings.warn("detected %s; aliasing chunkize to chunkize serial " % entity)

peak memory: 2102.64 MiB, increment: 1444.09 MiB

CPU times: user 11min 15s, sys: 1min 52s, total: 13min 8s

Wall time: 15min 21s

```
In [8]: | q = wiki corpus.get texts()
        print(next(q)[:10])
        print(next(g)[:10])
        print(next(q)[:10])
```

['歐幾里得', '西元前三世紀的古希臘數學家', '現在被認為是幾何之父', '此畫為拉 斐爾的作品','雅典學院','数学','是研究數量','屬於形式科學的一種','數學利 用抽象化和邏輯推理','從計數']

['蘇格拉底之死', '由雅克', '路易', '大卫所繪', '年', '哲學', '是研究普遍的

','基本问题的学科','包括存在','知识'] ['文學','在狭义上','是一种语言艺术','亦即使用语言文字为手段','形象化地反 映客观社会生活','表达主观作者思想感情的一种艺术','文学不仅强调传达思想观念', '更强调传达方式的独特性','且讲究辞章的美感','文学']

```
In [9]: WIKI_SEG_TXT = "wiki_seg.txt"
        generator = wiki_corpus.get_texts()
        with open(WIKI_SEG_TXT, "w", encoding='utf-8') as output:
            for texts num, tokens in enumerate(generator):
                output.write(" ".join(tokens) + "\n")
                if (texts_num + 1) % 100000 == 0:
                    print(f"[{str(datetime.now()):.19}] 已寫入 {texts_num} 篇
```

```
[2023-05-11 10:24:06] 已寫入 99999 篇斷詞文章
[2023-05-11 10:26:05] 已寫入 199999 篇斷詞文章
[2023-05-11 10:29:21] 已寫入 299999 篇斷詞文章
[2023-05-11 10:31:46] 已寫入 399999 篇斷詞文章
```

```
In [10]: %time
         from gensim.models import word2vec
         import multiprocessing
         max cpu counts = multiprocessing.cpu count()
         word_dim_size = 300
         print(f"Use {max_cpu_counts} workers to train Word2Vec (dim={word_d
         sentences = word2vec.LineSentence(WIKI SEG TXT)
         model = word2vec.Word2Vec(sentences, vector_size=word_dim_size, worl
         output_model = f"word2vec.zh.{word_dim_size}.model"
         model.save(output_model)
         Use 8 workers to train Word2Vec (dim=300)
         CPU times: user 29min 2s, sys: 1min 59s, total: 31min 1s
         Wall time: 7min 56s
In [11]: ! ls word2vec.zh*
                                              word2vec.zh.300.model.wv.vect
         word2vec.zh.300.model
         ors.npy
         word2vec.zh.300.model.syn1neg.npy
In [12]: ! du -sh word2vec.zh*
          57M
                 word2vec.zh.300.model
                 word2vec.zh.300.model.syn1neg.npy
         1.8G
                 word2vec.zh.300.model.wv.vectors.npy
         1.8G
In [13]: | print(model.wv.vectors.shape)
         model.wv.vectors
         (1578559, 300)
Out[13]: array([[-1.7076457e+00,
                                  1.7358593e+00, -3.4208825e-01, ...,
                  6.5092337e-01,
                                  6.4788365e-01,
                                                  2.2596502e-01],
                [-8.6641109e-01, 1.0497972e+00,
                                                  4.8340130e-01, ...,
                  2.2776024e-01, -5.6819314e-01,
                                                  3.1535363e-01],
                                                  3.3163098e-01, ...,
                [-1.3289380e+00, 1.2796842e+00,
                  6.8806994e-01, -4.8488963e-01,
                                                  5.2340209e-01],
                [-3.0601058e-02, 5.1672857e-02,
                                                  1.3539110e-02, ...,
                 -1.4815503e-02, 5.7537202e-02, -1.9980976e-02],
                [-3.5888821e-02, 4.1780226e-02,
                                                  6.7084683e-03, ...,
                 -2.1981372e-02, 7.2206617e-03, -9.2937918e-03],
                [-4.5324679e-02, -1.6357239e-02, -9.1987170e-02, ...,
                  4.6356138e-02, -4.8235804e-03, 1.1394625e-03]], dtype=fl
         oat32)
```

```
In [14]: |vec = model.wv['數學家']
         print(vec.shape)
         vec
         (300.)
Out[14]: array([ 0.53049093, 0.01827557, 0.2647456 , 0.30384144,
                                                                   0.8471
         7506.
                                          0.37918493, -0.4042048 ,
               -0.3160529 , -0.84586823,
                                                                   0.0628
         6512,
               -0.43317407, 0.15615006, 0.1487516, 0.7061684, -0.9255
         093 ,
               -0.9253154 , -0.759502 , 0.18732086 , -0.22279754 , -1.3600
         307,
               -0.13007402, 0.33115828, 0.2514567, 0.3103663, 0.5041
         195 ,
                0.5105871 , 0.275075 , -0.99158734, -0.8061154 , 0.6094
         9665,
               -0.97135323, -0.36245635, 0.5228062, -0.9929437, -0.3768
         951,
               -0.27796376, 0.32308862, 0.17225985, -0.15119103, -0.5225
         684 ,
                0.6778689 , 0.5649924 , -0.4203485 , -0.3324206 , -0.8110
         969
In [15]: word = "這肯定沒見過 "
         try:
            vec = model.wv[word]
         except KeyError as e:
            print(e)
         "Key '這肯定沒見過 ' not present"
In [16]: model.wv.most_similar("飲料", topn=10)
Out[16]: [('飲品', 0.9133813977241516),
          ('炸雞', 0.8783719539642334),
          ('冰淇淋', 0.8746067881584167),
          ('服飾', 0.8678401708602905),
          ('化妝品', 0.8615073561668396),
          ('零食', 0.8536527156829834),
          ('啤酒', 0.8466242551803589),
          ('珠寶', 0.8449875116348267),
          ('電子產品', 0.8367608785629272),
          ('食品', 0.8348352313041687)]
```

```
In [17]: model.wv.most_similar("car")
Out[17]: [('truck', 0.7824944257736206),
          ('brake', 0.7262388467788696),
          ('貨卡車', 0.716031551361084),
          ('motorcycle', 0.7159311771392822),
          ('motor', 0.7153626680374146),
          ('volkswagen', 0.7126927971839905),
          ('hybrid', 0.7122430205345154),
          ('saloon', 0.7067892551422119),
          ('cadillac', 0.7067204713821411),
          ('convertible', 0.7063345313072205)]
In [18]: model.wv.most similar("facebook")
Out[18]: [('instagram', 0.879693329334259),
          ('臉書', 0.8340373039245605),
          ('專頁', 0.7984201312065125),
          ('twitter', 0.779519259929657),
          ('xanga', 0.7731849551200867),
          ('facebook專頁', 0.7496058344841003),
          ('myspace', 0.7484684586524963),
          ('微博', 0.746212363243103),
          ('推特', 0.7453634142875671),
          ('新浪微博', 0.7418176531791687)]
In [19]: |model.wv.most_similar("詐欺")
Out[19]: [('恐嚇', 0.8686190843582153),
          ('盜竊', 0.8680432438850403),
('販毒', 0.8651722073554993),
          ('性騷擾', 0.8630328178405762),
          ('洗錢', 0.856837272644043),
          ('脅迫', 0.851883590221405),
          ('毆打', 0.848781943321228),
          ('的質疑', 0.8473358154296875),
          ('欺詐', 0.846764326095581),
          ('搶劫', 0.8438825607299805)]
In [20]: model.wv.most similar("合約")
         [('年內', 0.8128535151481628),
Out [20]:
          ('總值', 0.8031351566314697),
          ('據了解', 0.7691568732261658),
          ('卻在', 0.7609128355979919),
          ('耗資超過', 0.7589302659034729),
          ('鑑於', 0.7558416128158569),
          ('聯盟於', 0.751262366771698),
          ('億台幣', 0.7507293820381165),
          ('億美金', 0.74944007396698),
          ('往後', 0.7488248348236084)]
```

```
In [21]: model.wv.similarity("連結", "鏈結")
Out[21]: 0.5377023

In [22]: model.wv.similarity("連結", "陰天")
Out[22]: 0.30575657

In [23]: print(f"Loading {output_model}...")
new_model = word2vec.Word2Vec.load(output_model)
Loading word2vec.zh.300.model...

In [24]: model.wv.similarity("連結", "陰天") == new_model.wv.similarity("連結"
Out[24]: True
```