# Word2vec词向量

### In [1]:

```
#导入实验所需的工具包
import re
import jieba
import pandas as pd
from gensim.models.word2vec import LineSentence
from gensim.models import Word2Vec
import gensim
import logging
logging.basicConfig(format='%(asctime)s:%(levelname)s:%(message)s',level=logging.INFO)
```

# 数据预处理

### In [2]:

```
1  df = pd. read_csv("online_shopping_10_cats. csv")[:60000]
2  df. head()
```

### Out[2]:

	cat	label	review
0	书籍	1	做父母一定要有刘墉这样的心态,不断地学习,不断地进步,不断地给自己补充新鲜血液, 让自己保持
1	书籍	1	作者真有英国人严谨的风格,提出观点、进行论述论证,尽管本人对物理学了解不深,但是 仍然能感受到
2	书籍	1	作者长篇大论借用详细报告数据处理工作和计算结果支持其新观点。为什么荷兰曾经县有欧 洲最高的生产…
3	书籍	1	作者在战几时之前用了 " 拥抱 " 令人叫绝. 日本如果没有战败, 就有会有美军的占领, 没胡官僚主义的延
4	书籍	1	作者在少年时即喜阅读,能看出他精读了无数经典,因而他有一个庞大的内心世界。他的作品。 品最难能可贵…

### In [3]:

```
#使用re正则提取中文并用jieba分词提取词语语料
extract_chinese = re.compile(r'[\u4e00-\u9fa5]+')
chinese_corpus_raw = df['review'].tolist()
chinese_corpus_raw
df['chinese_corpus']=[jieba.lcut("".join(extract_chinese.findall(str(corpus)))) for corpus in
df.head()
```

Building prefix dict from the default dictionary ...

2022-06-03 15:30:34,023:DEBUG:Building prefix dict from the default dictionary ...

Loading model from cache C:\Users\23176\AppData\Local\Temp\jieba.cache

2022-06-03 15:30:34,026:DEBUG:Loading model from cache C:\Users\23176\AppData\Local \Temp\jieba.cache

Loading model cost 0.609 seconds.

2022-06-03 15:30:34,635:DEBUG:Loading model cost 0.609 seconds.

Prefix dict has been built successfully.

2022-06-03 15:30:34,637:DEBUG:Prefix dict has been built successfully.

### Out[3]:

	cat	label	review	chinese_corpus
0	书籍	1	做父母一定要有刘墉这样的心态,不断地学习,不断 地进步,不断地给自己补充新鲜血液,让自己保持…	[做, 父母, 一定, 要, 有, 刘墉, 这样, 的, 心态, 不断, 地, 学习, 不断
1	书籍	1	作者真有英国人严谨的风格,提出观点、进行论述论证,尽管本人对物理学了解不深,但是仍然能感受到	[作者, 真有, 英国人, 严谨, 的, 风格, 提出, 观点, 进行, 论述, 论证, 尽
2	书籍	1	作者长篇大论借用详细报告数据处理工作和计算结果 支持其新观点。为什么荷兰曾经县有欧洲最高的生产	[作者,长篇大论,借用,详细,报告,数据处理,工作,和,计算结果,支持,
3	书籍	1	作者在战几时之前用了 " 拥抱 " 令人叫绝. 日本如果 没有战败, 就有会有美军的占领, 没胡官僚主义的延	[作者, 在, 战, 几时, 之前, 用, 了, 拥抱, 令人, 叫绝, 日本, 如果, 没
4	书籍	1	作者在少年时即喜阅读,能看出他精读了无数经典, 因而他有一个庞大的内心世界。他的作品最难能可贵	[作者, 在, 少年, 时即, 喜, 阅读, 能, 看出, 他, 精读, 了, 无数, 经典

#### In [4]:

```
#将每条评论分词后整合到一个列表中,将每个词用空格隔开放入一个列表中
 1
 2
    words list = []
    corpus = []
 4
    for corpu in df['chinese corpus']. tolist():
 5
       words list.append(corpu)
       corpus.append(''.join(corpu))
 6
 7
    words list[0]
Out[4]:
```

```
['做',
'父母',
'一定',
'要',
'有',
 '刘墉'
'这样',
'的',
'心态'
'不断',
'地',
'学习'
'不断',
'地',
'进步'
'不断',
```

# skip-gram构建词向量

### In [5]:

'地', '给'.

```
#skip-gram构建词向量
  1
    skip model = Word2Vec(sentences=words list, sg=1, negative=10, vector size=500, window=5, min co
    skip model. save ("skip. model")
  3
    skip model = Word2Vec. load("skip. model")
2022-06-03 15:30:49, 120:INFO:collecting all words and their counts
2022-06-03 15:30:49,120:INFO:PROGRESS: at sentence #0, processed 0 words, keeping
0 word types
2022-06-03 15:30:49,177:INFO:PROGRESS: at sentence #10000, processed 379631 word
s, keeping 29235 word types
2022-06-03 15:30:49,228:INFO:PROGRESS: at sentence #20000, processed 666652 word
s, keeping 38407 word types
2022-06-03 15:30:49,269:INFO:PROGRESS: at sentence #30000, processed 867324 word
s, keeping 42195 word types
2022-06-03 15:30:49,302:INFO:PROGRESS: at sentence #40000, processed 1066310 word
s, keeping 47477 word types
2022-06-03 15:30:49,342:INFO:PROGRESS: at sentence #50000, processed 1242441 word
s, keeping 50332 word types
2022-06-03 15:30:49,430:INFO:collected 63744 word types from a corpus of 1722838
raw words and 60000 sentences
2022-06-03 15:30:49,431:INFO:Creating a fresh vocabulary
2022-06-03 15:30:49,650:INFO:Word2Vec lifecycle event {'msg': 'effective min coun
t=1 retains 63744 unique words (100.00% of original 63744, drops 0)', 'datetime':
'2022-06-03T15:30:49.650694', 'gensim': '4.2.0', 'python': '3.7.1 (default, Dec 1
```

```
In [6]:
```

```
1 #查看与给定词相似度最高的10个词
2 sims = skip_model.wv.most_similar('孩子', topn=10)
3 sims
```

### Out[6]:

```
[('宝宝', 0.818596601486206),
('小孩', 0.8021312355995178),
('大人', 0.7726424336433411),
('父母', 0.7704293727874756),
('儿子', 0.7631828188896179),
('小朋友', 0.7627369165420532),
('小孩子', 0.755376398563385),
('女儿', 0.7456046342849731),
('妈妈', 0.7416766881942749),
('家长', 0.7315623164176941)]
```

# 词向量可视化展示

### In [7]:

```
from collections import Counter
 2
   #统计词频
 3
   words_all_list = []
   for word list in words list:
 4
       for word in word list:
 5
 6
           if len(word)!=1:
 7
               words_all_list.append(word)
   counter = Counter(words all list)
 8
 9
   #按照词频降序排列取前100个
   words freq = sorted(list(counter.items()), key = lambda x:x[1], reverse=True)
10
   words top100=list(dict(words freq[:100]).keys())
11
   #获得词频前100的词的词向量
   vectors top100 = skip model.wv[words top100]
13
14
   vectors top100[0]
```

#### Out[7]:

```
array([-9.63533472e-04, 3.08847725e-01, 2.17242047e-01, -3.15669551e-02,
      -2.40764618e-01, -2.10673243e-01, -2.87210699e-02, 1.99883744e-01,
       1. 15101904e-01, 1. 56388119e-01, 4. 52579260e-02,
                                                          1.94912955e-01,
      -1.04395011e-02, -3.98569852e-02,
                                         9. 28105116e-02, -8. 20474476e-02,
       6. 08142540e-02, 9. 85509232e-02, -1. 42001584e-01, -1. 32267684e-01,
       7. 53863901e-02, 9. 34082025e-04, 3. 84868145e-01, 7. 39663467e-02,
       3. 69062126e-02, -4. 36517671e-02, -1. 07947372e-01, -1. 15842782e-02,
      -7. 91159198e-02, -5. 01922928e-02, 1. 25287294e-01, 4. 58126888e-02,
       5. 52382991e-02, 5. 05636893e-02,
                                         1. 57338724e-01, 3. 28775406e-01,
       1.00260787e-01, -9.23958495e-02, 2.21496373e-02, -8.03652927e-02,
                       1.96411274e-02, -3.11306477e-01,
       4. 12130617e-02,
                                                           3.50587405e-02,
      -2.39718720e-01, 5.64864650e-02, -7.53426924e-02, 8.76449421e-03,
      -6.42778203e-02, 1.32489577e-01, -1.44223310e-02, 1.07294023e-01,
      -2.49427810e-01, -7.08280429e-02, 1.63544506e-01, -1.14593111e-01,
       3. 75115126e-02, -1. 44282296e-01,
                                         3.70074391e-01, 8.74571577e-02,
      -1.84445940e-02, -7.38058053e-03, 5.43578938e-02, -1.78246841e-01,
      -1.15735322e-01, 9.05112028e-02, -8.09661075e-02, 1.44248590e-01,
       4. 79126573e-01. -1. 58096571e-02. -2. 98388034e-01. 4. 43404689e-02.
```

### In [8]:

```
#使用PCA降维得到2维词向量
 2
   from sklearn.decomposition import PCA
   pca = PCA(n_components=2)
 4
   word 2d = pca. fit transform(vectors top100)
 5
   #词向量可视化展示
   import matplotlib.pyplot as plt
 6
   plt.rcParams['font.sans-serif']=['SimHei'] #图中文字体设置为黑体
   plt.rcParams['axes.unicode_minus']=False #负值显示
   plt.figure(figsize=(20,10))
10
   plt.scatter(word_2d[:, 0], word_2d[:, 1])
   for i in range(word_2d.shape[0]):
       plt.text(word_2d[i,0]*1.05, word_2d[i,1]*1.05, words_top100[i], fontsize=10, color = "r",
12
   plt.savefig('word_plot.png', dpi=300, bbox_inches='tight')
13
```

