```
In [1]: WIKI_SEG_TXT = "wiki_seg.txt"
In [2]: from gensim.models import FastText
        from gensim.models.word2vec import LineSentence
        import multiprocessing
        max cpu counts = multiprocessing.cpu count()
        word dim size = 300 # 设置word vector维度
        print(f"Use {max cpu counts} workers to train FastText (dim={word dim size})")
       WIKI SEG TXT = "wiki seg.txt"
        # 读取训练语句
        sentences = LineSentence(WIKI SEG TXT)
        # 训练模型
        model = FastText(sentences, vector size=word dim size, workers=max cpu counts)
        # 保存模型
        output_model = f"fasttext.zh.{word_dim_size}.model"
        model.save(output model)
```

Use 12 workers to train FastText (dim=300)

```
In [3]: print(model.wv.vectors.shape)
       model.wv.vectors
       (1281108, 300)
Out[3]: array([[-2.1143034 , -3.5834088 , 0.9235744 , ..., -2.9513686 ,
               0.85165715, 0.7485893 ],
             [0.44912398, 1.157038, 0.17390107, ..., -1.5394628]
              -2.0758872 , 2.3475122 ],
             [-4.7415714, 4.1082935, 3.0406713, ..., -0.18160735,
               0.97919846, -3.1430063 ],
              [-0.42775387, -0.4923242, 0.3848211, ..., 0.34394506,
               0.08799452, -0.6257678 ],
              [0.3650192, -0.02768007, 0.23861456, ..., 0.26748458,
               0.39309955, 1.2193569 ],
             [-0.14890154, -0.08819858, -0.0111459, \ldots, 0.13148315,
              -0.0556561 , -0.13519895]], dtype=float32)
In [5]: # 加载并训练FastText模型
       model = FastText.load('fasttext.zh.300.model') # 替换成您的模型路径
       vocab = model.wv.key to index
       # 打印总共收录的词汇数
       print(f"總共收錄了 {len(vocab)} 個詞彙")
       # 打印前20个收录的词汇
       print("印出 20 個收錄詞彙:")
       print(list(vocab.keys())[:20])
       總共收錄了 1281108 個詞彙
       印出 20 個收錄詞彙:
       ['年','月','日','中','10','12','11','小行星','中國','時','-','日本','美國','20','香港','臺灣','15','位
```

於','30','站']

```
Out[6]: array([-0.36248013, 0.98012364, 0.55402476, 0.4230872, -0.08536739,
               0.37592587, -1.7252742, 0.35973006, 1.0355555, 0.20347224,
               0.89693195, 2.3107479, 1.1560698, 0.43550915, -1.8510116,
               1.8772343 . 2.1570227 . -0.8266401 . -0.4693025 . -2.9896805 .
              -0.76506203, 0.55268776, -0.6439924, -2.1026824, 0.82134354,
               -0.03820011. 1.9603138. 3.5589116. 1.4944109. -1.1495035.
               -2.2098894 , -2.3953311 , -0.26409993 , 0.7407733 , 0.37456003 ,
               -3.1395462 , 0.02966588, 1.4000149 , -0.7277578 , 0.29956517,
               0.23028906, -3.5200157, -1.7080253, 1.3862095, -0.80813515,
              -2.0764017 , 1.3242884 , 0.78814757 , 0.70669633 , 0.8410793 ,
               -2.373359 , 0.17978494, -1.7763336 , -0.5197104 , 0.02317583,
               -0.61163217. -0.24958536. 1.333366 . 0.6208196 . 0.62406814.
               0.20208071, -0.8206698 , -1.3902856 , -1.7098118 , 0.6752405 ,
               2.733623 , 1.2758677 , -0.6498876 , 0.2879006 , 0.38086182,
               0.99583673. -0.8345073. -0.37709767. 0.33063385. -0.92941964.
               0.00550079, 0.5862416, 0.87920475, 0.52489877, -3.6531105,
               0.4016076 , -0.6659497 , 0.82306755 , -3.8803885 , -1.2592509 ,
               1.3713334 , 1.7969382 , -0.40232936 , -0.06662682 , -1.0427252 ,
               1.8947021 , 0.33118635, -3.2141137 , 0.13048308, -2.2011406 ,
               0.77231276, 1.3026292, 1.8185749, -0.91952515, -1.3727586,
               1.6142715 , -0.35515103, -1.1541601 , 0.8725955 , 2.8665287 ,
               -1.7573503 , 0.9622064 , 2.2132113 , 2.2489228 , 1.295852 ,
               0.41261426, 1.413129 , -0.97843295, 0.48973542, 1.3581734 ,
               0.71574175, -3.1343539, -2.4122615, -2.9045713, 0.8909527,
               -1.7113088 , 1.910596 , 1.9032061 , 1.8908036 , 1.0927012 ,
               1.2950226 , 1.3321568 , 0.9011788 , -0.04983101 , -1.31874 ,
               1.4708176 , 0.09124085, -2.5455387 , 1.1389534 , 2.9139862 ,
               1.9673725 , 2.7447848 , -1.3354753 , 3.1814198 , 0.14665474,
               -1.8544286 . 0.8223278 . 3.2614264 . -0.8112384 . -0.09817249.
               1.4645436 , 0.75998557, -0.8995562 , 0.11551278, 1.5027435 ,
               1.8521479 , 1.0239952 , 1.5470933 , 1.8559628 , 1.3798646 ,
               3.0782177 , 4.1210227 , 1.4251592 , 0.9985844 , -2.2082667 ,
              -2.6183095 , -1.1970755 , -3.152174 , -2.6965501 , 0.17867514,
               -0.9372266 , -0.32081595 , -1.0925673 , 2.88764 , 0.38610265 ,
               0.05830714, -1.0056459, 1.7154334, -1.44236, -0.5403752,
               0.42215803, 1.2282223, 3.1132028, -1.5636071, 0.40505138,
               1.432704 . 2.3415618 . 2.3565447 . 2.7195437 . 0.8814994 .
               -0.4873589 . 1.1314331 . 0.57285845 . 0.74154717 . 0.41416332 .
               0.7483571 , 0.31172642 , 0.6211714 , 0.21189976 , 1.6088046 ,
               0.5581938 \cdot -1.8051902 \cdot -1.5454528 \cdot 2.6511977 \cdot -0.30181536
              -1.6230351 , 0.8690149 , -0.6757468 , 2.7721062 , 1.1747085 ,
```

```
1.4936074 . -1.3174665 . -0.05859082 . 3.9109533 . 0.04994953 .
 0.3170531 , 1.0511549 , -0.42471218 , -0.27838972 , 1.7440172 ,
 0.719544 . 0.64353424 .- 0.34134546 .- 1.178439 .- 0.9027263 .
 -2.2345853 , 1.2022699 , 1.0734307 , -0.41167647, 2.8903198 ,
 1.2145857 , 0.11285645, -2.2847824 , 0.1919113 , 0.16430385,
 0.9349448 , 1.0600609 , 0.52170885 ,-0.38075987 , 2.165146 ,
 -2.3794723 . -0.91174155. 1.0703781 . 2.0973678 . 0.55109
 0.52043056, -0.31678596, -0.1718689, 2.059049, -0.04883113,
 2.1690748 , 3.3277364 , 1.179526 , -1.1479416 , -0.26288223,
 -1.8426039 , 2.4506278 , -0.9216002 , -0.31827846 , -2.2192037 ,
-0.6037803 , 1.1627191 , -0.30118388 , 1.6683006 , 1.8911893 ,
 2.2625875 , 0.31399542, -0.2117326 , 0.24281022, 3.30119 ,
 2.025536 , 0.39673144, 0.04100621, 0.77655894, -0.6997056 ,
 -0.37861657, 0.8386439, -1.5636232, 0.66094923, -0.30948257,
 2.08865 , 1.3413908 , 0.11441649 , 0.10324325 , 0.53062344 ,
 -0.269244 , -0.46463028, 1.6779274 , -3.0914824 , -1.921438 ,
 1.3992082 , -0.23602292 , 0.7996966 , -1.2923689 , 0.68264467 ,
-2.070288 , -0.6330929 , 0.01733787 , 0.82230055 , -2.3566875 ,
 1.3539721 , -1.0342708 , 0.3492705 , 1.1793842 , 0.5706733 ],
dtvpe=float32)
```

```
In [7]: word = "這肯定沒見過 "

# 若強行取值會報錯

try:
    vec = model.wv[word]
    except KeyError as e:
        print(e)
```

```
In [8]: model.wv.most similar("飲料", topn=10)
 Out[8]: 「('輝劍', 0.9710693359375),
          ('名松', 0.9501043558120728),
          ('飲料類'、0.9315357208251953)、
          ('飲料機', 0.9279479384422302),
          ('飲料罐', 0.8953368663787842),
          ('軟飲料', 0.8831291198730469),
          ('茶飲料', 0.8725141882896423),
          ('經米濱'、0.8699420690536499)、
          ('飲品', 0.8453800082206726),
          ('飲料瓶', 0.7923927307128906)]
 In [9]: model.wv.most similar("car")
 Out[9]: [('hcar', 0.8572422862052917),
          ('carcar', 0.8509683012962341),
          ('ccar', 0.849014401435852),
          ('jetcar', 0.8113610148429871),
          ('tramcar', 0.8039993643760681),
          ('zipcar', 0.8031772971153259),
          ('motorcar', 0.8004679083824158),
          ('boxcar', 0.8001610636711121),
          ('indycar', 0.7994945645332336),
          ('cars', 0.7986459136009216)]
In [10]: model.wv.most similar("facebook")
Out[10]: [('voutubefacebook', 0.927127480506897),
          ('thefacebook', 0.8969534635543823),
          ('facebookpage', 0.8882699012756348),
          ('facebox', 0.8686223030090332),
          ('instagram', 0.7984750270843506),
          ('twitteryoutube', 0.7682653069496155),
          ('googleyoutube', 0.7594155669212341),
          ('twitter', 0.7524656057357788),
          ('youtube', 0.7465772032737732),
          ('lnstagram', 0.7246598601341248)]
```

```
In [11]: model.wv.most similar("詐欺")
Out[11]: 「('賈邱', 0.8884372115135193),
          ('赤坑鎮', 0.831690788269043),
          ('中境', 0.8139835000038147),
          ('越中境', 0.7967145442962646),
          ('詐欺罪', 0.7719191908836365),
          ('他魚', 0.7685927152633667),
          ('欺詐', 0.7375842332839966),
          ('抱出', 0.7236839532852173),
          ('欺詐案', 0.6801178455352783),
          ('義德堂', 0.6536234021186829)]
In [12]: model.wv.most similar("合約")
Out[12]: 「('德康', 0.9192003607749939),
          ('合同', 0.8019339442253113),
          ('綠蠅', 0.749859631061554),
          ('合同期', 0.7271698713302612),
          ('合同額', 0.717912495136261),
          ('合同商', 0.7071802616119385),
          ('簽約', 0.7068753242492676),
          ('續約', 0.7045730948448181),
          ('籤合同', 0.6919059753417969),
          ('合同制', 0.6848821640014648)]
In [13]: model.wv.similarity("連結", "鍵接")
Out[13]: 0.4270057
In [14]: model.wv.similarity("連結", "陰天")
Out[14]: -0.015750855
```

```
In [17]: print(f"Loading {output_model}...")
    new_model = FastText.load(output_model)

Loading fasttext.zh.300.model...

In [18]: model.wv.similarity("連結", "陰天") == new_model.wv.similarity("連結", "陰天")

Out[18]: True

In []:
```