04ClusteringPracticeWord2Vec

November 10, 2018

```
In [1]: import pandas as pd
    import numpy as np
    import nltk
    from pprint import pprint

    from nltk.corpus import stopwords
    # if this is first time you use nltk, please open terminal and type codes bellow
    # >>> import nltk
    # >>> nltk.download('stopwords')
    # >>> nltk.download('punkt')

    stops = set(stopwords.words('english'))
    import string
    puns = string.punctuation
```

1 Word2Vec

1. 簡介

Word2Vec 其實是 Word to Vector 的簡稱,意在將每一個字轉換成一條向量,並讓這字的語意透過這條向量描繪出來。早期做自然語言處理時,很難對讓電腦對詞背後的意思有更深一層的理解,因此詞與詞之間的關係很難被挖掘出來,像是相似詞、相反詞、對應詞等,因此Word2Vec 在這樣的背景下產生就顯得極其珍貴。

- 2. 作用 & 賣點
 - 1. 它可以找到相似的字。
 - 2. 它可以加減,像是 Taiwan-Taipei=Germany-Berlin。
- 3. 訓練方法: 參照這篇網誌
- 4. 已經訓練好的 model:
 - 1. Various Models
 - 2. 本課程使用之GloVe Word2Vec(時間考量已經壓縮過,只留下這個文件中會用到的字)。
- 5. 專案: 將 e-commerce 商品標籤分群

2 Load Data

```
In [2]: with open('all_categories.list', 'r', encoding='utf8') as f:
           all_categories = np.array(eval(f.read()))
       print("商品標籤個數:", len(all_categories))
       print("前 10 個商品標籤:", all_categories[:10])
商品標籤個數:910
前 10 個商品標籤: ['Small Animal' 'Kitchen' 'Fragrance' 'Track & Sweat Suits' 'Wallet'
 'Favors' 'Quilts' 'Sticker' 'Pets' 'Skirt']
In [3]: word_vec_mapping = {}
       path = "glove.twitter.27B.50d.txt"
       # 打開上述檔案,並將每一行中的第一個詞作為 key,後面的數字做為向量,加入到 word_vec_mapping
       with open(path, 'r', encoding='utf8') as f: ## 這個文檔的格式是一行一個字並配上他的向量
           for line in f:
               #=======your works starts=======#
               tokens =
               token =
               vec =
               word_vec_mapping[token] =
               #======your works ends======#
       vec_dimensions = len(word_vec_mapping.get('men'))
       print("vec_dimensions:", vec_dimensions)
       print("word_vec_mapping length:", len(list(word_vec_mapping.items())))
       pprint(list(word_vec_mapping.items())[:5])
       # vec dimensions: 50
       # word_vec_mapping length: 947
       # [('shoes',
          array([-0.75313002, -1.78719997, 0.14522 , -0.29681 , 0.12436
                -0.40922999, 1.22679996, 0.50806999, 0.27913001, 0.34277001,
                -0.013902 , 1.52499998, -3.44880009, 1.05630004, -0.49985
vec_dimensions: 50
word_vec_mapping length: 947
[('protection',
 array([-2.6041e-01, 9.3470e-03, -1.2779e+00, 9.3997e-01, 1.3464e-01,
       2.8652e-01, 4.7567e-01, -5.1847e-01, 6.8337e-01, -6.8621e-01,
       2.5913e-01, 2.9725e-01, -2.7242e+00, 3.8473e-01, 1.2560e+00,
       9.2542e-01, -1.0193e-01, 1.5966e-01, 2.2935e-03, -3.8759e-01,
      -1.0683e+00, -5.3661e-01, 6.1156e-03, 1.5376e-01, 3.5490e-01,
       5.9846e-01, 8.5329e-02, 8.7829e-01, 2.1870e-01, 1.0114e+00,
      -6.9214e-03, -5.1215e-01, -2.7296e-01, -8.3198e-01, 8.5664e-01,
      -5.2144e-01, -4.6561e-01, 9.8429e-01, -6.7122e-01, -9.6129e-01,
       7.8881e-01, -6.1323e-01, 2.6551e-01, -3.9457e-01, 5.8291e-01,
      -5.6443e-01, 3.0565e-01, 4.1577e-02, 8.3677e-01, -4.5295e-01],
```

```
dtype=float32)),
('headsets',
array([-0.026823 , -0.092859 , -1.0312 , 0.90884 , -0.49068 ,
               , 0.63185 , -0.61496 , 0.57759 , -1.0434
      0.3701
     -0.21304 . 1.0892
                          , -0.67756 , 0.39059 , 0.12729
     -0.39507 , 0.070779 , -0.53489 , 0.62134 , 0.35858
     -0.3978
             , -0.053658 , 0.59362 , -0.15411 , -0.50521
      0.64195
               , -0.34041 , 0.33592 , -0.085267 , 0.035649 ,
      0.53375 , 0.034903 , 0.54944 , -0.80055 , 0.72882
     -1.1041
             , 0.41168 , -0.53155 , -0.46107 , 1.2108
               , 1.1987 , -0.52795 , -0.0043981, -0.30671
      1.3971
     -0.059243 , -0.079573 , -0.33629 , 0.24877 , 0.44079 ],
    dtype=float32)),
('jackets',
array([ 8.5055e-02, -1.7499e+00, -1.7031e-01, -2.6836e-01, -4.0375e-01,
     -3.1689e-01, 9.3073e-01, 1.2195e-01, 5.3707e-01, -8.3221e-01,
      9.2084e-01, 1.1719e+00, -2.0878e+00, 6.3773e-01, -2.4071e-01,
      1.3865e-03, -1.3375e+00, -8.2863e-01, -6.2409e-02, 1.2643e-01,
      3.3143e-01, -4.6428e-02, 1.2935e+00, -8.8655e-01, -4.2175e-01,
      1.3437e+00, 3.2258e-01, 5.6446e-01, -3.7294e-01, -9.2323e-01,
      2.8631e-01, 1.0212e+00, 6.5575e-01, -1.7160e-01, 1.1947e+00,
     -1.6754e+00, -1.0792e+00, 2.9502e-01, 1.3830e-01, 9.4862e-01,
      5.9838e-01, -9.7969e-02, 2.2767e-01, -1.6997e-01, -6.8851e-02,
      1.1240e-01, 5.0171e-01, -1.7130e-01, -2.4996e-01, 2.0415e-01],
    dtype=float32)),
('cart',
array([-9.5941e-01, -1.3244e-01, -4.0781e-01, -1.6105e-01, 9.0079e-01,
     -6.2495e-01, 1.9808e-02, -3.3392e-01, 6.9751e-01, -3.2706e-01,
      7.6056e-01, 9.7973e-01, -2.0602e+00, 2.8303e-02, 4.0289e-01,
      3.9693e-01, 3.7966e-01, -6.2229e-01, -2.0016e-03, -6.7711e-02,
     -2.3231e-01, -2.5877e-01, 6.0567e-01, -5.0645e-03, -1.0308e-01,
      8.1790e-01, -8.9303e-01, 2.6405e-01, 1.9210e-01, -4.2307e-01,
      2.5479e-01, -6.8431e-02, -1.4063e-02, -1.3592e+00, 1.5808e-01,
      2.1494e-01, 6.8527e-01, 5.2884e-01, 3.3263e-01, 2.5476e-02,
     -2.1072e-01, 1.3368e+00, -8.4683e-01, -7.6833e-01, 8.5318e-01,
      4.0324e-01, -3.4235e-02, 2.7536e-01, 1.2719e-01, 4.8708e-01],
    dtype=float32)),
('seats',
array([ 0.26285 , -0.36878 , 0.95363 , 0.25466 , -0.21188 ,
     -0.50289 , 0.92677 , 0.46699 , 0.26938 , -0.73464
      0.68061 , -0.46538 , -2.9934
                                     , 0.15943 , 0.92958
      0.96445 , -0.19603 , -0.84996 , -0.57998 , 0.22855
     -0.80445 , -1.4504 , 0.77642 , -0.049717 , -0.37453
      0.99299 , -0.78962 , -0.33776 , -0.43973 , -0.35432
                                                , 1.4032
      0.66495 , 0.72955 , 0.43549 , 0.49489
     -0.42153 , 0.0055356, 0.2083 , 0.24984 , 0.24588
      0.60279 , 1.9941 , -0.25869
                                     , -0.47908 , 0.1719
     -0.30145 , 0.67101 , -0.20577 , 0.54319 , 0.74157 ],
```

3 Tokenize & Doc2Vec

```
In [4]: #將每一個句子 (商品類別) 的詞彙
       # 1. 切割開來
       # 2. 去掉停用字 (set(stopwords.words('english')))
       # 3. 去掉標點符號 (string.punctuation)
       # 4. 轉小寫
       def tokenize(Doc):
           if pd.notnull(Doc):
              # 使用 nltk.wordpunct_tokenize 將 Doc 切開
              # 去掉停用字與標點符號,並轉小寫
              #======your works starts======#
              tokens =
              words =
              #=======your works ends=======#
              return words
           else:
              return None
       print("before tokenize:", all_categories[0])
       print("after tokenize:", tokenize(all_categories[0]))
       print("before tokenize:", all_categories[3])
       print("after tokenize:", tokenize(all_categories[3]))
       # before tokenize: Small Animal
       # after tokenize: ['small', 'animal']
       # before tokenize: Track & Sweat Suits
       # after tokenize: ['track', 'sweat', 'suits']
before tokenize: Small Animal
after tokenize: ['small', 'animal']
before tokenize: Track & Sweat Suits
after tokenize: ['track', 'sweat', 'suits']
In [5]: test_arr = np.array([
           [1,2,3,4],
           [4,5,6,7],
           [7,8,9,10]
       ])
       # 請將 test_arr 中的三個 array 做 element-wise 的平均
       #=======your works starts========#
       test_out =
       #=======your works ends========#
```

```
# [4. 5. 6. 7.]
[4. 5. 6. 7.]
In [6]: def doc2vec(doc, word2vec=word_vec_mapping):
          if pd.notnull(doc):
              # 使用剛剛定義好的 tokenize 函式 tokenize doc, 並指派到 trems
              # 找出每一個詞彙的代表向量 (word_vec_mapping)
              # 並平均 (element-wise) 所有出現的詞彙向量 (注意 axis=0)·作為 doc 的代表向量
              #======your works starts=======#
              terms =
              termvecs =
              docvec =
              if np.sum(np.isnan(docvec)) > 0:
              ## 若找不到對應的詞向量,則給一條全部為零的向量,長度為原詞彙代表向量的長度 (vec a
              #======your works starts======#
              docvec=
              #=======your works ends=======#
          return docvec
       print("before tokenize:", all_categories[3])
       print("output shape", doc2vec(all_categories[3]).shape)
       print("after tokenize:", doc2vec(all_categories[3])[:5])
       print("before tokenize:", all_categories[70])
       print("output shape", doc2vec(all_categories[70]).shape)
       print("after tokenize:", doc2vec(all_categories[70])[:5])
       # before tokenize: Track & Sweat Suits
       # output shape (50,)
       # after tokenize: [-0.76383996 -0.49650002 0.23154134 -0.16717 0.42855397]
       # before tokenize: Teethers
       # output shape (50,)
       # after tokenize: [0. 0. 0. 0. 0.]
before tokenize: Track & Sweat Suits
output shape (50,)
after tokenize: [-0.76383996 -0.49650002 0.23154134 -0.16717 0.42855397]
before tokenize: Teethers
output shape (50,)
after tokenize: [0. 0. 0. 0. 0.]
 avg = a.mean(axis)
 ret = ret.dtype.type(ret / rcount)
```

print(test_out)

```
In [7]: #將 doc2vec 應用到 all_categories 中的每一個元素上
       #=======your works starts=======#
       cat_vecs =
       print("cat_vecs length:", len(cat_vecs))
       print(cat_vecs[3][:5])
       print(cat_vecs.shape)
       # cat_vecs length: 910
       # [-0.76383996 -0.49650002 0.23154134 -0.16717 0.42855397]
       # (910, 50)
cat_vecs length: 910
[-0.76383996 -0.49650002 0.23154134 -0.16717 0.42855397]
(910, 50)
   Clustering
In [8]: from sklearn.cluster import KMeans
       from sklearn.cluster import AgglomerativeClustering
       from sklearn.cluster import DBSCAN
       from collections import Counter
       X = cat_vecs
       n_clusters= 20
4.1 K means
In [9]: # 請使用 kmeans 將商品類別分成 20 類
       #=======your works starts========#
       kmeans =
       all_categories_labels_kmeans =
       #========your works ends========#
       for i in range(3):
          print(";;".join(all_categories[all_categories_labels_kmeans==i]))
          print("======")
Athletic Training;;Golf;;Golf Apparel;;Track & Field;;Boxing & MMA;;Lacrosse;;Hockey;;Badminto:
Small Animal;; Pets;; Animals;; Toddler;; Animal;; Art Doll;; Kids;; Human Figure Doll;; Afghan;; Doll
Girls;;Face;;Fan Shop;;Dress Up & Pretend Play;;Sleep Positioners;;Watch;;Boys;;50 To 75 Years
```

4.2 Hireachy(single link)

```
In [10]: #請使用 hierachical(single link) 將商品類別分成 20 類
        #======your works starts=======#
        hierachy =
        all_categories_labels_single =
        #=======your works ends=======#
        for i in range(3):
            print(";;".join(all_categories[all_categories_labels_single==i]))
            print("=======")
Athletic Training;; Motorcycle;; Golf;; Golf Apparel;; Hiking & Camping;; Water Sports;; Bike & Skat-
Historical, Military;; Exercise;; Arts & Crafts;; Education & Teaching;; Writing;; Potty Training;;
Quilts;; Calendars;; Diapering;; Plush;; Knitting Supplies;; Baskets & Bins;; Yarn;; Baguette;; Painti:
   Hireachy(average link)
In [11]: #請使用 hierachical (average link) 將商品類別分成 20 類
        #======your works starts======#
        hierachy =
        all_categories_labels_average =
        #=======your works ends=======#
        for i in range(3):
            print(";;".join(all_categories[all_categories_labels_average==i]))
            print("=======")
Athletic Training;; Boxing & MMA;; Lacrosse;; Hockey;; Badminton;; Bowl;; Varsity;; Team Sports;; Athletic Training;
Dusting;; Waxing;; Thermometers;; Oils & Fluids;; Toothbrushes;; Sponges;; Brushes & Applicators;; Br
Eyes;;Feet;;Sleep Positioners;;Lips;;Relaxed
```

4.4 Hireachy(complete link)

```
In [12]: # 請使用 hierachical(complete link) 將商品類別分成 20 類
#========your works starts=======#

hierachy =
all_categories_labels_complete =
#========your works ends=======#
```

```
for i in range(3):
            print(";;".join(all_categories[all_categories_labels_complete==i]))
            print("======")
Kitchen;; Photo Albums & Frames;; Plush;; Storage & Organization;; Knitting Supplies;; Puzzles;; Bas:
Pets;; Athletic Training;; Animals;; Golf;; Water Sports;; Exercise;; Track & Field;; Education & Tea
Charm; ;Beads; ;Felted; ;Bracelet; ;Handmade; ;Vintage & Collectibles; ;Geekery; ;Bead; ;Glassware; ;Wr
4.5 DBSCAN
In [13]: # 請使用 hierachical(complete link) 將商品類別分成 20 類 (eps=0.3)
         #======your works starts=======#
        dbscan =
        all_categories_labels_dbscan =
         #======your works ends=======#
        for i in [-1, 0]:
            print(";;".join(all_categories[all_categories_labels_dbscan==i]))
            print("=======")
==========
Teethers;;Playards;;Epilators;;Sweatercoat;;Rainwear;;Needlecraft;;Bedspreads & Coverlets;;Deh
In [14]: all_categories_labels_dbscam = DBSCAN().fit_predict(X)
        Counter(all_categories_labels_dbscam)
         # Counter({-1: 898, 0: 12})
Out[14]: Counter({-1: 898, 0: 12})
In [15]: df_cat = pd.DataFrame(all_categories_labels_dbscam, index=all_categories, columns=['16]
        print(list(df_cat[df_cat['label'] == 0].index))
         # ['Teethers', 'Playards', 'Epilators', 'Sweatercoat', 'Rainwear', 'Needlecraft', 'Be
['Teethers', 'Playards', 'Epilators', 'Sweatercoat', 'Rainwear', 'Needlecraft', 'Bedspreads & '
```

5 PCA

```
from sklearn.decomposition import PCA
         def draw_PCA(X, Y, title):
             fig = plt.figure(1, figsize=(8, 6))
             ax = Axes3D(fig, elev=-150, azim=110)
             X_reduced = PCA(n_components=3).fit_transform(X)
             ax.scatter(X_reduced[:, 0], X_reduced[:, 1], X_reduced[:, 2], c=Y,
                        cmap=plt.cm.Set1, edgecolor='k', s=40)
             ax.set_title(title)
             ax.set_xlabel("1st eigenvector")
             ax.w_xaxis.set_ticklabels([])
             ax.set_ylabel("2nd eigenvector")
             ax.w_yaxis.set_ticklabels([])
             ax.set_zlabel("3rd eigenvector")
             ax.w_zaxis.set_ticklabels([])
             plt.show()
In [17]: draw_PCA(X, all_categories_labels_kmeans, 'kmeans')
In [18]: draw_PCA(X, all_categories_labels_single, 'single link')
In [19]: draw_PCA(X, all_categories_labels_average, 'average link')
```

```
In [20]: draw_PCA(X, all_categories_labels_complete, 'complete link')
```



In [21]: draw_PCA(X, all_categories_labels_dbscan, 'dbscan')



6 Evaluation

```
In [22]: df_cat = pd.DataFrame(all_categories_labels_kmeans, index=all_categories, columns=['labels_kmeans, index=all_categories, index=all_categories, columns=['labels_kmeans, index=all_categories, in
                    for i in range(len(set(all_categories_labels_kmeans))):
                              cats = list(df_cat[df_cat['label'] == i].index)
                              print("cluster " + str(i) + ": ")
                              print(list(cats))
                              print("======="")
                              print("======="")
cluster 0:
['Athletic Training', 'Golf', 'Golf Apparel', 'Track & Field', 'Boxing & MMA', 'Lacrosse', 'Ho
_____
_____
cluster 1:
['Small Animal', 'Pets', 'Animals', 'Toddler', 'Animal', 'Art Doll', 'Kids', 'Human Figure Dol'
_____
_____
cluster 2:
['Girls', 'Face', 'Fan Shop', 'Dress Up & Pretend Play', 'Sleep Positioners', 'Watch', 'Boys',
['Teethers', 'Playards', 'Humor', 'Epilators', 'Cargo', 'Sweatercoat', 'Rainwear', 'Needlecraf
_____
cluster 4:
['Crewneck', 'Turtleneck', 'Vest', 'Gloves', 'Sweatshirt, Pullover', 'Blazers & Sport Coats',
```

```
['Fragrance', 'Plate', 'Coffee & Tea Accessories', 'Pot Holder', 'Bubble Bath', 'Baguette', 'D
_____
_____
cluster 6:
['Dusting', 'Monitors', 'Thermometers', 'Oils & Fluids', 'Garbage Disposals', 'Amplifiers & Ef
_____
cluster 7:
_____
_____
['Motorcycle', 'Storage & Organization', 'Medical Supplies & Equipment', 'Car Electronics & Ac
_____
['Books', 'Puzzles', 'Arts & Crafts', 'Paintings', 'Crafting', 'Christian Books & Bibles', 'No.
_____
_____
cluster 10:
['Quilts', 'Diapering', 'Plush', 'Baskets & Bins', 'Yarn', 'Teapot', 'Bakeware', 'Grooming', '
_____
_____
['Historical, Military', 'Trading Cards', 'Exercise', 'Education & Teaching', 'Health', 'Busing
_____
cluster 12:
['Skirt', 'Dresses', 'Leggings', 'Jeans', 'Fashion Sneakers', 'Jerseys', 'Sweater', 'Dress Sui
_____
_____
['Luggage', 'Pouch', 'Felted', 'Blouse', 'Lightweight', 'Tiered', 'Double Breasted', 'Frame',
  _____
-----
cluster 14:
['Track & Sweat Suits', 'Wallet', 'Full-Length', 'Shams, Bed Skirts & Bed Frame Draperies', 'S
_____
_____
cluster 15:
['Charm', 'Beads', 'Bracelet', 'Handmade', 'Vintage & Collectibles', 'Bead', 'Earrings', 'Cabo
_____
_____
```

['TV, Audio & Surveillance', 'Laptop', 'DVD & Blu-ray Players', 'Radio', 'Patch', 'Pin', 'Magi

```
______
cluster 17:
['Eyes', 'Waxing', 'Hair Relaxers', 'Nail Care', 'Pregnancy & Maternity', 'Scrubs & Body Treats
_____
______
['Sticker', 'Photo Albums & Frames', "Men's Accessories", 'Changing Kits', 'GPS Accessories & N
_____
        _____
cluster 19:
['Kitchen', 'Hiking & Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', 'Home Decor', 'Kitchen Safety', 'Kitchen Storage Camping', 'Water Sports', '
_____
 -------
In [23]: df_cat = pd.DataFrame(all_categories_labels_single, index=all_categories, columns=['labels_single, index=all_c
                     for i in range(len(set(all_categories_labels_kmeans))):
                               cats = list(df_cat[df_cat['label'] == i].index)
                               print("cluster " + str(i) + ": ")
                               print(sorted(list(cats)))
                               print("======="")
                               print("======="")
cluster 0:
['Athletic', 'Athletic Apparel', 'Athletic Training', 'Badminton', 'Ballet', 'Band & Orchestra
_____
 -----
cluster 1:
['Afghan', 'Arts & Crafts', 'Bomber', 'Education & Teaching', 'Educational', 'Exercise', 'Fitne
_____
_____
cluster 2:
['Aceo', 'Baguette', 'Bakeware', 'Baskets', 'Baskets & Bins', 'Bedding', 'Bookmark', 'Bouquets
_____
cluster 3:
['Backpacks & Carriers', 'Batteries', 'Binoculars & Telescopes', 'Brushes', 'Brushes & Applica
_____
_____
['Apron', 'Asymmetrical', 'Asymmetrical Hem', 'Beading', 'Capri, Cropped', 'Capris, Cropped',
_____
_____
cluster 5:
['Apparel', 'Belt', 'Blazer', 'Blazers & Sport Coats', 'Blouse', 'Boots', 'Bottoms', 'Buckle',
_____
_____
cluster 6:
```

```
['Accessory', 'Action Figures & Statues', 'Activity Centers & Entertainers', 'Art', 'Arts & Pho
_____
_____
cluster 7:
['100 Years or Older', '50 To 75 Years', '75 To 100 Years', 'All Other Sports', 'Baby', 'Baby'
_____
_____
cluster 8:
cluster 9:
['Accessories', 'Bathing Accessories', 'Beach Accessories', 'Custom', 'Dolls & Accessories', '
_____
_____
['A-Line', 'Action Figure', 'Action, Adventure', 'Baby Gyms & Playmats', 'Baby Seats', 'Block'
_____
cluster 11:
['Baseball', 'Baseball & Softball', 'Basketball', 'Bowl', 'Bowls', 'Football', 'Game', 'Games'
_____
_____
['Antique', 'Bead', 'Beads', 'Bracelet', 'Bracelets', 'Brooch', 'Cabochon', 'Charm', 'Collecti'
_____
cluster 13:
['Air Conditioners', 'Air Fresheners', 'Air Purifiers', 'Amplifiers & Effects', 'Breastfeeding
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cluster 14:
['Basic Supplies', 'Bathroom', 'Bathroom Accessories', 'Bathroom Furniture Sets', 'Bathroom Sa
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cluster 15:
['Above Knee, Mini', 'Backpack', 'Backpack Style', 'Baggy, Loose', 'Boot Cut', 'Box', 'Button'
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cluster 16:
['Advertisement', 'Animation', 'Artwork', 'Biographies & Memoirs', 'Biography', 'Book', 'Books
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['Bedspreads & Coverlets', 'Dehumidifiers', 'Epilators', 'Humidifiers', 'Needlecraft', 'Paperm
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['Animal', 'Animals', 'Art Doll', 'Baby & Toddler Toys', 'Building Toys', 'Child Friendly', 'C

cluster 18: