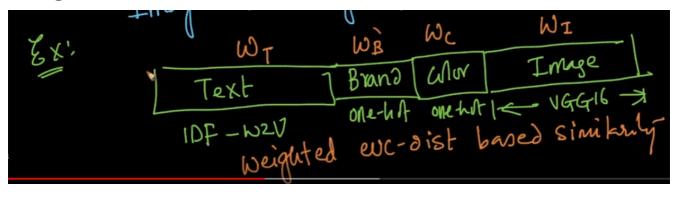
Assignment



In [1]:

```
#import all the necessary packages.
# For Image Processing
from PIL import Image
import requests
from io import BytesIO
# For Text Processing
from bs4 import BeautifulSoup
from nltk.corpus import stopwords
from nltk.tokenize import word tokenize
import nltk
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
from collections import Counter # To count each of the list
# Plotting purpose
import plotly
import plotly.figure_factory as ff
from plotly.graph_objs import Scatter, Layout
import matplotlib.pyplot as plt # For plotting purpose
import seaborn as sns # Plot purpose
from matplotlib import gridspec # To divide the region to plot
from sklearn.metrics.pairwise import cosine_similarity # Cal. cosine sim.
from sklearn.metrics import pairwise distances # Cal. distance b/w of each pair of datapts
import numpy as np # fast computation purpose
import pandas as pd # Print the data in tabular form
import warnings
import os # environment
import math # Apply some math concept
import time # To see execution time
import re # Regular Expression
from scipy.sparse import hstack # Merge all features
import pickle # Read file from pickle
import itertools
plotly.offline.init notebook mode(connected=True)
warnings.filterwarnings("ignore")
```

In [2]:

```
# Load Asin, CNN features of images
asins = np.load('16k_data_cnn_feature_asins.npy')
cnn_feature = np.load('16k_data_cnn_features.npy')
print('Shape of asins: {}\nShape of cnn_feature: {}'.format(asins.shape,cnn_feature.shape))

Shape of asins: (16042,)
Shape of cnn_feature: (16042, 25088)
```

In [3]:

```
# Load original dataset
data = pd.read_pickle('pickels/16k_apperal_data_preprocessed')
# Store asins values from original data
d_asins = list(data['asin'])
print('Shape of data: {}'.format(data.shape))
```

Shape of data: (16042, 7)

In [4]:

```
# Display data to see everything is working fine.
data.head()
```

Out[4]:

	asin	brand	color	medium_image_url	product_type_name	title	formatted_price
4	B004GSI2OS	FeatherLite	Onyx Black/ Stone	https://images-na.ssl-images- amazon.com/images	SHIRT	featherlite ladies long sleeve stain resistant	\$26.26
6	B012YX2ZPI	HX-Kingdom Fashion T- shirts	White	https://images-na.ssl-images- amazon.com/images	SHIRT	womens unique 100 cotton special olympics wor	\$9.99
15	B003BSRPB0	FeatherLite	White	https://images-na.ssl-images- amazon.com/images	SHIRT	featherlite ladies moisture free mesh sport sh	\$20.54
27	B014ICEJ1Q	FNC7C	Purple	https://images-na.ssl-images- amazon.com/images	SHIRT	supernatural chibis sam dean castiel neck tshi	\$7.39
46	B01NACPBG2	Fifth Degree	Black	https://images-na.ssl-images- amazon.com/images	SHIRT	fifth degree womens gold foil graphic tees jun	\$6.95

Dealing with Text

In [5]:

```
# Import w2v file
with open('word2vec_model', 'rb') as handle:
    model = pickle.load(handle)
```

In [6]:

```
# Featurize TFIDF...
# We are going to need idf values and their feature names only!!
title_tfidf = TfidfVectorizer()
title_idf_feature = title_tfidf.fit_transform(data['title'])
```

In [7]:

```
# Create dict store key as feature names and value as idf_score
title_idf = {}
for index, fname in enumerate(title_tfidf.get_feature_names()):
    title_idf[fname] = title_tfidf.idf_[index]
```

In [8]:

```
# https://stackoverflow.com/questions/29216889/slicing-a-dictionary
dict(itertools.islice(title_idf.items(), 5))
```

```
{'00': 9.584415607682967,
 '000': 9.98988071579113,
 '00008066': 9.98988071579113,
 '0000844': 9.98988071579113,
 '000085200': 9.98988071579113}
In [9]:
np.array([1,2,3,4])/2
Out[9]:
array([0.5, 1. , 1.5, 2. ])
In [10]:
def idf_w2v(model, sentence):
    vec word = np.zeros(300,)
   vec_idf = 0
    # Iterate each word
    for i in sentence.split():
        if i in model.keys() and i in title_idf.keys():
            vec word += title idf[i]*model[i]
            vec_idf += title_idf[i]
    return vec word/vec idf
In [11]:
title w2v = np.zeros((data['title'].shape[0],300))
for i in range(data.shape[0]):
    title w2v[i] = idf w2v(model,data['title'].iloc[i])
```

Dealing with Brand and Color

In [12]:

Out[8]:

```
# One hot encoding using countervizer

# Need to replace Null with string "NULL"
data['brand'].fillna(value="Not given", inplace=True )

# replace spaces with hypen
brands = [x.replace(" ", "-") for x in data['brand'].values]
colors = [x.replace(" ", "-") for x in data['color'].values]

# Brand
brand_vec = CountVectorizer(binary=True)
brand_onehot = brand_vec.fit_transform(brands)
brand_feat = brand_vec.get_feature_names()

# Color
color_vec = CountVectorizer(binary=True)
color_onehot = color_vec.fit_transform(colors)
color_feat = color_vec.get_feature_names()
```

Dealing with Images

```
In [13]:
```

```
# Already pretrained cnn_feature[:5]
```

Λ--<u>-</u>-Γ101.

```
Out[13]:
array([[0.11759627, 0.
                        , 0.
                                 , ..., 0.
                                                , 0.67754525,
      0.
              ],
      [0.06240697, 0.
                        , 0.
                                  , ..., 0.
                                                , 0.6599147 ,
      0.
             ],
                        , 0.
      [0.39736205, 0.
                                 , ..., 0.
                                                , 0.6154604 ,
      0.
             ],
      [0.17635377, 0.
                                 , ..., 0.
                                               , 0.53826314,
                        , 0.
          ],
      0.
      [0.4577136 , 0.
                                 , ..., 0. , 0.5774003 ,
                        , 0.
           ]], dtype=float32)
Recomd
In [14]:
```

```
#Display an image
def display_img(url,ax,fig):

    # we get the url of the apparel and download it
    response = requests.get(url)
    img = Image.open(BytesIO(response.content))

# we will display it in notebook
    plt.imshow(img)
```

In [15]:

```
def get distance(vec1, vec2):
   # vec1 = np.array(#number of words title1 * 300), each row is a vector of length 300 corresponds to
each word in give title
   # vec2 = np.array(#number of words title2 * 300), each row is a vector of length 300 corresponds to
each word in give title
   final dist = []
    # for each vector in vec1 we caluclate the distance (euclidean) to all vectors in vec2
   for i in vec1:
       dist = []
       for j in vec2:
           # np.linalq.norm(i-j) will result the euclidean distance between vectors i, j
           dist.append(np.linalg.norm(i-j))
       final dist.append(np.array(dist))
    # final dist = np.array(#number of words in title1 * #number of words in title2)
    # final dist[i,j] = euclidean distance between vectors i, j
   return np.array(final dist)
```

In [16]:

```
title_idf_feature = title_idf_feature.astype(np.float)
vocab = model.keys()
def get_word_vec(sentence, doc_id, m_name):
   # sentence : title of the apparel
    # doc id: document id in our corpus
    # m name: model information it will take two values
       # if m name == 'avg', we will append the model[i], w2v representation of word i
       # if m name == 'weighted', we will multiply each w2v[word] with the idf(word)
   vec = []
   for i in sentence.split():
       if i in vocab:
           if m name == 'weighted' and i in title tfidf.vocabulary :
               vec.append(title idf feature[doc id, title tfidf.vocabulary [i]] * model[i])
           elif m name == 'avg':
               vec.append(model[i])
       else:
            # if the word in our courpus is not there in the google word2vec corpus, we are just ignori
ng it
           vec.append(np.zeros(shape=(300,)))
    \# we will return a numpy array of shape (\#number of words in title * 300 ) 300 = len(w2v model[word
```

each row represents the word2vec representation of each word (weighted/avg) in given sentance return np.array(vec)

In [17]:

```
def heat map w2v brand (sentance1, sentance2, url, doc id1, doc id2, df id1, df id2, model):
    # sentance1 : title1, input appare1
    # sentance2 : title2, recommended apparel
    # url: apparel image url
    # doc idl: document id of input apparel
   # doc id2: document id of recommended apparel
   # df idl: index of document1 in the data frame
   # df id2: index of document2 in the data frame
    # model: it can have two values, 1. avg 2. weighted
   #s1 vec = np.array(#number of words title1 * 300), each row is a vector(weighted/avg) of length 300
corresponds to each word in give title
   s1 vec = get word vec(sentance1, doc id1, model)
   #s2 vec = np.array(#number of words title2 * 300), each row is a vector(weighted/avg) of length 300
corresponds to each word in give title
   s2_vec = get_word_vec(sentance2, doc_id2, model)
    # s1_s2_dist = np.array(#number of words in title1 * #number of words in title2)
    # s1 s2 dist[i,j] = euclidean distance between words i, j
   s1 s2 dist = get distance(s1 vec, s2 vec)
   data matrix = [['Asin', 'Brand', 'Color', 'title'],
               [data['asin'].loc[df id1], brands[doc id1], colors[doc id1], sentance1], # input appare1'
s features
               [data['asin'].loc[df id2], brands[doc id2], colors[doc id2], sentance2]] # recommonded ap
parel's features
   colorscale = [[0, '#1d004d'],[.5, '#f2e5ff'],[1, '#f2e5d1']] # to color the headings of each column
    # we create a table with the data matrix
   table = ff.create table(data matrix, index=True, colorscale=colorscale)
    # plot it with plotly
   plotly.offline.iplot(table, filename='simple table')
    # devide whole figure space into 25 * 1:10 grids
   gs = gridspec.GridSpec(25, 15)
    fig = plt.figure(figsize=(25,5))
    # in first 25*10 grids we plot heatmap
   ax1 = plt.subplot(gs[:, :-5])
    # ploting the heap map based on the pairwise distances
   ax1 = sns.heatmap(np.round(s1 s2 dist,6), annot=True)
    # set the x axis labels as recommended apparels title
   ax1.set xticklabels(sentance2.split())
    # set the y axis labels as input apparels title
   ax1.set_yticklabels(sentance1.split())
    # set title as recommended apparels title
   ax1.set title(sentance2)
    # in last 25 * 10:15 grids we display image
   ax2 = plt.subplot(gs[:, 10:16])
    # we dont display grid lins and axis labels to images
   ax2.grid(False)
   ax2.set xticks([])
   ax2.set yticks([])
    # pass the url it display it
   display_img(url, ax2, fig)
   plt.show()
```

In [18]:

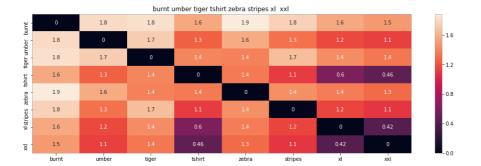
```
def idf_w2v_brand_color(index, wt,wb,wc,wi,n_feat):
    # pairwise distance whereas this X parameter act as corpus idf titles and Y parameter act as one id
f title
    title dist = pairwise distances(title w2v_title w2v_t
```

```
citte_aist = pairwise_aistances(titte_wzv,titte_wzv[inaex].tesnape(t,-i))
# pairwise distance whereas this X parameter act as corpus brand and Y parameter act as one brand
brand dist = pairwise distances (brand onehot, brand onehot[index].reshape(1,-1))
# pairwise distance whereas this X parameter act as corpus color and Y parameter act as one color
color dist = pairwise distances(color onehot, color onehot[index].reshape(1,-1))
# pairwise distance whereas this X parameter act as corpus images and Y parameter act as one image
imq dist = pairwise distances (cnn feature, cnn feature[index].reshape (1,-1))
pw dist = (wt * title dist + wb * brand dist + wc * color dist + wi * img dist)/float(wt+wb+wc+wi)
# Sort in ascending because we are using euc. dist.
ind sort = np.argsort(pw dist.flatten())[0:n feat]
# Based on index sort, get their distance score
pw dist = pw dist[ind sort].flatten()
# Also store title with index sort
df_indices = list(data.index[ind_sort])
# Based on we got ind_sort value, we want to plot the recomment using heatmap
for i in range(len(ind sort)):
    heat map w2v brand(data['title'].loc[df indices[0]], \
                       data['title'].loc[df_indices[i]], \
                       data['medium image url'].loc[df indices[i]], \
                       ind sort[0], \
                       ind sort[i], \
                       df indices[0],
                       df indices[i], \
                       'weighted')
    print('ASIN :',data['asin'].loc[df_indices[i]])
    print('Brand :',data['brand'].loc[df_indices[i]])
    print('euclidean distance from input :', pw_dist[i])
    print('='*125)
```

When I Put Title weight highest

```
In [20]:
```

```
idf_w2v_brand_color(12566,20,1,1,1,5)
```





ASIN: B00JXQB5FQ Brand: Si Row

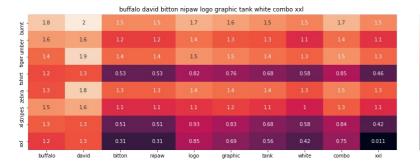
euclidean distance from input : 3.524037113137361e-07

pink tiger tshirt zebra stripes xl xxl burnt 1.9 xistripes zebra tshirt tiger umber 1.9 1.7 2 0.15 1.7 1.8 1.7 1.8 tshirt pink zebra stripes tiger



ASIN: B00JXQASS6 Brand: Si Row

euclidean distance from input : 2.847047198575608

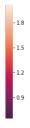




ASIN : B018H5AZXQ Brand : Buffalo

euclidean distance from input : 3.009161385528684

			womens cro	chet trim shirts	olive tree large	xhilaration		
burnt		2	1.7	1.7	1.8	2		1.5
umber b	1.2					1.8		1.2
tiger um		1.9	1.6		1.8	1.8		1.4
tshirt ti	0.54		1.2	0.8		1.6	0.76	0.53
zebra ts		1.7			1.7	1.9		1.3
stripes ze	11			11		1.9	1.2	11
£	0.53			0.00			0.50	0.53









ASIN: B06XBHNM7J Brand: Xhilaration

euclidean distance from input : 3.0851707128696004





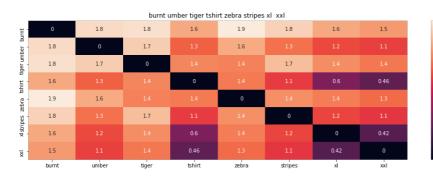
ASIN: B074LTBWSW Brand: Liz Lange

euclidean distance from input: 3.128606119620645

When I Put brand weight highest

In [21]:

idf_w2v_brand_color(12566,1,20,1,1,5)





ASIN : B00JXQB5FQ Brand : Si Row

euclidean distance from input : 3.277844015152052e-07

pink tiger tshirt zebra stripes xl xxl burnt 1.9 zebra tshirt tiger umber 1.9 0.4 0.33 1.7 xistripes 1.7 1.8 pink zebra tiger tshirt stripes



- 2.0

- 1.2

- 0.8

ASIN: B00JXQASS6 Brand: Si Row

euclidean distance from input : 2.204431651909357





ASIN : B00JXQCUIC Brand : Si Row

euclidean distance from input : 2.3873902274282846

			yellow tiger tank	op tiger stripes		
purut .		3.1			3.1	
umber b	1.2	3.1	1.3	1.2	3.1	
tiger um			1.4	1.4		
shirt ti	0.91	2.9	0.72	0.58	2.9	12
ъ						

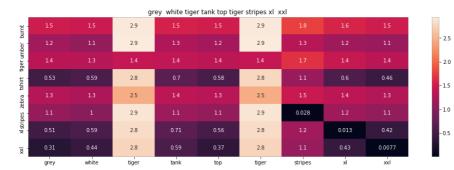






ASIN: BOOJXQAUWA Brand: Si Row

euclidean distance from input : 2.443557690368855





ASIN : B00JXQAFZ2 Brand : Si Row

euclidean distance from input : 2.4480314034792743

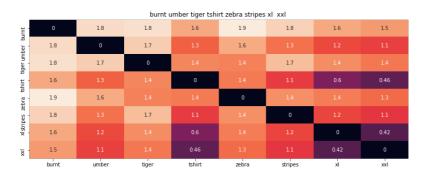
- 1.6

- 1.2

When I Put color weight highest

In [22]:

idf_w2v_brand_color(12566,1,1,20,1,5)





ASIN: B00JXQB5FQ

- ' - -

Brand : Si Row

euclidean distance from input : 3.277844015152052e-07

hip latter crochet back womens small hilow blouse brown 1.9 1.8 1.9 burnt umber 1.8 1.6 tiger shirt 1.7 1.6 1.7 zebra xlstripes × .



ASIN : B074MJN1K9

Brand : Hip

euclidean distance from input : 2.1189781032405866

- 1.2

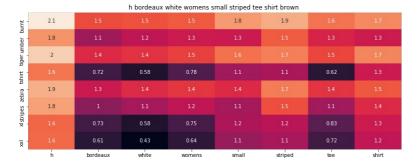
- 0.9

- 2.1

- 1.8

- 1.5

- 1.2





ASIN: B072BVB47Z Brand: H By Bordeaux

euclidean distance from input : 2.1365242256396018

			leisure	vest moda	al tank top	loose cond	dole belt la	arge size b	acking shi	rtkhakı		
ŧ.			1.9				2.1	1.8				
ber bu							2.1					
Ē	1.6	1.5	1.7	1.4	1.4	1.4	2.2	1.6	1.4	1.4	1.7	1.4



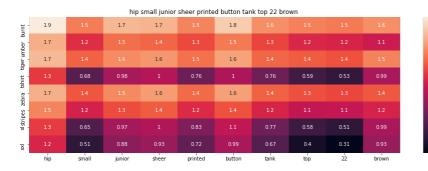




ASIN : B014OUHUZY

Brand: Black Temptation

euclidean distance from input: 2.224791790770683





ASIN: B071LDTQ1F Brand: Hip

euclidean distance from input: 2.3278722357995307

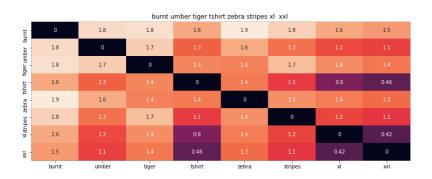
- 1.5

- 0.8

When I Put image weight highest

In [25]:

idf_w2v_brand_color(12566,1,1,1,20,5)





ASIN : B00JXQB5FQ Brand : Si Row

euclidean distance from input : 6.531068483246085e-06

womens crochet trim shirts olive tree large xhilaration 1.7 1.8 2 burnt tiger umber 1.6 1.9 1.8 1.8 1.6 1.7 1.7 1.9 zebra 1.9 1.7 crochet shirts large olive tree xhilaration



ASIN: B06XBHNM7J Brand: Xhilaration

euclidean distance from input: 32.55369071482225

- 1.5

- 0.9

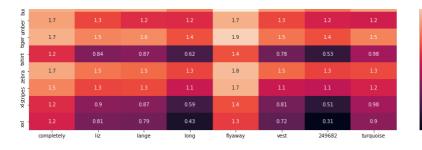
			breast cance	r awareness jur	iors vneck shir	t fight cancer		
print		2.6						2.6
		2.5			1.2			2.5
tiger umber		2.5						2.5
shirt 6		2.3		0.71	0.53	0.44		2.3
zebra ts		2.6			1.3			2.6
		2.6		12	11			2.6
xlstripes		2.3		0.68	0.51	0.59		2.3
<u> x</u>		2.3		0.56	0.31	0.42	11	2.3
_	breast	cancer	awareness	juniors	vneck	shirt	fight	cancer



ASIN: B016CU40IY Brand: Juiceclouds

euclidean distance from input : 33.54351664926501

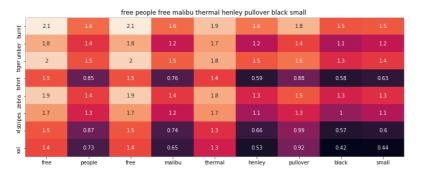
19





ASIN: B074LTBWSW Brand: Liz Lange

euclidean distance from input: 34.44161415781852







ASIN: B074MXY984
Brand: We The Free

euclidean distance from input: 34.4553628863689

Observation:

- 1. Giving the more weightage in color or title didn't given the good recommended
- 2. Apart from other weightage (color or brand), it gives better resultant output

Conclusion

- 1. Load the preprocessed data and cnn images feature data
- 2. Perform IDF W2V on title, OneHotEncoding on brand and color and already pretrained images features using cnn
- 3. We define a display_img function to show the image
- 4. We define a get_distance function to get the distance between two vectors to inorder to display output
- 5. We define a **get_word_vec** function to get the evaluation of idf w2v inorder to display score in output
- 6. We define a **heat_map_w2v_brand** function to get overall figure configuration and distance calculated between 2 vectors for every index sort values
- 7. we define idf_w2v_brand_color function to to evaluate weighted euc. dist similarity of 4 features
- 8. Got the display output