Amazon Clothing Recommendation Engine ¶

Import necessary packages.

```
In [249]:
```

```
from PIL import Image
import requests
from io import BytesIO
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import warnings
from bs4 import BeautifulSoup
from nltk.corpus import stopwords
from nltk.tokenize import word tokenize
import math
import time
import re
import os
import seaborn as sns
from collections import Counter
from sklearn.feature extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.metrics.pairwise import cosine similarity
from sklearn.metrics import pairwise distances
from matplotlib import gridspec
from scipy.sparse import hstack
import plotly
import plotly.figure factory as ff
from plotly.graph_objs import Scatter, Layout
plotly.offline.init notebook mode(connected=True)
warnings.filterwarnings("ignore")
```

```
In [250]:
```

```
from pprint import pprint
```

Load Original Data

```
In [251]:
```

```
data = pd.read_json('/ilab/users/ak1648/akshat/women_clothing_amazon.json')
```

We will be using the following features

```
1. asin ( Amazon standard identification number)
2. brand ( brand to which the product belongs to )
3. color ( Color information of apparel, it can contain many colors as a v alue ex: red and black stripes )
4. product_type_name (type of the apperal, ex: SHIRT/TSHIRT )
5. medium_image_url ( url of the image )
6. title (title of the product.)
7. formatted price (price of the product)
```

In [252]:

Number of data points: 183138 Number of features: 7

Out[252]:

	asin	brand	color	medium_image_url	product_type_name	title	form
0	B016l2TS4W	FNC7C	None	https://images-na.ssl- images- amazon.com/images	SHIRT	Minions Como Superheroes Ironman Long Sleeve R	
1	B01N49Al08	FIG Clothing	None	https://images-na.ssl- images- amazon.com/images	SHIRT	FIG Clothing Womens Izo Tunic	
2	B01JDPCOHO	FIG Clothing	None	https://images-na.ssl- images- amazon.com/images	SHIRT	FIG Clothing Womens Won Top	
3	B01N19U5H5	Focal18	None	https://images-na.ssl- images- amazon.com/images	SHIRT	Focal18 Sailor Collar Bubble Sleeve Blouse Shi	
4	B004GSI2OS	FeatherLite	Onyx Black/ Stone	https://images-na.ssl- images- amazon.com/images	SHIRT	Featherlite Ladies' Long Sleeve Stain Resistan	

In [253]:

```
data = pd.read_pickle('/ilab/users/ak1648/akshat/180k_apparel_data')
```

Data Preprocessing

Removing Null Values

In [254]:

```
data = data.loc[-data['formatted_price'].isnull()]
print('Number of products fter eliminating NULL price :', data.shape[0])
data =data.loc[-data['color'].isnull()]
print('Number of data points After eliminating NULL color :', data.shape[0])
```

Number of products fter eliminating NULL price: 28395 Number of data points After eliminating NULL color: 28385

Removing the products with short description

```
In [255]:
```

```
data_sorted = data[data['title'].apply(lambda x: len(x.split())>4)]
print("After removal of products with short description:", data_sorted.shape[0])
```

After removal of products with short description: 27949

Removing the products with duplicate title

```
In [256]:
```

```
data = pd.read pickle('/ilab/users/ak1648/akshat/28k apparel data')
print('The number of entries with duplicate title is %d'%sum(data.duplicated('title
# Sort the whole data based on title
data sorted.sort values('title',inplace=True, ascending=False)
indices = []
for i,row in data sorted.iterrows():
    indices.append(i)
import itertools
stage1 dedupe asins = []
i = 0
j = 0
num_data_points = data_sorted.shape[0]
while i < num data points and j < num data points:
    previous i = i
    a = data['title'].loc[indices[i]].split()
    j = i+1
    while j < num data points:</pre>
        b = data['title'].loc[indices[j]].split()
        length = max(len(a), len(b))
        count = 0
        for k in itertools.zip longest(a,b):
            if (k[0] == k[1]):
                count += 1
        # Ignore the products if the number of words in which both strings differ a
        if (length - count) > 2: # number of words in which both sensences differ
            stage1 dedupe asins.append(data sorted['asin'].loc[indices[i]])
            if j == num data points-1: stage1 dedupe asins.append(data sorted['asin
            i = j
            break
        else:
            j += 1
    if previous i == i:
        break
data = data.loc[data['asin'].isin(stage1 dedupe asins)]
print('Number of data points now is: ', data.shape[0])
```

The number of entries with duplicate title is 2325 Number of data points now is: 17593

Text Preprocessing

```
In [257]:
```

```
data = pd.read_pickle('/ilab/users/ak1648/akshat/16k_apperal_data')
```

In [258]:

```
import nltk
nltk.download('stopwords')

[nltk_data] Downloading package stopwords to
[nltk_data] /ilab/users/ak1648/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

Out[258]:

True

In [259]:

list of stop words: {'hers', 'but', 'not', 'above', "wasn't", 'or', 'o', 'most', 'yours', 'he', 'some', 'again', 'be', "doesn't", 'on', "y ou're", 'below', 'why', 't', 'weren', 'because', 'being', 'any', 'fo r', 'ain', 'what', 'shan', "mustn't", "should've", 'mightn', "aren't", 'of', 'very', 'than', 'ma', "wouldn't", 'over', 'it', 'up', 'did', "ha ven't", 'have', 'didn', 'haven', 'against', 'had', 'wasn', 'his', "wo n't", 'its', 'yourself', 'before', 'will', 'then', 'they', 'so', 've', 'whom', 'from', "mightn't", "you'll", 'm', 'couldn', 'the', 'only', 'b een', 'between', 'itself', 'this', 'ours', 'was', "it's", 'more', 'wh o', 'is', 'there', 'few', 'too', 'until', 'theirs', 'y', 'should', 'al l', 'through', 'both', 's', 'after', 'were', 'and', 'with', 'doesn', 'in', 'themselves', 'once', 'are', 'other', 'if', "that'll", 'him', 'w hen', 'nor', 'no', "don't", 'can', 'needn', "weren't", 'i', "she's", 'me', 'those', 'hadn', 'off', 'how', 'under', 'you', 'at', 'into', "co uldn't", "you'd", 'your', "didn't", 'has', 'having', 'her', 'myself', 'won', 'these', 'hasn', 'my', 'we', 'here', 'd', 'about', "needn't", 'does', "shan't", 'our', 'ourselves', 'now', 'do', 'mustn', 'don', 'wh ile', 'a', 'that', 'such', "shouldn't", 'by', "hadn't", 'doing', 'as', 'own', 'himself', 'she', 'where', 'their', 'just', 'isn', 'to', 're', 'which', 'each', 'aren', 'further', 'll', 'wouldn', 'same', 'them', 'down', 'out', "isn't", 'an', 'yourselves', 'herself', 'during', 'shouldn', "you've", "hasn't", 'an', 'yourselves', 'herself', 'during', 'shouldn', "you've", "hasn't", 'an', 'yourselves', 'herself', 'during', 'shouldn', "you've", "hasn't", 'am'}

In [260]:

```
start_time = time.clock()
# we take each title and we text-preprocess it.
for index, row in data.iterrows():
    nlp_preprocessing(row['title'], index, 'title')
# we print the time it took to preprocess whole titles
print(time.clock() - start_time, "seconds")
data.head()
```

9.90000000000034 seconds

Out[260]:

	asin	brand	color	medium_image_url	product_type_name	title	for
4	B004GSl2OS	FeatherLite	Onyx Black/ Stone	https://images-na.ssl- images- amazon.com/images	SHIRT	featherlite ladies long sleeve stain resistant	
6	B012YX2ZPI	HX- Kingdom Fashion T- shirts	White	https://images-na.ssl- images- amazon.com/images	SHIRT	womens unique 100 cotton special olympics wor	
15	B003BSRPB0	FeatherLite	White	https://images-na.ssl- images- amazon.com/images	SHIRT	featherlite ladies moisture free mesh sport sh	
27	B014ICEJ1Q	FNC7C	Purple	https://images-na.ssl- images- amazon.com/images	SHIRT	supernatural chibis sam dean castiel neck tshi	
46	B01NACPBG2	Fifth Degree	Black	https://images-na.ssl- images- amazon.com/images	SHIRT	fifth degree womens gold foil graphic tees jun	

Utility Functions

In [261]:

```
#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)
#plotting code to understand the decision
def plot image(keys, values, labels, url, text):
        gs = gridspec.GridSpec(2, 2, width_ratios=[4,1], height_ratios=[4,1])
        fig = plt.figure(figsize=(25,3))
        # plotting image of the the apparel
        ax = plt.subplot(gs[0])
        ax.grid(False)
        ax.set xticks([])
        ax.set yticks([])
        display img(url, ax, fig)
        plt.show()
def plot condition(doc id, vec1, vec2, url, text, model):
    # doc id : index of the title1
    # vec1 : input apparels's vector, it is of a dict type {word:count}
    # vec2 : recommended apparels's vector, it is of a dict type {word:count}
    # url : apparels image url
    # text: title of recomonded apparel (used to keep title of image)
    # model, Bag of Words or TF-IDF
    intersection = set(vec1.keys()) & set(vec2.keys())
    for i in vec2:
        if i not in intersection:
            vec2[i]=0
    keys = list(vec2.keys())
    values = [vec2[x] for x in vec2.keys()]
        # if model == 'bag of words': labels(i) = values(i)
        # if model == 'tfidf weighted bag of words':labels(i) = tfidf(keys(i))
    if model == 'bag of words':
        labels = values
    elif model == 'tfidf':
        labels = []
        for x in vec2.keys():
            if x in tfidf title vectorizer.vocabulary :
                labels.append(tfidf title features[doc id, tfidf title vectorizer.ve
            else:
                labels.append(0)
    plot image(keys, values, labels, url, text)
# this function gets a list of wrods along with the frequency of each word given "te
def text to vector(text):
    word = re.compile(r'\w+')
    words = word.findall(text)
    return Counter(words)
def get result(doc id, content a, content b, url, model):
    text1 = content_a
    text2 = content b
    vector1 = text to vector(text1)
```

```
vector2 = text_to_vector(text2)
plot_condition(doc_id, vector1, vector2, url, text2, model)
```

Textual Recommendation based on Product Titles

Bag of Words model on Product Titles

```
In [262]:
```

```
from sklearn.feature extraction.text import CountVectorizer
title_vectorizer = CountVectorizer()
title features
               = title vectorizer.fit transform(data['title'])
title features.get shape() # get number of rows and columns in feature matrix.
bag of words euclidean=[]
def bag of words model(doc id, num results):
    # doc id: apparel's id in given corpus
    # pairwise dist will store the distance from given input apparel to all remaining
    # cosine distance for similarity
    pairwise dist = pairwise distances(title features, title features[doc id])
    # np.argsort will return indices of the smallest distances
    indices = np.argsort(pairwise dist.flatten())[0:num results]
    # pdists will store the smallest distances
    pdists = np.sort(pairwise dist.flatten())[0:num results]
    df indices = list(data.index[indices])
    for i in range(0,len(indices)):
        # Pass 1. doc id, 2. title1, 3. title2, url, model
        get result(indices[i],data['title'].loc[df indices[0]], data['title'].loc[df
       print ('Title:', data['title'].loc[df indices[i]])
       bag of words euclidean.append(pdists[i])
        print ('Euclidean similarity with the input image :', pdists[i])
    print('Average euclidean distance is ',sum(bag of words euclidean)/num results)
```

Top 10 Recommendation using Bag of Words Model

In [263]:

print('First one is input image and the rest 10 are recommended clothing')
bag_of_words_model(12562, 10)

First one is input image and the rest 10 are recommended clothing



Title: im huckleberry doc holliday 34sleeve raglan long sleeve Euclidean similarity with the input image: 0.0



Title: incredibles 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.23606797749979



Title: moo im goat 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.23606797749979



Title: pizza 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.23606797749979



Title: maxpower 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.23606797749979





Title: get nothing 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.449489742783178



Title: love husband 34sleeve raglan long sleeve
Euclidean similarity with the input image: 2.449489742783178





Title: metallica justice 34sleeve raglan long sleeve Euclidean similarity with the input image : 2.449489742783178

Average euclidean distance is 2.119172062391505

TF-IDF Model on Product Titles

```
In [264]:

ctorizer = TfidfVectorizer(min_df = 0)
atures = tfidf_title_vectorizer.fit_transform(data['title'])
an=[]
l(doc_id, num_results):
ist = pairwise_distances(tfidf_title_features,tfidf_title_features[doc_id])

np.argsort(pairwise_dist.flatten())[0:num_results]
np.sort(pairwise_dist.flatten())[0:num_results]
= list(data.index[indices])

ange(0,len(indices)):
sult(indices[i], data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]],
('Title:', data['title'].loc[df_indices[i]])
euclidean.append(pdists[i])
('Eucliden distance from the given image :', pdists[i])
rage euclidean distance is',sum(tf_idf_euclidean)/num_results)
```

Top 10 Recommendations using TF-IDF Model

In [265]:

print('First one is input image and the rest 10 are recommended clothing')
tfidf_model(12562, 10)

First one is input image and the rest 10 are recommended clothing



Title: im huckleberry doc holliday 34sleeve raglan long sleeve Eucliden distance from the given image: 0.0



Title: im weird im unicorn 34sleeve raglan long sleeve Eucliden distance from the given image: 1.0649487461630645



Title: moo im goat 34sleeve raglan long sleeve Eucliden distance from the given image: 1.133575440305773



Title: pizza 34sleeve raglan long sleeve Eucliden distance from the given image : 1.1627003249587693



Title: incredibles 34sleeve raglan long sleeve Eucliden distance from the given image: 1.1872291620915076



Title: maxpower 34sleeve raglan long sleeve Eucliden distance from the given image: 1.1872291620915076



Title: love husband 34sleeve raglan long sleeve Eucliden distance from the given image: 1.1960847619438466



Title: revolution logo 34sleeve raglan long sleeve Eucliden distance from the given image: 1.2023792573811647



Title: get nothing 34sleeve raglan long sleeve Eucliden distance from the given image: 1.2117794081962616



Title: flower life galaxy 34sleeve raglan long sleeve Eucliden distance from the given image: 1.21205545761381 Average euclidean distance is 1.0557981720745704

Visual Recommendation based on Product Images

In []:

```
import numpy as np
from keras.preprocessing.image import ImageDataGenerator
from keras.models import Sequential
from keras.layers import Dropout, Flatten, Dense
from keras import applications
from sklearn.metrics import pairwise_distances
import matplotlib.pyplot as plt
import requests
from PIL import Image
import pandas as pd
import pickle
```

Feature Extraction from Product Images using VGG16 CNN

In []:

```
# each image is converted into 25088 length dense-vector
# dimensions of our images.
img width, img height = 224, 224
top model weights path = '/ilab/users/ak1648/akshat/images/bottleneck fc model.h5'
train data dir = '/ilab/users/ak1648/akshat/images/'
nb train samples = 16042
epochs = 50
batch size = 1
def save_bottlebeck_features():
    #Function to compute VGG-16 CNN for image feature extraction.
    asins = []
    datagen = ImageDataGenerator(rescale=1. / 255)
    # build the VGG16 network
    model = applications.VGG16(include top=False, weights='imagenet')
    generator = datagen.flow from directory(
        train data dir,
        target size=(img width, img height),
        batch size=batch size,
        class mode=None,
        shuffle=False)
    for i in generator.filenames:
        asins.append(i[2:-5])
    bottleneck features train = model.predict generator(generator, nb train samples
    bottleneck features train = bottleneck features train.reshape((16042,25088))
    np.save(open('/ilab/users/ak1648/akshat/16k data cnn features.npy', 'wb'), bottl
    np.save(open('/ilab/users/ak1648/akshat/16k data cnn feature asins.npy', 'wb'),
save_bottlebeck_features()
```

VGG16 CNN model on Product Images

In [267]:

```
#load the features and corresponding ASINS info.
bottleneck features train = np.load('/ilab/users/ak1648/akshat/16k data cnn features
asins = np.load('/ilab/users/ak1648/akshat/16k data cnn feature asins.npy')
asins = list(asins)
# load the original 16K dataset
data = pd.read pickle('/ilab/users/ak1648/akshat/16k apperal data preprocessed')
df asins = list(data['asin'])
from IPython.display import display, Image, SVG, Math, YouTubeVideo
cnn euclidean=[]
#get similar products using CNN features (VGG-16)
def get similar products cnn(doc id, num results):
    doc id = asins.index(df asins[doc id])
    pairwise_dist = pairwise_distances(bottleneck_features_train, bottleneck_feature)
    indices = np.argsort(pairwise dist.flatten())[0:num results]
   pdists = np.sort(pairwise dist.flatten())[0:num results]
    for i in range(len(indices)):
        rows = data[['medium image url','title']].loc[data['asin']==asins[indices[i
        for indx, row in rows.iterrows():
            display(Image(url=row['medium_image_url'], embed=True))
            print('Product Title: ', row['title'])
            cnn euclidean.append(pdists[i])
            print('Euclidean Distance from input image:', pdists[i])
    print('Average euclidean distance is ',sum(cnn euclidean)/num results)
```

Top 10 Recommendations using VGG16 CNN Model

In [268]:

print('First one is input image and the rest 10 are recommended clothing')
get_similar_products_cnn(12562, 10)

First one is input image and the rest 10 are recommended clothing



Product Title: im huckleberry doc holliday 34sleeve raglan long sleev e
Euclidean Distance from input image: 0.0625



Product Title: chubby unicorn need love 34sleeve raglan long sleeve Euclidean Distance from input image: 17.071169



Product Title: bibliophile literary like party 34sleeve raglan long s leeve

Euclidean Distance from input image: 18.06369



Product Title: rip harambe cartoon 17th birthday killed 34sleeve ragl an long sleeve

Euclidean Distance from input image: 18.658527



Product Title: jesus drank wine 34sleeve raglan long sleeve Euclidean Distance from input image: 18.854345



Product Title: kanye west pinterest 34sleeve raglan long sleeve Euclidean Distance from input image: 18.991518



Product Title: xjbd womens peaky drama series long sleeve blended bas eball shirt size xl
Euclidean Distance from input image: 19.783867



Product Title: engineer good math 34sleeve raglan long sleeve Euclidean Distance from input image: 20.048672



Product Title: tell time pm 34sleeve raglan long sleeve Euclidean Distance from input image: 20.094067



Product Title: choose violence 34sleeve raglan long sleeve Euclidean Distance from input image: 20.477503

Average euclidean distance is 17.21058578491211

Performance Comparison using Eucledian Distance

In [269]:

```
euclidean_distance=[]
num_results=10
euclidean_distance.append(sum(bag_of_words_euclidean)/num_results)
euclidean_distance.append(sum(tf_idf_euclidean)/num_results)
euclidean_distance.append(sum(cnn_euclidean)/num_results)
x=euclidean_distance
```

In [270]:

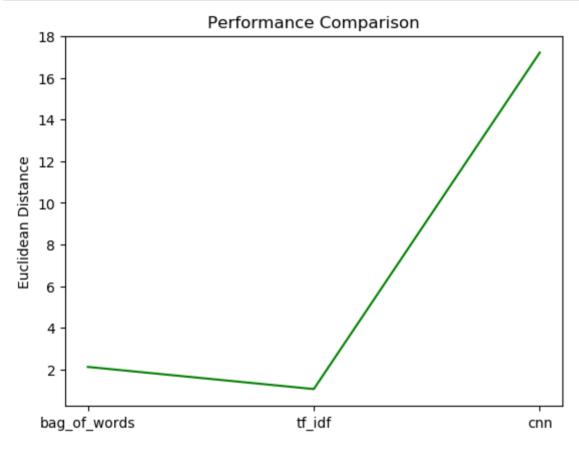
euclidean_distance

Out[270]:

[2.119172062391505, 1.0557981720745704, 17.21058578491211]

In [280]:

```
import matplotlib.pyplot as plt; plt.rcdefaults()
import numpy as np
import matplotlib.pyplot as plt
objects = ('bag_of_words', 'tf_idf','cnn')
y_pos = np.arange(len(objects))
plt.plot(y_pos,x,color='green')
plt.xticks(y_pos, objects)
plt.ylabel('Euclidean Distance')
plt.title('Performance Comparison')
plt.show()
```

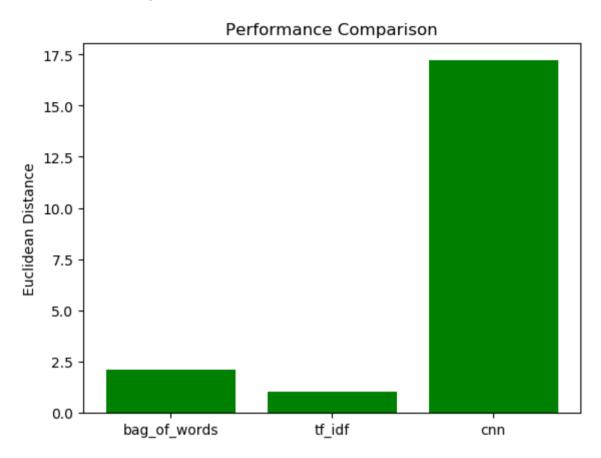


In [281]:

```
plt.ylabel('Euclidean Distance')
plt.title('Performance Comparison')
plt.bar(objects,x,color = 'green')
```

Out[281]:

<BarContainer object of 3 artists>



In []: