

Amazon Apparel Recommendations

[4.2] Data and Code:

<https://drive.google.com/open?id=0BwNkduBnePt2VWhCYXhMV3p4dTg>

[4.3] Overview of the data

In [1]:

```
#import all the necessary packages.

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 nltk
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 [0]:

```
# we have give a json file which consists of all information about
# the products
# loading the data using pandas' read_json file.
data = pd.read_json('tops_fashion.json')
```

In [0]:

```
print ('Number of data points : ', data.shape[0], \
      'Number of features/variables:', data.shape[1])
```

Number of data points : 183138 Number of features/variables: 19

Terminology:

What is a dataset?
Rows and columns
Data-point
Feature/variable

```
In [0]:
```

```
# each product/item has 19 features in the raw dataset.  
data.columns # prints column-names or feature-names.
```

```
Out[0]:
```

```
Index(['asin', 'author', 'availability', 'availability_type', 'brand', 'color',  
       'editorial_review', 'editorial_review', 'formatted_price',  
       'large_image_url', 'manufacturer', 'medium_image_url', 'model',  
       'product_type_name', 'publisher', 'reviews', 'sku', 'small_image_url',  
       'title'],  
      dtype='object')
```

Of these 19 features, we will be using only 6 features in this workshop.

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 value ex: red and black stripes)
4. product_type_name (type of the apparel, 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 [0]:
```

```
data = data[['asin', 'brand', 'color', 'medium_image_url', 'product_type_name', 'title', 'formatted_price']]
```

```
In [0]:
```

```
print ('Number of data points : ', data.shape[0], \  
      'Number of features:', data.shape[1])  
data.head() # prints the top rows in the table.
```

Number of data points : 183138 Number of features: 7

```
Out[0]:
```

	asin	brand	color	medium_image_url	product_type_name	title	formatted_price
0	B016I2TS4W	FNC7C	None	https://images-na.ssl-images-amazon.com/images...	SHIRT	Minions Como Superheroes Ironman Long Sleeve R...	None
1	B01N49AI08	FIG Clothing	None	https://images-na.ssl-images-amazon.com/images...	SHIRT	FIG Clothing Womens Izo Tunic	None
2	B01JDPCOHO	FIG Clothing	None	https://images-na.ssl-images-amazon.com/images...	SHIRT	FIG Clothing Womens Won Top	None
3	B01N19U5H5	Focal18	None	https://images-na.ssl-images-amazon.com/images...	SHIRT	Focal18 Sailor Collar Bubble Sleeve Blouse Shi...	None
4	B004GSI2OS	FeatherLite	Onyx Black/ Stone	https://images-na.ssl-images-amazon.com/images...	SHIRT	Featherlite Ladies' Long Sleeve Stain Resistan...	\$26.26

[5.1] Missing data for various features.

Basic stats for the feature: product_type_name

```
In [0]:
```

```
# We have total 72 unique type of product_type_names  
print(data['product_type_name'].describe())  
  
# 91.62% (167794/183138) of the products are shirts,
```

```
count      183138
unique       72
top        SHIRT
freq      167794
Name: product_type_name, dtype: object
```

In [0]:

```
# names of different product types
print(data['product_type_name'].unique())
```

```
['SHIRT' 'SWEATER' 'APPAREL' 'OUTDOOR_RECREATION_PRODUCT'
 'BOOKS_1973_AND_LATER' 'PANTS' 'HAT' 'SPORTING_GOODS' 'DRESS' 'UNDERWEAR'
 'SKIRT' 'OUTERWEAR' 'BRA' 'ACCESSORY' 'ART_SUPPLIES' 'SLEEPWEAR'
 'ORCA_SHIRT' 'HANDBAG' 'PET_SUPPLIES' 'SHOES' 'KITCHEN' 'ADULT_COSTUME'
 'HOME_BED_AND_BATH' 'MISC_OTHER' 'BLAZER' 'HEALTH_PERSONAL_CARE'
 'TOYS_AND_GAMES' 'SWIMWEAR' 'CONSUMER_ELECTRONICS' 'SHORTS' 'HOME'
 'AUTO_PART' 'OFFICE_PRODUCTS' 'ETHNIC_WEAR' 'BEAUTY'
 'INSTRUMENT_PARTS_AND_ACCESSORIES' 'POWERSPORTS_PROTECTIVE_GEAR' 'SHIRTS'
 'ABIS_APPAREL' 'AUTO_ACCESSORY' 'NONAPPARELMISC' 'TOOLS' 'BABY_PRODUCT'
 'SOCKSHOSIERY' 'POWERSPORTS RIDING SHIRT' 'EYEWEAR' 'SUIT'
 'OUTDOOR_LIVING' 'POWERSPORTS RIDING JACKET' 'HARDWARE' 'SAFETY_SUPPLY'
 'ABIS_DVD' 'VIDEO_DVD' 'GOLF_CLUB' 'MUSIC_POPULAR_VINYL'
 'HOME_FURNITURE_AND_DECOR' 'TABLET_COMPUTER' 'GUILD_ACCESSORIES'
 'ABIS_SPORTS' 'ART_AND_CRAFT_SUPPLY' 'BAG' 'MECHANICAL_COMPONENTS'
 'SOUND_AND_RECORDING_EQUIPMENT' 'COMPUTER_COMPONENT' 'JEWELRY'
 'BUILDING_MATERIAL' 'LUGGAGE' 'BABY_COSTUME' 'POWERSPORTS_VEHICLE_PART'
 'PROFESSIONAL_HEALTHCARE' 'SEEDS_AND_PLANTS' 'WIRELESS_ACCESSORY']
```

In [0]:

```
# find the 10 most frequent product_type_names.
product_type_count = Counter(list(data['product_type_name']))
product_type_count.most_common(10)
```

Out[0]:

```
[('SHIRT', 167794),
 ('APPAREL', 3549),
 ('BOOKS_1973_AND_LATER', 3336),
 ('DRESS', 1584),
 ('SPORTING_GOODS', 1281),
 ('SWEATER', 837),
 ('OUTERWEAR', 796),
 ('OUTDOOR_RECREATION_PRODUCT', 729),
 ('ACCESSORY', 636),
 ('UNDERWEAR', 425)]
```

Basic stats for the feature: brand

In [0]:

```
# there are 10577 unique brands
print(data['brand'].describe())

# 183138 - 182987 = 151 missing values.
```

```
count      182987
unique      10577
top        Zago
freq       223
Name: brand, dtype: object
```

In [0]:

```
brand_count = Counter(list(data['brand']))
brand_count.most_common(10)
```

Out[0]:

Out[0]:

```
[('Zago', 223),  
 ('XQS', 222),  
 ('Yayun', 215),  
 ('YUNY', 198),  
 ('XiaoTianXin-women clothes', 193),  
 ('Generic', 192),  
 ('Boohoo', 190),  
 ('Alion', 188),  
 ('Abetteric', 187),  
 ('TheMogan', 187)]
```

Basic stats for the feature: color

In [0]:

```
print(data['color'].describe())  
  
# we have 7380 unique colors  
# 7.2% of products are black in color  
# 64956 of 183138 products have brand information. That's approx 35.4%.
```

```
count      64956  
unique     7380  
top        Black  
freq       13207  
Name: color, dtype: object
```

In [0]:

```
color_count = Counter(list(data['color']))  
color_count.most_common(10)
```

Out[0]:

```
[(None, 118182),  
 ('Black', 13207),  
 ('White', 8616),  
 ('Blue', 3570),  
 ('Red', 2289),  
 ('Pink', 1842),  
 ('Grey', 1499),  
 ('*', 1388),  
 ('Green', 1258),  
 ('Multi', 1203)]
```

Basic stats for the feature: formatted_price

In [0]:

```
print(data['formatted_price'].describe())  
  
# Only 28,395 (15.5% of whole data) products with price information
```

```
count      28395  
unique     3135  
top        $19.99  
freq       945  
Name: formatted_price, dtype: object
```

In [0]:

```
price_count = Counter(list(data['formatted_price']))  
price_count.most_common(10)
```

Out[0]:

```
[None, 154743),  
('$19.99', 945),  
('$9.99', 749),  
('$9.50', 601),  
('$14.99', 472),  
('$7.50', 463),  
('$24.99', 414),  
('$29.99', 370),  
('$8.99', 343),  
('$9.01', 336)]
```

Basic stats for the feature: title

In [0]:

```
print(data['title'].describe())  
  
# All of the products have a title.  
# Titles are fairly descriptive of what the product is.  
# We use titles extensively in this workshop  
# as they are short and informative.
```

```
count                183138  
unique              175985  
top      Nakoda Cotton Self Print Straight Kurti For Women  
freq                  77  
Name: title, dtype: object
```

In [0]:

```
data.to_pickle('pickels/180k_apparel_data')
```

We save data files at every major step in our processing in "pickle" files. If you are stuck anywhere (or) if some code takes too long to run on your laptop, you may use the pickle files we give you to speed things up.

In [0]:

```
# consider products which have price information  
# data['formatted_price'].isnull() => gives the information  
# about the dataframe row's which have null values price == None|Null  
data = data.loc[~data['formatted_price'].isnull()]  
print('Number of data points After eliminating price=NULL :', data.shape[0])
```

```
Number of data points After eliminating price=NULL : 28395
```

In [0]:

```
# consider products which have color information  
# data['color'].isnull() => gives the information about the dataframe row's which have null values  
# price == None|Null  
data = data.loc[~data['color'].isnull()]  
print('Number of data points After eliminating color=NULL :', data.shape[0])
```

```
Number of data points After eliminating color=NULL : 28385
```

We brought down the number of data points from 183K to 28K.

We are processing only 28K points so that most of the workshop participants can run this code on their laptops in a reasonable amount of time.

For those of you who have powerful computers and some time to spare, you are recommended to use all of the 183K images.

In [0]:

```
data.to_pickle('pickels/28k_apparel_data')
```

In [0]:

```
# You can download all these 28k images using this code below.  
# You do NOT need to run this code and hence it is commented.  
  
'''  
from PIL import Image  
import requests  
from io import BytesIO  
  
for index, row in images.iterrows():  
    url = row['large_image_url']  
    response = requests.get(url)  
    img = Image.open(BytesIO(response.content))  
    img.save('images/28k_images/'+row['asin']+'.jpeg')  
  
'''
```

Out [0]:

```
"\nfrom PIL import Image\nimport requests\nfrom io import BytesIO\n\nfor index, row in  
images.iterrows():\n    url = row['large_image_url']\n    response = requests.get(url)\n    img = Image.open(BytesIO(response.content))\n    img.save('workshop/images/28k_images/'+row['asin']+'.jpeg')\n\n"
```

[5.2] Remove near duplicate items

[5.2.1] Understand about duplicates.

In [0]:

```
# read data from pickle file from previous stage  
data = pd.read_pickle('pickels/28k_apparel_data')  
  
# find number of products that have duplicate titles.  
print(sum(data.duplicated('title')))  
# we have 2325 products which have same title but different color
```

2325

These shirts are exactly same except in size (S, M,L,XL)

:B00AQ4GMCK	:B00AQ4GMTS
:B00AQ4GMLQ	:B00AQ4GN3I

These shirts exactly same except in color

:B00G278GZ6	:B00G278W6O
:B00G278Z2A	:B00G2786X8

In our data there are many duplicate products like the above examples, we need to de-dupe them for better results.

[5.2.2] Remove duplicates : Part 1

In [0]:

```
# read data from pickle file from previous stage  
data = pd.read_pickle('pickels/28k_apparel_data')
```

In [0]:

```
data.head()
```

Out[0]:

	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 Resistan...	\$26.26
6	B012YX2ZPI	HX-Kingdom Fashion T-shirts	White	https://images-na.ssl-images-amazon.com/images...	SHIRT	Women's Unique 100% Cotton T - Special Olympic...	\$9.99
11	B001LOUGE4	Fitness Etc.	Black	https://images-na.ssl-images-amazon.com/images...	SHIRT	Ladies Cotton Tank 2x1 Ribbed Tank Top	\$11.99
15	B003BSRPB0	FeatherLite	White	https://images-na.ssl-images-amazon.com/images...	SHIRT	FeatherLite Ladies' Moisture Free Mesh Sport S...	\$20.54
21	B014ICEDNA	FNC7C	Purple	https://images-na.ssl-images-amazon.com/images...	SHIRT	Supernatural Chibis Sam Dean And Castiel Short...	\$7.50

In [0]:

```
# Remove All products with very few words in title
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

In [0]:

```
# Sort the whole data based on title (alphabetical order of title)
data_sorted.sort_values('title', inplace=True, ascending=False)
data_sorted.head()
```

Out[0]:

	asin	brand	color	medium_image_url	product_type_name	title	formatted_price
61973	B06Y1KZ2WB	Éclair	Black/Pink	https://images-na.ssl-images-amazon.com/images...	SHIRT	Éclair Women's Printed Thin Strap Blouse Black...	\$24.99
133820	B010RV33VE	xiaoming	Pink	https://images-na.ssl-images-amazon.com/images...	SHIRT	xiaoming Womens Sleeveless Loose Long T-shirts...	\$18.19
81461	B01DDSDLNS	xiaoming	White	https://images-na.ssl-images-amazon.com/images...	SHIRT	xiaoming Women's White Long Sleeve Single Brea...	\$21.58
75995	B00X5LYO9Y	xiaoming	Red Anchors	https://images-na.ssl-images-amazon.com/images...	SHIRT	xiaoming Stripes Tank Patch/Bear Sleeve Anchor...	\$15.91
151570	B00WPJG35K	xiaoming	White	https://images-na.ssl-images-amazon.com/images...	SHIRT	xiaoming Sleeve Sheer Loose Tassel Kimono Woma...	\$14.32

Some examples of duplicate titles that differ only in the last few words.

Titles 1:

16. woman's place is in the house and the senate shirts for Womens XXL White
17. woman's place is in the house and the senate shirts for Womens M Grey

Title 2:

25. tokidoki The Queen of Diamonds Women's Shirt X-Large
26. tokidoki The Queen of Diamonds Women's Shirt Small
27. tokidoki The Queen of Diamonds Women's Shirt Large

```
Title 3:  
61. psychedelic colorful Howling Galaxy Wolf T-shirt/Colorful Rainbow Animal Print Head  
Shirt for woman Neon Wolf t-shirt  
62. psychedelic colorful Howling Galaxy Wolf T-shirt/Colorful Rainbow Animal Print Head  
Shirt for woman Neon Wolf t-shirt  
63. psychedelic colorful Howling Galaxy Wolf T-shirt/Colorful Rainbow Animal Print Head  
Shirt for woman Neon Wolf t-shirt  
64. psychedelic colorful Howling Galaxy Wolf T-shirt/Colorful Rainbow Animal Print Head  
Shirt for woman Neon Wolf t-shirt
```

In [0]:

```
indices = []  
for i, row in data_sorted.iterrows():  
    indices.append(i)
```

In [0]:

```
import itertools  
stage1_deduplicate_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  
  
    # store the list of words of ith string in a, ex: a = ['tokidoki', 'The', 'Queen', 'of', 'Diamonds', 'Women's', 'Shirt', 'X-Large']  
    a = data['title'].loc[indices[i]].split()  
  
    # search for the similar products sequentially  
    j = i+1  
    while j < num_data_points:  
  
        # store the list of words of jth string in b, ex: b = ['tokidoki', 'The', 'Queen', 'of', 'Diamonds', 'Women's', 'Shirt', 'Small']  
        b = data['title'].loc[indices[j]].split()  
  
        # store the maximum length of two strings  
        length = max(len(a), len(b))  
  
        # count is used to store the number of words that are matched in both strings  
        count = 0  
  
        # itertools.zip_longest(a,b): will map the corresponding words in both strings, it will  
        appened None in case of unequal strings  
        # example: a = ['a', 'b', 'c', 'd']  
        # b = ['a', 'b', 'd']  
        # itertools.zip_longest(a,b): will give [('a', 'a'), ('b', 'b'), ('c', 'd'), ('d', None)]  
        for k in itertools.zip_longest(a, b):  
            if (k[0] == k[1]):  
                count += 1  
  
        # if the number of words in which both strings differ are > 2 , we are considering it as those two apperals are different  
        # if the number of words in which both strings differ are < 2 , we are considering it as those two apperals are same, hence we are ignoring them  
        if (length - count) > 2: # number of words in which both sensences differ  
            # if both strings are differ by more than 2 words we include the 1st string index  
            stage1_deduplicate_asins.append(data_sorted['asin'].loc[indices[i]])  
  
            # if the comprision between is between num_data_points, num_data_points-1 strings and they differ in more than 2 words we include both  
            if j == num_data_points-1: stage1_deduplicate_asins.append(data_sorted['asin'].loc[indices[j]])  
  
            # start searching for similar apperals corresponds 2nd string  
            i = j  
            break  
        else:  
            j += 1
```

```
if previous_i == i:  
    break
```

In [0]:

```
data = data.loc[data['asin'].isin(stage1_dedupe_asins)]
```

We removed the duplicates which differ only at the end.

In [0]:

```
print('Number of data points : ', data.shape[0])
```

```
Number of data points : 17593
```

In [0]:

```
data.to_pickle('pickels/17k_apperal_data')
```

[5.2.3] Remove duplicates : Part 2

In the previous cell, we sorted whole data in alphabetical order of titles. Then, we removed titles which are adjacent and very similar title

But there are some products whose titles are not adjacent but very similar.

Examples:

Titles-1

```
86261. UltraClub Women's Classic Wrinkle-Free Long Sleeve Oxford Shirt, Pink, XX-Large  
115042. UltraClub Ladies Classic Wrinkle-Free Long-Sleeve Oxford Light Blue XXL
```

Titles-2

```
75004. EVALY Women's Cool University Of UTAH 3/4 Sleeve Raglan Tee  
109225. EVALY Women's Unique University Of UTAH 3/4 Sleeve Raglan Tees  
120832. EVALY Women's New University Of UTAH 3/4-Sleeve Raglan Tshirt
```

In [0]:

```
data = pd.read_pickle('pickels/17k_apperal_data')
```

In [0]:

```
# This code snippet takes significant amount of time.  
# O(n^2) time.  
# Takes about an hour to run on a decent computer.  
  
indices = []  
for i, row in data.iterrows():  
    indices.append(i)  
  
stage2_dedupe_asins = []  
while len(indices)!=0:  
    i = indices.pop()  
    stage2_dedupe_asins.append(data['asin'].loc[i])  
    # consider the first apperal's title  
    a = data['title'].loc[i].split()  
    # store the list of words of ith string in a, ex: a = ['tokidoki', 'The', 'Queen', 'of', 'Diamonds', 'Women's', 'Shirt', 'X-Large']  
    for j in indices:  
        b = data['title'].loc[j].split()
```

```

# store the list of words of jth string in b, ex: b = ['tokidoki', 'The', 'Queen', 'of', 'Diamonds', 'Women's', 'Shirt', 'X-Large']

length = max(len(a), len(b))

# count is used to store the number of words that are matched in both strings
count = 0

# itertools.zip_longest(a,b): will map the corresponding words in both strings, it will
# append None in case of unequal strings
# example: a =['a', 'b', 'c', 'd']
# b = ['a', 'b', 'd']
# itertools.zip_longest(a,b): will give [('a','a'), ('b','b'), ('c',None), ('d', None)]
for k in itertools.zip_longest(a,b):
    if (k[0]==k[1]):
        count += 1

# if the number of words in which both strings differ are < 3 , we are considering it as those two apperals are same, hence we are ignoring them
if (length - count) < 3:
    indices.remove(j)

```

In [0]:

```

# from whole previous products we will consider only
# the products that are found in previous cell
data = data.loc[data['asin'].isin(stage2_dedupe_asins)]

```

In [0]:

```

print('Number of data points after stage two of dedupe: ',data.shape[0])
# from 17k apperals we reduced to 16k apperals

```

Number of data points after stage two of dedupe: 16042

In [0]:

```

data.to_pickle('pickels/16k_apperal_data')
# Storing these products in a pickle file
# candidates who wants to download these files instead
# of 180K they can download and use them from the Google Drive folder.

```

6. Text pre-processing

In [0]:

```

data = pd.read_pickle('pickels/16k_apperal_data')

# NLTK download stop words. [RUN ONLY ONCE]
# goto Terminal (Linux/Mac) or Command-Prompt (Window)
# In the temrinal, type these commands
# $python3
# $import nltk
# $nltk.download()

```

In [0]:

```

# we use the list of stop words that are downloaded from nltk lib.
stop_words = set(stopwords.words('english'))
print ('list of stop words:', stop_words)

def nlp_preprocessing(total_text, index, column):
    if type(total_text) is not int:
        string = ""
        for words in total_text.split():
            # remove the special chars in review like ''#$@!%^&*()_+-~?>< etc.
            word = ("").join(e for e in words if e.isalnum())
            # Cover all letters to lower-case
            word = word.lower()

```

```

# stop-word removal
if not word in stop_words:
    string += word + " "
data[column][index] = string

```

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

In [0]:

```

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")

```

3.5727220000000006 seconds

In [0]:

```
data.head()
```

Out[0]:

	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

In [0]:

```
data.to_pickle('pickels/16k_apperial_data_preprocessed')
```

Stemming

In [0]:

```

from nltk.stem.porter import *
stemmer = PorterStemmer()
print(stemmer.stem('arguing'))
print(stemmer.stem('fishing'))

# We tried using stemming on our titles and it didnot work very well.

```

```
argu  
fish
```

[8] Text based product similarity

In [3]:

```
data = pd.read_pickle('pickels/16k_apperal_data_preprocessed')  
data.head()
```

Out [3]:

	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

In [2]:

```
# Utility Functions which we will use through the rest of the workshop.  
  
#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 algorithm's decision.  
def plot_heatmap(keys, values, labels, url, text):  
    # keys: list of words of recommended title  
    # values: len(values) == len(keys), values(i) represents the occurence of the word  
    keys(i)  
    # labels: len(labels) == len(keys), the values of labels depends on the model we are using  
    # if model == 'bag of words': labels(i) = values(i)  
    # if model == 'tfidf weighted bag of words':labels(i) = tfidf(keys(i))  
    # if model == 'idf weighted bag of words':labels(i) = idf(keys(i))  
    # url : apparel's url  
  
    # we will devide the whole figure into two parts  
    gs = gridspec.GridSpec(2, 2, width_ratios=[4,1], height_ratios=[4,1])  
    fig = plt.figure(figsize=(25,3))  
  
    # 1st, plotting heat map that represents the count of commonly occurred words in title2  
    ax = plt.subplot(gs[0])  
    # it displays a cell in white color if the word is intersection(lis of words of title1 and  
    # list of words of title2), in black if not  
    ax = sns.heatmap(np.array([values]), annot=np.array([labels]))  
    ax.set_xticklabels(keys) # set that axis labels as the words of title  
    ax.set_title(text) # apparel title  
  
    # 2nd, plotting image of the apparel  
    ax = plt.subplot(gs[1])  
    # we don't want any grid lines for image and no labels on x-axis and y-axis  
    ax.grid(False)
```

```

ax.set_xticks([])
ax.set_yticks([])

# we call dispaly_img based with paramete url
display_img(url, ax, fig)

# displays combine figure ( heat map and image together)
plt.show()

def plot_heatmap_image(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, it can be any of the models,
    # 1. bag_of_words
    # 2. tfidf
    # 3. idf

    # we find the common words in both titles, because these only words contribute to the distance
    between two title vec's
    intersection = set(vec1.keys()) & set(vec2.keys())

    # we set the values of non intersecting words to zero, this is just to show the difference in
    heatmap
    for i in vec2:
        if i not in intersection:
            vec2[i]=0

    # for labeling heatmap, keys contains list of all words in title2
    keys = list(vec2.keys())
    # if ith word in intersection(lis of words of title1 and list of words of title2):
    values(i)=count of that word in title2 else values(i)=0
    values = [vec2[x] for x in vec2.keys()]

    # labels: len(labels) == len(keys), the values of labels depends on the model we are using
    # if model == 'bag of words': labels(i) = values(i)
    # if model == 'tfidf weighted bag of words':labels(i) = tfidf(keys(i))
    # if model == 'idf weighted bag of words':labels(i) = idf(keys(i))

    if model == 'bag_of_words':
        labels = values
    elif model == 'tfidf':
        labels = []
        for x in vec2.keys():
            # tfidf_title_vectorizer.vocabulary_ it contains all the words in the corpus
            # tfidf_title_features[doc_id, index_of_word_in_corpus] will give the tfidf value of w
            rd in given document (doc_id)
            if x in tfidf_title_vectorizer.vocabulary_:
                labels.append(tfidf_title_features[doc_id, tfidf_title_vectorizer.vocabulary_[x]])
            else:
                labels.append(0)
    elif model == 'idf':
        labels = []
        for x in vec2.keys():
            # idf_title_vectorizer.vocabulary_ it contains all the words in the corpus
            # idf_title_features[doc_id, index_of_word_in_corpus] will give the idf value of word
            in given document (doc_id)
            if x in idf_title_vectorizer.vocabulary_:
                labels.append(idf_title_features[doc_id, idf_title_vectorizer.vocabulary_[x]])
            else:
                labels.append(0)

    plot_heatmap(keys, values, labels, url, text)

    # this function gets a list of wrods along with the frequency of each
    # word given "text"
    def text_to_vector(text):
        word = re.compile(r'\w+')
        words = word.findall(text)
        # words stores list of all words in given string, you can try 'words = text.split()' this will
        also gives same result
        return Counter(words) # Counter counts the occurence of each word in list, it returns dict
        type object {word1:count}

```

```

def get_result(doc_id, content_a, content_b, url, model):
    text1 = content_a
    text2 = content_b

    # vector1 = dict{word11:#count, word12:#count, etc.}
    vector1 = text_to_vector(text1)

    # vector2 = dict{word21:#count, word22:#count, etc.}
    vector2 = text_to_vector(text2)

    plot_heatmap_image(doc_id, vector1, vector2, url, text2, model)

```

[8.2] Bag of Words (BoW) on product titles.

In [0]:

```

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.
# title_features.shape = #data_points * #words_in_corpus
# CountVectorizer().fit_transform(corpus) returns
# the a sparse matrix of dimensions #data_points * #words_in_corpus

# What is a sparse vector?

# title_features[doc_id, index_of_word_in_corpus] = number of times the word occurred in that doc

```

Out[0]:

(16042, 12609)

In [0]:

```

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 apparels
    # the metric we used here is cosine, the coside distance is measured as K(X, Y) = <X, Y> / (||X
    || * ||Y||)
    # http://scikit-learn.org/stable/modules/metrics.html#cosine-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]

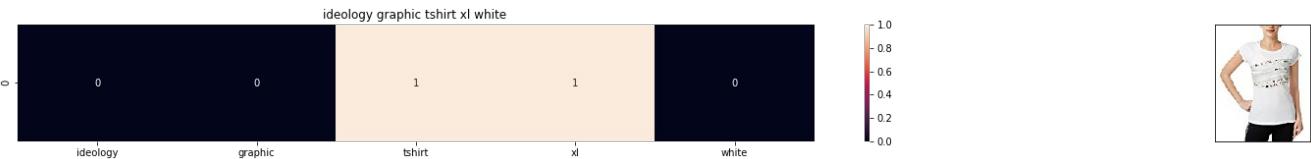
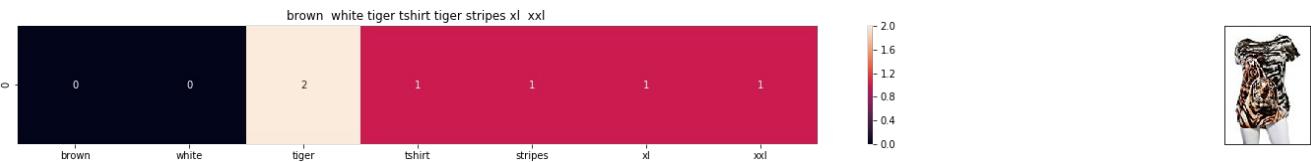
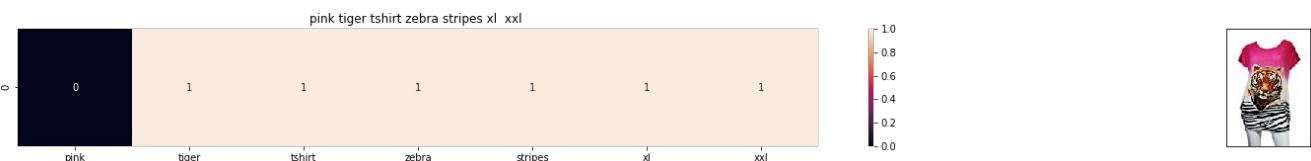
    #data frame indices of the 9 smallest distace's
    df_indices = list(data.index[indices])

    for i in range(0,len(indices)):
        # we will pass 1. doc_id, 2. title1, 3. title2, url, model
        get_result(indices[i],data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], 'bag_of_words')
        print('ASIN :',data['asin'].loc[df_indices[i]])
        print ('Brand:', data['brand'].loc[df_indices[i]])
        print ('Title:', data['title'].loc[df_indices[i]])
        print ('Euclidean similarity with the query image :', pdists[i])
        print('='*60)

    #call the bag-of-words model for a product to get similar products.
    bag_of_words_model(12566, 20) # change the index if you want to.
    # In the output heat map each value represents the count value
    # of the label word, the color represents the intersection
    # with inputs title.

#try 12566
#try 931

```

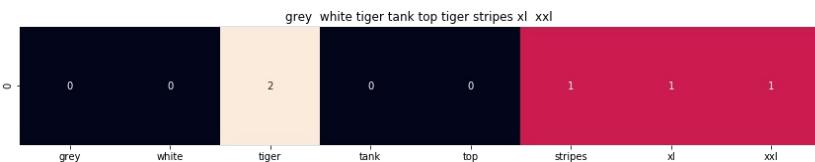


ASIN : B01NB0NKRO

Brand: Ideology

Title: ideology graphic tshirt xl white

Euclidean similarity with the query image : 3.0

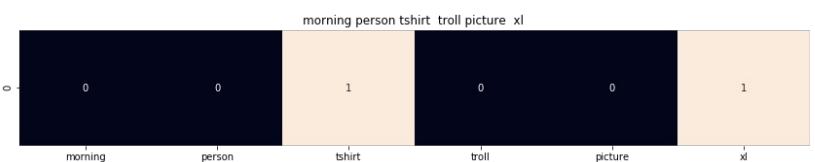


ASIN : B00JXQAFZ2

Brand: Si Row

Title: grey white tiger tank top tiger stripes xl xxl

Euclidean similarity with the query image : 3.0

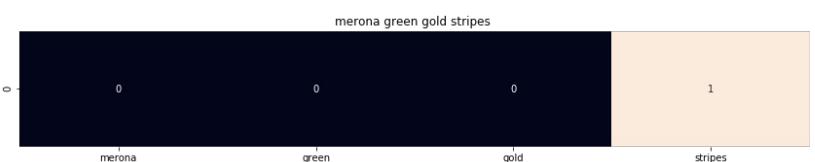


ASIN : B01CLS8LMW

Brand: Awake

Title: morning person tshirt troll picture xl

Euclidean similarity with the query image : 3.16227766017

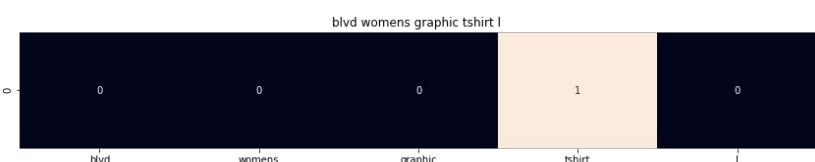


ASIN : B01KVZUB6G

Brand: Merona

Title: merona green gold stripes

Euclidean similarity with the query image : 3.16227766017

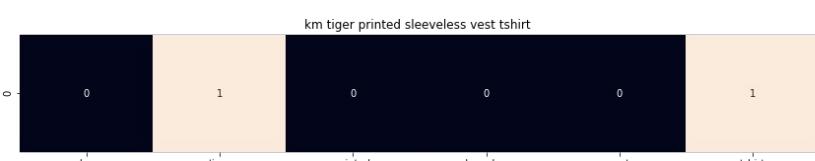


ASIN : B0733R2CJK

Brand: BLVD

Title: blvd womens graphic tshirt l

Euclidean similarity with the query image : 3.16227766017



ASIN : B012VQLT6Y

Brand: KM T-shirt

Title: km tiger printed sleeveless vest tshirt

Euclidean similarity with the query image : 3.16227766017



ASIN : B00JXQC8L6

Brand: Si Row

Title: blue peacock print tshirt l

Euclidean similarity with the query image : 3.16227766017

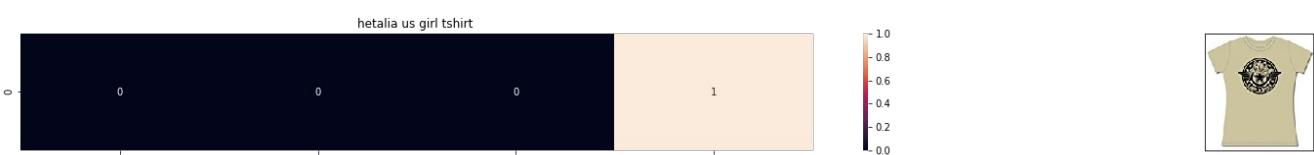


ASIN : B06XC3CZF6

Brand: Fjallraven

Title: fjallraven womens ovik tshirt plum xxl

Euclidean similarity with the query image : 3.16227766017

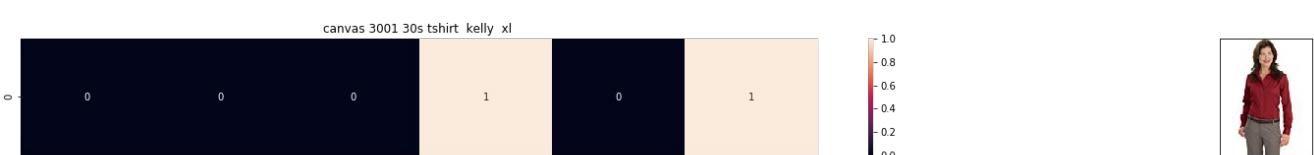


ASIN : B005IT8OBA

Brand: Hetalia

Title: hetalia us girl tshirt

Euclidean similarity with the query image : 3.16227766017

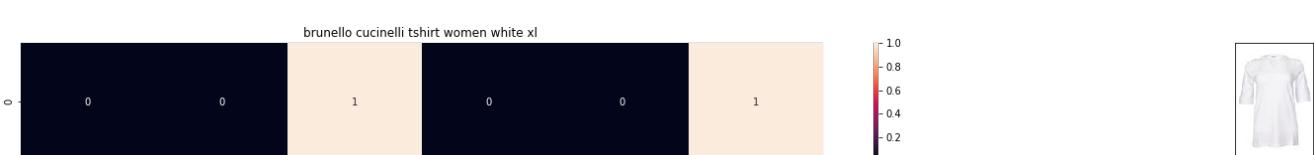


ASIN : B0088PNOLA

Brand: Red House

Title: canvas 3001 30s tshirt kelly xl

Euclidean similarity with the query image : 3.16227766017

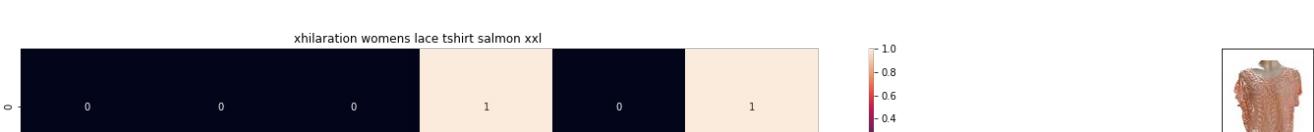


ASIN : B06X99V6WC

Brand: Brunello Cucinelli

Title: brunello cucinelli tshirt women white xl

Euclidean similarity with the query image : 3.16227766017





ASIN : B06Y1JPW1Q

Brand: Xhilaration

Title: xhilaration womens lace tshirt salmon xxl

Euclidean similarity with the query image : 3.16227766017



ASIN : B06X6GX6WG

Brand: Animal

Title: animal oceania tshirt yellow

Euclidean similarity with the query image : 3.16227766017

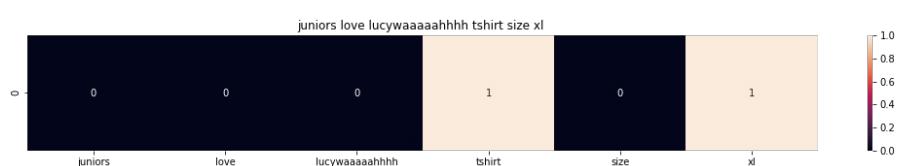


ASIN : B017X8PW9U

Brand: Diesel

Title: diesel tserraf tshirt black

Euclidean similarity with the query image : 3.16227766017



ASIN : B00IAA4JIQ

Brand: I Love Lucy

Title: juniors love lucywaaaaahhhh tshirt size xl

Euclidean similarity with the query image : 3.16227766017

[8.5] TF-IDF based product similarity

In [0]:

```
tfidf_title_vectorizer = TfidfVectorizer(min_df = 0)
tfidf_title_features = tfidf_title_vectorizer.fit_transform(data['title'])
# tfidf_title_features.shape = #data_points * #words_in_corpus
# CountVectorizer().fit_transform(courpus) returns the a sparase matrix of dimensions #data_points
* #words_in_corpus
# tfidf_title_features[doc_id, index_of_word_in_corpus] = tfidf values of the word in given doc
```

In [0]:

```
def tfidf_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 apparels
    # the metric we used here is cosine, the coside distance is mesured as K(X, Y) = <X, Y> / (||X|| * ||Y||)
    # http://scikit-learn.org/stable/modules/metrics.html#cosine-similarity
    pairwise_dist = pairwise_distances(tfidf_title_features,tfidf_title_features[doc_id])
```

```

# np.argsort will return indices of 9 smallest distances
indices = np.argsort(pairwise_dist.flatten())[0:num_results]
#pdists will store the 9 smallest distances
pdists = np.sort(pairwise_dist.flatten())[0:num_results]

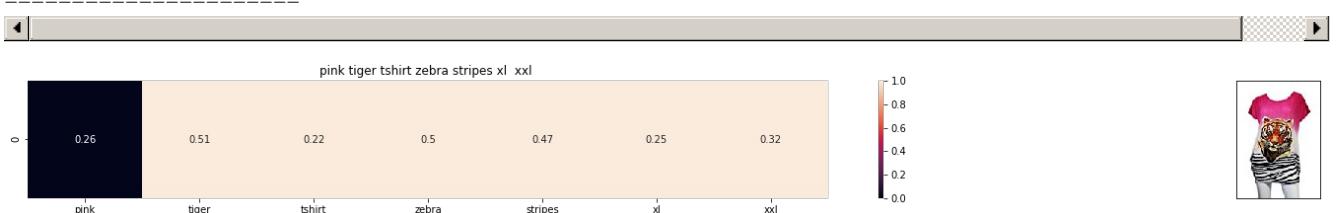
#data frame indices of the 9 smallest distace's
df_indices = list(data.index[indices])

for i in range(0,len(indices)):
    # we will pass 1. doc_id, 2. title1, 3. title2, url, model
    get_result(indices[i], data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]],
data['medium_image_url'].loc[df_indices[i]], 'tfidf')
        print('ASIN :',data['asin'].loc[df_indices[i]])
        print('BRAND :',data['brand'].loc[df_indices[i]])
        print ('Eucliden distance from the given image :', pdists[i])
        print('='*125)
tfidf_model(12566, 20)
# in the output heat map each value represents the tfidf values of the label word, the color represents the intersection with inputs title

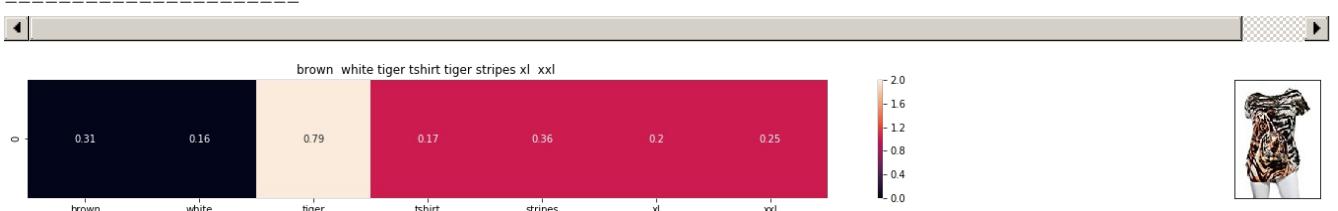
```



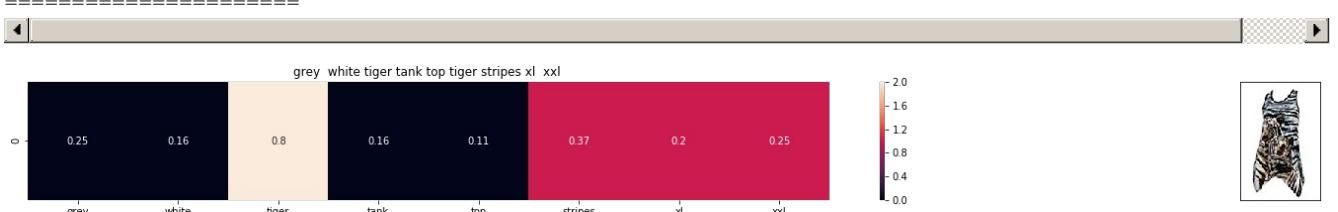
ASIN : B00JXQB5FQ
 BRAND : Si Row
 Eucliden distance from the given image : 0.0



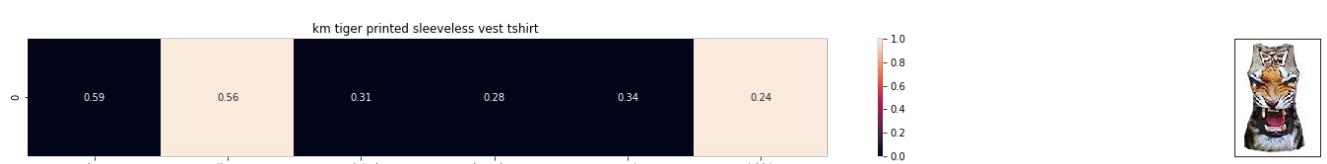
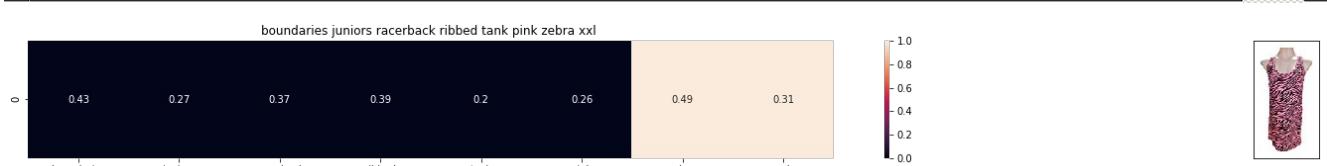
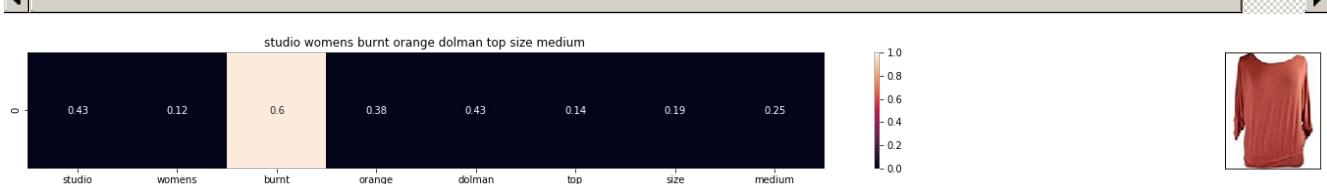
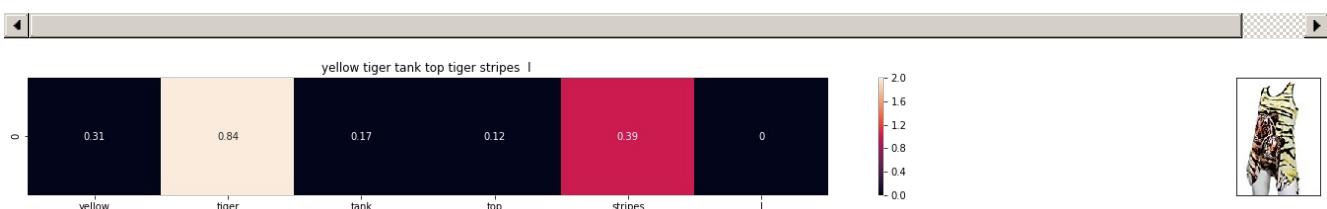
ASIN : B00JXQASS6
 BRAND : Si Row
 Eucliden distance from the given image : 0.753633191245



ASIN : B00JXQCWT0
 BRAND : Si Row
 Eucliden distance from the given image : 0.935764394377



ASIN : B00JXQAFZ2
 BRAND : Si Row
 Eucliden distance from the given image : 0.95861535242



km

tiger

printed

sleeveless

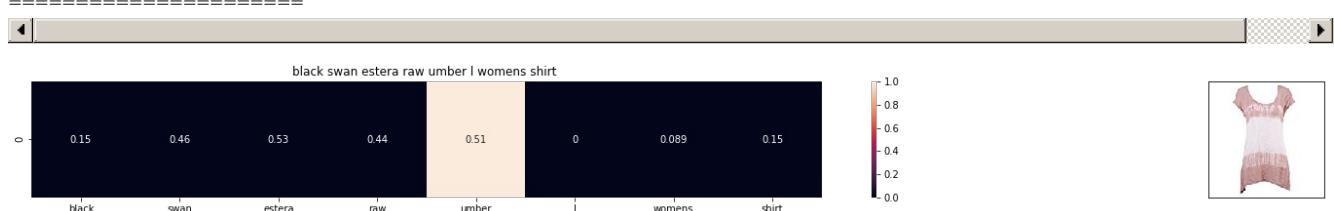
vest

tshirt

ASIN : B012VQLT6Y

BRAND : KM T-shirt

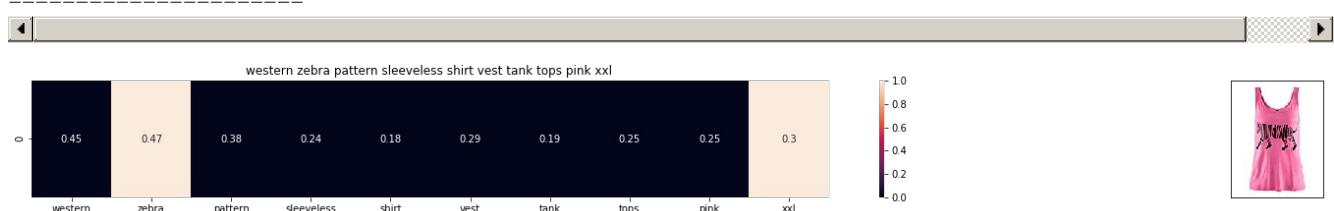
Euclidean distance from the given image : 1.21979064028



ASIN : B06Y1VN8WQ

BRAND : Black Swan

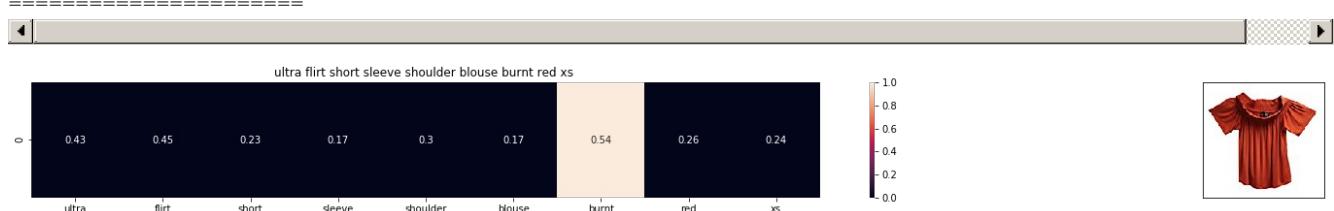
Euclidean distance from the given image : 1.220684966



ASIN : B00Z6HEXWI

BRAND : Black Temptation

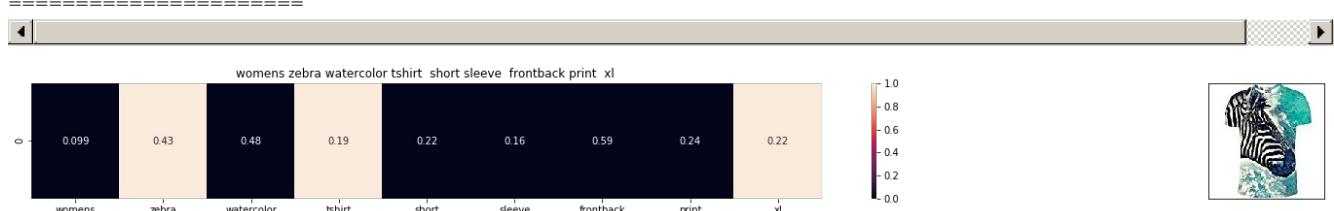
Euclidean distance from the given image : 1.22128139212



ASIN : B074TR12BH

BRAND : Ultra Flirt

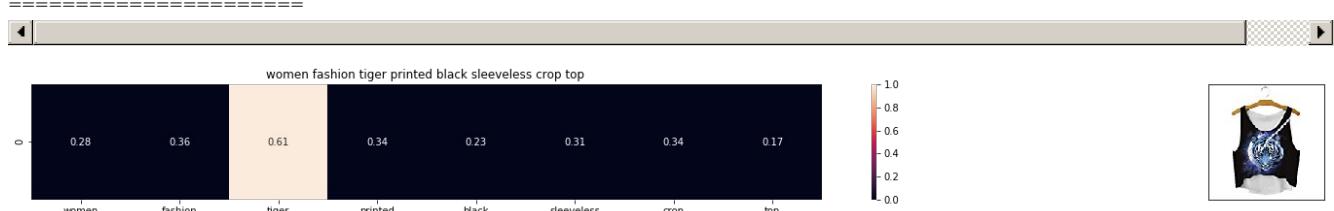
Euclidean distance from the given image : 1.23133640946



ASIN : B072R2JXKW

BRAND : WHAT ON EARTH

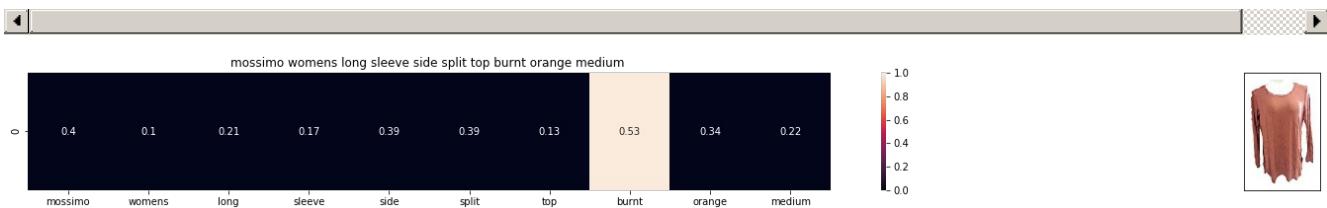
Euclidean distance from the given image : 1.23184519726



ASIN : B074T8ZYGX

BRAND : MKP Crop Top

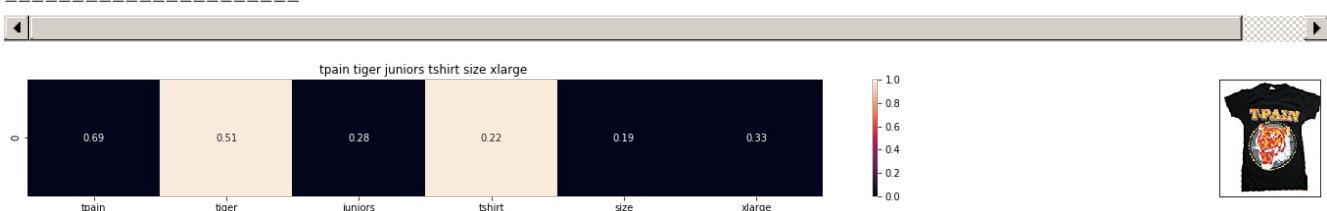
Euclidean distance from the given image : 1.23406074574



ASIN : B071ZDF6T2

BRAND : Mossimo

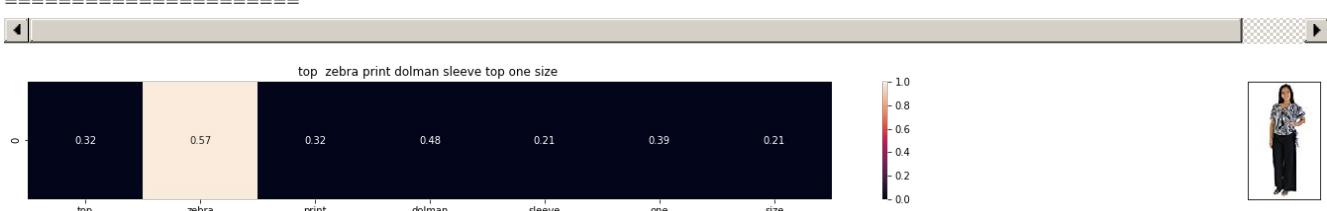
Euclidean distance from the given image : 1.23527855777



ASIN : B01K0H02OG

BRAND : Tultex

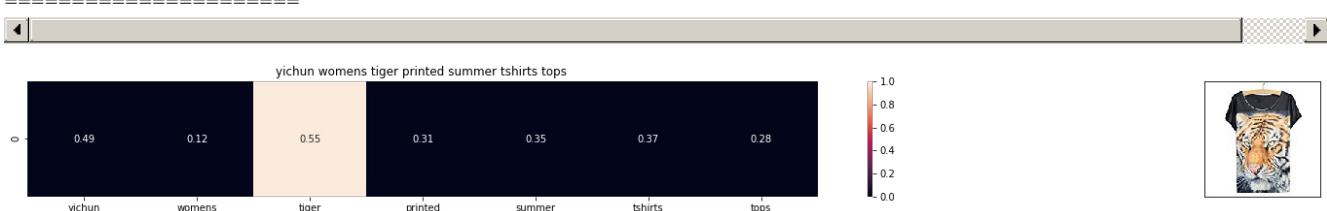
Euclidean distance from the given image : 1.23645729881



ASIN : B00H8A6ZLI

BRAND : Vivian's Fashions

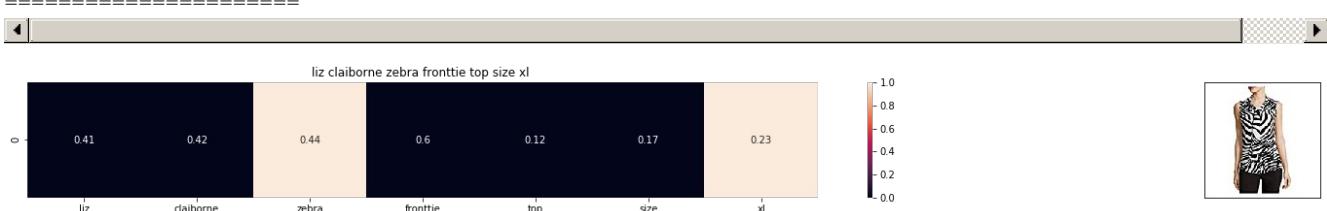
Euclidean distance from the given image : 1.24996155053



ASIN : B010NN9RXO

BRAND : YICHUN

Euclidean distance from the given image : 1.25354614209



ASIN : B06XBY5QXL

BRAND : Liz Claiborne

Euclidean distance from the given image : 1.25388329384

[8.5] IDF based product similarity

```
In [12]:
```

```
idf_title_vectorizer = CountVectorizer()
idf_title_features = idf_title_vectorizer.fit_transform(data['title'])

# idf_title_features.shape = #data_points * #words_in_corpus
# CountVectorizer().fit_transform(courpus) returns the a sparase matrix of dimensions #data_points
* #words_in_corpus
# idf_title_features[doc_id, index_of_word_in_corpus] = number of times the word occured in that document
```

```
In [13]:
```

```
def nContaining(word):
    # return the number of documents which had the given word
    return sum(1 for blob in data['title'] if word in blob.split())

def idf(word):
    # idf = log(#number of docs / #number of docs which had the given word)
    return math.log(data.shape[0] / (nContaining(word)))
```

```
In [14]:
```

```
# we need to convert the values into float
idf_title_features = idf_title_features.astype(np.float)

for i in idf_title_vectorizer.vocabulary_.keys():
    # for every word in whole corpus we will find its idf value
    idf_val = idf(i)

    # to calculate idf_title_features we need to replace the count values with the idf values of the word
    # idf_title_features[:, idf_title_vectorizer.vocabulary_[i]].nonzero()[0] will return all documents in which the word i present
    for j in idf_title_features[:, idf_title_vectorizer.vocabulary_[i]].nonzero()[0]:
        # we replace the count values of word i in document j with idf_value of word i
        # idf_title_features[doc_id, index_of_word_in_corpus] = idf value of word
        idf_title_features[j,idf_title_vectorizer.vocabulary_[i]] = idf_val
```

```
In [0]:
```

```
def idf_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 apparels
    # the metric we used here is cosine, the coside distance is mesured as K(X, Y) = <X, Y> / (||X|| * ||Y||)
    # http://scikit-learn.org/stable/modules/metrics.html#cosine-similarity
    pairwise_dist = pairwise_distances(idf_title_features,idf_title_features[doc_id])

    # np.argsort will return indices of 9 smallest distances
    indices = np.argsort(pairwise_dist.flatten())[0:num_results]
    #pdists will store the 9 smallest distances
    pdists = np.sort(pairwise_dist.flatten())[0:num_results]

    #data frame indices of the 9 smallest distace's
    df_indices = list(data.index[indices])

    for i in range(0,len(indices)):
        get_result(indices[i],data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], 'idf')
        print('ASIN :',data['asin'].loc[df_indices[i]])
        print('Brand :',data['brand'].loc[df_indices[i]])
        print ('euclidean distance from the given image :', pdists[i])
        print('='*125)

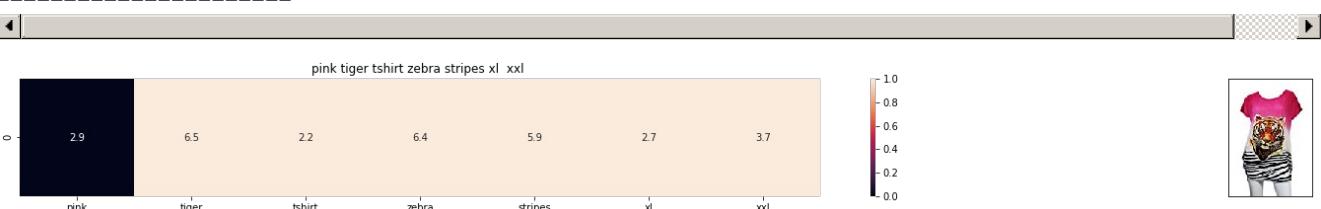
    idf_model(12566,20)
    # in the output heat map each value represents the idf values of the label word, the color represents the intersection with inputs title
```



ASIN : B00JXQB5FQ

Brand : Si Row

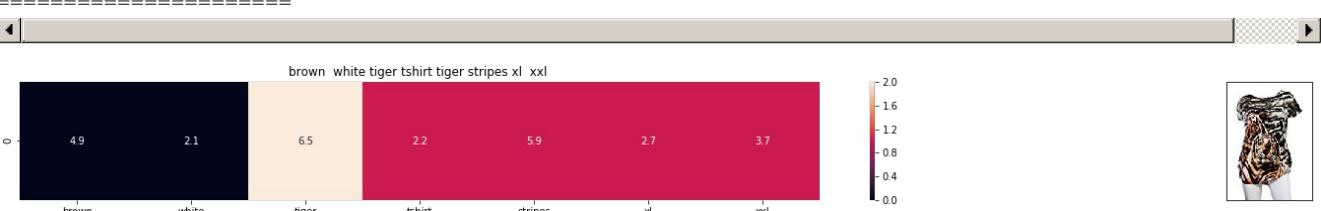
euclidean distance from the given image : 0.0



ASIN : B00JXQASS6

Brand : Si Row

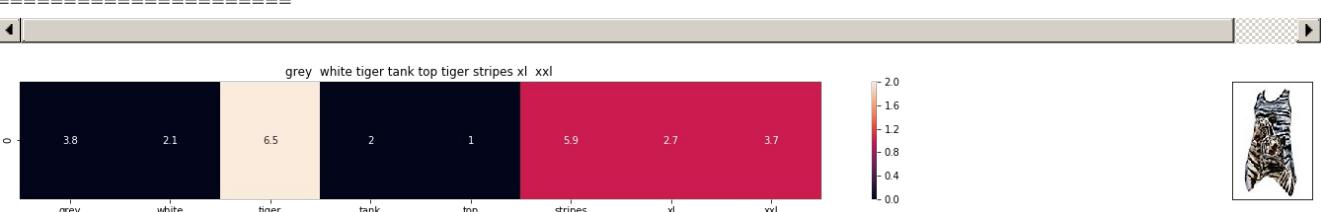
euclidean distance from the given image : 12.2050713112



ASIN : B00JXQCWT0

Brand : Si Row

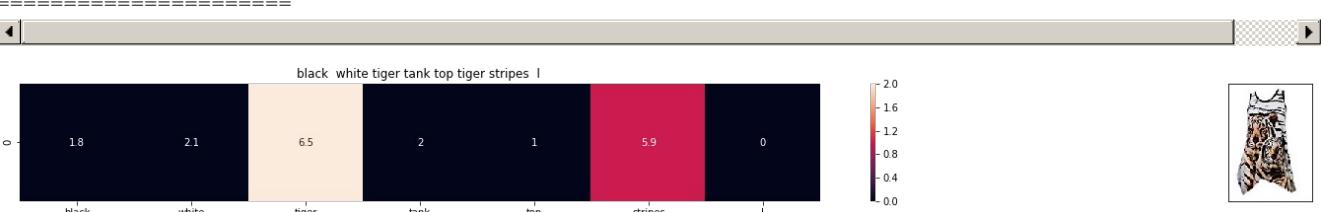
euclidean distance from the given image : 14.4683626856



ASIN : B00JXQAFZ2

Brand : Si Row

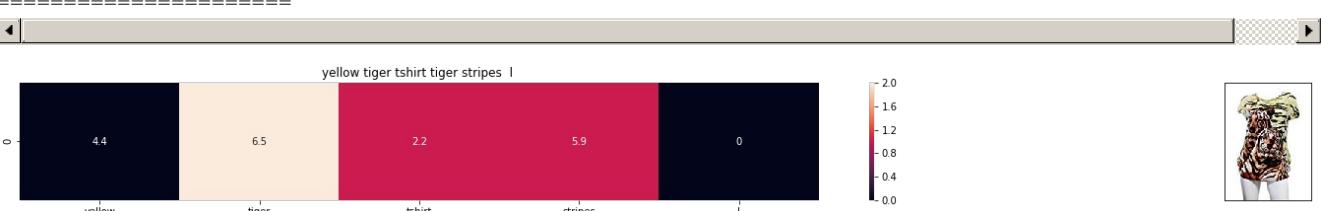
euclidean distance from the given image : 14.4868329248



ASIN : B00JXQA094

Brand : Si Row

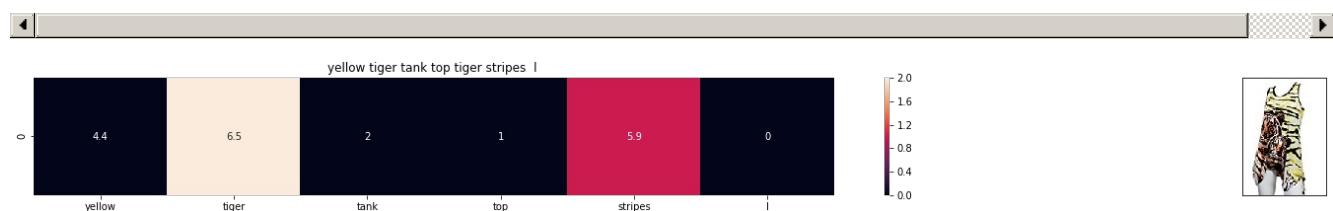
euclidean distance from the given image : 14.8333929667



ASIN : B00JXQCUIC

Brand : Si Row

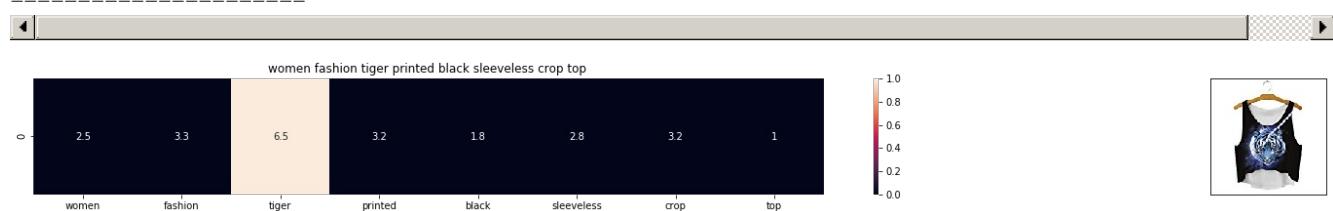
euclidean distance from the given image : 14.8987445167



ASIN : B00JXQAUWA

Brand : Si Row

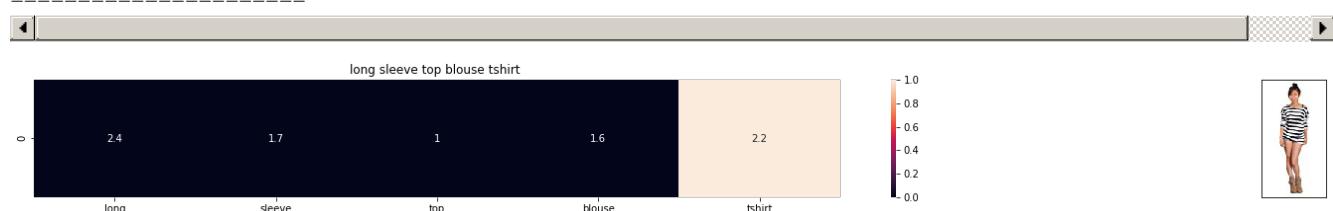
euclidean distance from the given image : 15.2244582873



ASIN : B074T8ZYGX

Brand : MKP Crop Top

euclidean distance from the given image : 17.0808129556



ASIN : B00KF2N5PU

Brand : Vietsbay

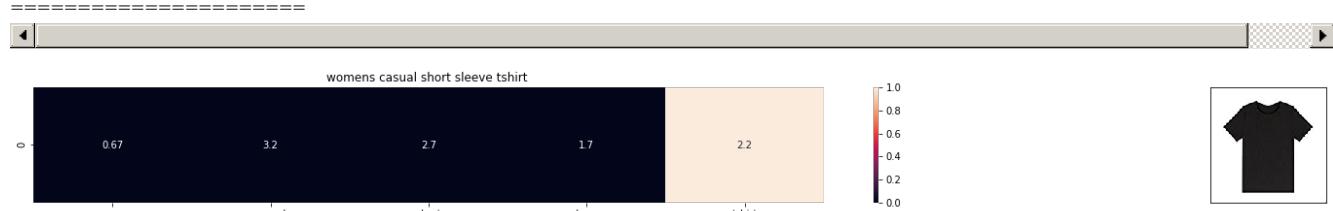
euclidean distance from the given image : 17.0901681256



ASIN : B00JPOZ9GM

Brand : Sofra

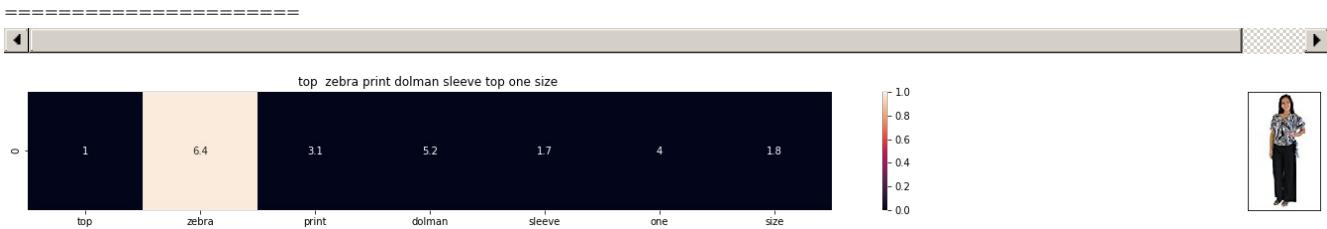
euclidean distance from the given image : 17.1532153376



ASIN : B074T9KG9Q

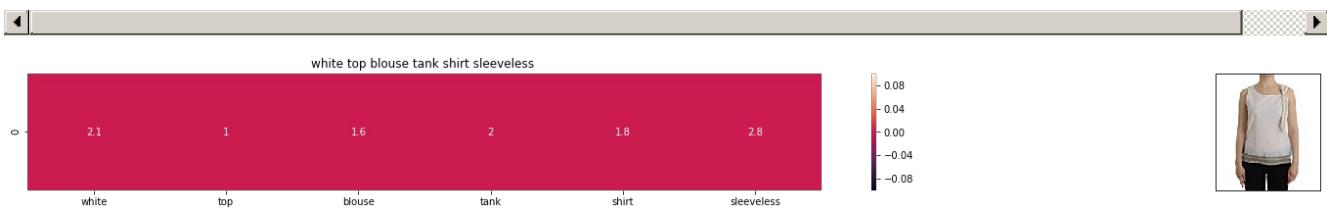
Brand : Rain

euclidean distance from the given image : 17.3367152387



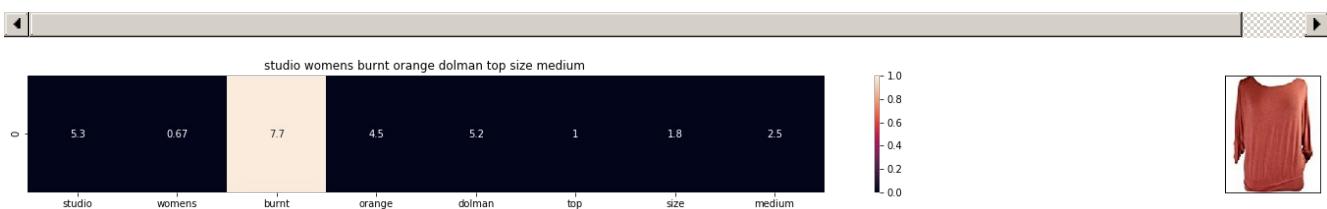
ASIN : B00H8A6ZLI
 Brand : Vivian's Fashions
 euclidean distance from the given image : 17.410075941

=====



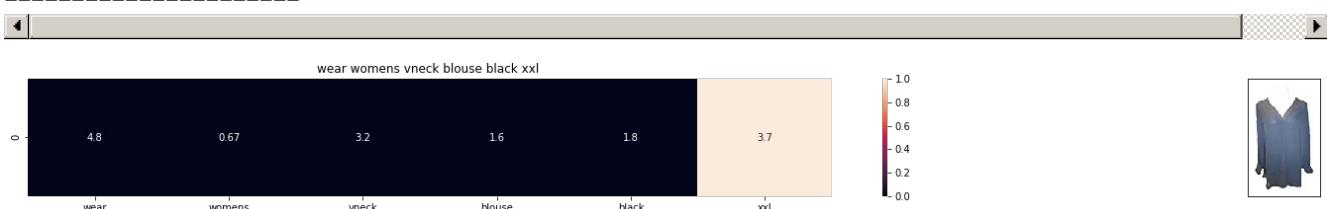
ASIN : B074G5G5RK
 Brand : ERMANNO SCERVINO
 euclidean distance from the given image : 17.5399213355

=====



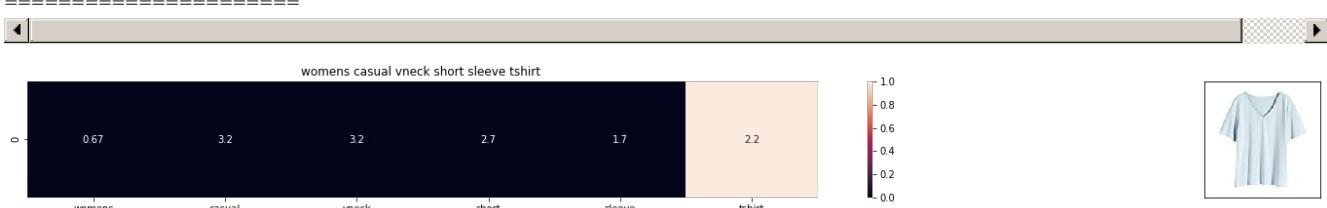
ASIN : B06XSCVFT5
 Brand : Studio M
 euclidean distance from the given image : 17.6127585437

=====



ASIN : B06Y6FH453
 Brand : Who What Wear
 euclidean distance from the given image : 17.6237452825

=====



ASIN : B074V45DCX
 Brand : Rain
 euclidean distance from the given image : 17.6343424968

=====





ASIN : B07583CQFT

Brand : Very J

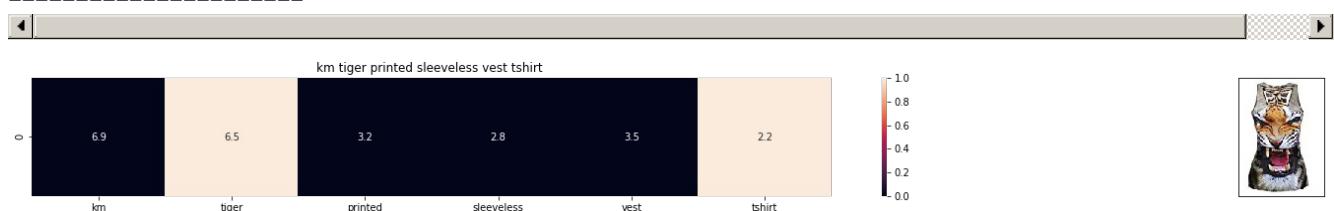
euclidean distance from the given image : 17.6375371274



ASIN : B073GJGVBN

Brand : Ivan Levi

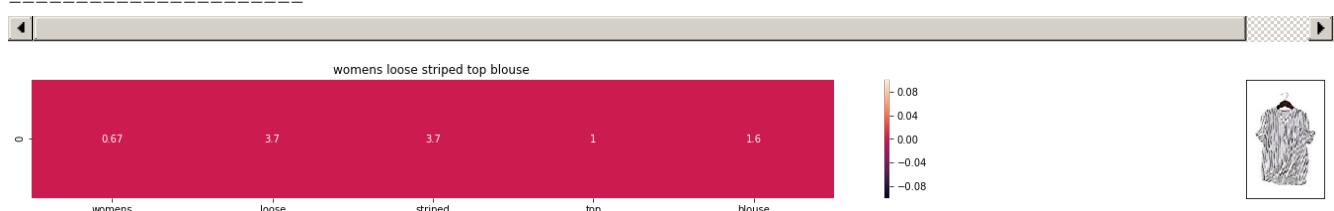
euclidean distance from the given image : 17.7230738913



ASIN : B012VQLT6Y

Brand : KM T-shirt

euclidean distance from the given image : 17.7625885612



ASIN : B00ZZMYBRG

Brand : HP-LEISURE

euclidean distance from the given image : 17.7795368647



[9] Text Semantics based product similarity

In [0]:

```
# credits: https://www.kaggle.com/c/word2vec-nlp-tutorial#part-2-word-vectors
# Custom Word2Vec using your own text data.
# Do NOT RUN this code.
# It is meant as a reference to build your own Word2Vec when you have
# lots of data.

'''

# Set values for various parameters
num_features = 300      # Word vector dimensionality
min_word_count = 1        # Minimum word count
num_workers = 4            # Number of threads to run in parallel
context = 10               # Context window size

downsampling = 1e-3       # Downsample setting for frequent words

# Initialize and train the model (this will take some time)
from gensim.models import word2vec
print ("Training model...")
```

```
model = word2vec.Word2Vec(sen_corpus, workers=num_workers, \
    size=num_features, min_count = min_word_count, \
    window = context)
```

```
'''
```

In [6]:

```
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle

# in this project we are using a pretrained model by google
# its 3.3G file, once you load this into your memory
# it occupies ~9Gb, so please do this step only if you have >12G of ram
# we will provide a pickle file which contains a dict ,
# and it contains all our corpus words as keys and model[word] as values
# To use this code-snippet, download "GoogleNews-vectors-negative300.bin"
# from https://drive.google.com/file/d/0B7XkCwpI5KDYNlNUTtSS2lPQmM/edit
# it's 1.9GB in size.

'''
model = KeyedVectors.load_word2vec_format('GoogleNews-vectors-negative300.bin', binary=True)
'''

#if you do NOT have RAM >= 12GB, use the code below.
with open('word2vec_model', 'rb') as handle:
    model = pickle.load(handle)
```

In [7]:

```
# Utility functions

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 idf_title_vectorizer.vocabulary_:
                vec.append(idf_title_features[doc_id, idf_title_vectorizer.vocabulary_[i]] * model[i])
            elif m_name == 'avg':
                vec.append(model[i])
        else:
            # if the word in our corpus is not there in the google word2vec corpus, we are just
            # ignoring 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)

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 calculate the distance(euclidean) to all vectors in vec2
    for i in vec1:
        dist = []
        for j in vec2:
            # np.linalg.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)
```

```

def heat_map_w2v(sentence1, sentence2, url, doc_id1, doc_id2, model):
    # sentance1 : title1, input apparel
    # sentance2 : title2, recommended apparel
    # url: apparel image url
    # doc_id1: document id of input apparel
    # doc_id2: document id of recommended apparel
    # 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(sentence1, doc_id1, model)
    s2_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
    s2_vec = get_word_vec(sentence2, 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)

# devide whole figure into 2 parts 1st part displays heatmap 2nd part displays image of apparel
1
gs = gridspec.GridSpec(2, 2, width_ratios=[4,1],height_ratios=[2,1])
fig = plt.figure(figsize=(15,15))

ax = plt.subplot(gs[0])
# ploting the heap map based on the pairwise distances
ax = sns.heatmap(np.round(s1_s2_dist,4), annot=True)
# set the x axis labels as recommended apparels title
ax.set_xticklabels(sentence2.split())
# set the y axis labels as input apparels title
ax.set_yticklabels(sentence1.split())
# set title as recommended apparels title
ax.set_title(sentence2)

ax = plt.subplot(gs[1])
# we remove all grids and axis labels for image
ax.grid(False)
ax.set_xticks([])
ax.set_yticks([])
display_img(url, ax, fig)

plt.show()

```

In [8]:

```

# vocab = stores all the words that are there in google w2v model
# vocab = model.wv.vocab.keys() # if you are using Google word2Vec

vocab = model.keys()
# this function will add the vectors of each word and returns the avg vector of given sentance
def build_avg_vec(sentence, num_features, doc_id, m_name):
    # sentace: its title of the apparel
    # num_features: the lenght of word2vec vector, its values = 300
    # 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)

featureVec = np.zeros((num_features,), dtype="float32")
# we will intialize a vector of size 300 with all zeros
# we add each word2vec(wordi) to this fatureVec
nwords = 0

for word in sentence.split():
    nwords += 1
    if word in vocab:
        if m_name == 'weighted' and word in idf_title_vectorizer.vocabulary_:
            featureVec = np.add(featureVec, idf_title_features[doc_id, idf_title_vectorizer.vocabulary_[word]] * model[word])
        elif m_name == 'avg':
            featureVec = np.add(featureVec, model[word])
if(nwords>0):
    featureVec = np.divide(featureVec, nwords)
    # returns the avg vector of given sentance, its of shape (1, 300)

```

```
    return featureVec
```

[9.2] Average Word2Vec product similarity.

In [10]:

```
doc_id = 0
w2v_title = []
# for every title we build a avg vector representation
for i in data['title']:
    w2v_title.append(build_avg_vec(i, 300, doc_id, 'avg'))
    doc_id += 1

# w2v_title = np.array(# number of doc in courpus * 300), each row corresponds to a doc
w2v_title = np.array(w2v_title)
```

In [0]:

```
def avg_w2v_model(doc_id, num_results):
    # doc_id: apparel's id in given corpus

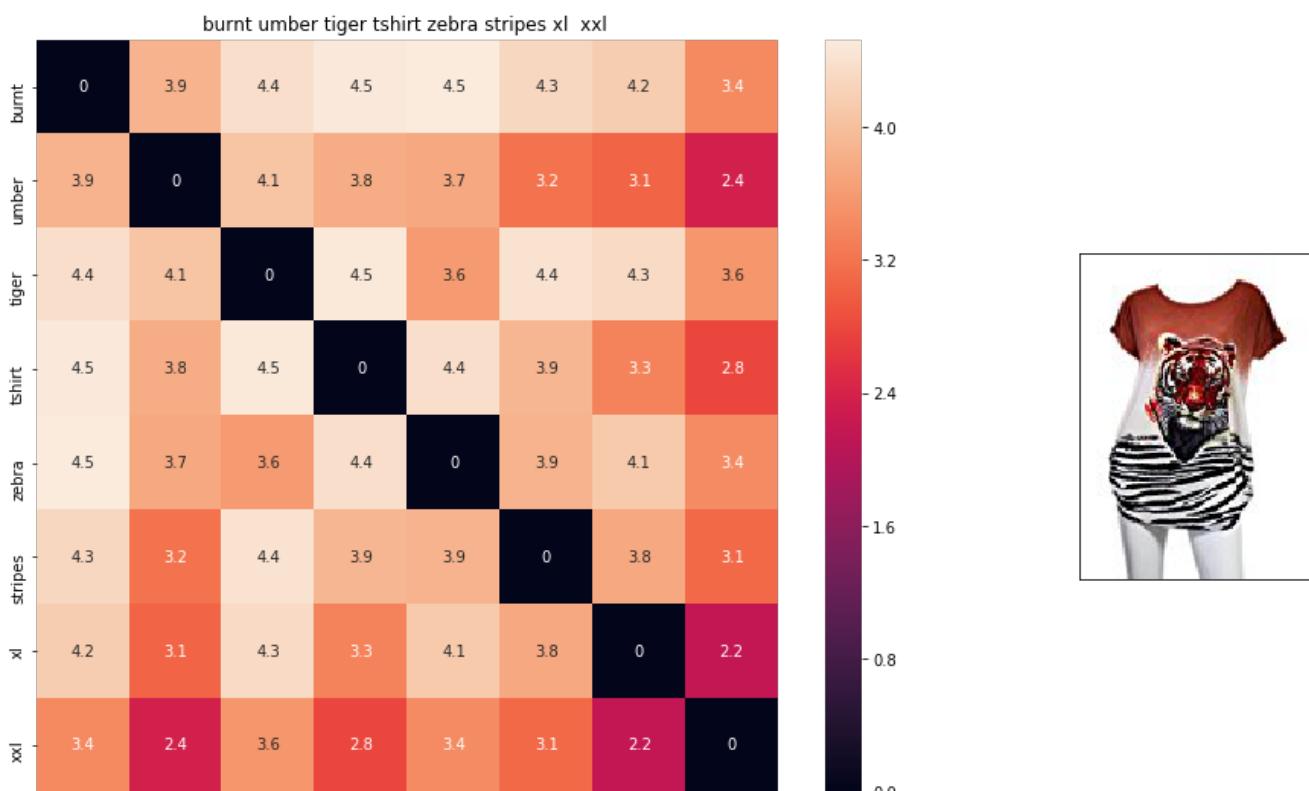
    # dist(x, y) = sqrt(dot(x, x) - 2 * dot(x, y) + dot(y, y))
    pairwise_dist = pairwise_distances(w2v_title, w2v_title[doc_id].reshape(1,-1))

    # np.argsort will return indices of 9 smallest distances
    indices = np.argsort(pairwise_dist.flatten())[0:num_results]
    #pdists will store the 9 smallest distances
    pdists = np.sort(pairwise_dist.flatten())[0:num_results]

    #data frame indices of the 9 smallest distance's
    df_indices = list(data.index[indices])

    for i in range(0, len(indices)):
        heat_map_w2v(data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], indices[0], indices[i], 'avg')
        print('ASIN :', data['asin'].loc[df_indices[i]])
        print('BRAND :', data['brand'].loc[df_indices[i]])
        print ('euclidean distance from given input image :', pdists[i])
        print('='*125)

avg_w2v_model(12566, 20)
# in the give heat map, each cell contains the euclidean distance between words i, j
```



burnt umber tiger tshirt zebra stripes xl xxl

ASIN : B00JXQB5FQ

BRAND : Si Row

euclidean distance from given input image : 0.000690534

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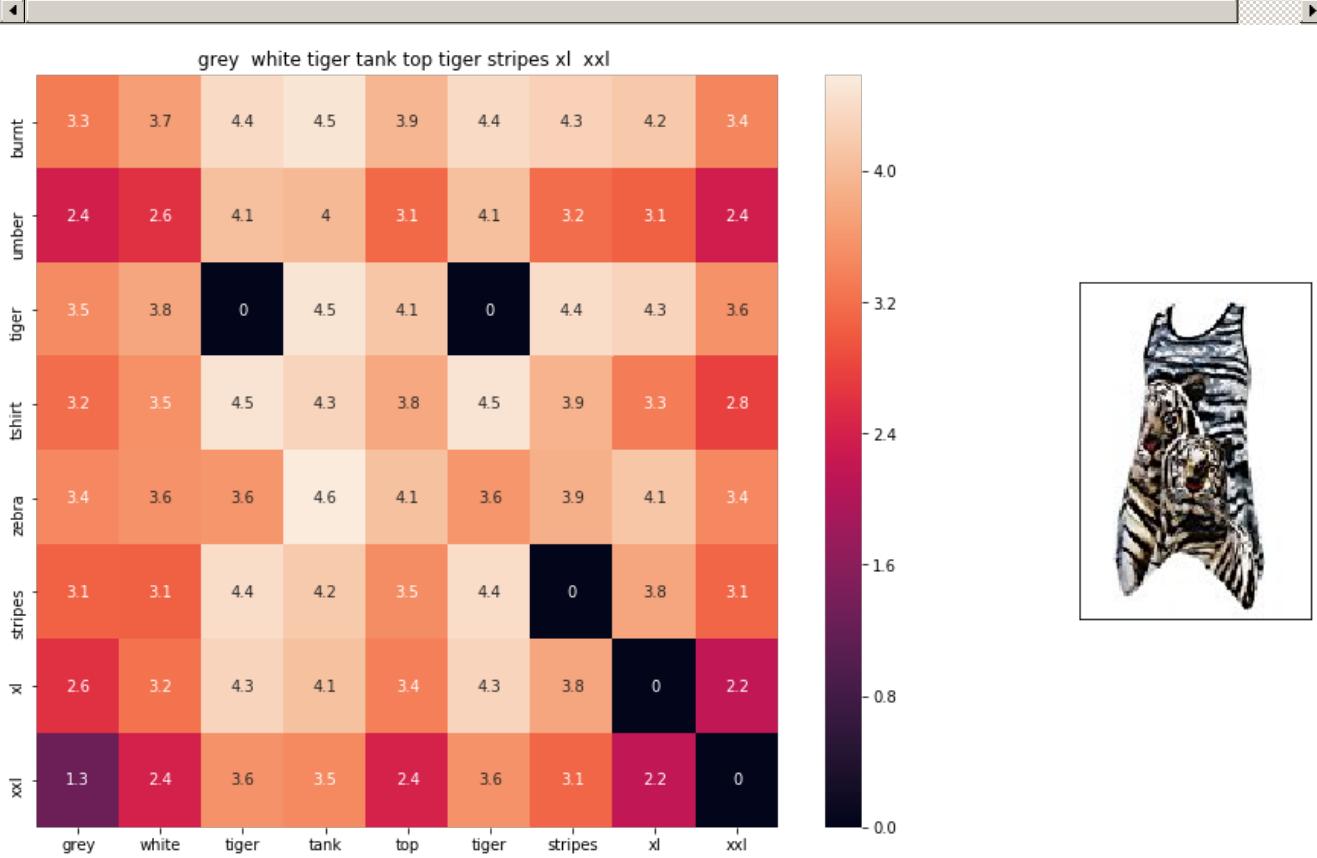
=====

brown white tiger tshirt tiger stripes xl xxl

ASIN : B00JXQCWT0

BRAND : Si Row

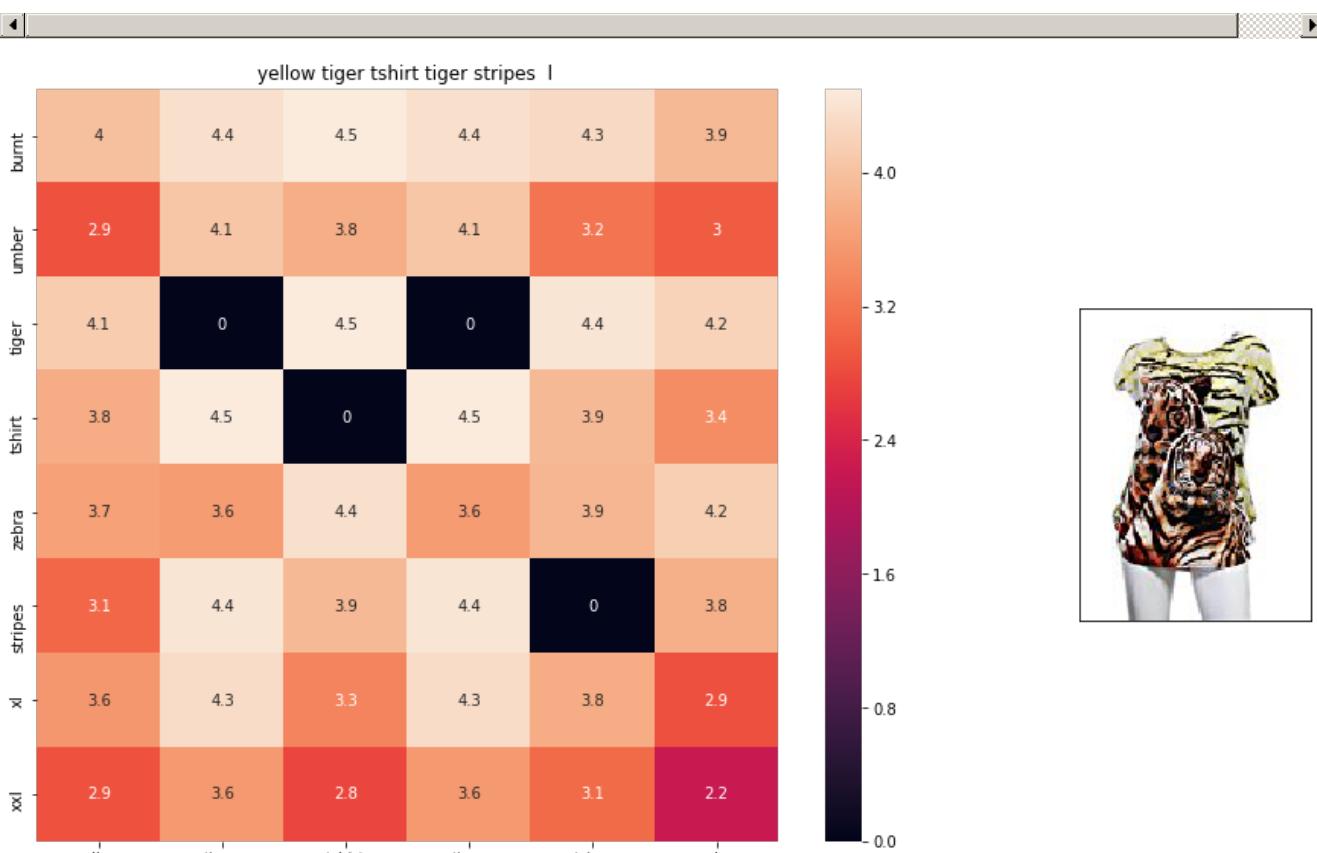
euclidean distance from given input image : 0.700344



ASIN : B00JXQAFZ2

BRAND : Si Row

euclidean distance from given input image : 0.89284



yellow tiger tshirt tiger stripes |

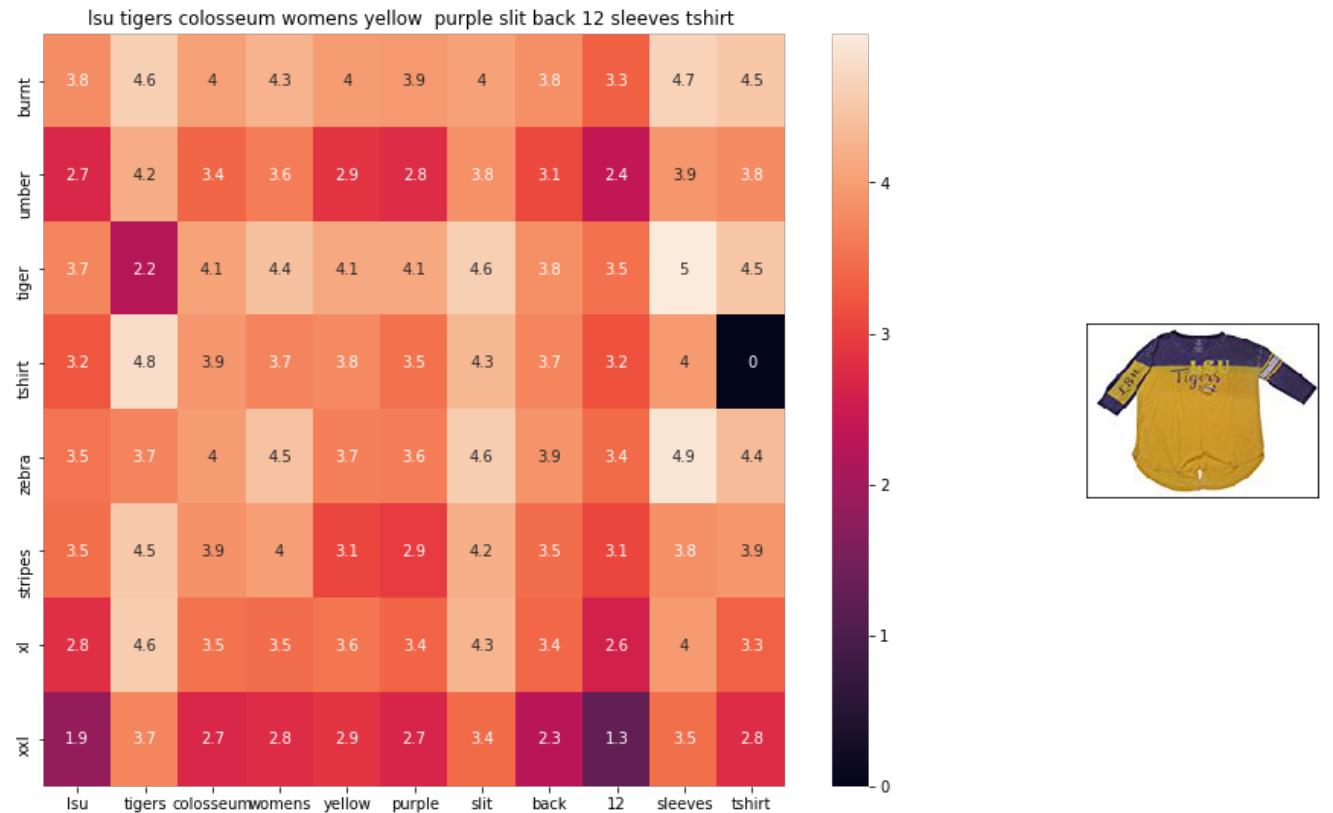
ASIN : B00JXQCUIC

BRAND : Si Row

euclidean distance from given input image : 0.956013

=====

=====



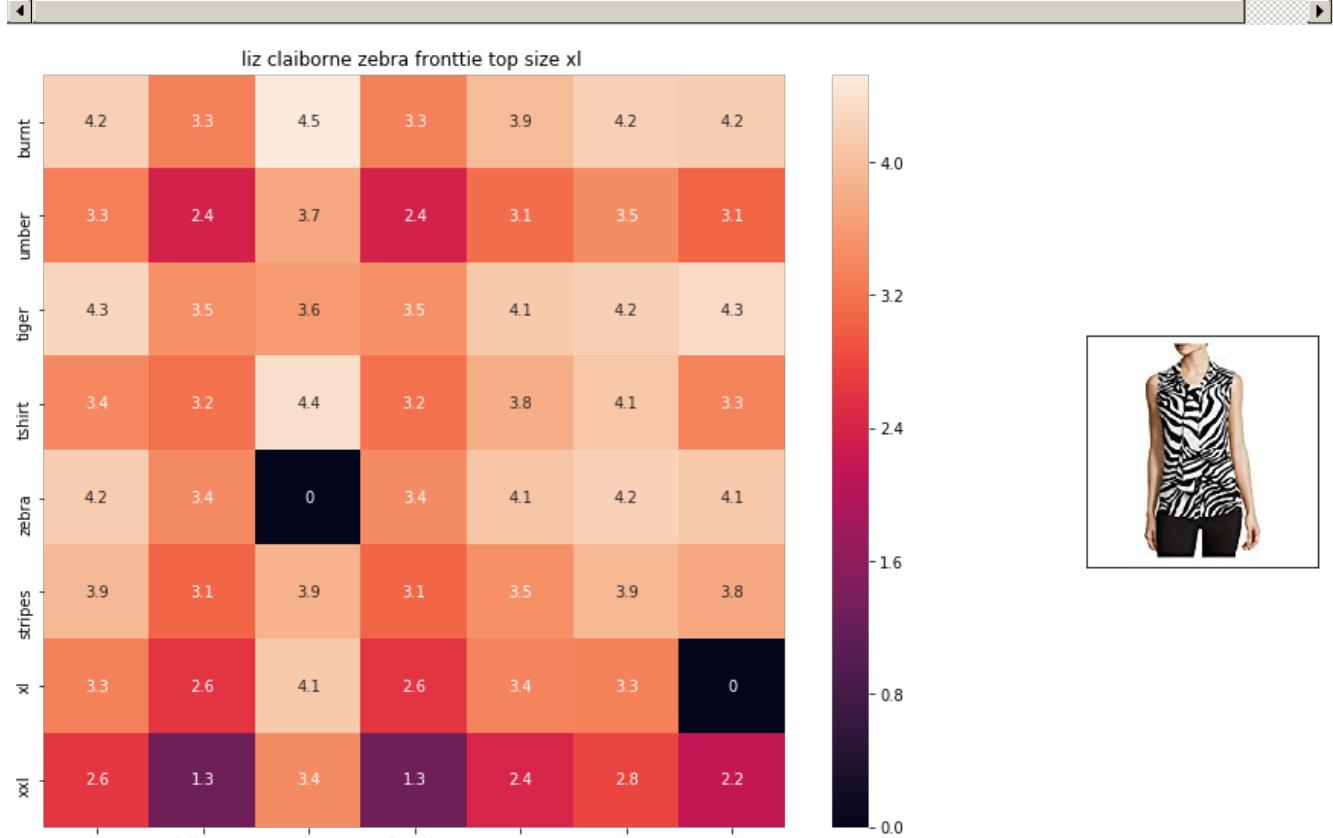
ASIN : B073R5Q8HD

BRAND : Colosseum

euclidean distance from given input image : 1.02297

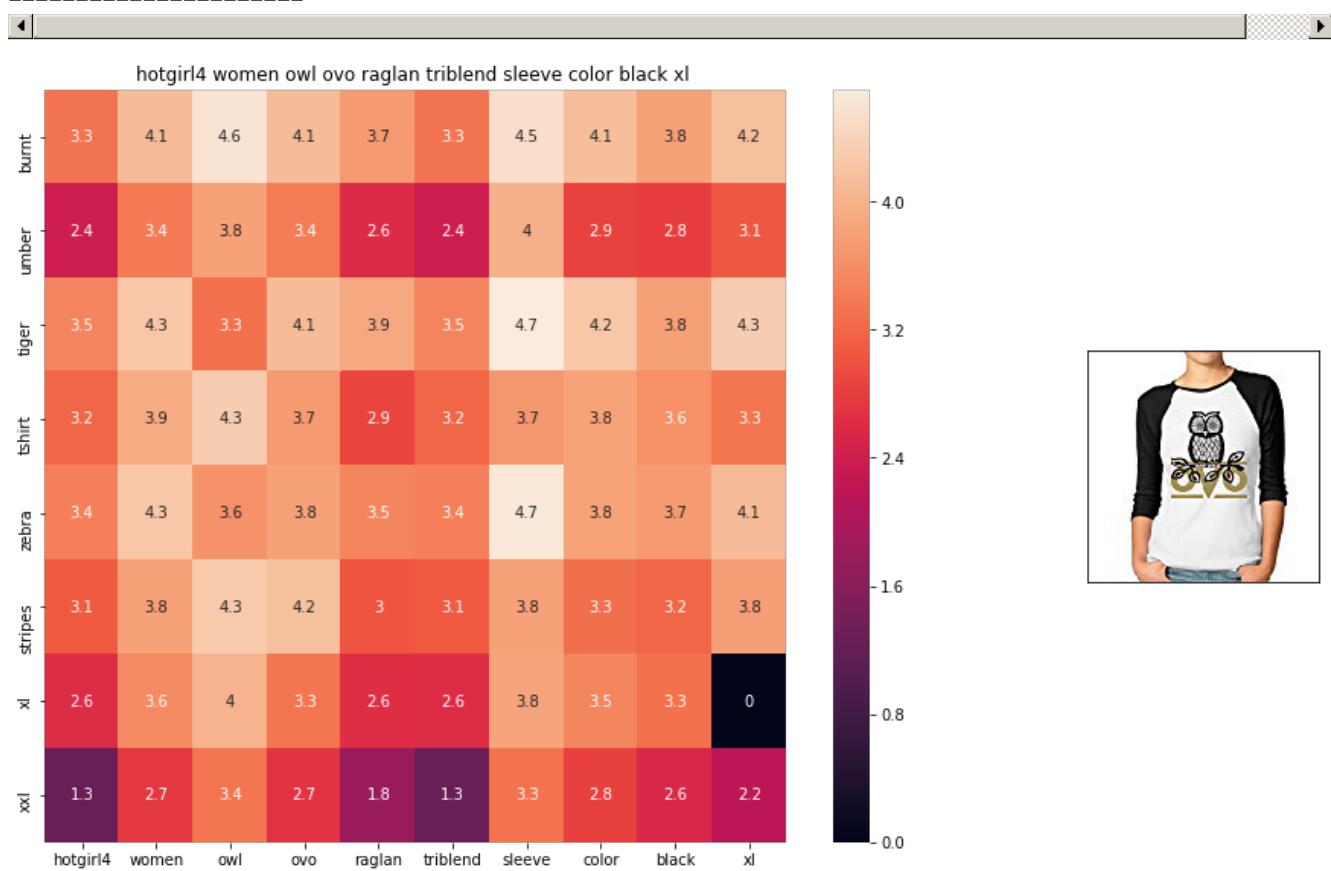
=====

=====

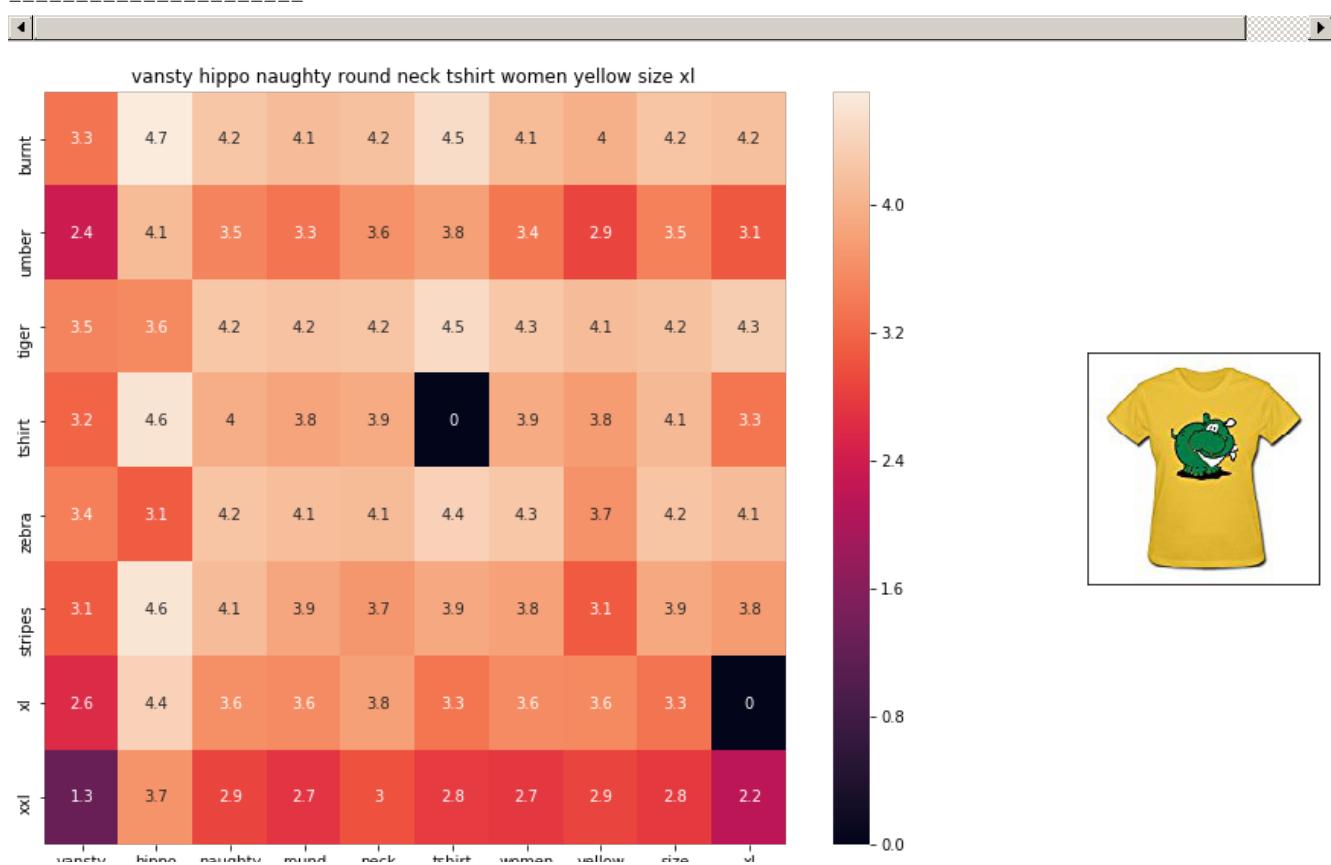


liz clairene zebra fronttie top size xl

ASIN : B06XBY5QXL
BRAND : Liz Claiborne
euclidean distance from given input image : 1.06693



ASIN : B01L8L73M2
BRAND : Hotgirl4 Raglan Design
euclidean distance from given input image : 1.07314

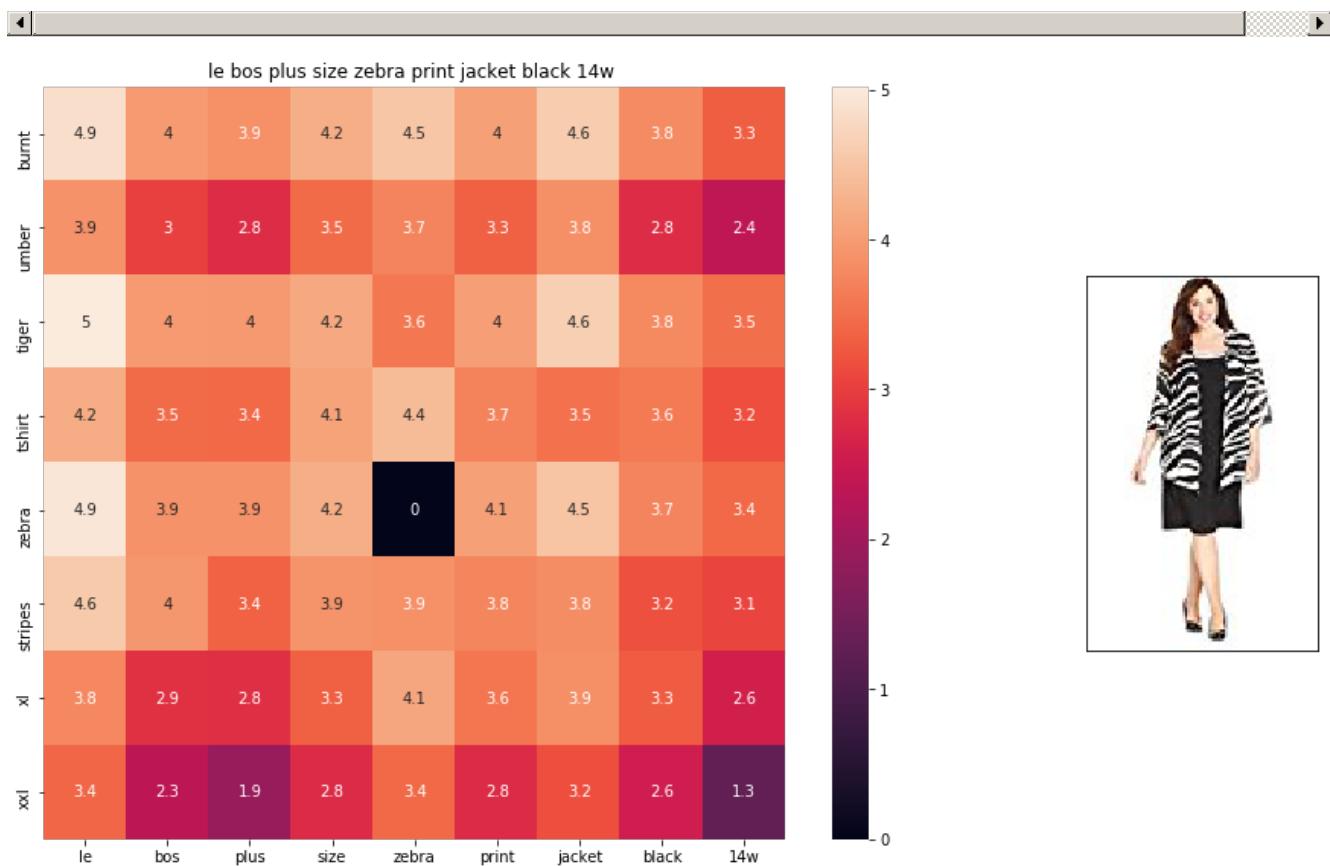


variety nippes naughty round neck shirt women yellow size xl

ASIN : B01EJS5H06

BRAND : Vansty

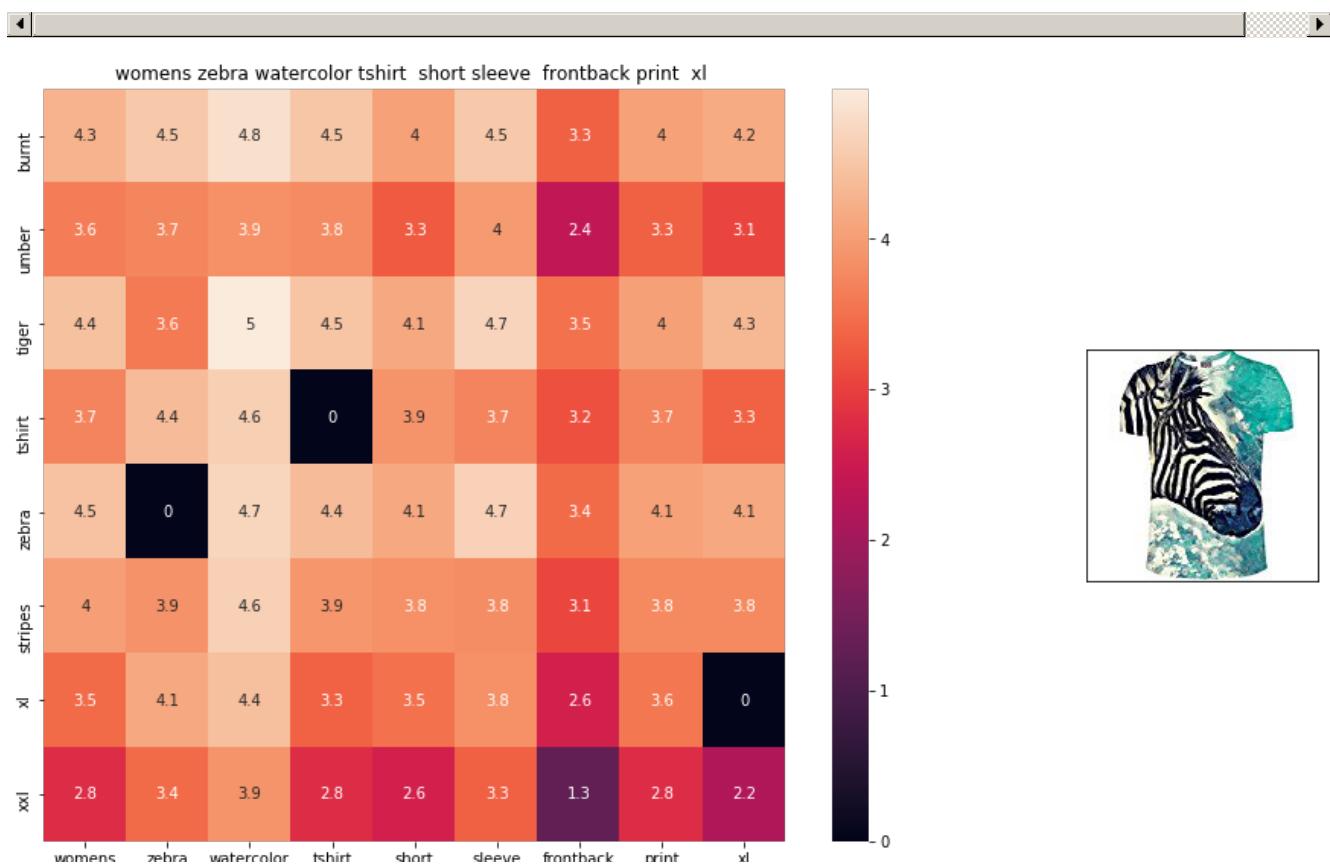
euclidean distance from given input image : 1.07572



ASIN : B01BO1XRK8

BRAND : Le Bos

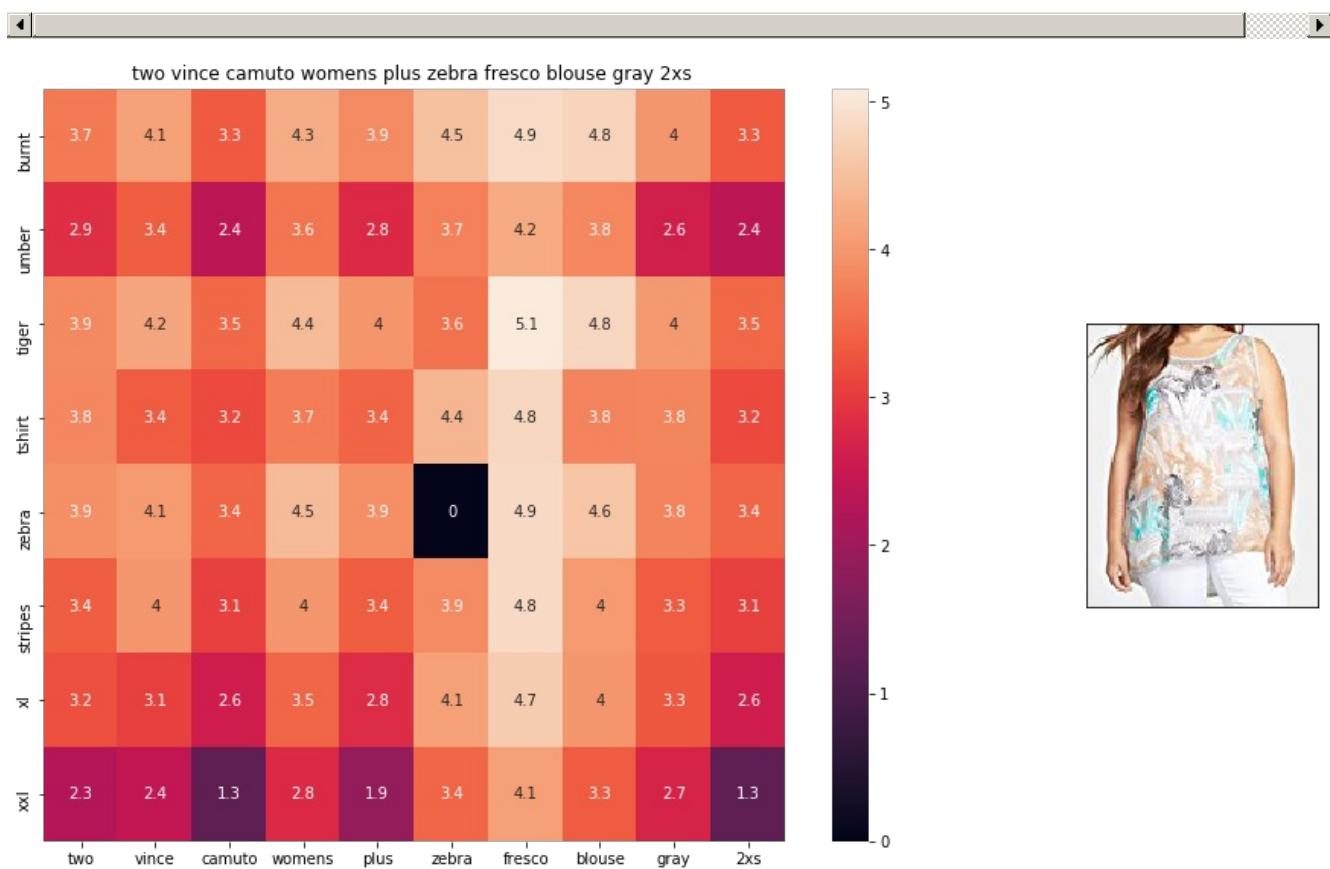
euclidean distance from given input image : 1.084



ASIN : B072R2JXKW

BRAND : WHAT ON EARTH

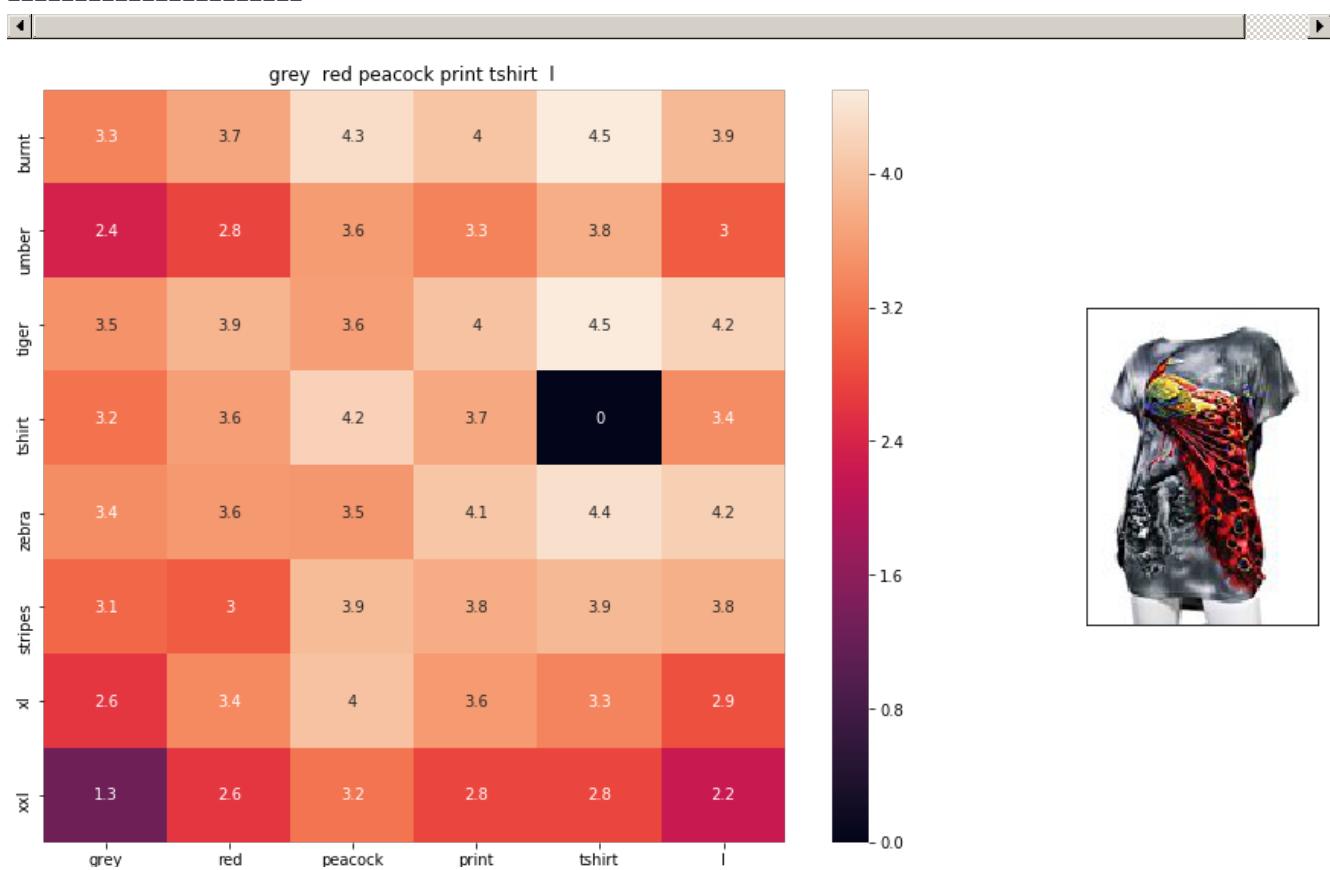
euclidean distance from given input image : 1.08422



ASIN : B074MJRGW6

BRAND : Two by Vince Camuto

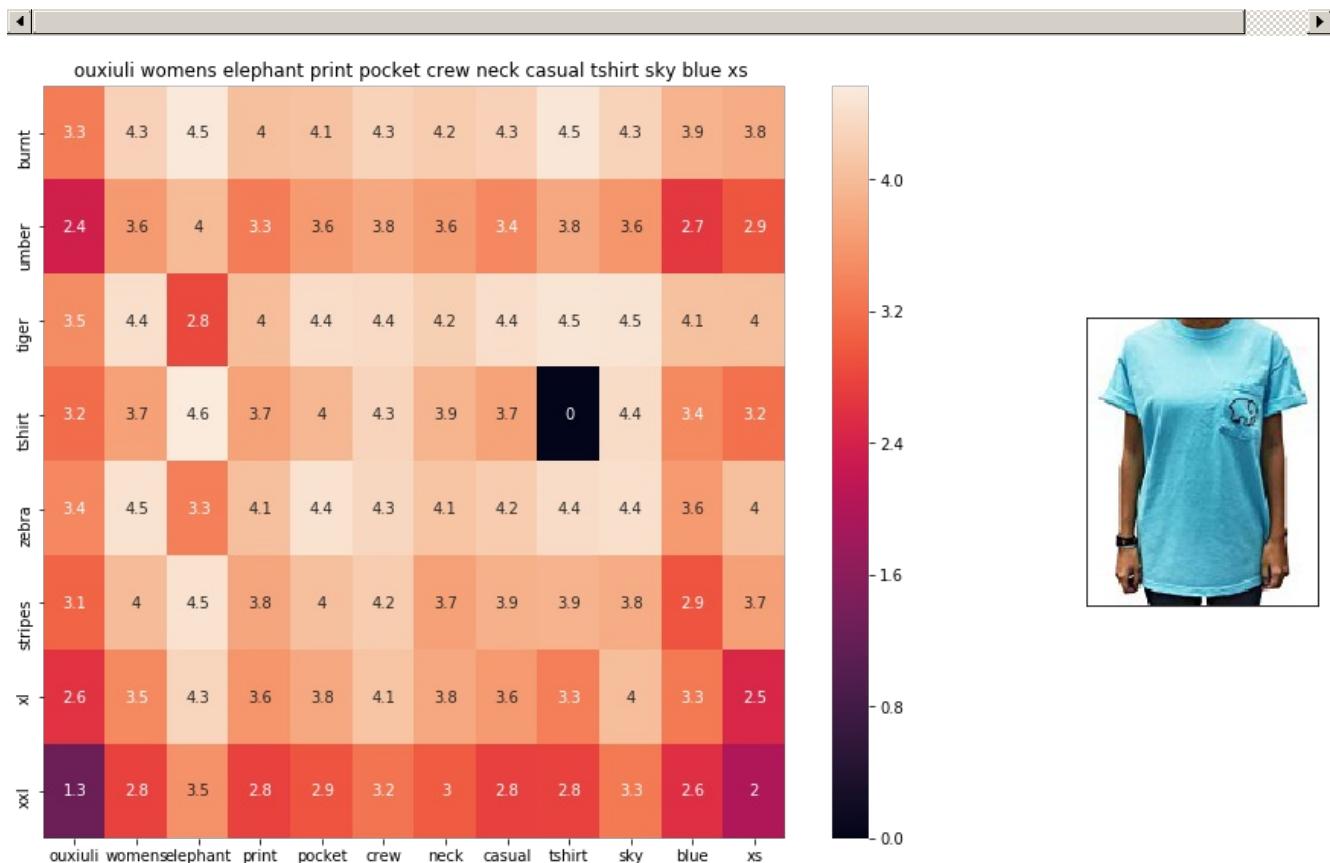
euclidean distance from given input image : 1.0895



ASIN : B00JXQCFRS

BRAND : Si Row

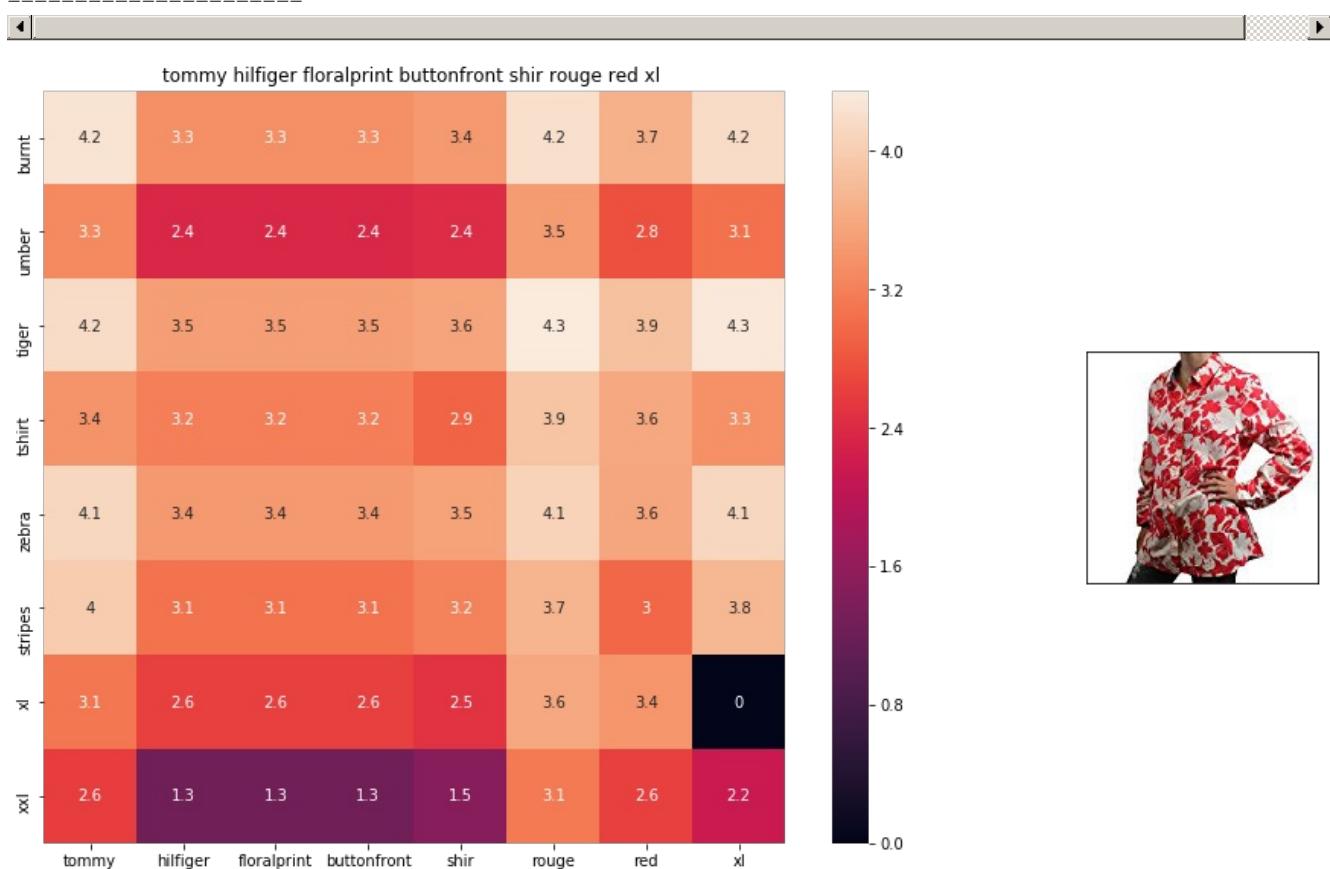
euclidean distance from given input image : 1.09006



ASIN : B01I53HU6K

BRAND : ouxiuli

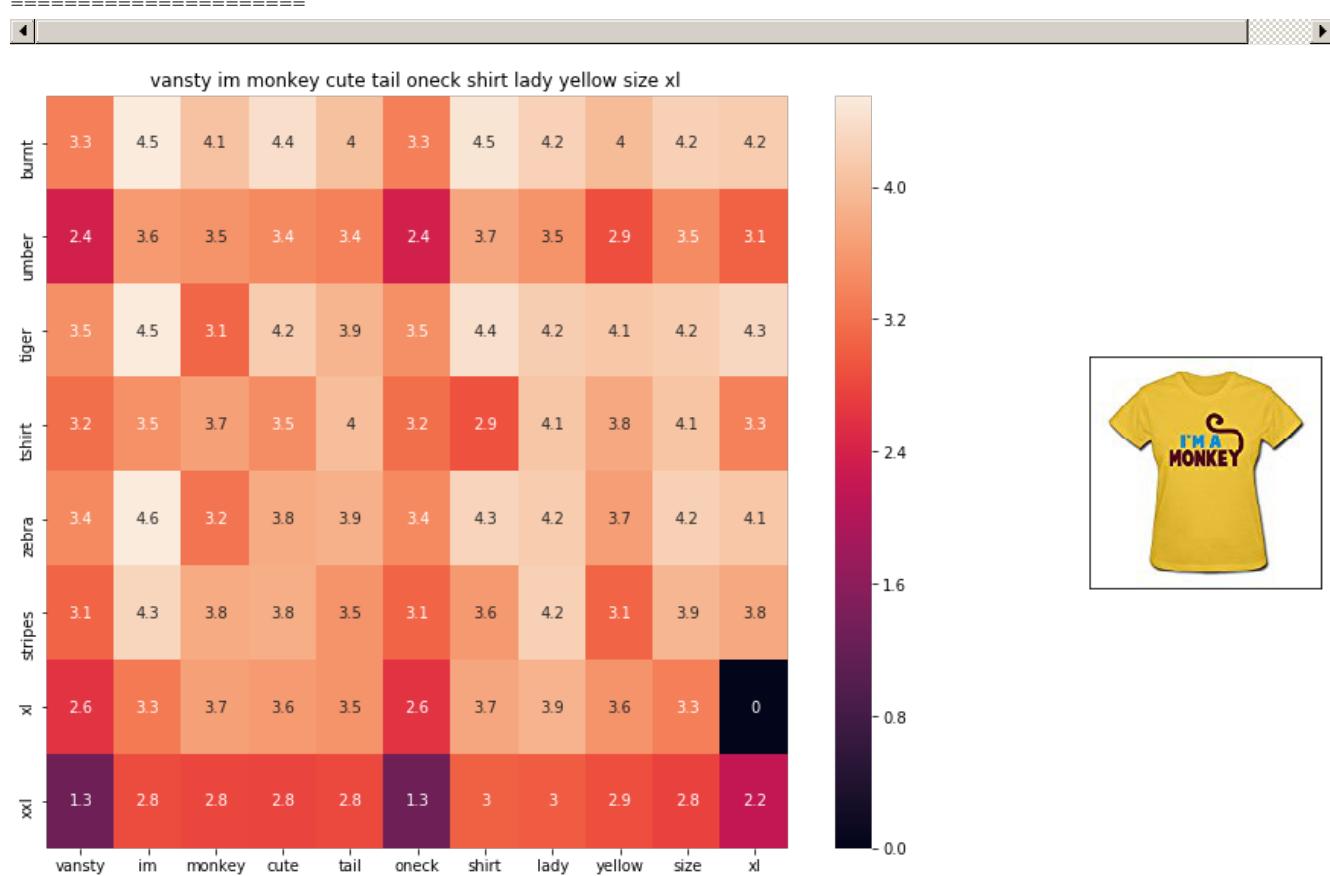
euclidean distance from given input image : 1.09201



ASIN : B0711NGTQM

BRAND : THILFIGER RTW

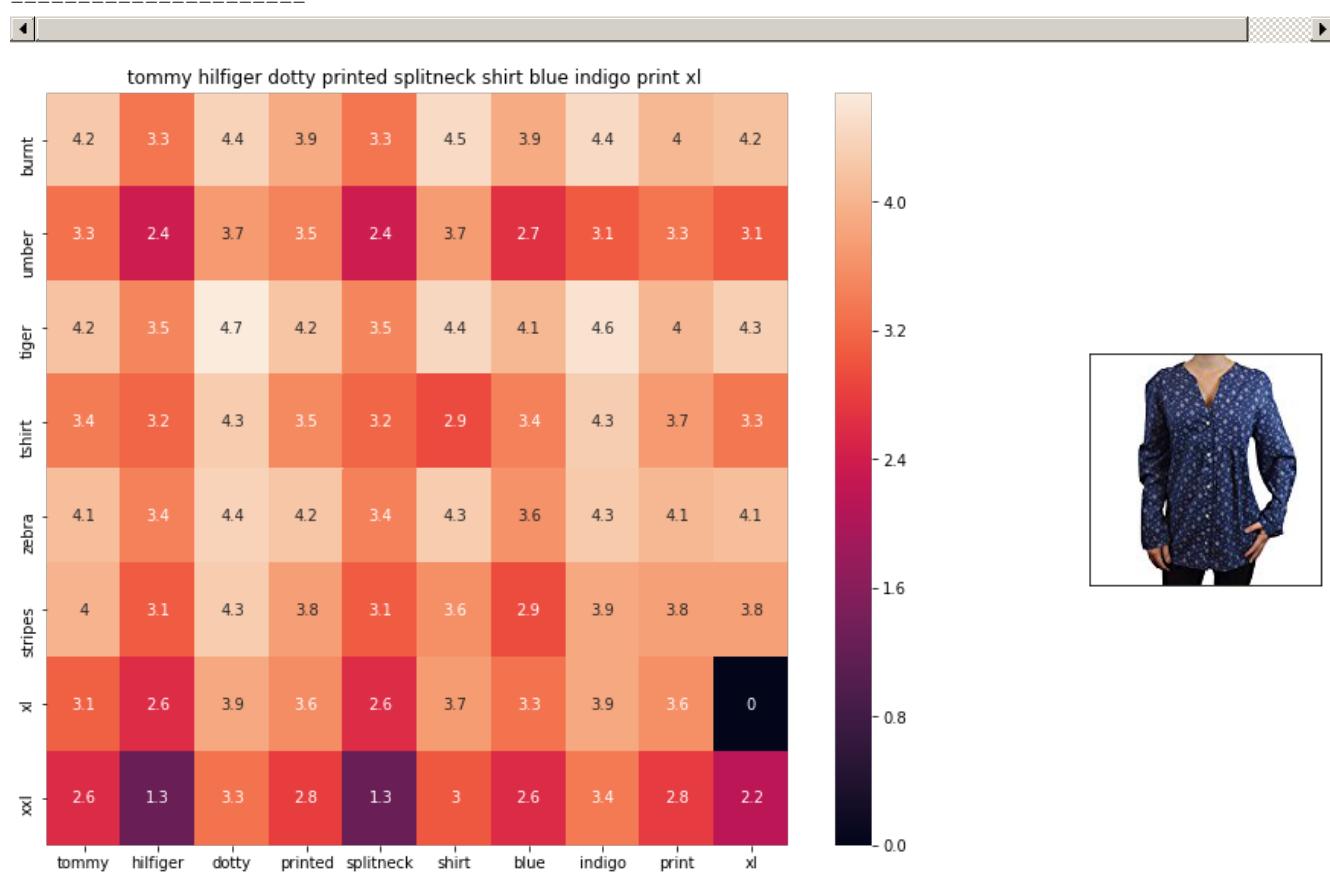
euclidean distance from given input image : 1.09234



ASIN : B01EFSLO8Y

BRAND : Vansty

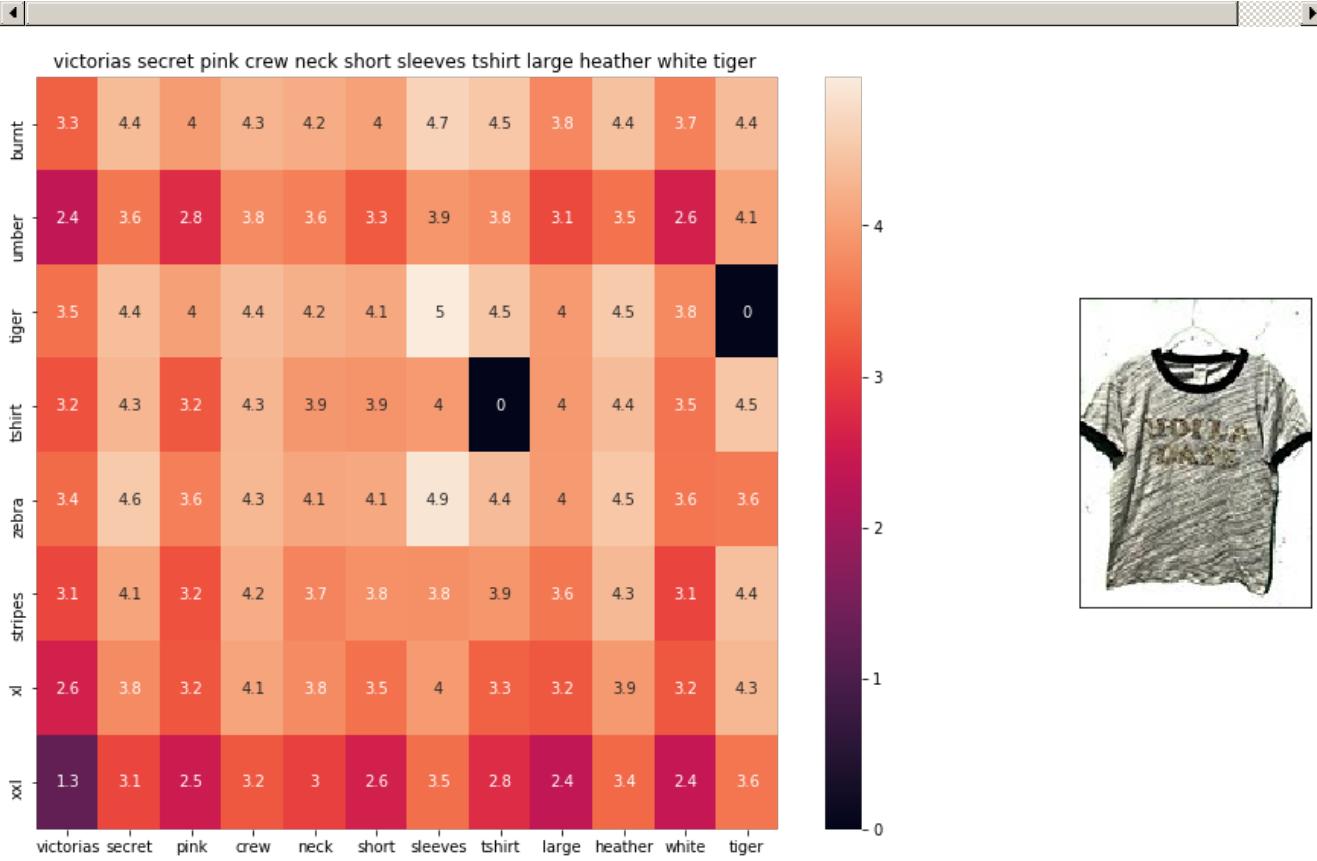
euclidean distance from given input image : 1.0934



ASIN : B0716TVWQ4

BRAND : THILFIGER RTW

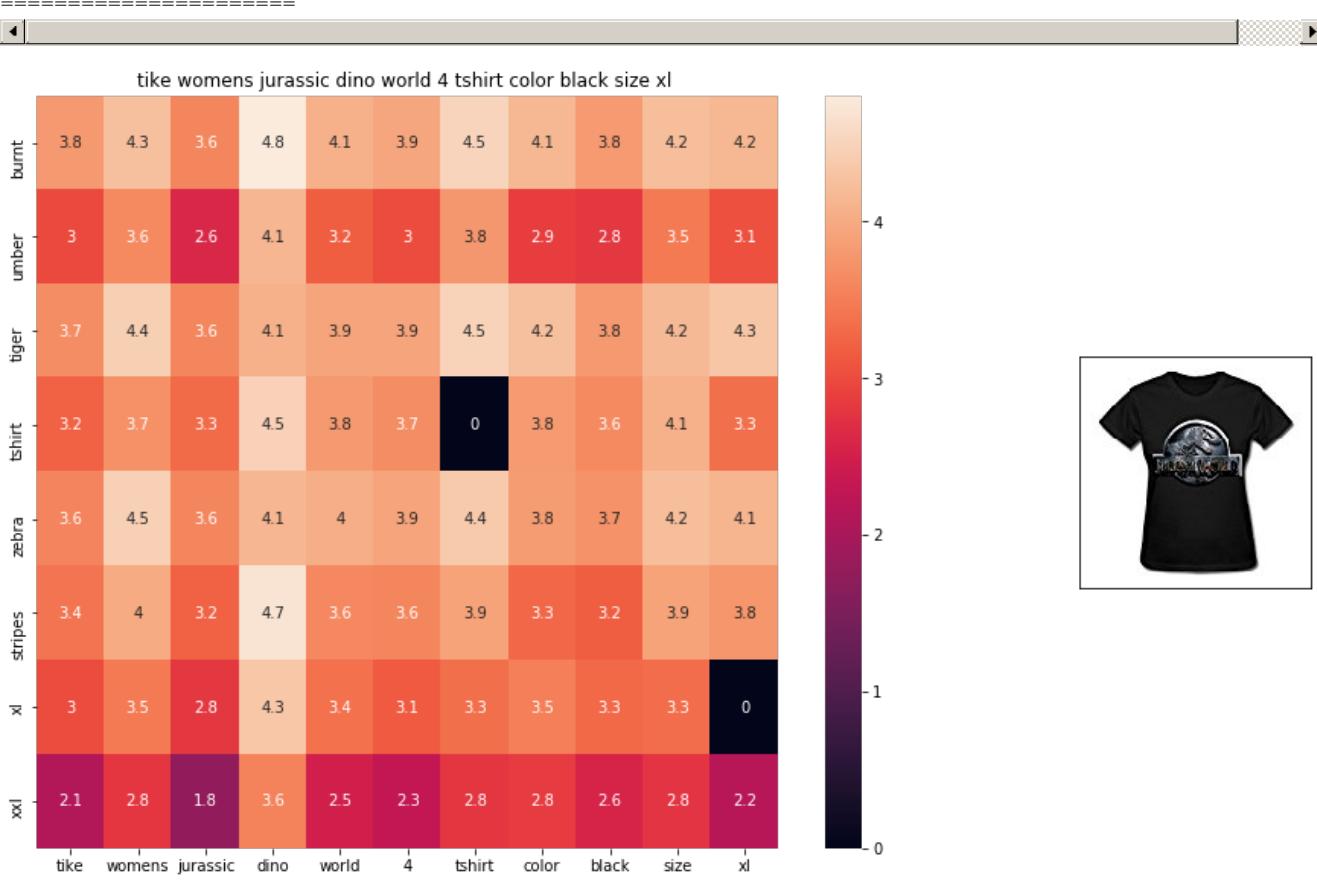
euclidean distance from given input image : 1.0942



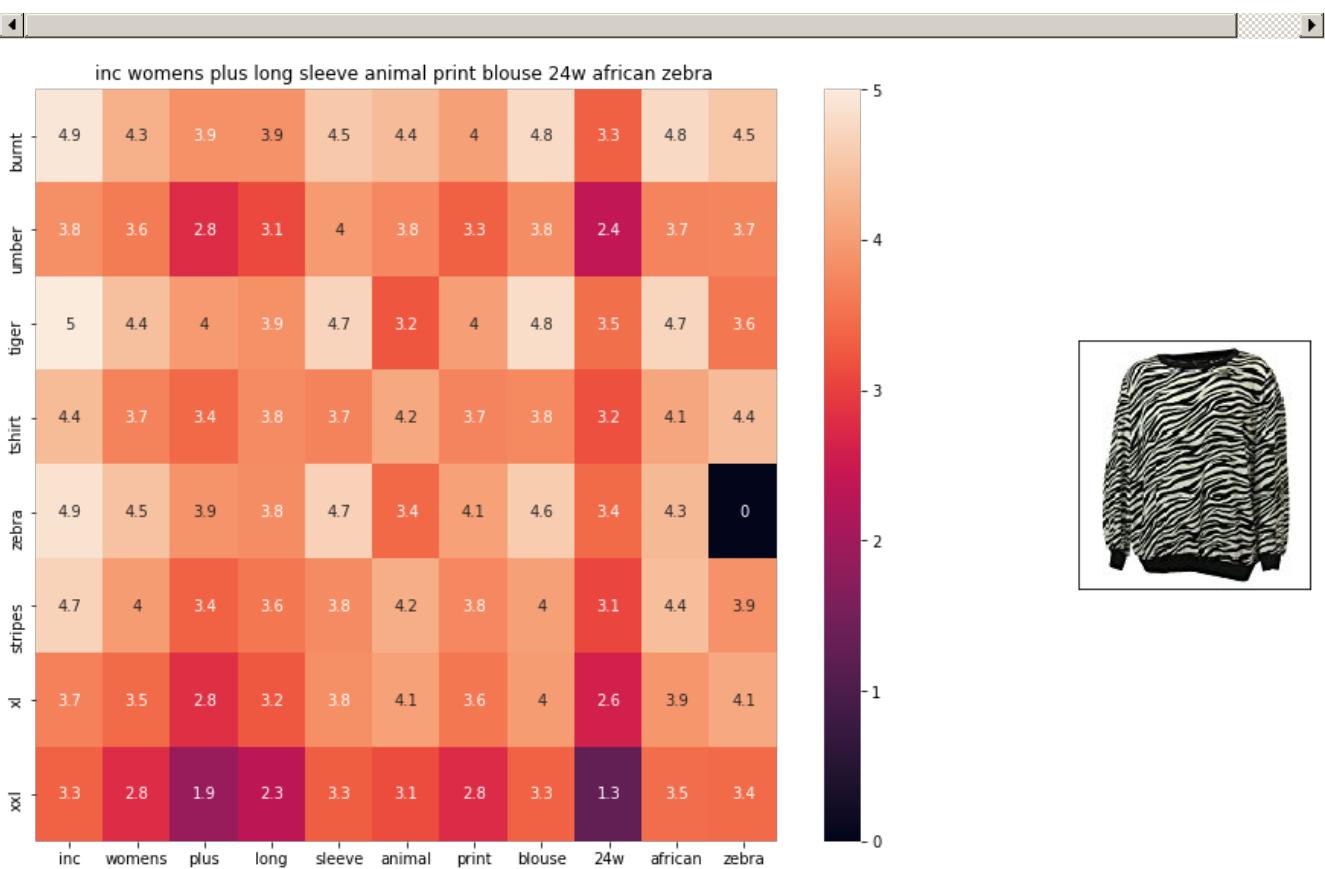
ASIN : B0716MVPGV

BRAND : V.Secret

euclidean distance from given input image : 1.09483



ASIN : B016OPN40I
 BRAND : TIKE Fashions
 euclidean distance from given input image : 1.09513



ASIN : B018WDJCUA
 BRAND : INC - International Concepts Woman
 euclidean distance from given input image : 1.09669

[9.4] IDF weighted Word2Vec for product similarity

In [15]:

```
doc_id = 0
w2v_title_weight = []
# for every title we build a weighted vector representation
for i in data['title']:
    w2v_title_weight.append(build_avg_vec(i, 300, doc_id, 'weighted'))
    doc_id += 1
# w2v_title = np.array(# number of doc in courpus * 300), each row corresponds to a doc
w2v_title_weight = np.array(w2v_title_weight)
```

In [0]:

```
def weighted_w2v_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 apparels
    # the metric we used here is cosine, the coside distance is mesured as K(X, Y) = <X, Y> / (||X|| * ||Y||)
    # http://scikit-learn.org/stable/modules/metrics.html#cosine-similarity
    pairwise_dist = pairwise_distances(w2v_title_weight, w2v_title_weight[doc_id].reshape(1,-1))

    # np.argsort will return indices of 9 smallest distances
    indices = np.argsort(pairwise_dist.flatten())[0:num_results]
    #pdists will store the 9 smallest distances
    pdists = np.sort(pairwise_dist.flatten())[0:num_results]
```

```

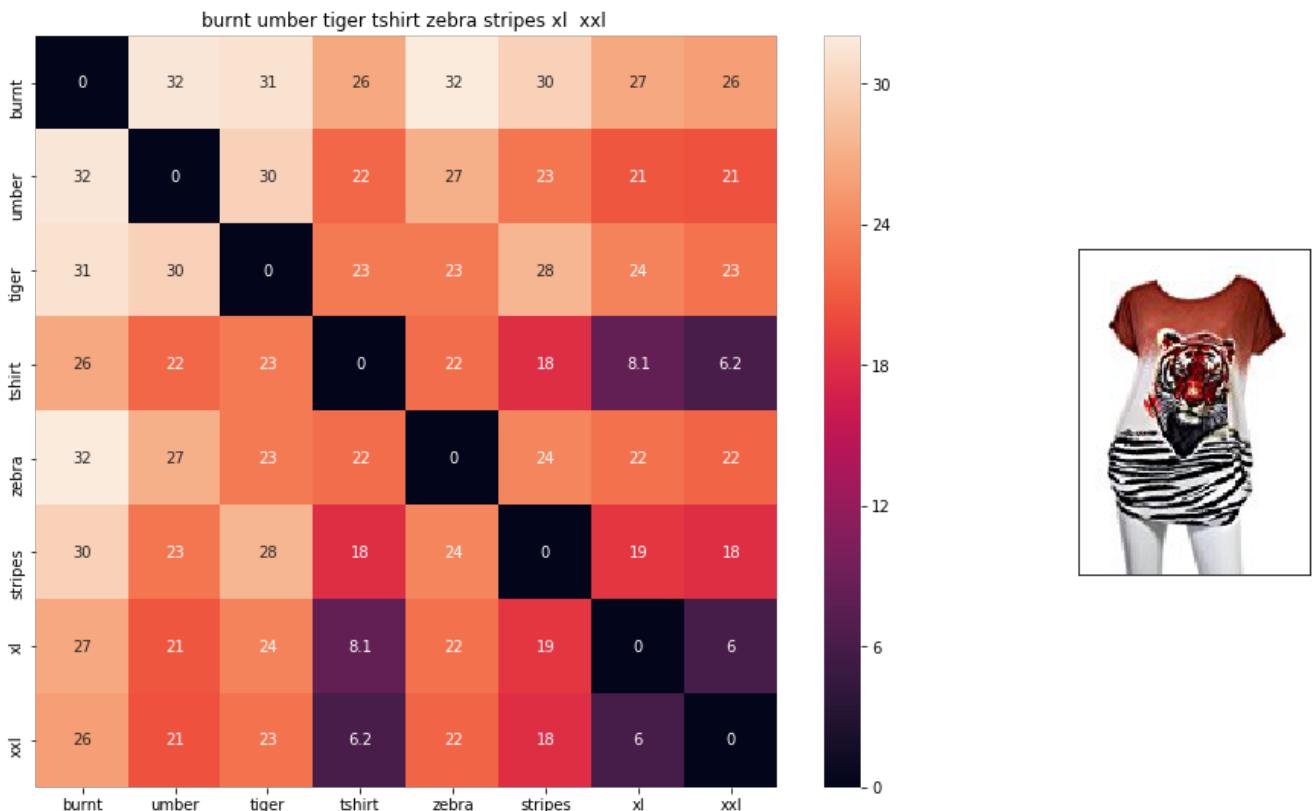
pdistss = np.sqrt(pairwise_euclidean() / len(results))

#data frame indices of the 9 smallest distance's
df_indices = list(data.index[indices])

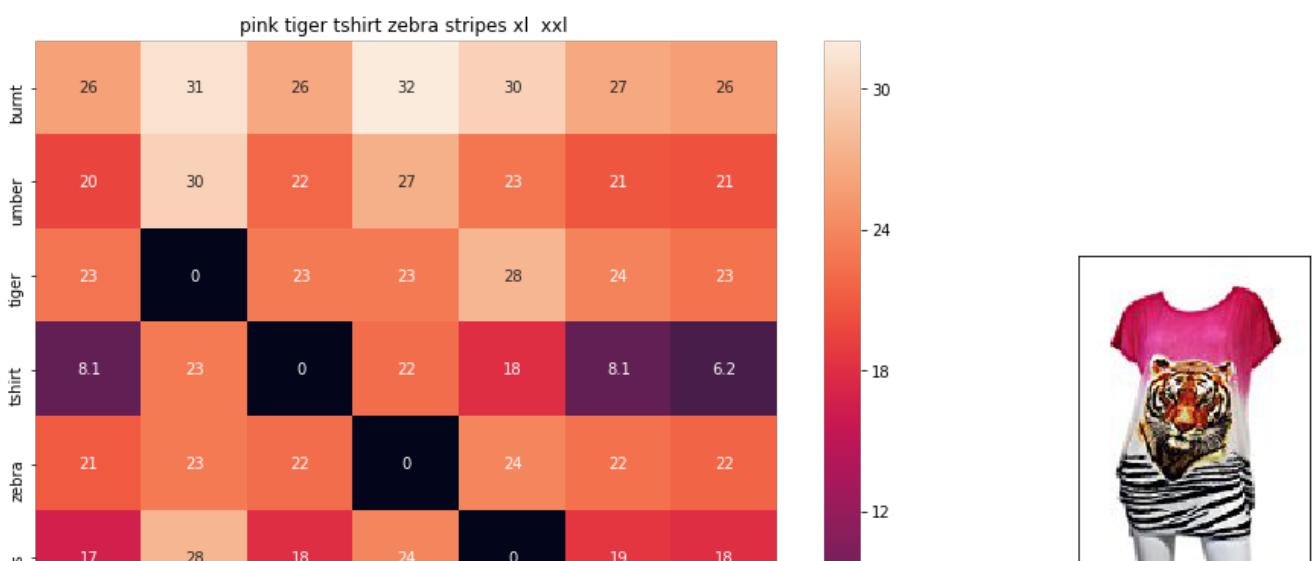
for i in range(0, len(indices)):
    heat_map_w2v(data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], indices[0], indices[i], 'weighted')
    print('ASIN :', data['asin'].loc[df_indices[i]])
    print('Brand :', data['brand'].loc[df_indices[i]])
    print('euclidean distance from input :', pdists[i])
    print('='*125)

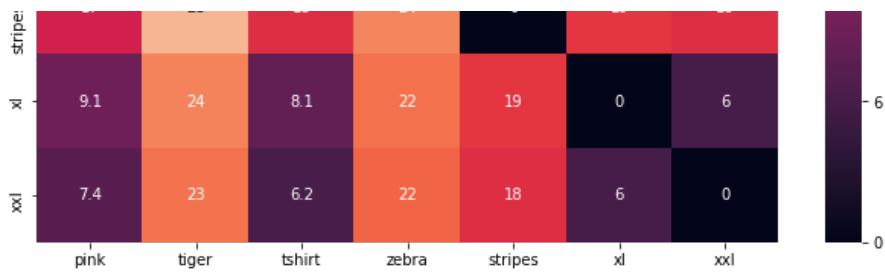
weighted_w2v_model(12566, 20)
#931
#12566
# in the give heat map, each cell contains the euclidean distance between words i, j

```



ASIN : B00JXQB5FQ
 Brand : Si Row
 euclidean distance from input : 0.00390625

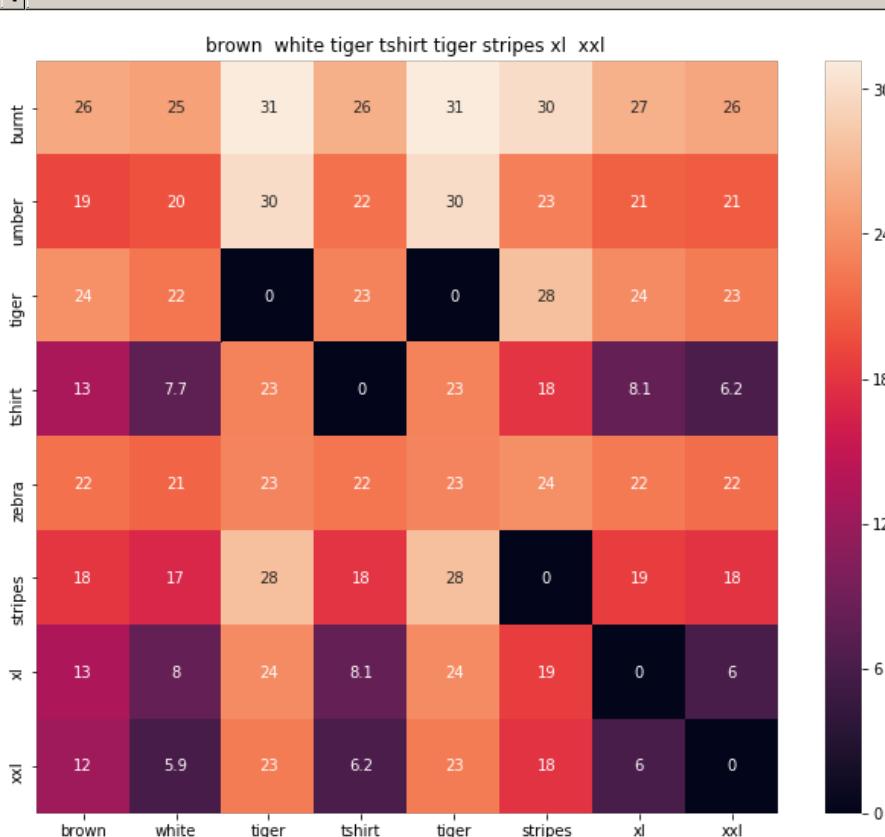




ASIN : B00JXQASS6

Brand : Si Row

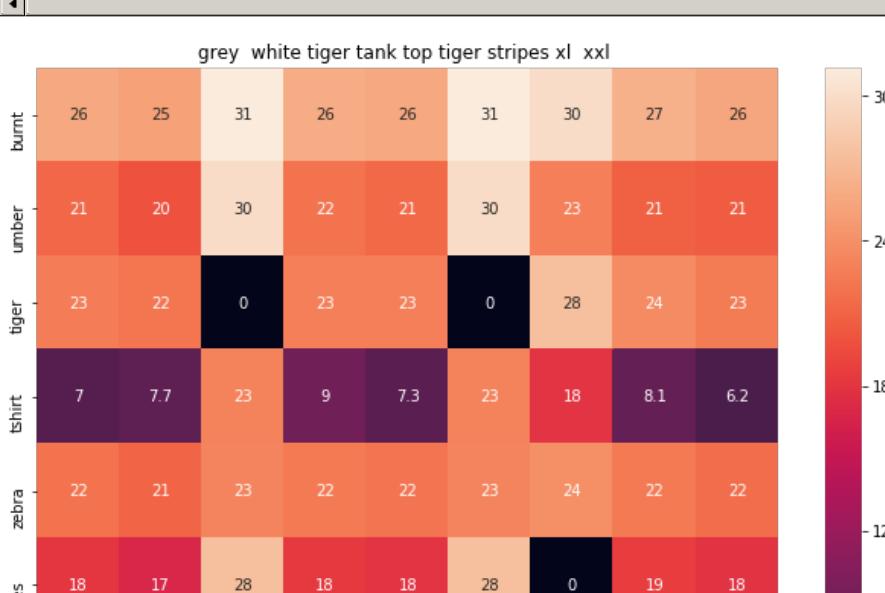
euclidean distance from input : 4.06389

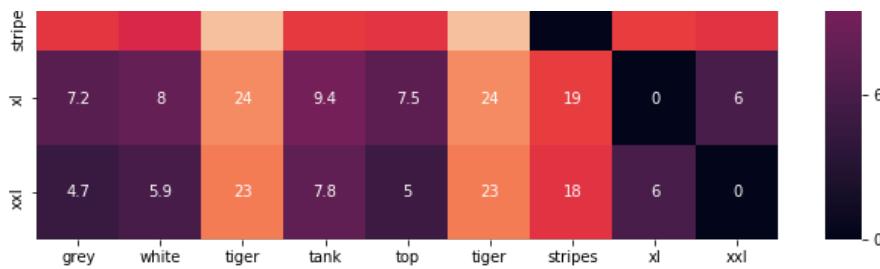


ASIN : B00JXQCWTO

Brand : Si Row

euclidean distance from input : 4.77094

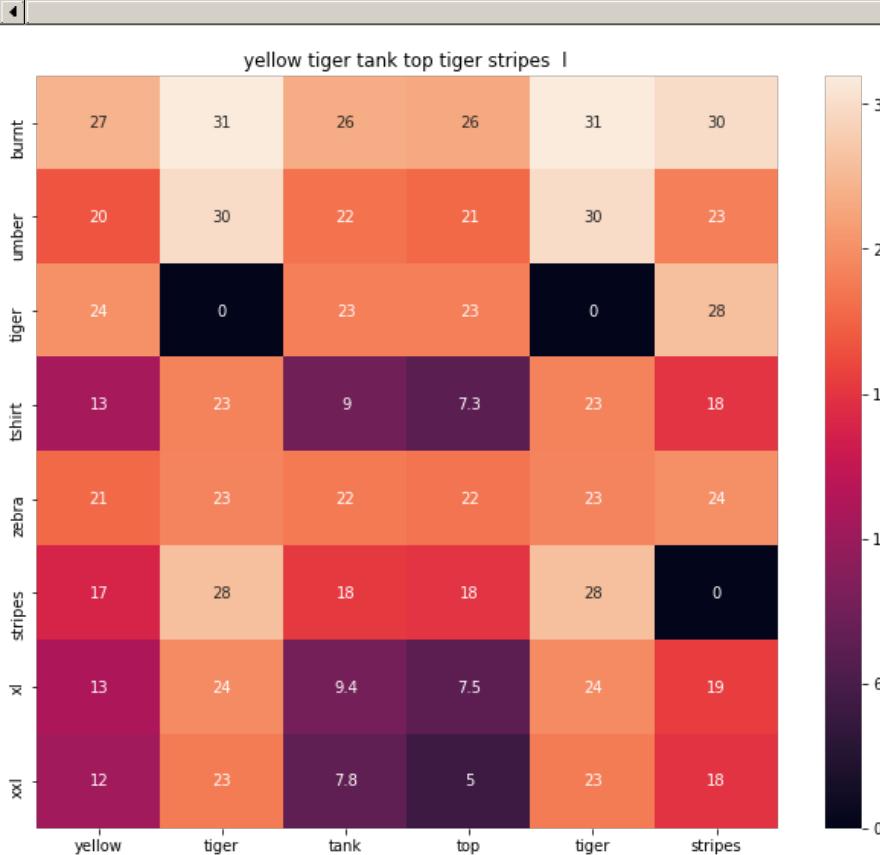




ASIN : B00JXQAFZ2

Brand : Si Row

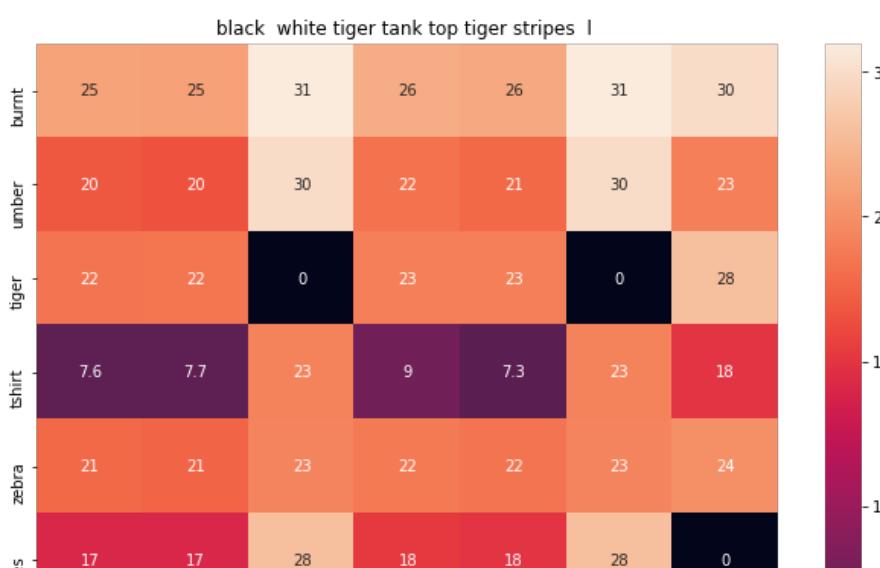
euclidean distance from input : 5.36016

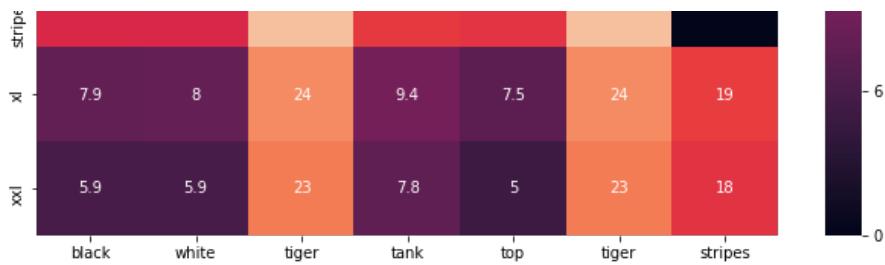


ASIN : B00JXQAUWA

Brand : Si Row

euclidean distance from input : 5.68952

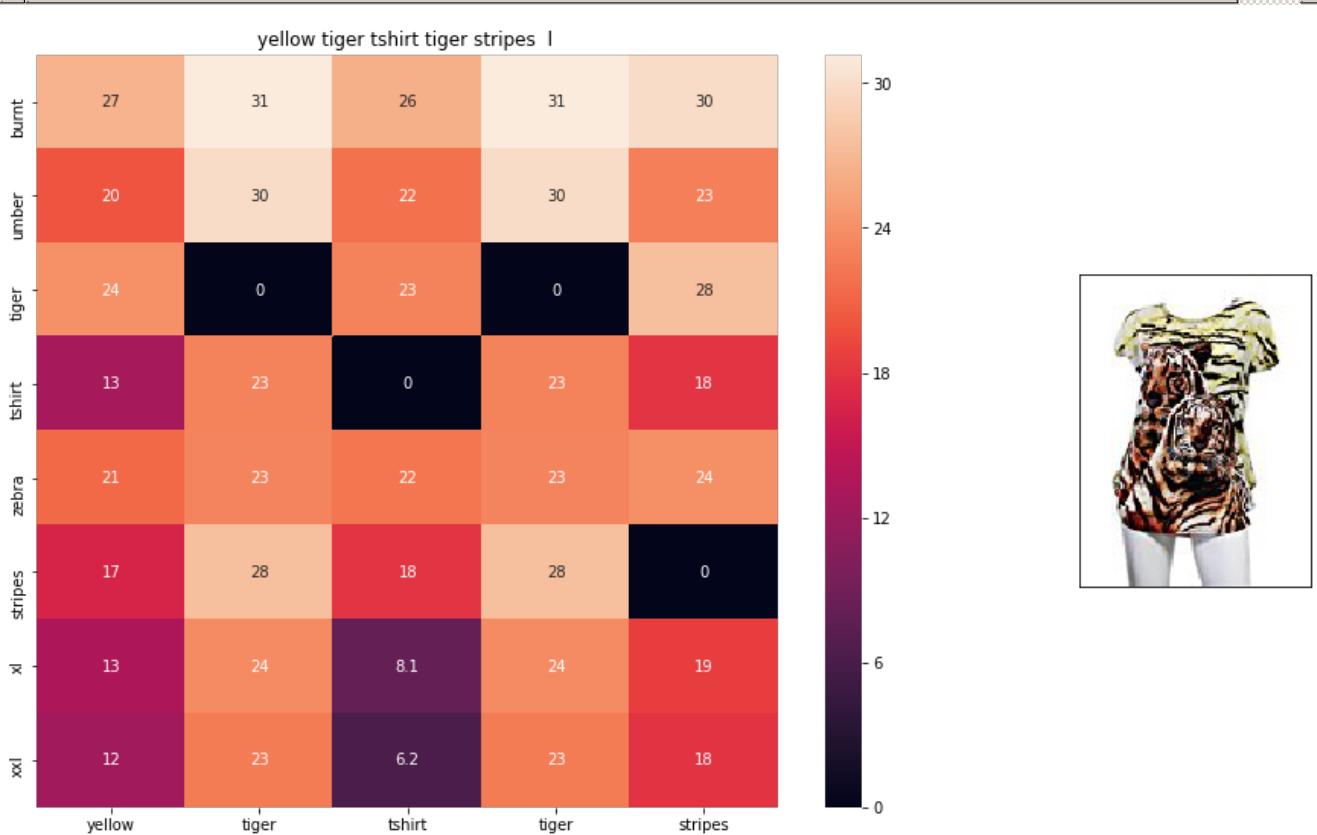




ASIN : B00JXQAO94

Brand : Si Row

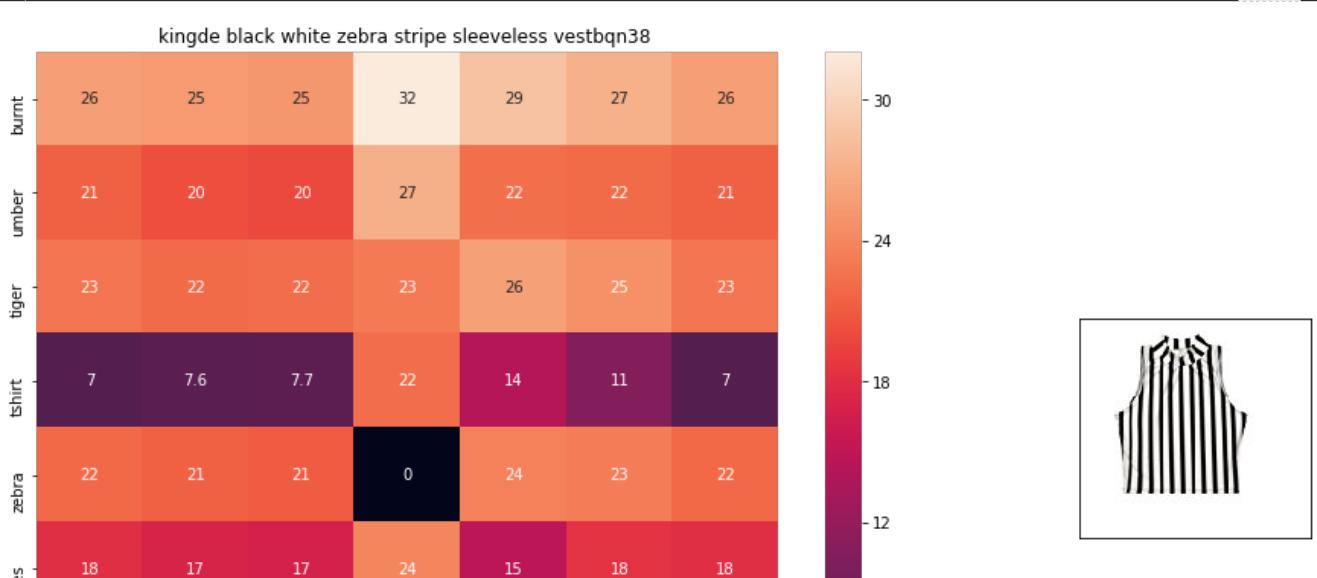
euclidean distance from input : 5.69302

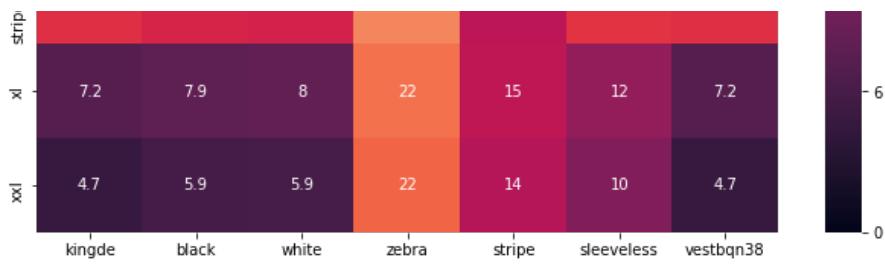


ASIN : B00JXQCUIC

Brand : Si Row

euclidean distance from input : 5.89344

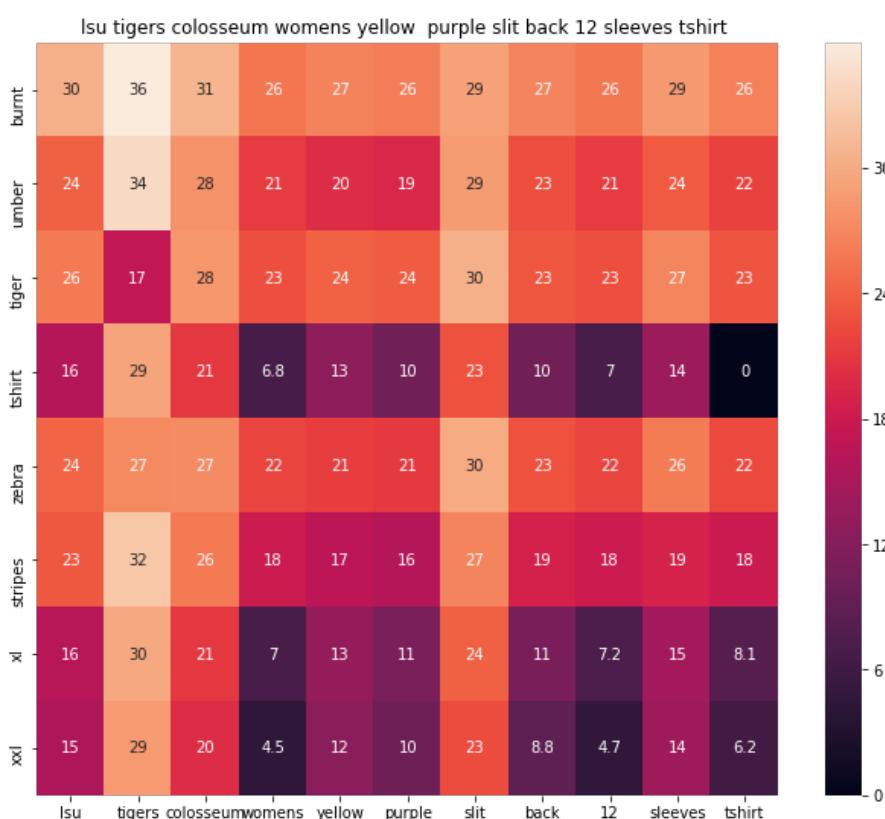




ASIN : B015H41F6G

Brand : KINGDE

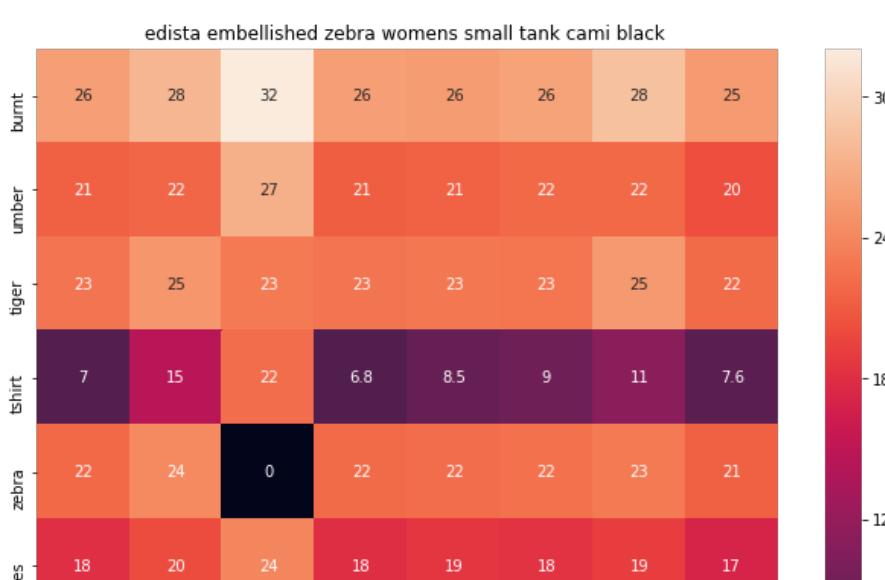
euclidean distance from input : 6.13299

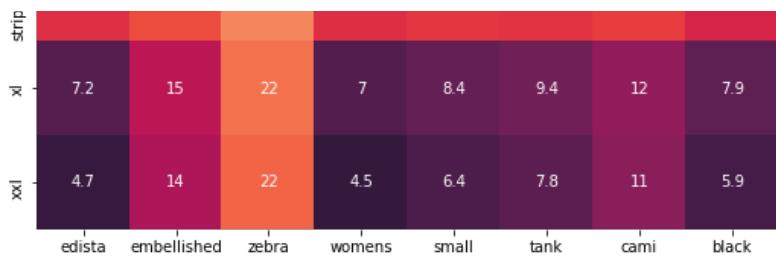


ASIN : B073R5Q8HD

Brand : Colosseum

euclidean distance from input : 6.25671





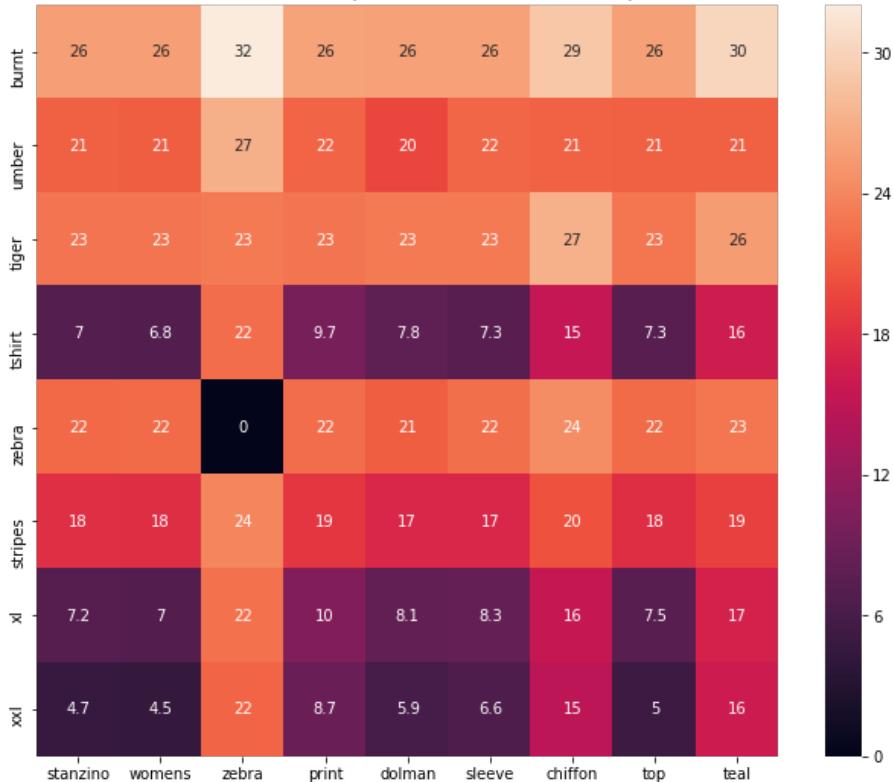
ASIN : B074P8MD22

Brand : Edista

euclidean distance from input : 6.3922



stanzino womens zebra print dolman sleeve chiffon top teal



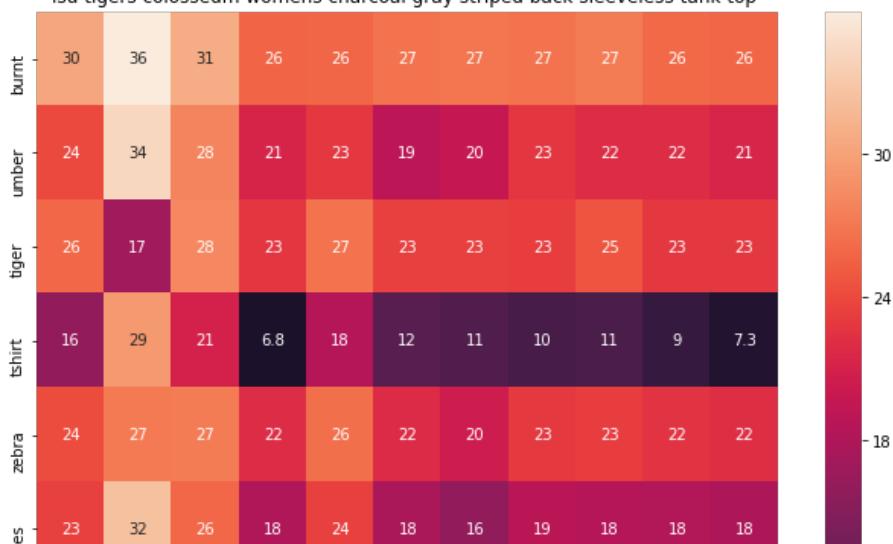
ASIN : B00C0I3U3E

Brand : Stanzino

euclidean distance from input : 6.4149



lsu tigers colosseum womens charcoal gray striped back sleeveless tank top





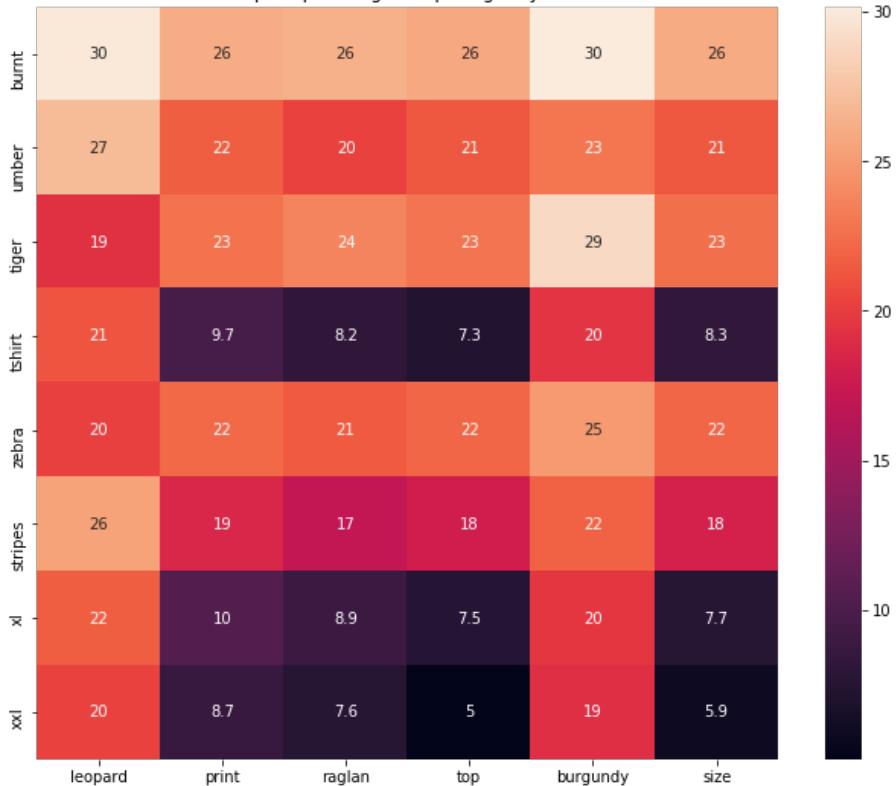
ASIN : B073R4ZM7Y

Brand : Colosseum

euclidean distance from input : 6.45096



leopard print raglan top burgundy size



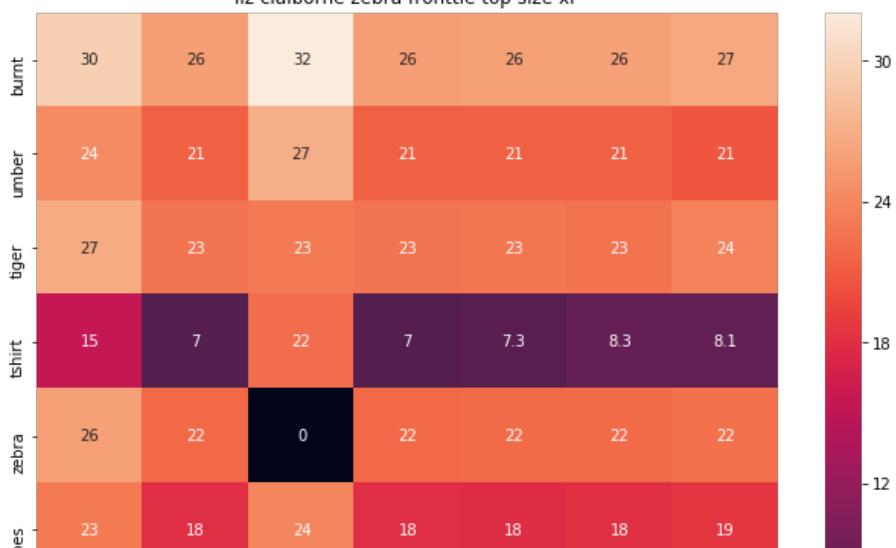
ASIN : B01C60RLDQ

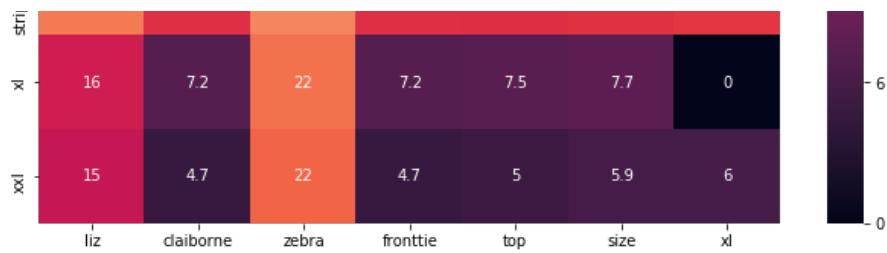
Brand : 1 Mad Fit

euclidean distance from input : 6.46341



liz claiborne zebra fronttie top size xl





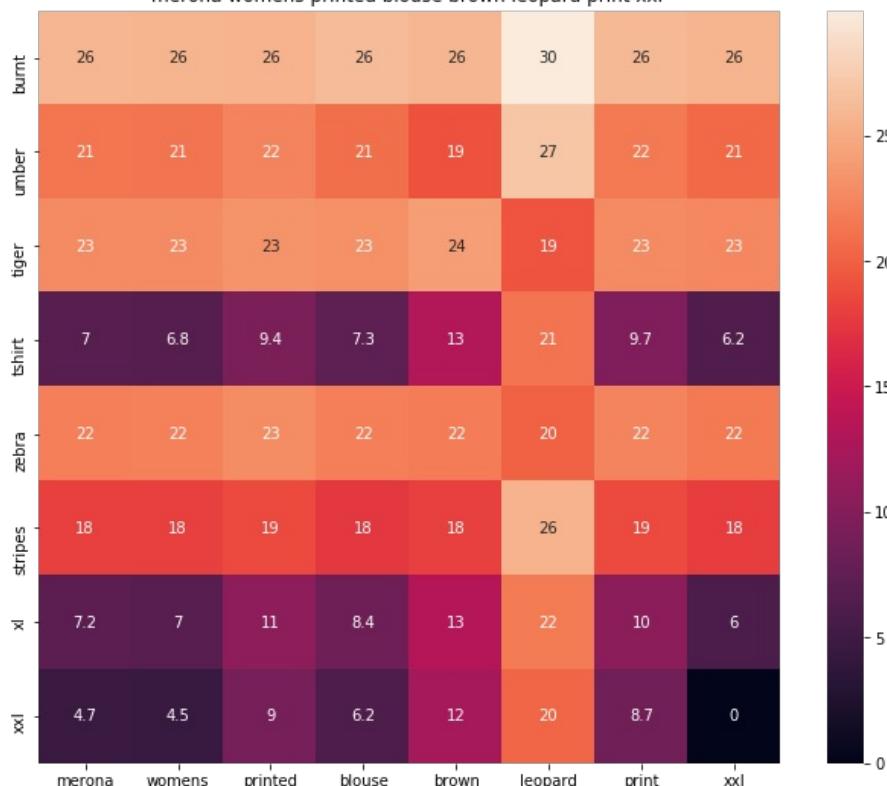
ASIN : B06XBY5QXL

Brand : Liz Claiborne

euclidean distance from input : 6.53922



merona womens printed blouse brown leopard print xxl



ASIN : B071YF3WDD

Brand : Merona

euclidean distance from input : 6.5755



top zebra print dolman sleeve top one size

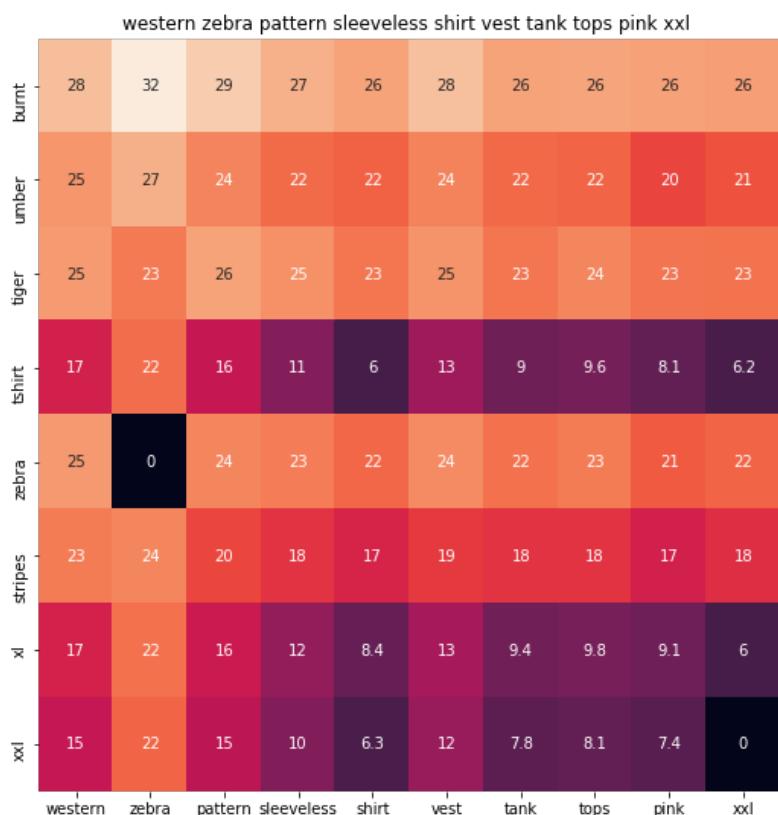




ASIN : B00H8A6ZLI

Brand : Vivian's Fashions

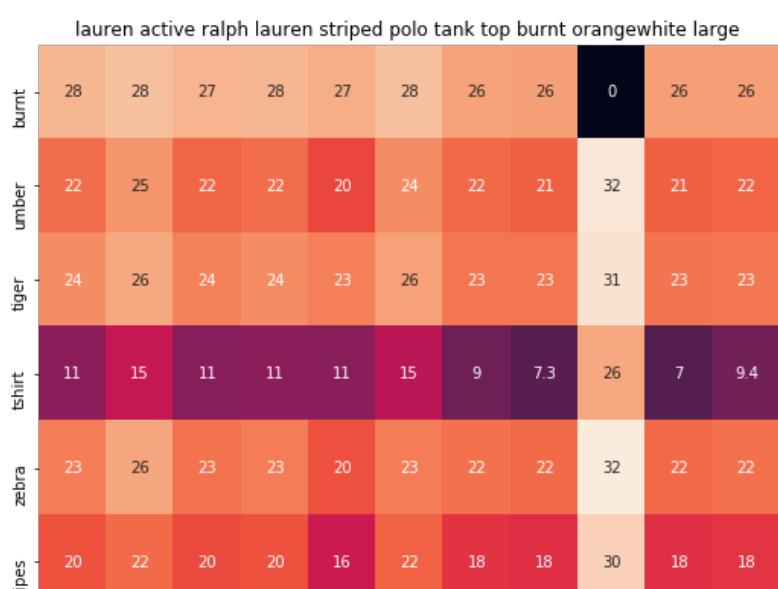
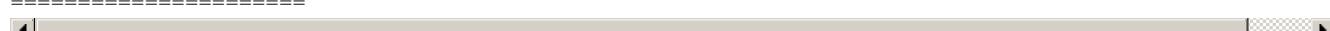
euclidean distance from input : 6.63821

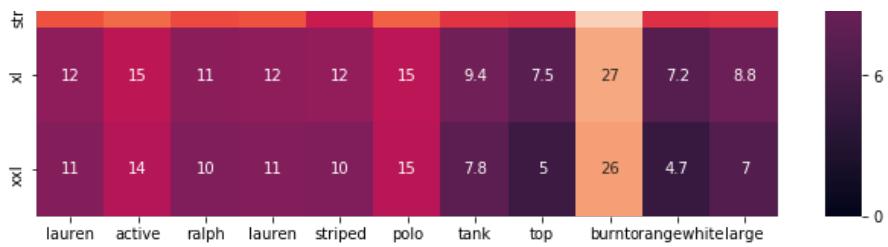


ASIN : B00Z6HEXWI

Brand : Black Temptation

euclidean distance from input : 6.66074

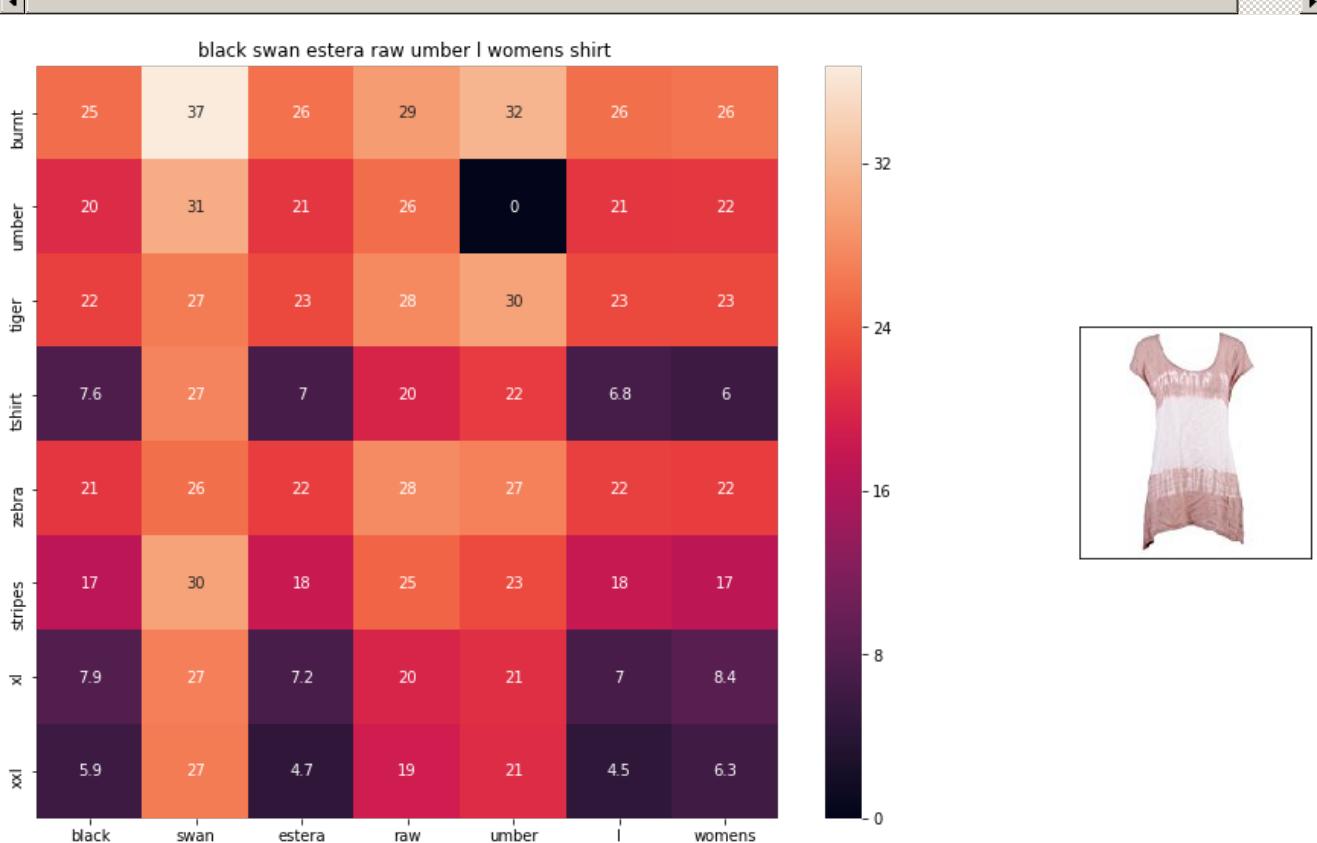




ASIN : B00ILGH5OY

Brand : Ralph Lauren Active

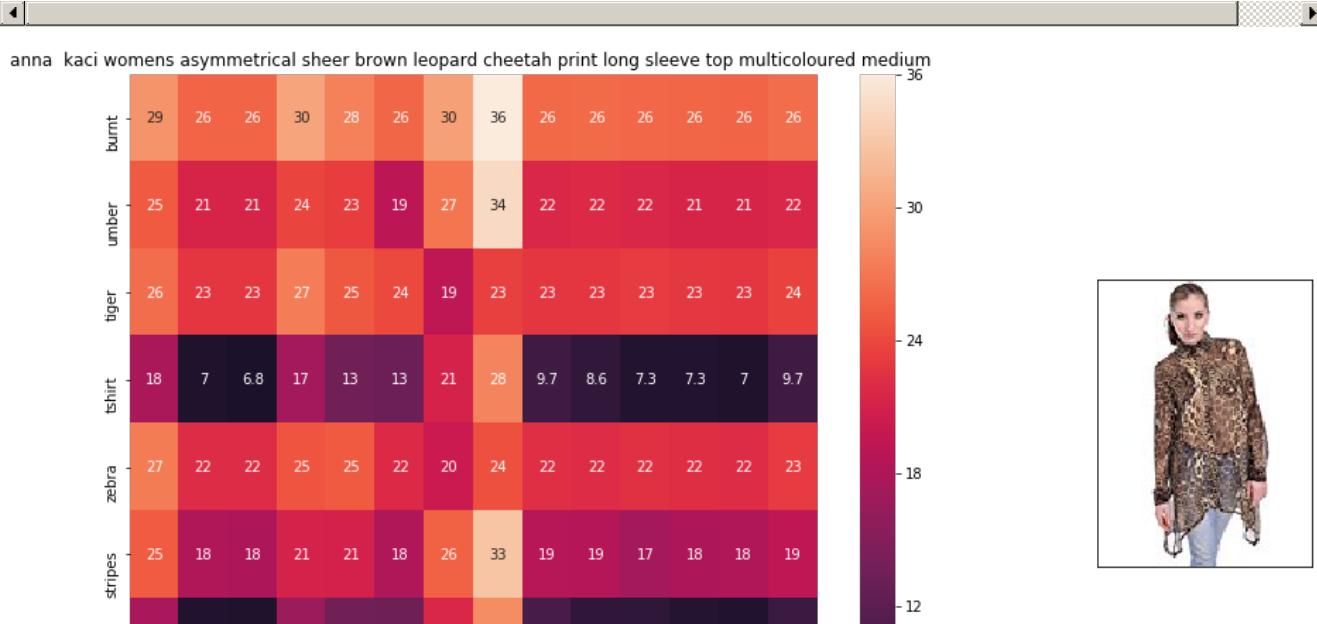
euclidean distance from input : 6.68391



ASIN : B06Y1VN8WQ

Brand : Black Swan

euclidean distance from input : 6.70576





ASIN : B00KSNTY7Y

Brand : Anna-Kaci

euclidean distance from input : 6.70612

=====

=====

[9.6] Weighted similarity using brand and color.

In [16]:

```
# some of the brand values are empty.
# Need to replace Null with string "NULL"
data['brand'].fillna(value="Not given", inplace=True)

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

brand_vectorizer = CountVectorizer()
brand_features = brand_vectorizer.fit_transform(brands)

type_vectorizer = CountVectorizer()
type_features = type_vectorizer.fit_transform(types)

color_vectorizer = CountVectorizer()
color_features = color_vectorizer.fit_transform(colors)

extra_features = hstack((brand_features, type_features, color_features)).tocsr()
```

In [0]:

```
def heat_map_w2v_brand(sentance1, sentance2, url, doc_id1, doc_id2, df_id1, df_id2, model):

    # sentance1 : title1, input apparel
    # sentance2 : title2, recommended apparel
    # url: apparel image url
    # doc_id1: document id of input apparel
    # doc_id2: document id of recommended apparel
    # df_id1: 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', 'Product type'],
                  [data['asin'].loc[df_id1], brands[doc_id1], colors[doc_id1], types[doc_id1]], # input
apparel's features
                  [data['asin'].loc[df_id2], brands[doc_id2], colors[doc_id2], types[doc_id2]]] # recommended apparel'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)
    # <table> + <div> with <table>...
```

```

# plot it with priority
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])
# plotting 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 lines 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 [0]:

```

def idf_w2v_brand(doc_id, w1, w2, num_results):
    # doc_id: apparel's id in given corpus
    # w1: weight for w2v features
    # w2: weight for brand and color features

    # pairwise_dist will store the distance from given input apparel to all remaining apparels
    # the metric we used here is cosine, the coside distance is mesured as  $K(X, Y) = \langle X, Y \rangle / (\|X\| * \|Y\|)$ 
    # http://scikit-learn.org/stable/modules/metrics.html#cosine-similarity
    idf_w2v_dist = pairwise_distances(w2v_title_weight, w2v_title_weight[doc_id].reshape(1,-1))
    ex_feat_dist = pairwise_distances(extra_features, extra_features[doc_id])
    pairwise_dist = (w1 * idf_w2v_dist + w2 * ex_feat_dist)/float(w1 + w2)

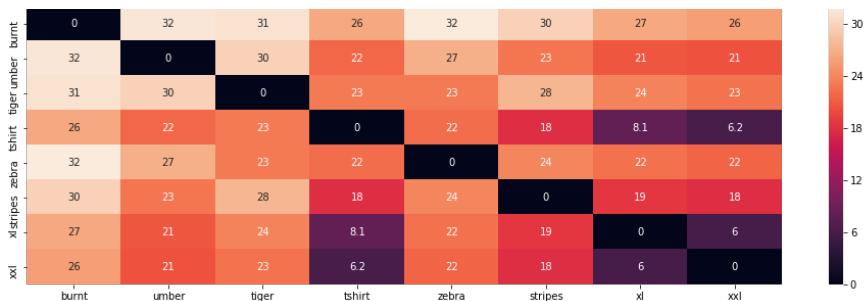
    # np.argsort will return indices of 9 smallest distances
    indices = np.argsort(pairwise_dist.flatten())[0:num_results]
    #pdists will store the 9 smallest distances
    pdists = np.sort(pairwise_dist.flatten())[0:num_results]

    #data frame indices of the 9 smallest distace's
    df_indices = list(data.index[indices])

    for i in range(0, len(indices)):
        heat_map_w2v_brand(data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], indices[0], indices[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 :', pdists[i])
        print('='*125)

idf_w2v_brand(12566, 5, 5, 20)
# in the give heat map, each cell contains the euclidean distance between words i, j

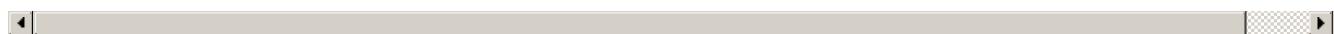
```



ASIN : B00JXQB5FQ

Brand : Si Row

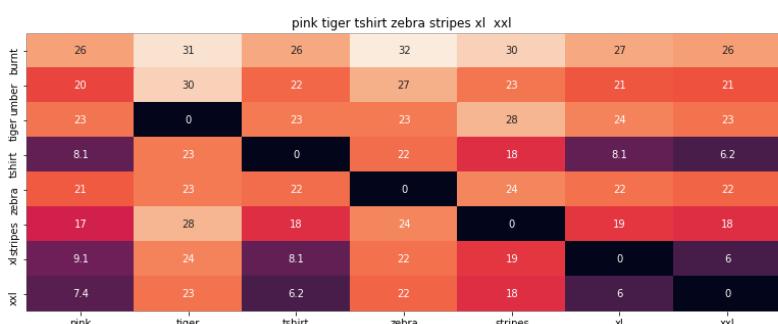
euclidean distance from input : 0.001953125



ASIN : B00JXQCWTO

Brand : Si Row

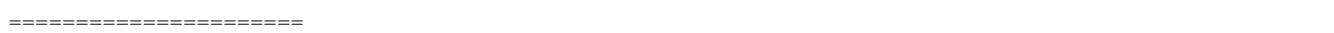
euclidean distance from input : 2.38547115326

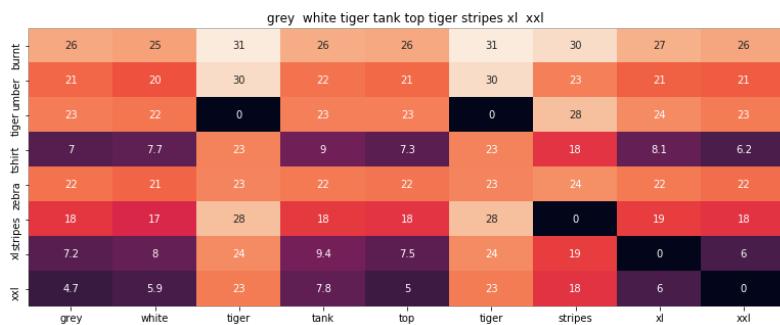


ASIN : B00JXQASS6

Brand : Si Row

euclidean distance from input : 2.73905105609

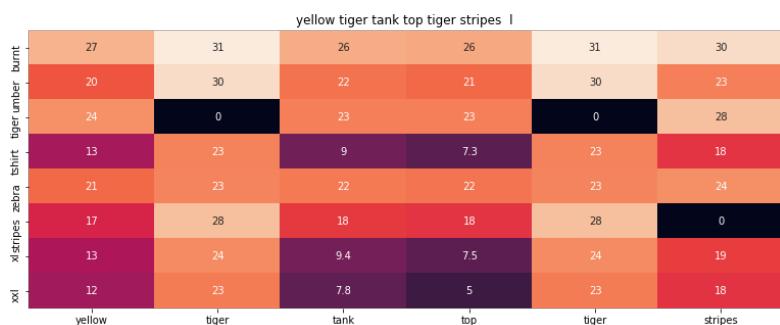




ASIN : B00JXQAFZ2

Brand : Si Row

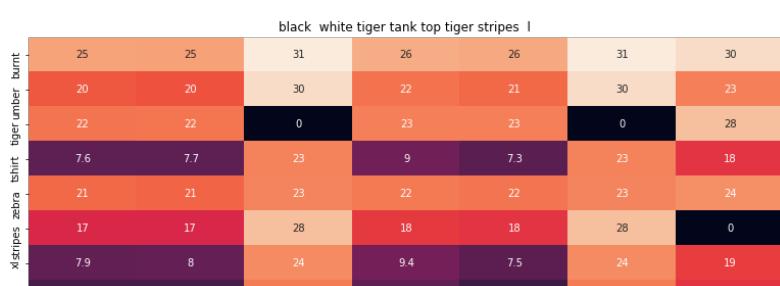
euclidean distance from input : 3.387187195

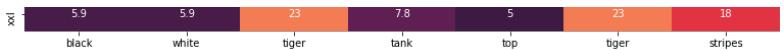


ASIN : B00JXQAUWA

Brand : Si Row

euclidean distance from input : 3.5518684389

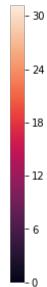
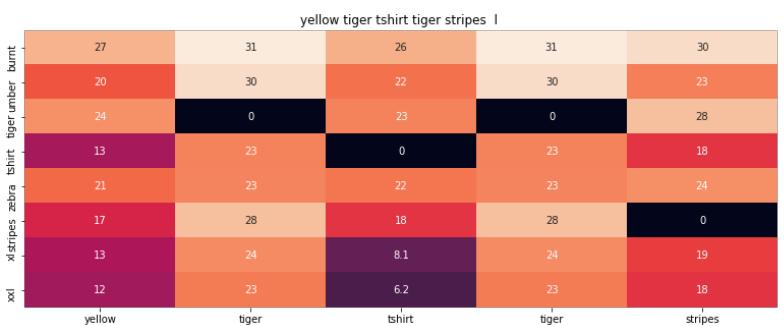




ASIN : B00JXQAO94

Brand : Si Row

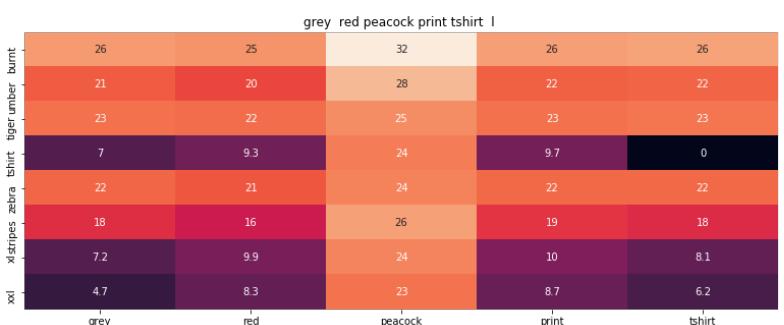
euclidean distance from input : 3.5536174776



ASIN : B00JXQCUIC

Brand : Si Row

euclidean distance from input : 3.65382804889



ASIN : B00JXQCFRS

Brand : Si Row

euclidean distance from input : 4.12881164569

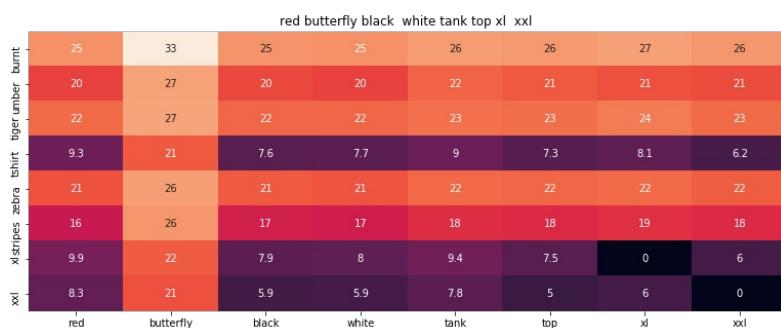




ASIN : B00JXQC8L6

Brand : Si Row

euclidean distance from input : 4.20390052813



ASIN : B00JV63CW2

Brand : Si Row

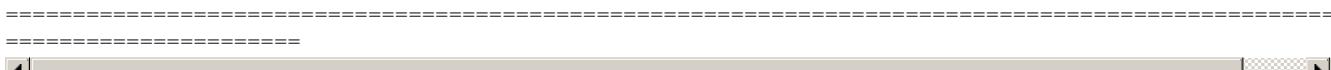
euclidean distance from input : 4.28658676166

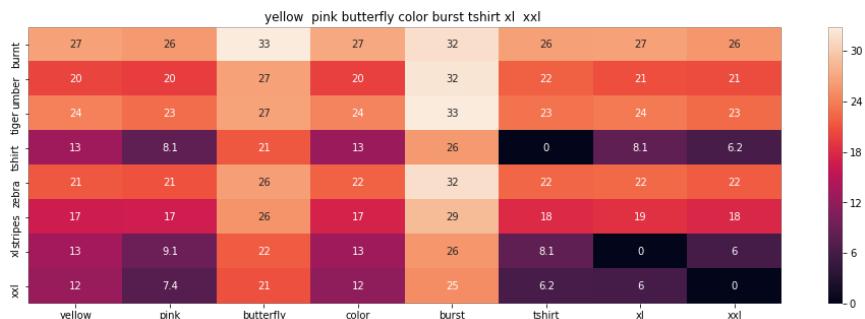


ASIN : B015H41F6G

Brand : KINGDE

euclidean distance from input : 4.38937078798

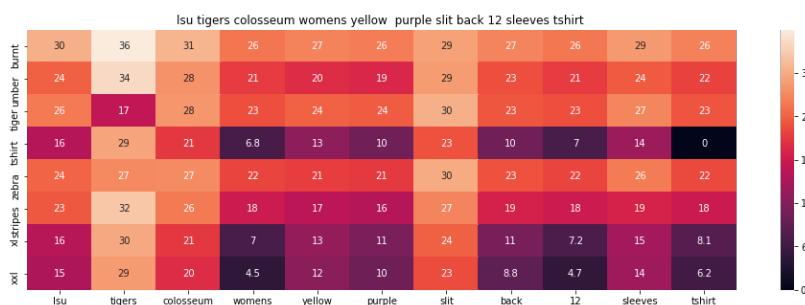




ASIN : B00JXQBBMI

Brand : Si Row

euclidean distance from input : 4.39790992755



ASIN : B073R5Q8HD

Brand : Colosseum

euclidean distance from input : 4.45122858369





ASIN : B074P8MD22

Brand : Edista

euclidean distance from input : 4.51897779787

=====



red pink floral heel sleeveless shirt xl xxl



ASIN : B00JV63QQE

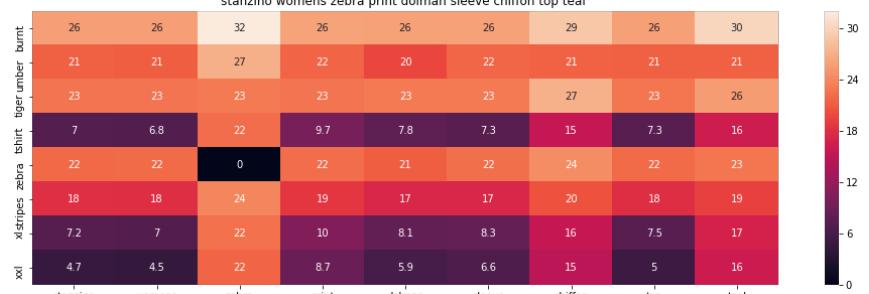
Brand : Si Row

euclidean distance from input : 4.52937545794

=====



stanzino womens zebra print dolman sleeve chiffon top teal



ASIN : B00C0I3U3E

Brand : Stanzino

euclidean distance from input : 4.53032614076

=====



		26	26	28	28	31	32	30	35	26	26	26	26
		21	21	24	24	25	29	23	32	22	21	22	21
		23	23	25	25	29	18	28	33	23	23	23	23
		7	6.8	14	15	19	23	18	25	9	7.3	8.8	8.3
		22	22	25	24	27	22	24	32	22	22	22	22
		18	18	22	23	24	28	0	28	18	18	19	18
		7.2	7	14	15	20	23	19	27	9.4	7.5	9.1	7.7
xxl	gupobou168	womens	girls	lady	boho	elephant	stripes	bandage	tank	top	one	size	



ASIN : B01ER18406

Brand : GuPoBoU168

euclidean distance from input : 4.54681702403



		lsu tigers colosseum womens charcoal gray striped back sleeveless tank top											
		30	36	31	26	26	27	27	27	26	26	26	
		24	34	28	21	23	19	20	23	22	22	21	
		26	17	28	23	27	23	23	23	25	23	23	
		16	29	21	6.8	18	12	11	10	11	9	7.3	
		24	27	27	22	26	22	20	23	23	22	22	
		23	32	26	18	24	18	16	19	18	18	18	
		16	30	21	7	19	11	12	11	12	9.4	7.5	
xxl		15	29	20	4.5	18	11	10	8.8	10	7.8	5	
	lsu	tigers	colosseum	womens	charcoal	gray	striped	back	sleeveless	tank	top		



ASIN : B073R4ZM7Y

Brand : Colosseum

euclidean distance from input : 4.54835554445



		merona womens printed blouse brown leopard print xxl											
		26	26	26	26	26	30	26	26				
		21	21	22	21	19	27	22	21				
		23	23	23	23	24	19	23	23				
		7	6.8	9.4	7.3	13	21	9.7	6.2				
		22	22	23	22	22	20	22	22				
		18	18	19	18	18	26	19	18				
		7.2	7	11	8.4	13	22	10	6				
xxl	merona	womens	printed	blouse	brown	leopard	print		xxl				

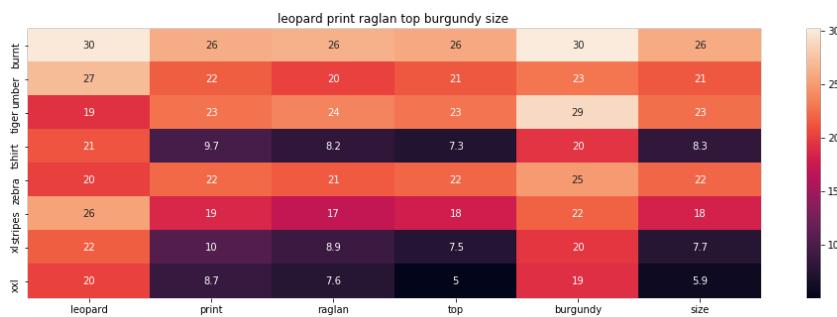


ASIN : B071YF3WDD

Brand : Merona

euclidean distance from input : 4.61062742555





ASIN : B01C60RLDQ

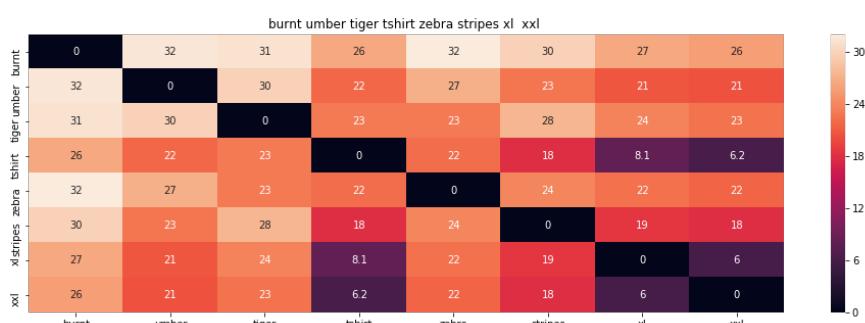
Brand : 1 Mad Fit

euclidean distance from input : 4.64591789282

In [0]:

```
# brand and color weight =50
# title vector weight = 5

idf_w2v_brand(12566, 5, 50, 20)
```



ASIN : B00JXQB5FQ

Brand : Si Row

euclidean distance from input : 0.000355113636364

In [0]:





ASIN : B00JXQCWTO

Brand : Si Row

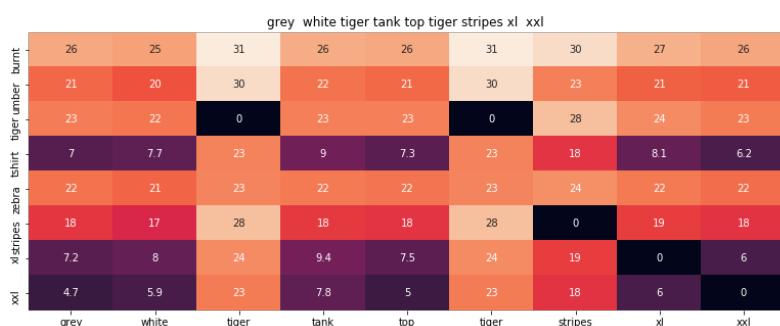
euclidean distance from input : 0.433722027865



ASIN : B00JXQASS6

Brand : Si Row

euclidean distance from input : 1.65509310669

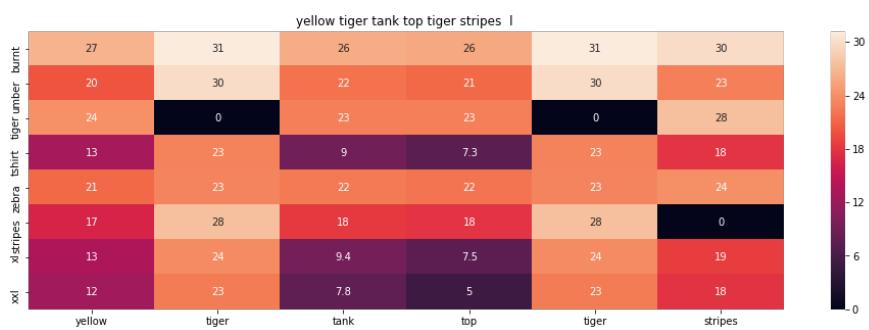


ASIN : B00JXQAFZ2

Brand : Si Row

euclidean distance from input : 1.77293604103

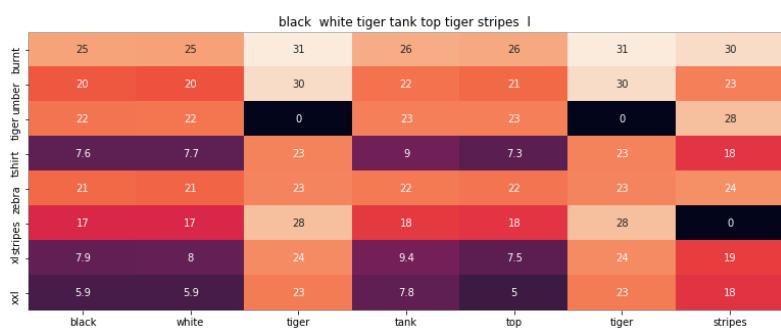
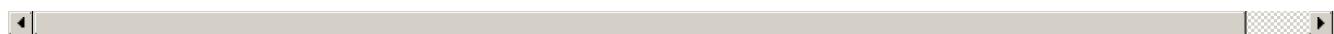




ASIN : B00JXQAUWA

Brand : Si Row

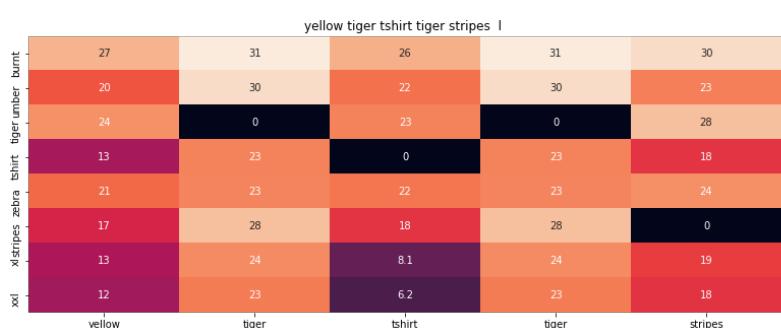
euclidean distance from input : 1.80287808538



ASIN : B00JXQAO94

Brand : Si Row

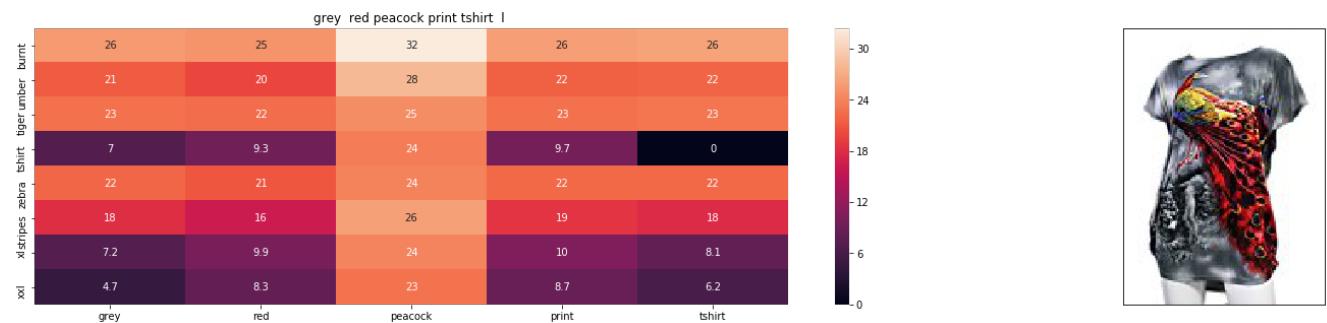
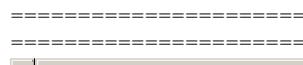
euclidean distance from input : 1.80319609241



ASIN : B00JXQCUIC

Brand : Si Row

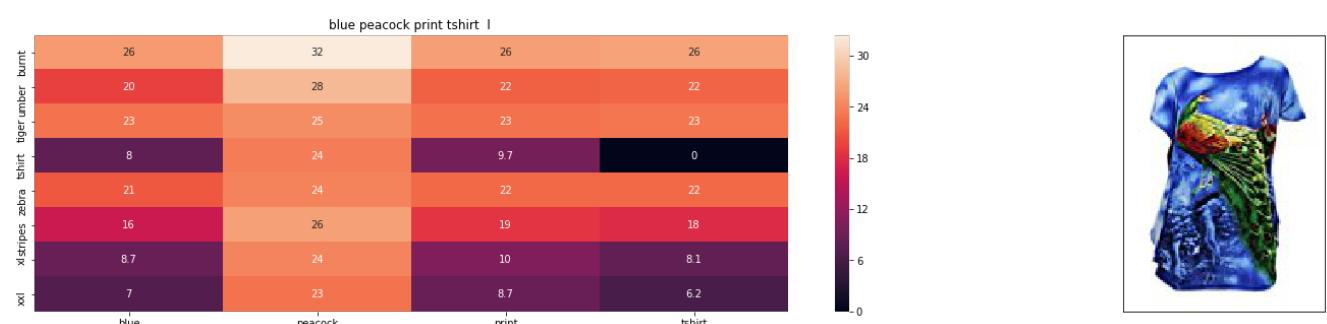
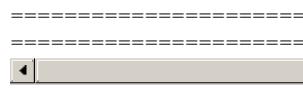
euclidean distance from input : 1.82141619628



ASIN : B00JXQCFRS

Brand : Si Row

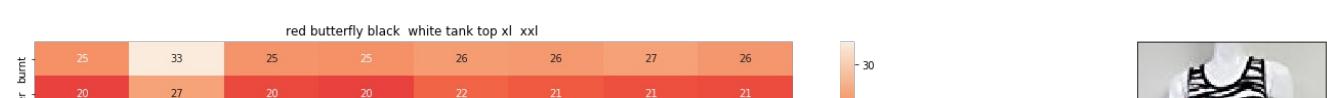
euclidean distance from input : 1.90777685025

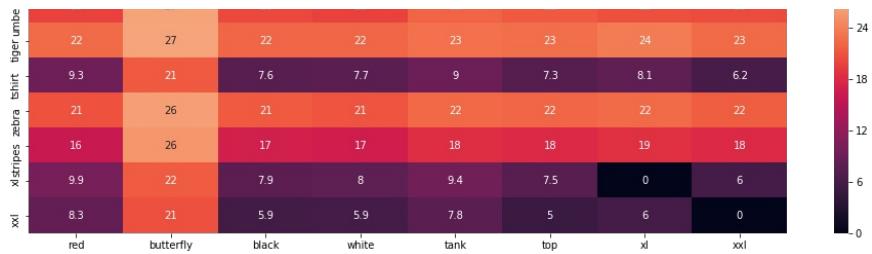


ASIN : B00JXQC8L6

Brand : Si Row

euclidean distance from input : 1.92142937433

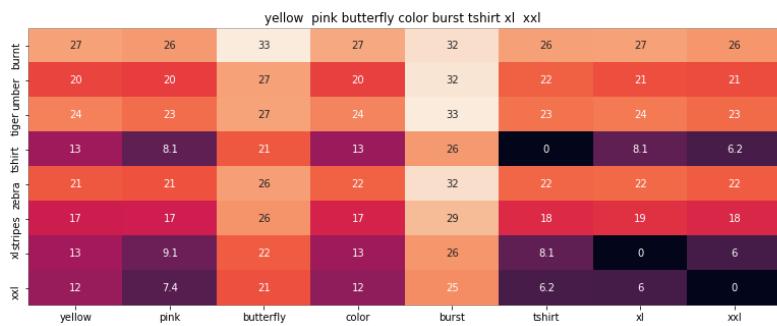




ASIN : B00JV63CW2

Brand : Si Row

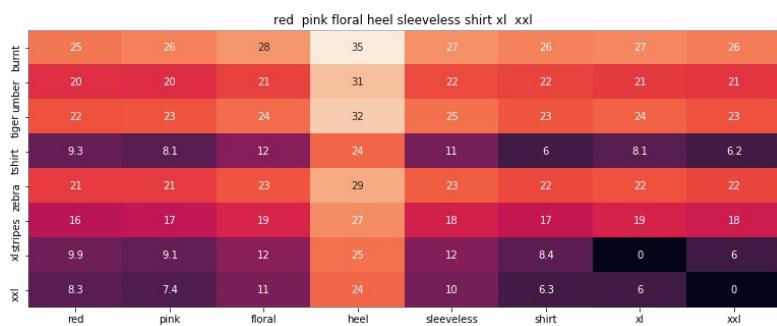
euclidean distance from input : 1.93646323497



ASIN : B00JXQBBMI

Brand : Si Row

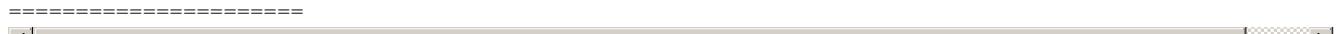
euclidean distance from input : 1.95670381059

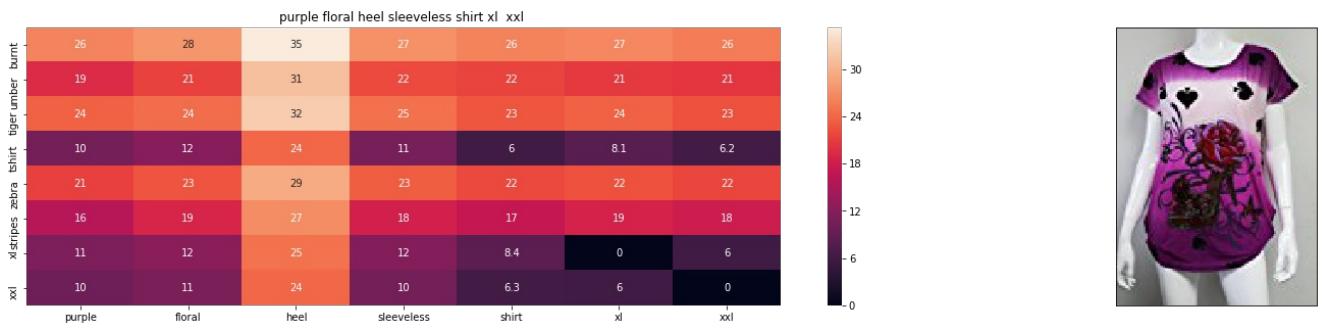


ASIN : B00JV63QQE

Brand : Si Row

euclidean distance from input : 1.9806066343

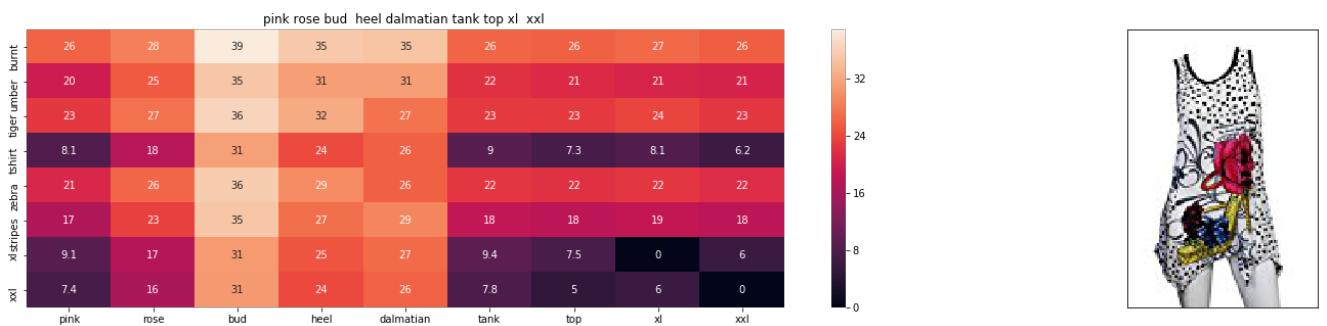




ASIN : B00JV63VC8

Brand : Si Row

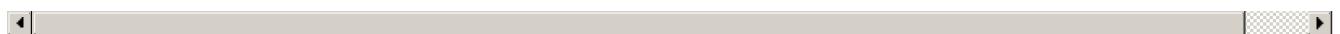
euclidean distance from input : 2.01218559992



ASIN : B00JXQAX2C

Brand : Si Row

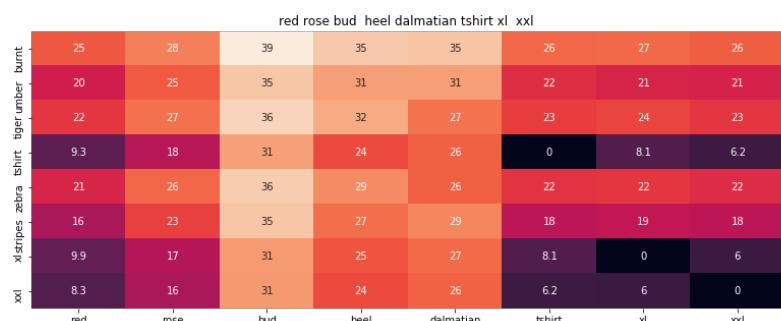
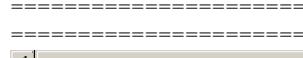
euclidean distance from input : 2.01335178755



ASIN : B00JXQC0C8

Brand : Si Row

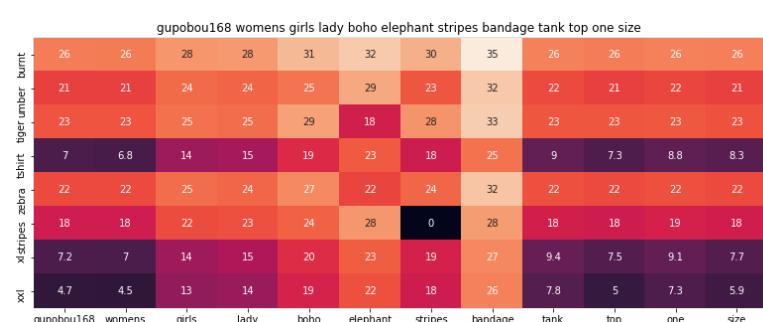
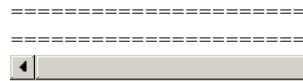
euclidean distance from input : 2.01388334827



ASIN : B00JXQABBO

Brand : Si Row

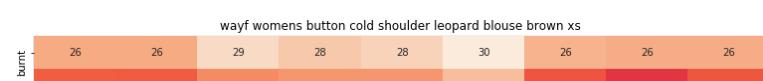
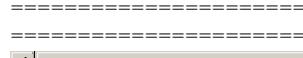
euclidean distance from input : 2.0367257555



ASIN : B01ER18406

Brand : GuPoBoU168

euclidean distance from input : 2.65620416778



		24	24	24	23	23	23	23	23	23
	tiger	23	23	25	26	25	19	23	24	23
	stripes	7	6.8	13	15	13	21	7.3	13	7.7
	zebra	22	22	25	25	24	20	22	22	23
	x-stripes	18	18	20	22	19	26	18	18	19
	xxl	7.2	7	14	15	14	22	8.4	13	6.9
	wayf	4.7	4.5	12	14	12	20	6.2	12	5.8
wayf	womens		button	cold	shoulder	leopard	blouse	brown		xs



ASIN : B01LZ7BQ4H

Brand : WAYF

euclidean distance from input : 2.6849067823



		yabina womens pullover leopard print long sleeve hoodie us10 brown									
	tiger	26	26	31	30	26	26	26	32	26	26
	stripes	21	21	25	27	22	22	22	28	21	19
	zebra	23	23	26	19	23	23	23	28	23	24
	x-stripes	7	6.8	14	21	9.7	8.6	7.3	17	7	13
	xxl	22	22	25	20	22	22	22	27	22	22
	yabina	18	18	21	26	19	19	17	23	18	18
	wayf	7.2	7	16	22	10	8.5	8.3	19	7.2	13
wayf	womens	4.7	4.5	15	20	8.7	6.6	6.6	18	4.7	12
		burnt				long	sleeve	hoodie	us10		brown



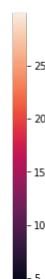
ASIN : B01KJUM6JI

Brand : YABINA

euclidean distance from input : 2.68583819266



		wayf womens small button leopard cold shoulder blouse brown									
	tiger	26	26	26	29	30	28	28	26	26	
	stripes	21	21	21	24	27	25	25	21	19	
	zebra	23	23	23	25	19	26	25	23	24	
	x-stripes	7	6.8	8.5	13	21	15	13	7.3	13	
	xxl	22	22	22	25	20	25	24	22	22	
	yabina	18	18	19	20	26	22	19	18	18	
	wayf	7.2	7	8.4	14	22	15	14	8.4	13	
wayf	womens	4.7	4.5	6.4	12	20	14	12	6.2	12	
		burnt				cold	shoulder	blouse	brown		



ASIN : B01M06V4X1

Brand : WAYF

euclidean distance from input : 2.69476194865



[10.4] Keras and Tensorflow to extract features

In [0]:

```
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
```

Using TensorFlow backend.

In [0]:

```
# https://gist.github.com/fchollet/f35fbc80e066a49d65f1688a7e99f069
# Code reference: https://blog.keras.io/building-powerful-image-classification-models-using-very-little-data.html

# This code takes 40 minutes to run on a modern GPU (graphics card)
# like Nvidia 1050.
# GPU (Nvidia 1050): 0.175 seconds per image

# This code takes 160 minutes to run on a high end i7 CPU
# CPU (i7): 0.615 seconds per image.

#Do NOT run this code unless you want to wait a few hours for it to generate output

# each image is converted into 25088 length dense-vector

'''

# dimensions of our images.
img_width, img_height = 224, 224

top_model_weights_path = 'bottleneck_fc_model.h5'
train_data_dir = 'images2/'
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 // batch_size)
    bottleneck_features_train = bottleneck_features_train.reshape((16042,25088))

    np.save(open('16k_data_cnn_features.npy', 'wb'), bottleneck_features_train)
    np.save(open('16k_data_cnn_feature_asins.npy', 'wb'), np.array(asins))
```

```
save_bottleneck_features()  
'''
```

[10.3] Visual features based product similarity.

In [0]:

```
#load the features and corresponding ASINS info.  
bottleneck_features_train = np.load('16k_data_cnn_features.npy')  
asins = np.load('16k_data_cnn_feature_asins.npy')  
asins = list(asins)  
  
# load the original 16K dataset  
data = pd.read_pickle('pickels/16k_apperial_data_preprocessed')  
df_asins = list(data['asin'])  
  
from IPython.display import display, Image, SVG, Math, YouTubeVideo  
  
#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_features_train[doc_id].reshape(1,-1))  
  
    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 idx, row in rows.iterrows():  
            display(Image(url=row['medium_image_url'], embed=True))  
            print('Product Title: ', row['title'])  
            print('Euclidean Distance from input image:', pdists[i])  
            print('Amazon Url: www.amazon.com/dp/' + asins[indices[i]])  
  
get_similar_products_cnn(12566, 20)
```



Product Title: burnt umber tiger tshirt zebra stripes xl xxl
Euclidean Distance from input image: 0.0
Amazon Url: www.amazon.com/dp/B00JXQB5FQ



Product Title: pink tiger tshirt zebra stripes xl xxl
Euclidean Distance from input image: 30.0501
Amazon Url: www.amazon.com/dp/B00JXQASS6





Product Title: yellow tiger tshirt tiger stripes l
Euclidean Distance from input image: 41.2611
Amazon Url: www.amazon.com/dp/B00JXQCUIC



Product Title: brown white tiger tshirt tiger stripes xl xxl
Euclidean Distance from input image: 44.0002
Amazon Url: www.amazon.com/dp/B00JXQCWT0



Product Title: kawaii pastel tops tees pink flower design
Euclidean Distance from input image: 47.3825
Amazon Url: www.amazon.com/dp/B071FCWD97



Product Title: womens thin style tops tees pastel watermelon print
Euclidean Distance from input image: 47.7184
Amazon Url: www.amazon.com/dp/B01JUNHBRM



Product Title: kawaii pastel tops tees baby blue flower design
Euclidean Distance from input image: 47.9021
Amazon Url: www.amazon.com/dp/B071SBCY9W





Product Title: edv cheetah run purple multi xl
Euclidean Distance from input image: 48.0465
Amazon Url: www.amazon.com/dp/B01CUPYBM0



Product Title: danskin womens vneck loose performance tee xsmall pink ombre
Euclidean Distance from input image: 48.1019
Amazon Url: www.amazon.com/dp/B01F7PHXY8



Product Title: summer alpaca 3d pastel casual loose tops tee design
Euclidean Distance from input image: 48.1189
Amazon Url: www.amazon.com/dp/B01I80A93G



Product Title: miss chievous juniors striped peplum tank top medium shadowpeach
Euclidean Distance from input image: 48.1313
Amazon Url: www.amazon.com/dp/B0177DM70S



Product Title: red pink floral heel sleeveless shirt xl xxl
Euclidean Distance from input image: 48.1695
Amazon Url: www.amazon.com/dp/B00JV63QOE



Product Title: moana logo adults hot v neck shirt black xxl
Euclidean Distance from input image: 48.2568
Amazon Url: www.amazon.com/dp/B01LX6H43D



Product Title: abaday multicolor cartoon cat print short sleeve longline shirt large
Euclidean Distance from input image: 48.2657
Amazon Url: www.amazon.com/dp/B01CR57YY0



Product Title: kawaii cotton pastel tops tees peach pink cactus design
Euclidean Distance from input image: 48.3626
Amazon Url: www.amazon.com/dp/B071WYLBZS



Product Title: chicago chicago 18 shirt women pink
Euclidean Distance from input image: 48.3836
Amazon Url: www.amazon.com/dp/B01GXAZTRY



Product Title: yichun womens tiger printed summer tshirts tops
Euclidean Distance from input image: 48.4493
Amazon Url: www.amazon.com/dp/B010NN9RXO



Product Title: nancy lopez whimsy short sleeve whiteblacklemon drop xs
Euclidean Distance from input image: 48.4788
Amazon Url: www.amazon.com/dp/B01MPX6IDX



Product Title: womens tops tees pastel peach ice cream cone print
Euclidean Distance from input image: 48.558
Amazon Url: www.amazon.com/dp/B0734GRKZL



Product Title: uswomens mary j blige without tshirts shirt
Euclidean Distance from input image: 48.6144
Amazon Url: www.amazon.com/dp/B01M0XXFKK

Assignment

In [17]:

```
#load the features and corresponding ASINS info.
bottleneck_features_train = np.load('16k_data_cnn_features.npy')
asins = np.load('16k_data_cnn_feature_asins.npy')
asins = list(asins)
```

In [56]:

```
def function (doc_id,num_results,w1=1,w2=1,w3=1,w4=1):
    brand_dist = pairwise_distances(brand_features, brand_features[doc_id])
    color_dist = pairwise_distances(color_features, color_features[doc_id])
    image_dist = pairwise_distances(bottleneck_features_train, bottleneck_features_train[doc_id].reshape(1,-1))
    text_dist = pairwise_distances(w2v_title_weight, w2v_title_weight[doc_id].reshape(1,-1))
        # np.argsort will return indices of 9 smallest distances
    print(text_dist.shape,
    image_dist.shape,
    color_dist.shape,
    brand_dist.shape)
    pairwise_dist = ((w1 * text_dist) +(w2 * brand_dist) +(w3 * color_dist) + (w4 * image_dist))/float(w1 + w2 + w3 + w4)
    indices = np.argsort(pairwise_dist.flatten())[0:num_results]
    print(pairwise_dist.shape)
    #ndists will store the 9 smallest distances
```

```

    "Prints will occur one at a time for distances
pdists = np.sort(pairwise_dist.flatten())[0:num_results]

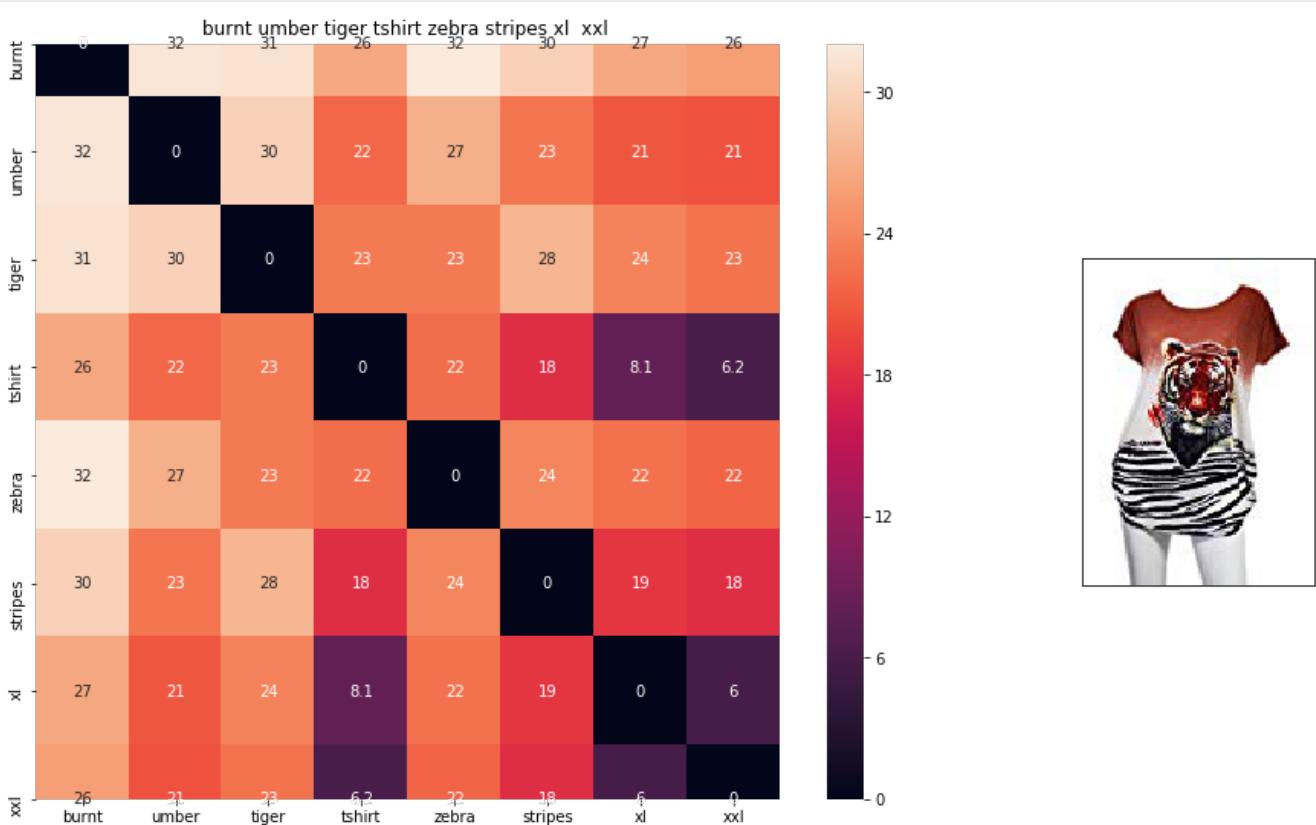
#data frame indices of the 9 smallest distance's
df_indices = list(data.index[indices])
print(indices)
print(data.index[indices])
print(df_indices)

for i in range(0, len(indices)):
    heat_map_w2v(data['title'].loc[df_indices[0]], data['title'].loc[df_indices[i]], data['medium_image_url'].loc[df_indices[i]], indices[0], indices[i], 'weighted')
    print('ASIN :', data['asin'].loc[df_indices[i]])
    print('Brand :', data['brand'].loc[df_indices[i]])
    print('euclidean distance from input :', pdists[i])
    print('='*125)

```

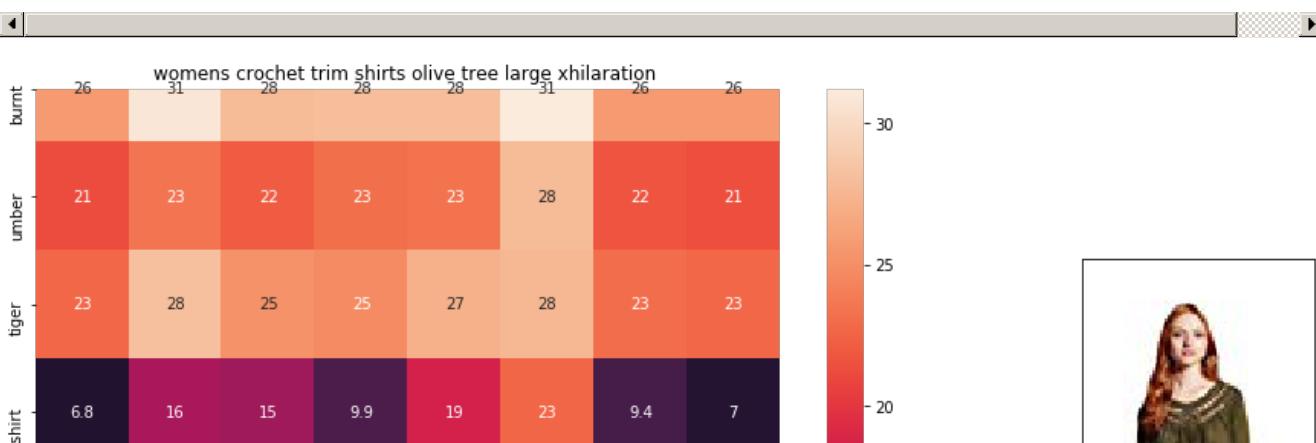
[48]:

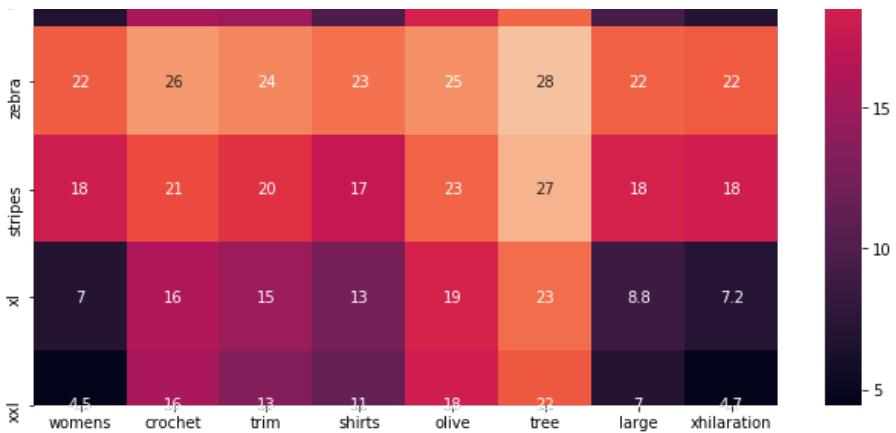
```
function(12566,10)
```



```

ASIN : B00JXQB5FQ
Brand : Si Row
euclidean distance from input : 4.214684778958144e-08
=====
```

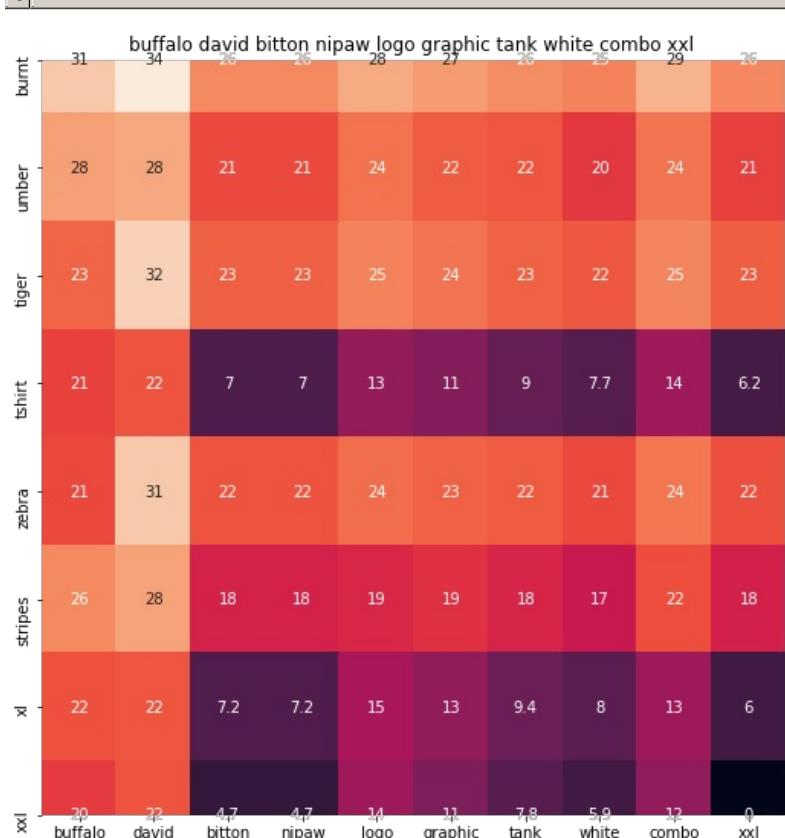




ASIN : B06XBHNM7J

Brand : Xhilaration

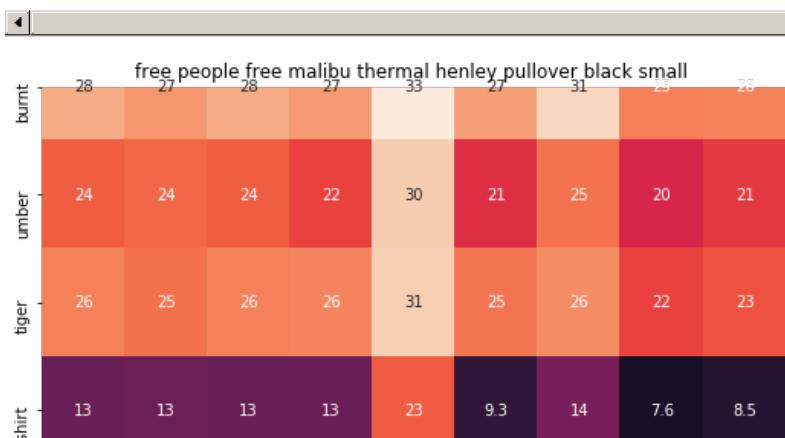
euclidean distance from input : 12.145854071425124

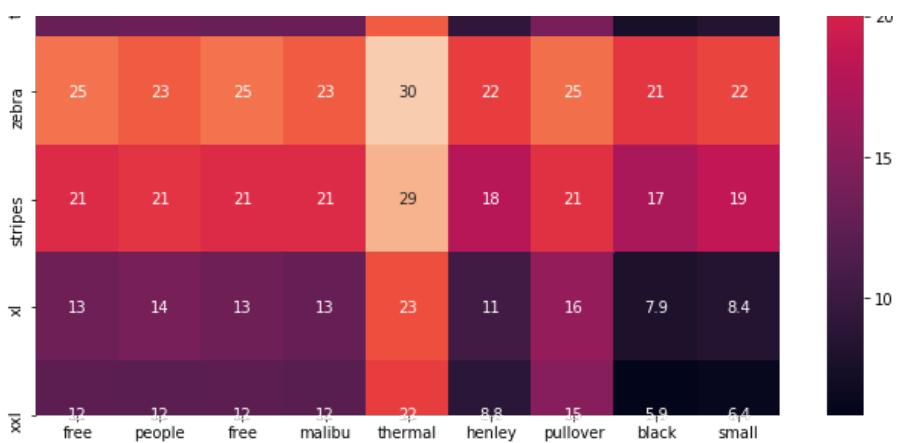


ASIN : B018H5AZXQ

Brand : Buffalo

euclidean distance from input : 12.611845806883498

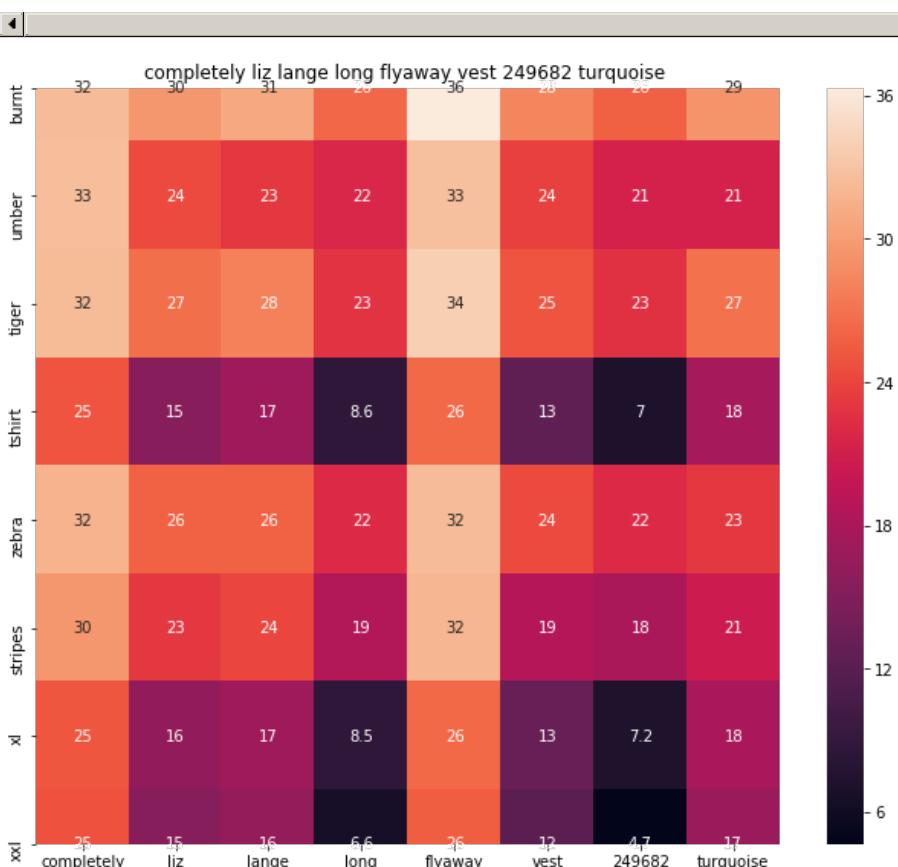




ASIN : B074MXY984

Brand : We The Free

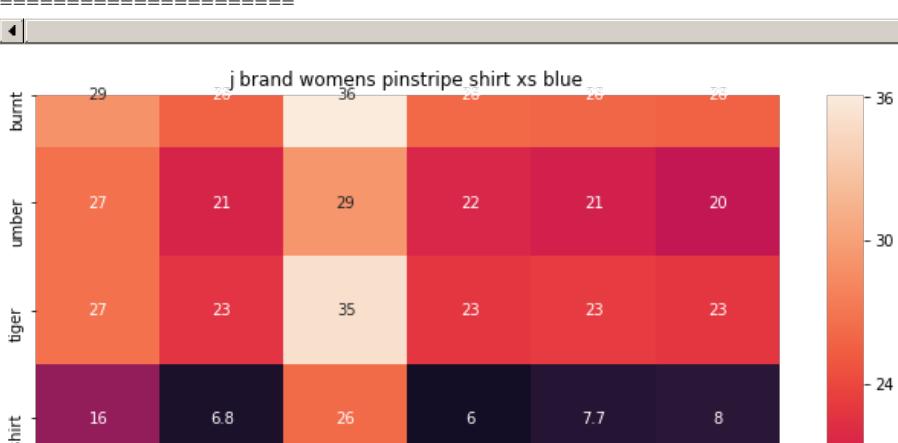
euclidean distance from input : 12.861257699955404

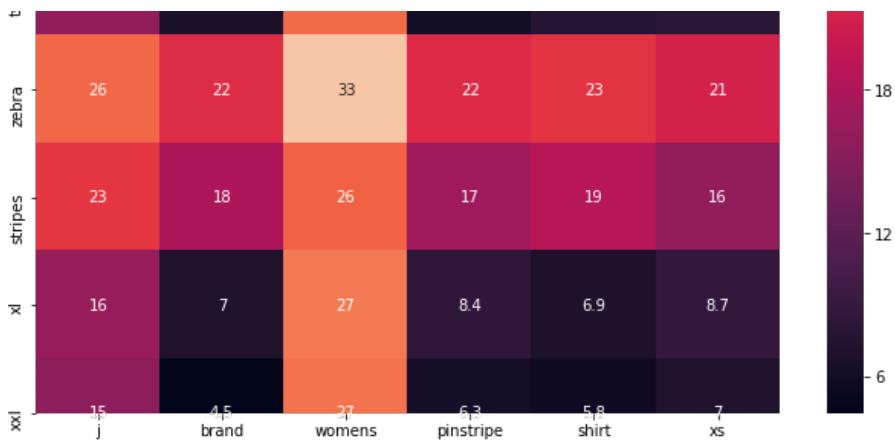


ASIN : B074LTBWSW

Brand : Liz Lange

euclidean distance from input : 12.864222860426647

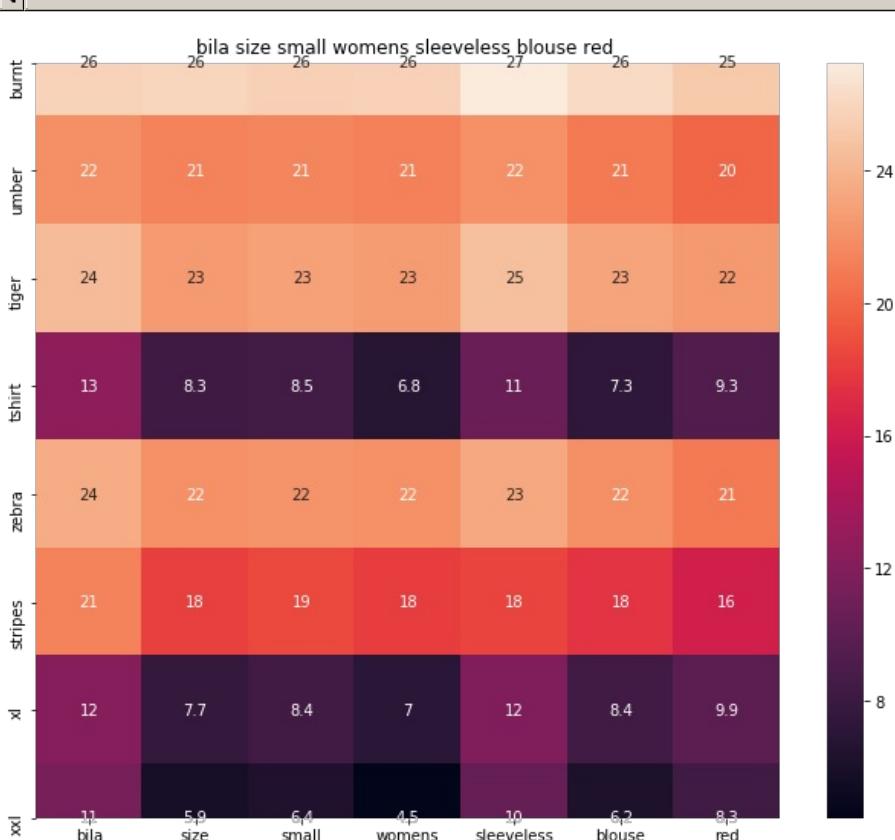




ASIN : B06XYP1X1F

Brand : J Brand Jeans

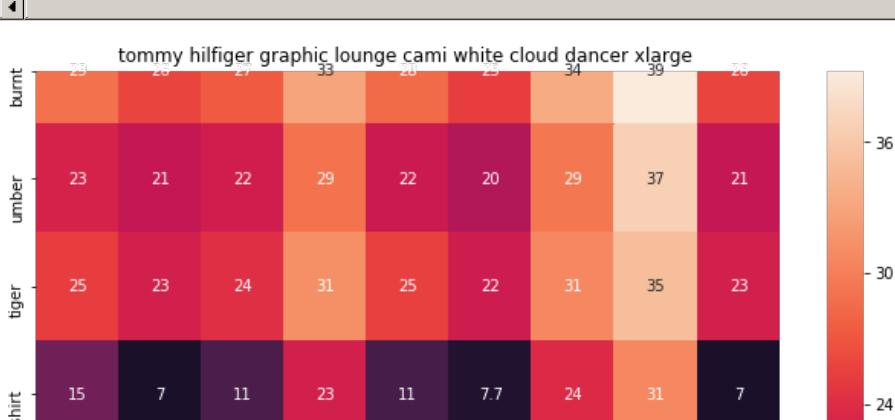
euclidean distance from input : 12.926071882247925

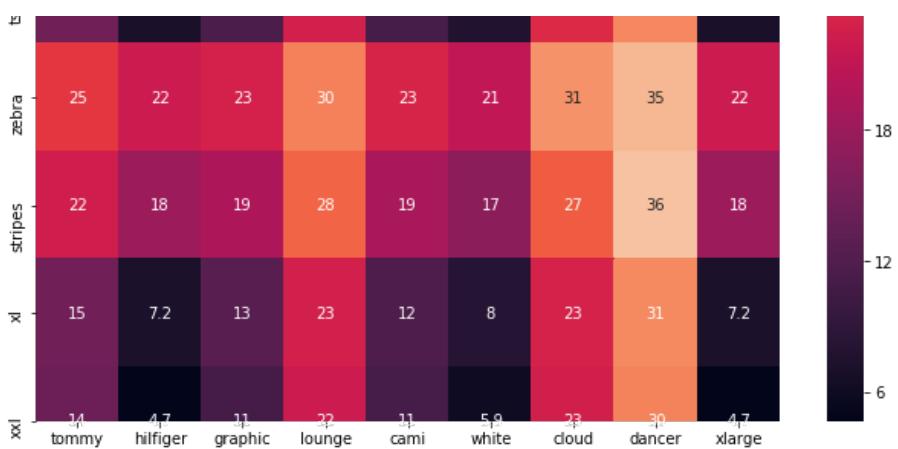


ASIN : B01L7ROZNC

Brand : Bila

euclidean distance from input : 12.94621927928359

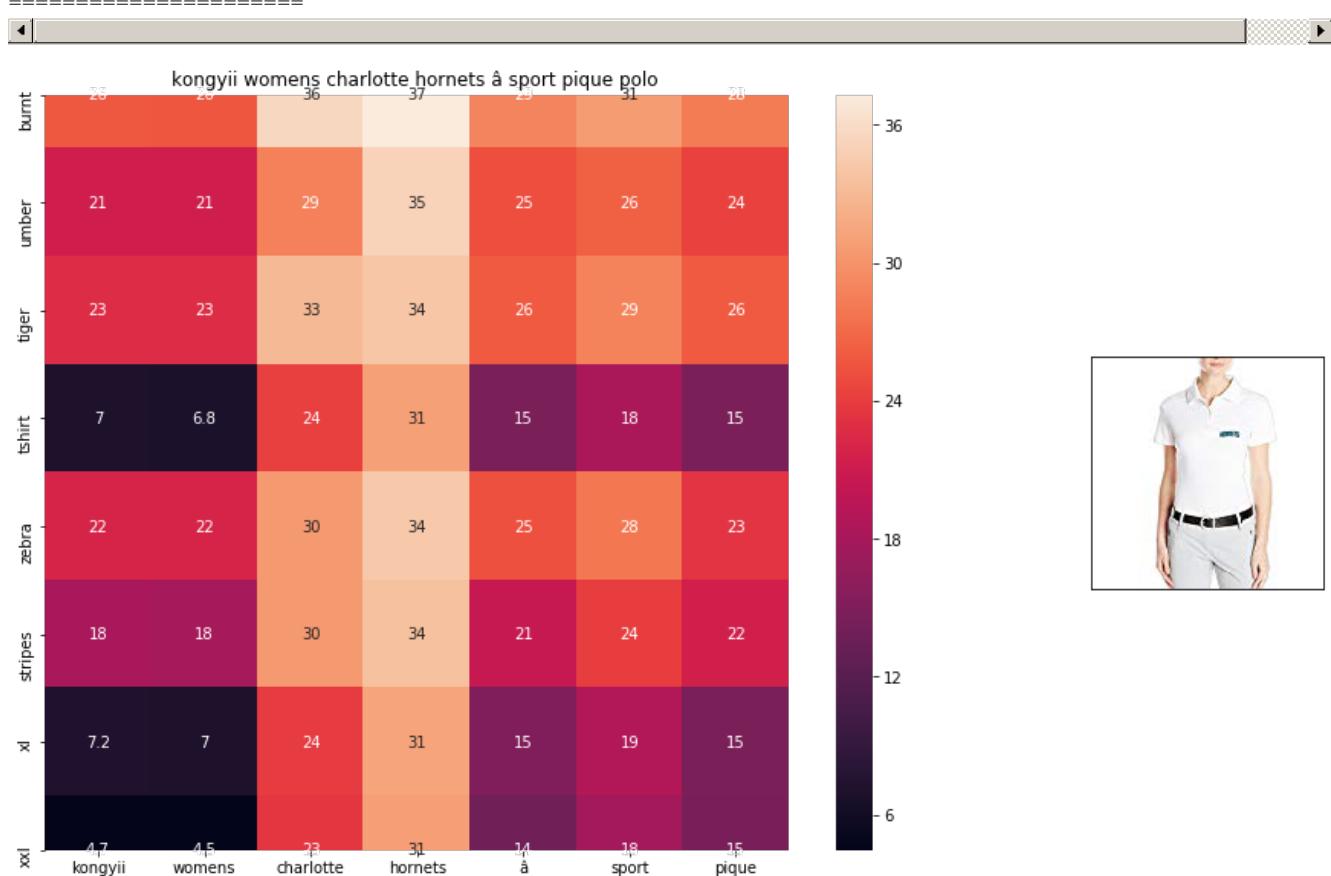




ASIN : B01BMSFYW2

Brand : igertommy hilf

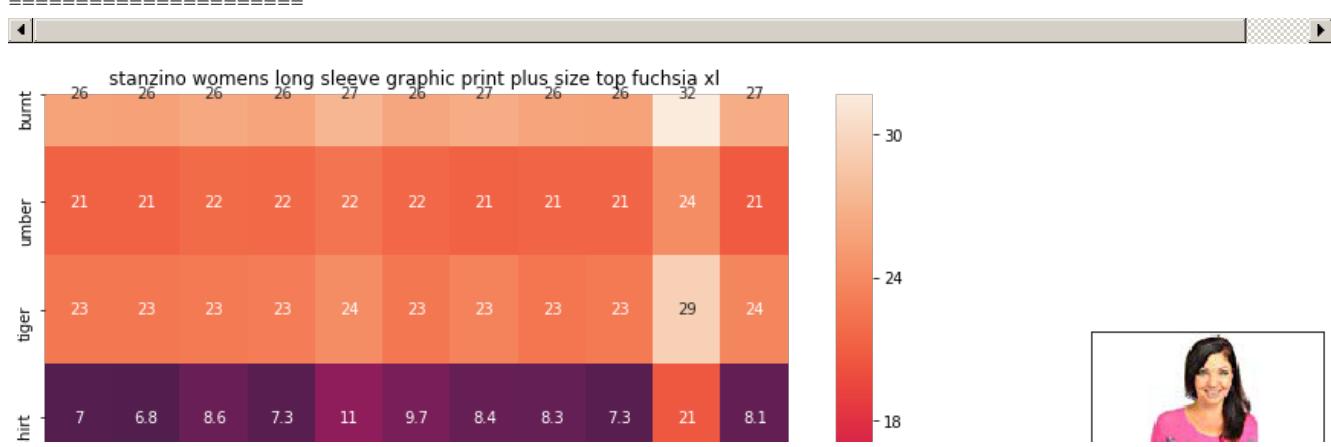
euclidean distance from input : 13.17028522491455

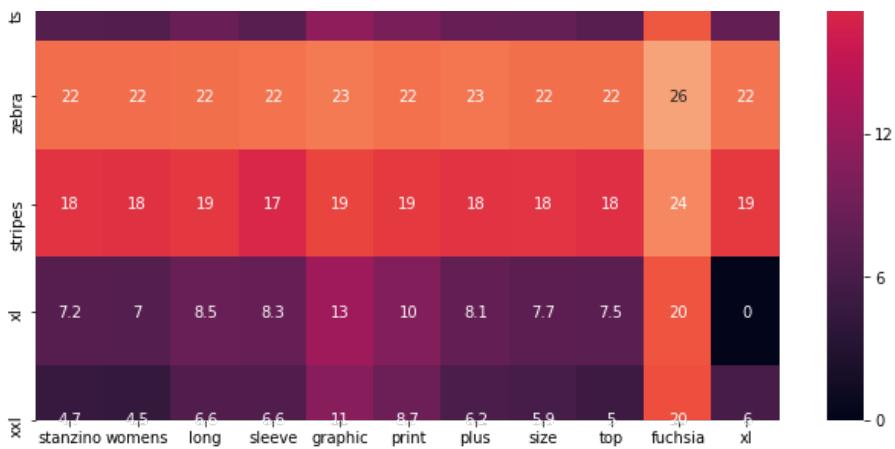


ASIN : B01FJVZST2

Brand : KONGYII

euclidean distance from input : 13.288894011491841





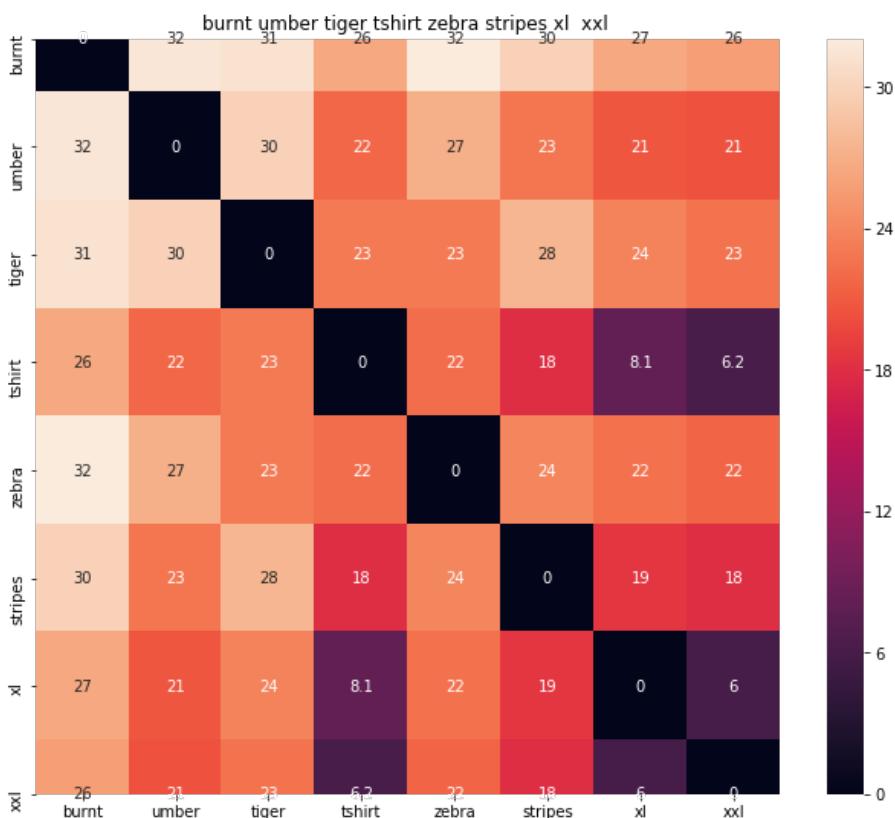
ASIN : B00DP4VHWI

Brand : Stanzino

euclidean distance from input : 13.31425590705306

In [49]:

```
function(12566,10,100,10,20,150)
```



ASIN : B00JXQB5FQ

Brand : Si Row

euclidean distance from input : 6.020978357160597e-08

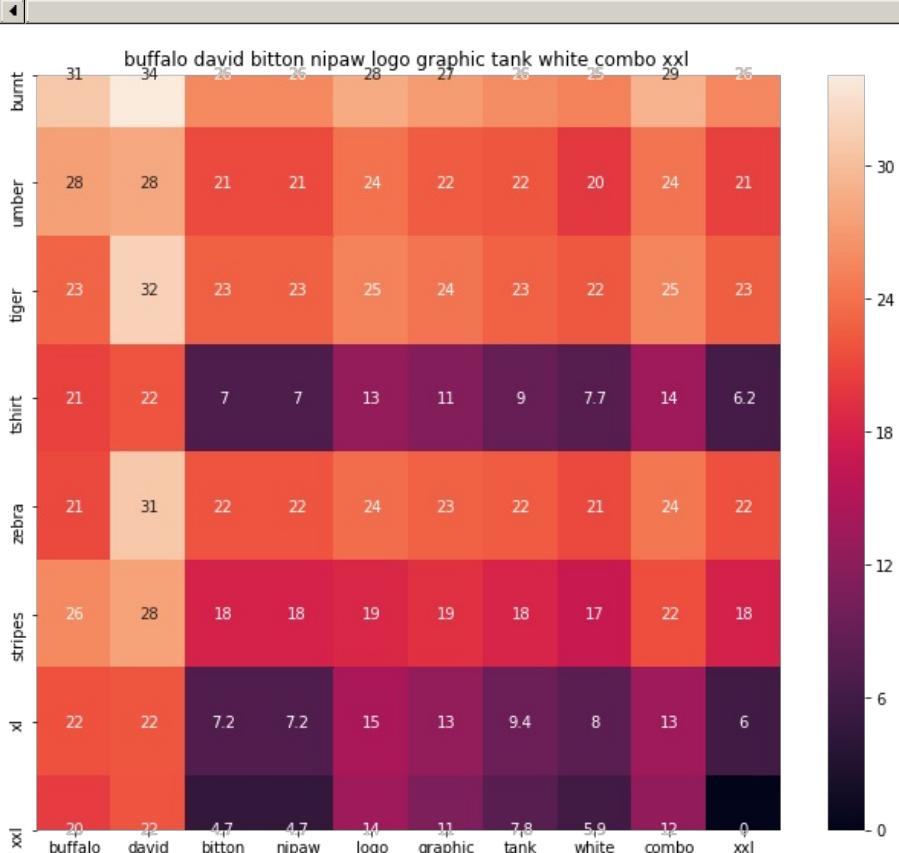




ASIN : B06XBHNM7J

Brand : Xhilaration

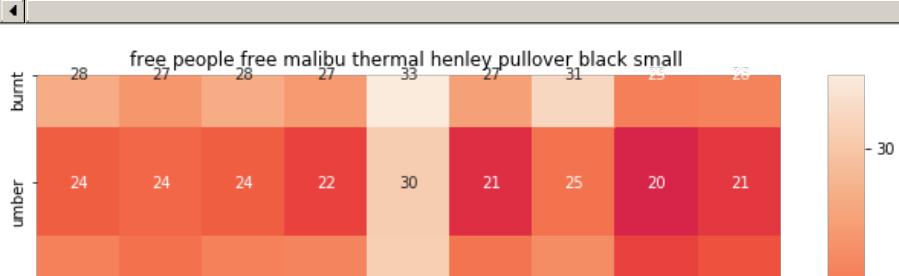
euclidean distance from input : 22.94029074814633

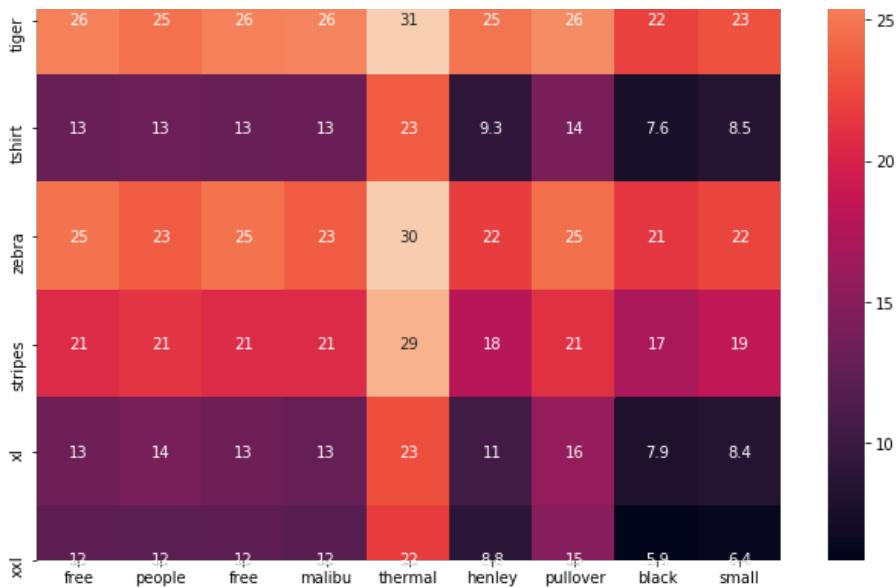


ASIN : B018H5AZXQ

Brand : Buffalo

euclidean distance from input : 24.029977332618987

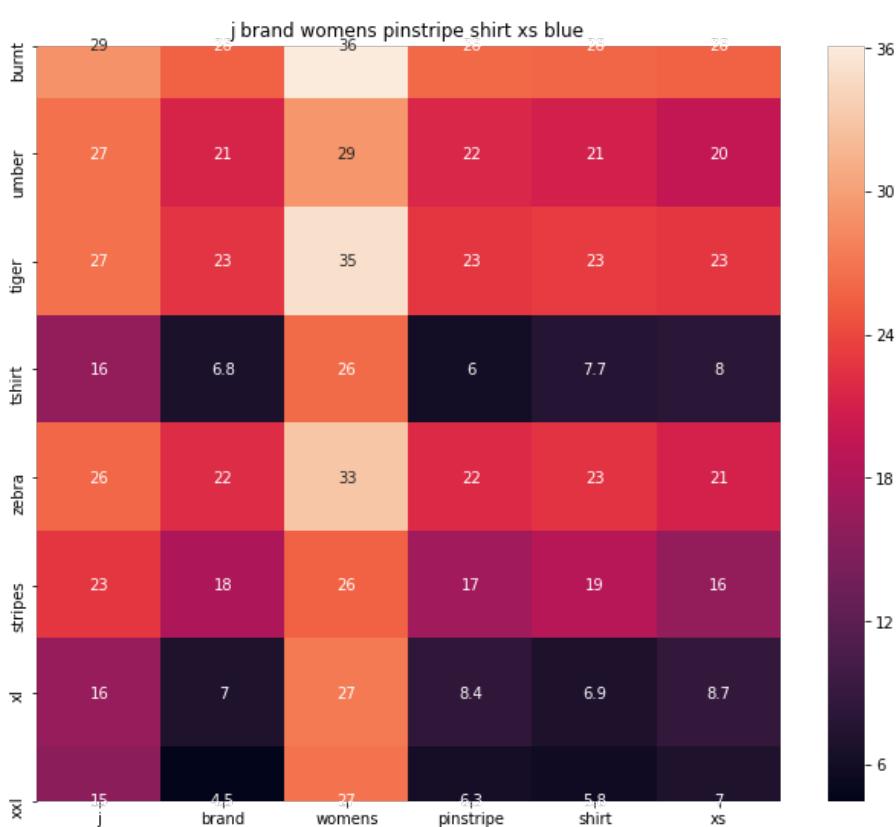




ASIN : B074MXY984

Brand : We The Free

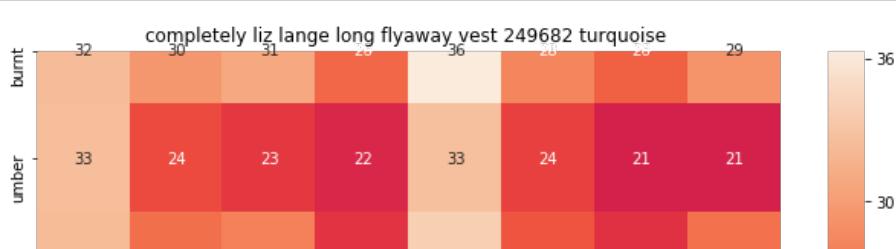
euclidean distance from input : 24.279748508464703

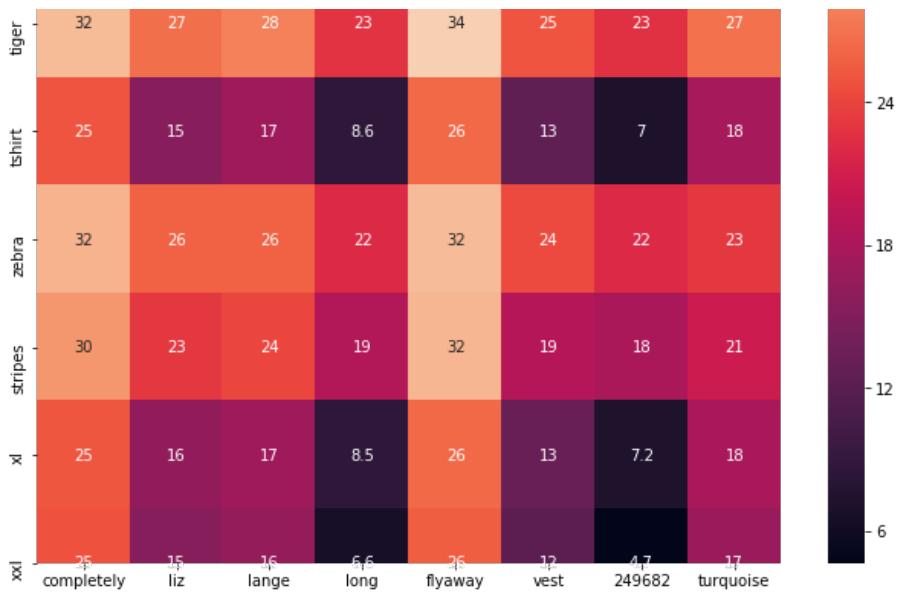


ASIN : B06XYP1X1F

Brand : J Brand Jeans

euclidean distance from input : 24.32128186907087

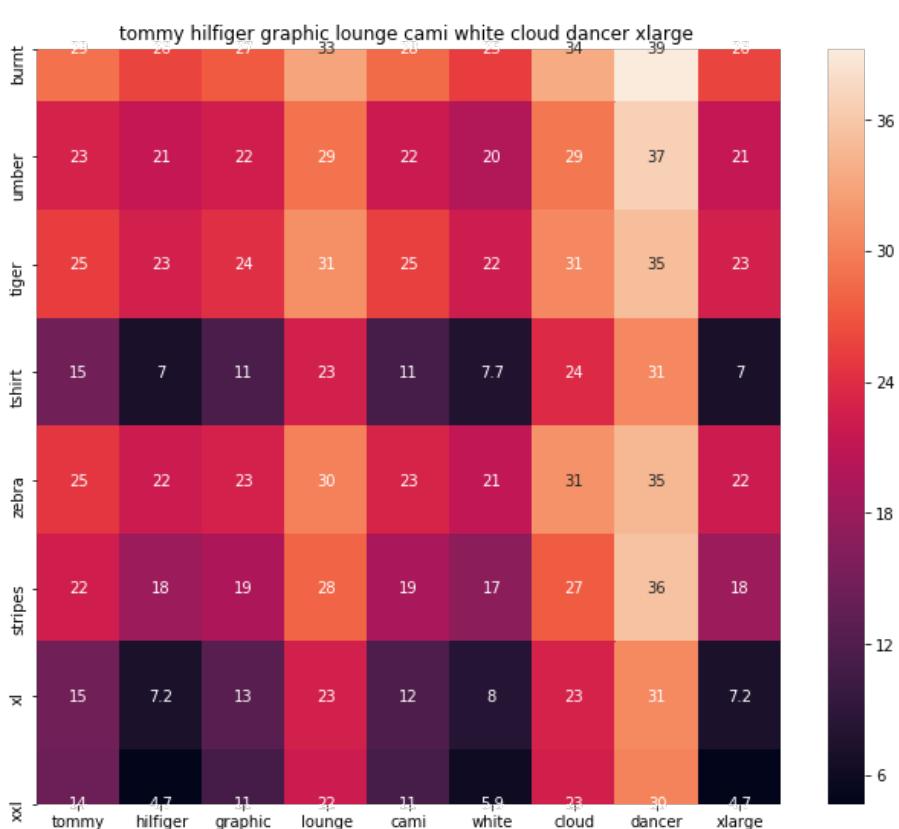




ASIN : B074LTBWSW

Brand : Liz Lange

euclidean distance from input : 24.359863390267165

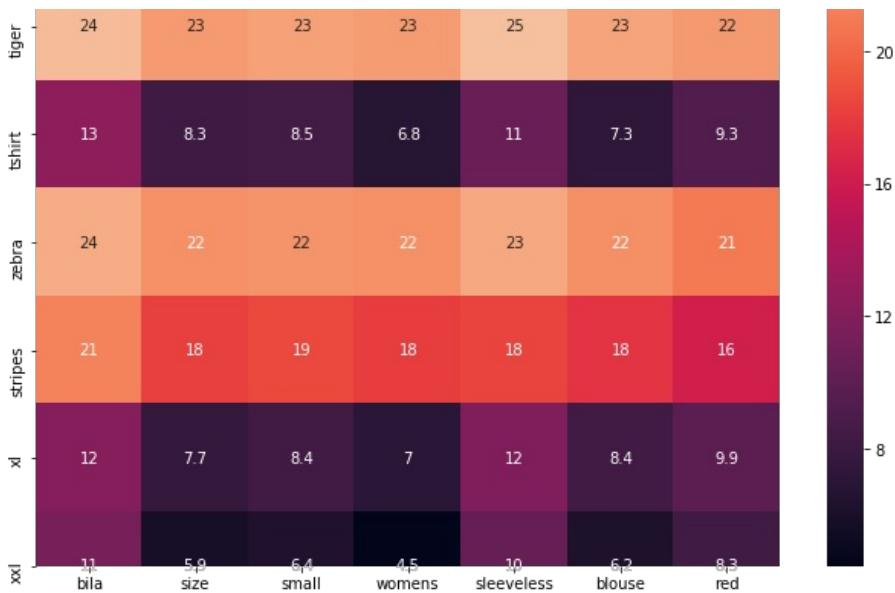


ASIN : B01BMSFYW2

Brand : igertommy hilf

euclidean distance from input : 24.695704868861608

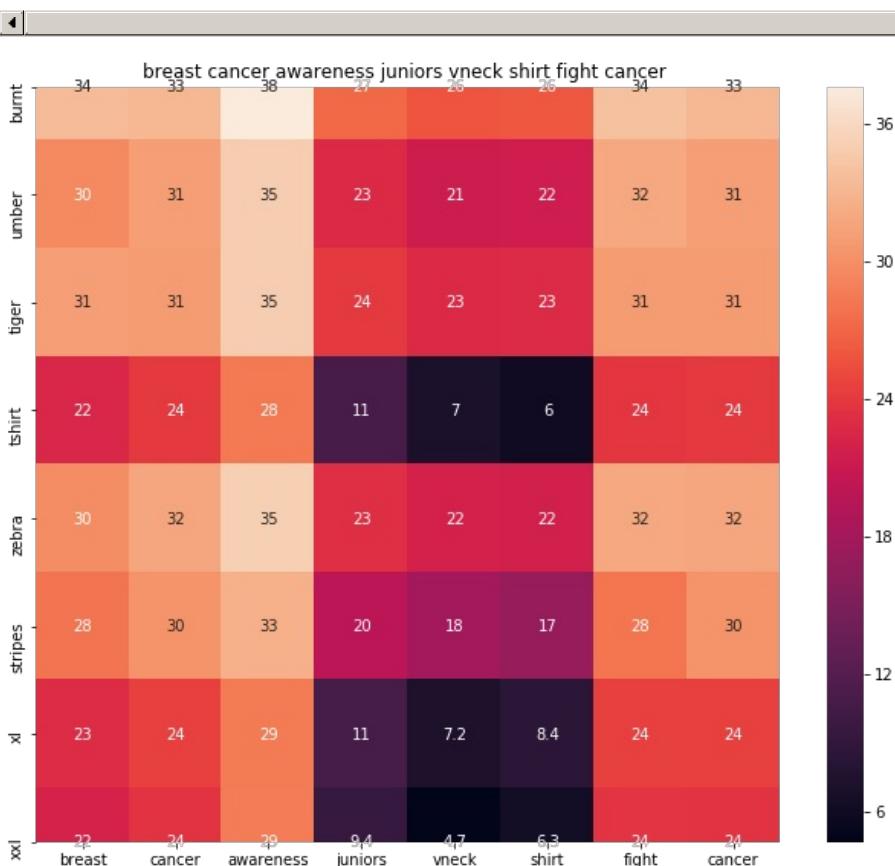




ASIN : B01L7ROZNC

Brand : Bila

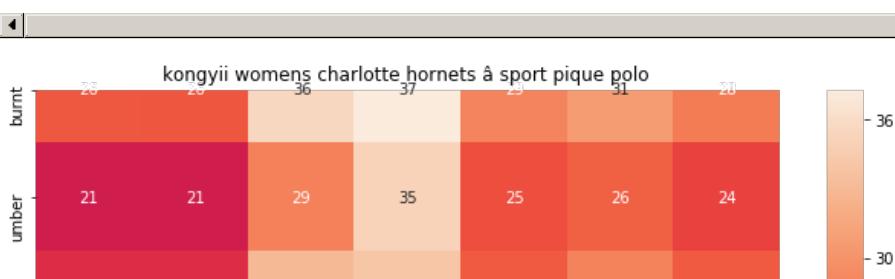
euclidean distance from input : 24.861743971564266

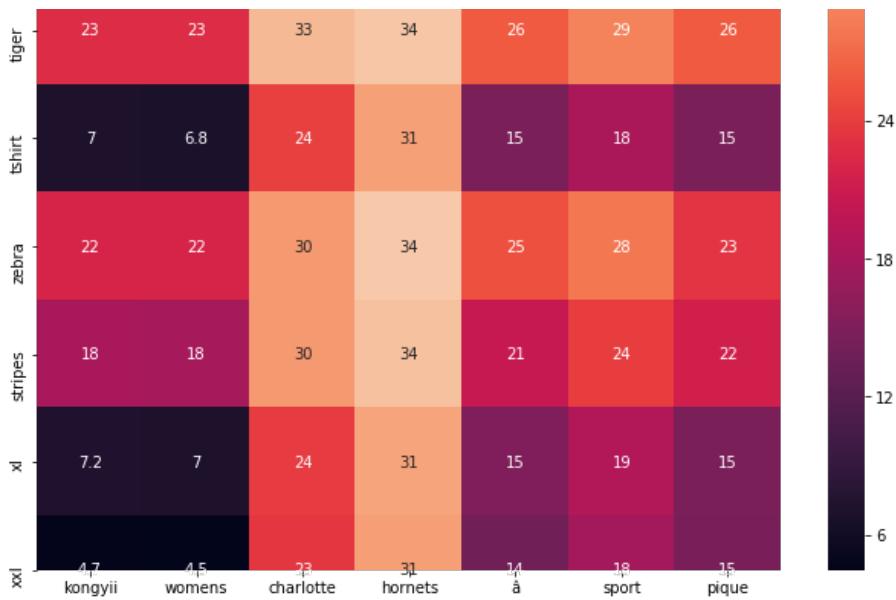


ASIN : B016CU40IY

Brand : Juiceclouds

euclidean distance from input : 25.034124909413038

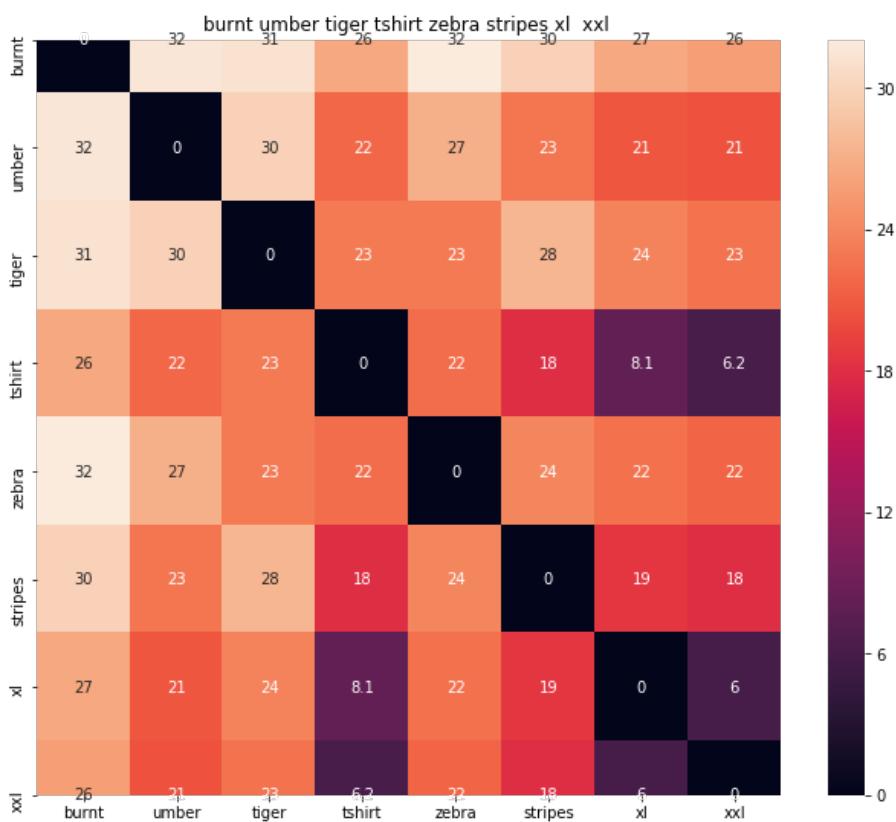




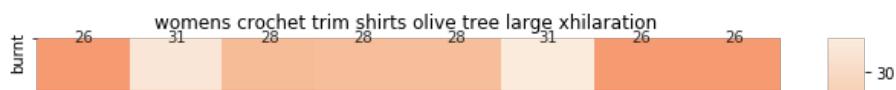
ASIN : B01FJVZST2
 Brand : KONGYII
 euclidean distance from input : 25.296121587356676

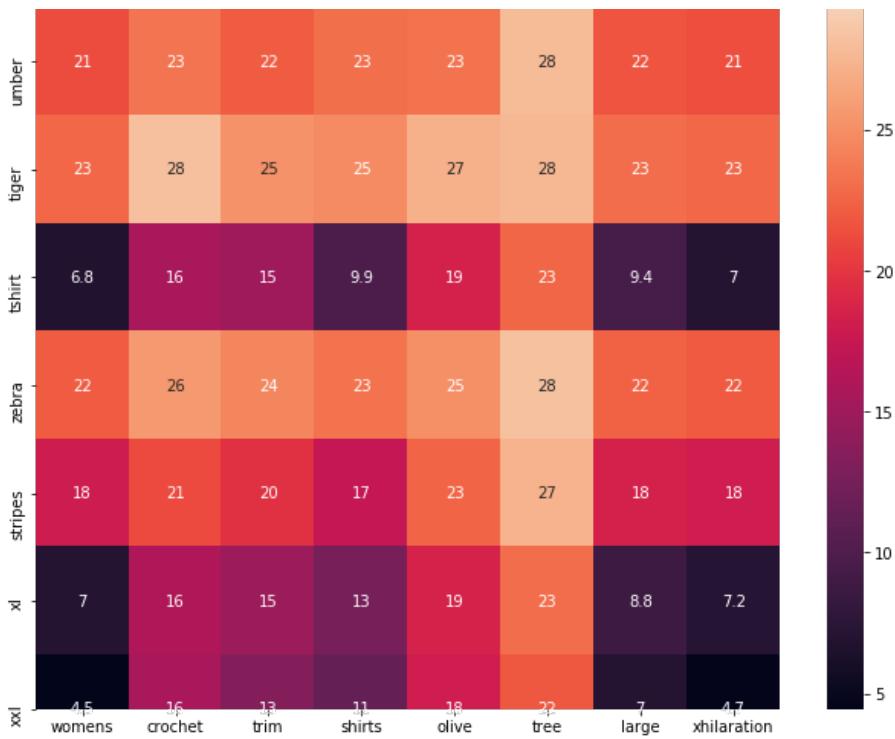
In [50]:

```
function(12566,10,100,1,50,150)
```



ASIN : B00JXQB5FQ
 Brand : Si Row
 euclidean distance from input : 5.600910099684276e-08

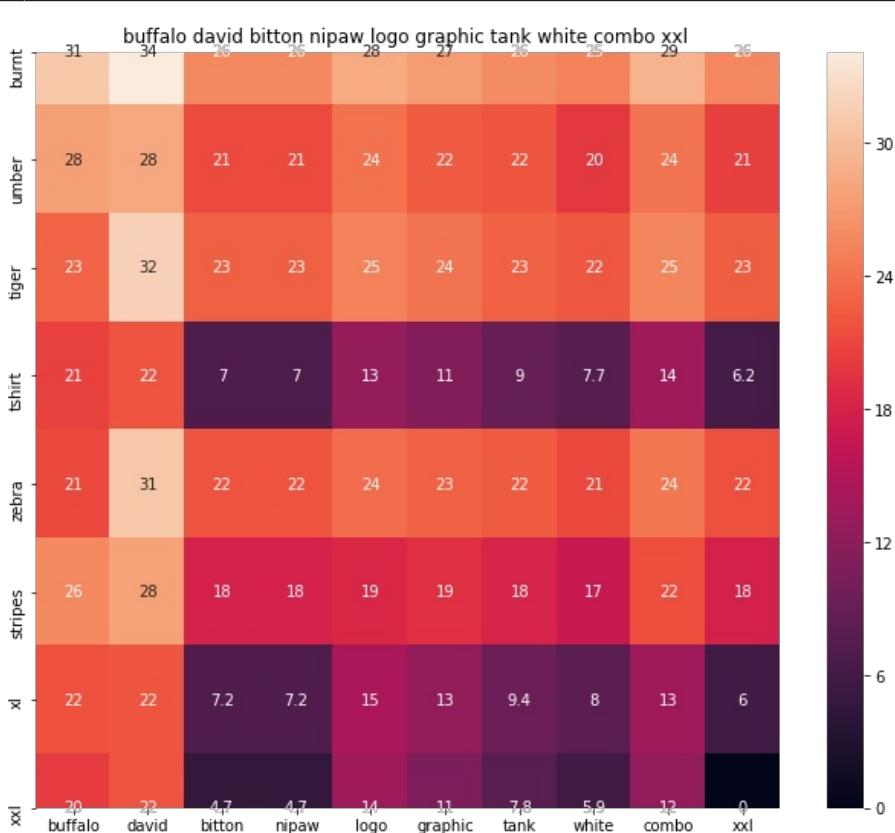




ASIN : B06XBHNM7J

Brand : Xhilaration

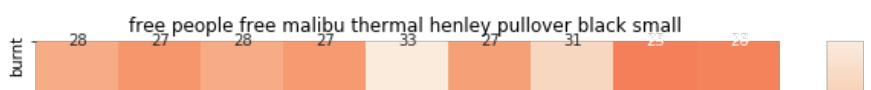
euclidean distance from input : 21.4606461011293

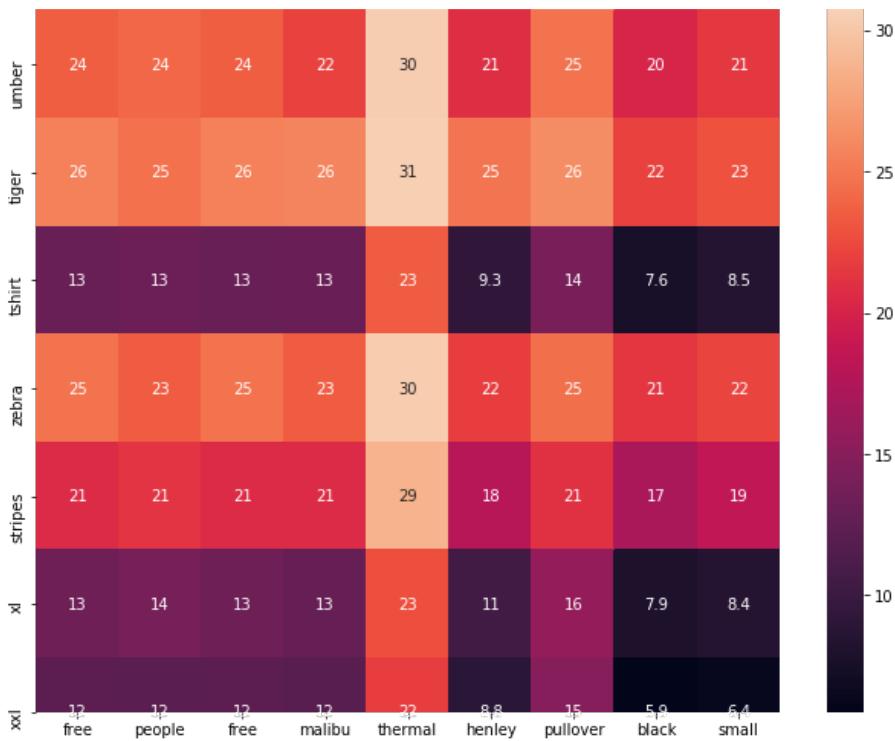


ASIN : B018H5AZXQ

Brand : Buffalo

euclidean distance from input : 22.47430804017363

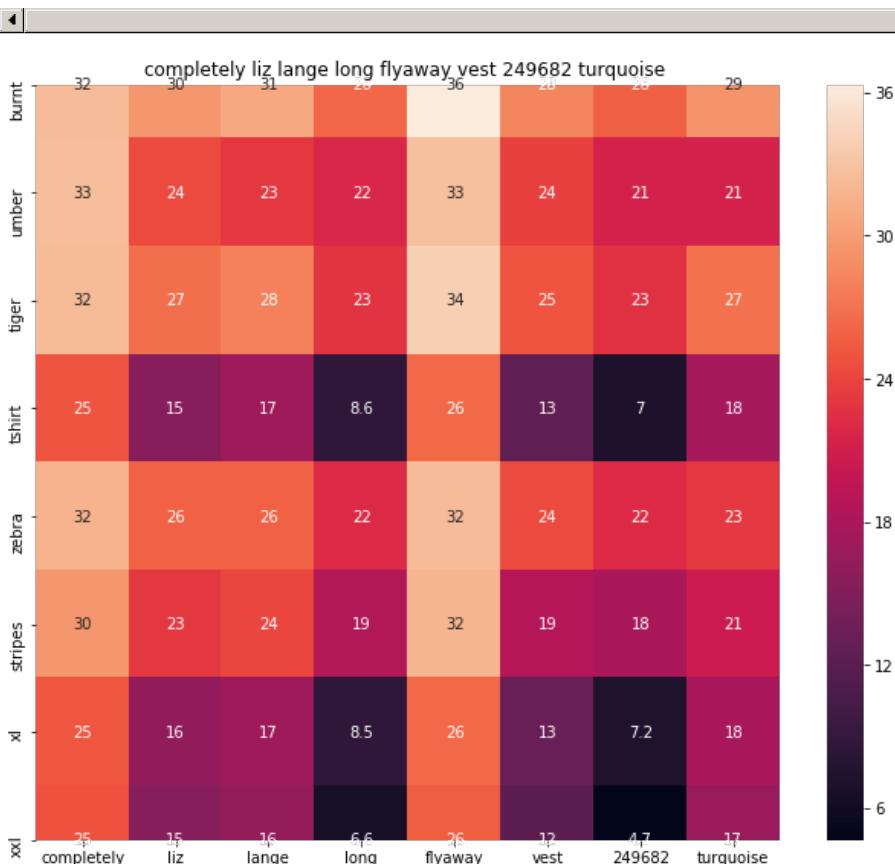




ASIN : B074MXY984

Brand : We The Free

euclidean distance from input : 22.6599049084512

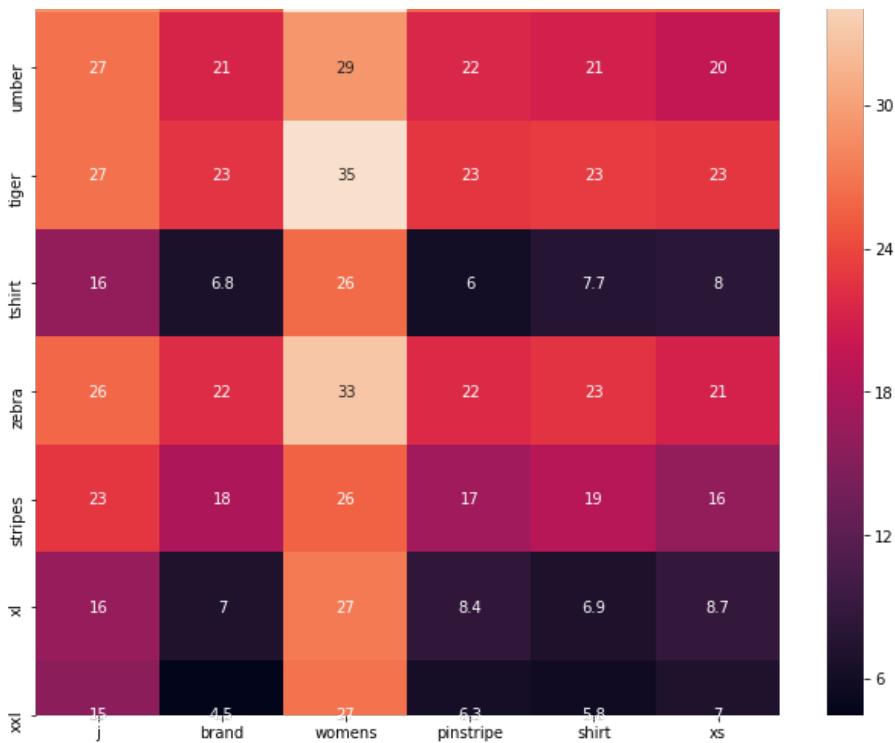


ASIN : B074LTBWSW

Brand : Liz Lange

euclidean distance from input : 22.741488890850494

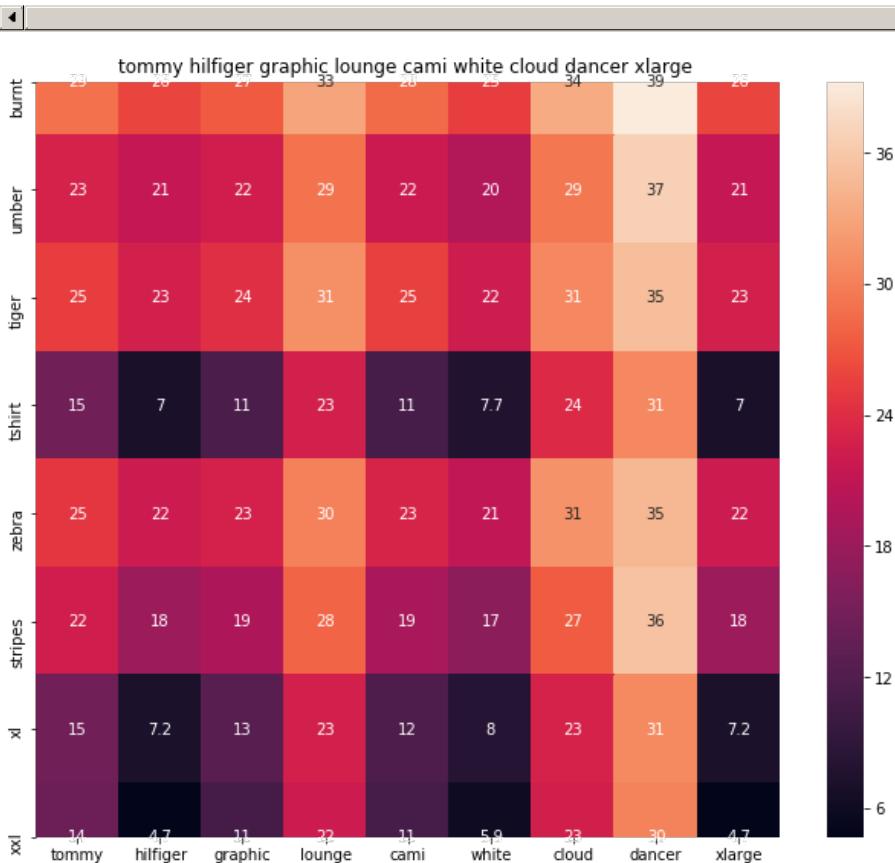




ASIN : B06XYP1X1F

Brand : J Brand Jeans

euclidean distance from input : 22.763983134019416

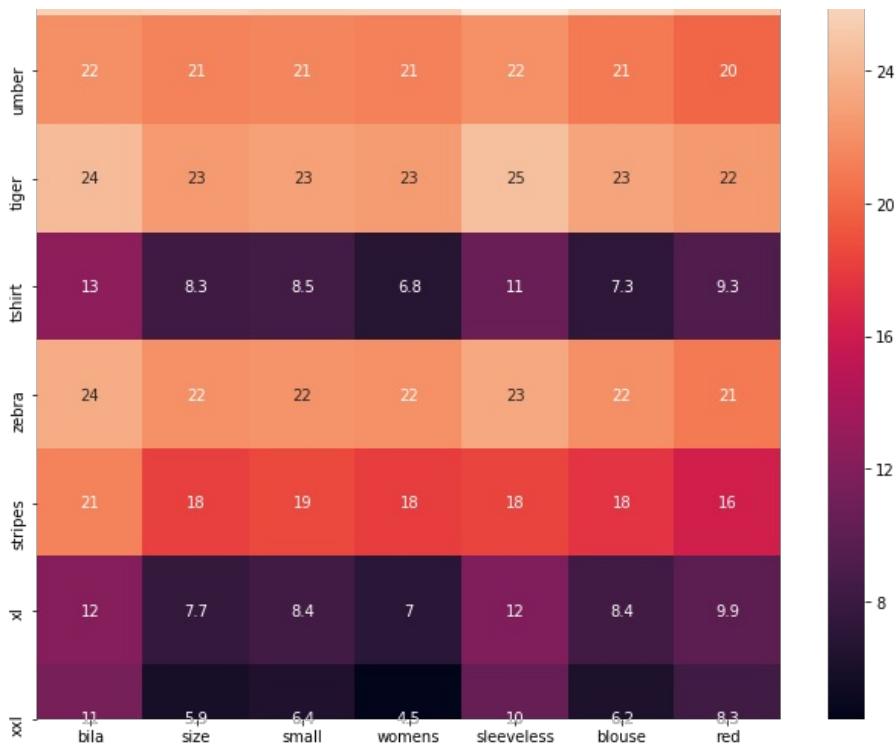


ASIN : B01BMSFYW2

Brand : igertommy hilf

euclidean distance from input : 23.11228359894103

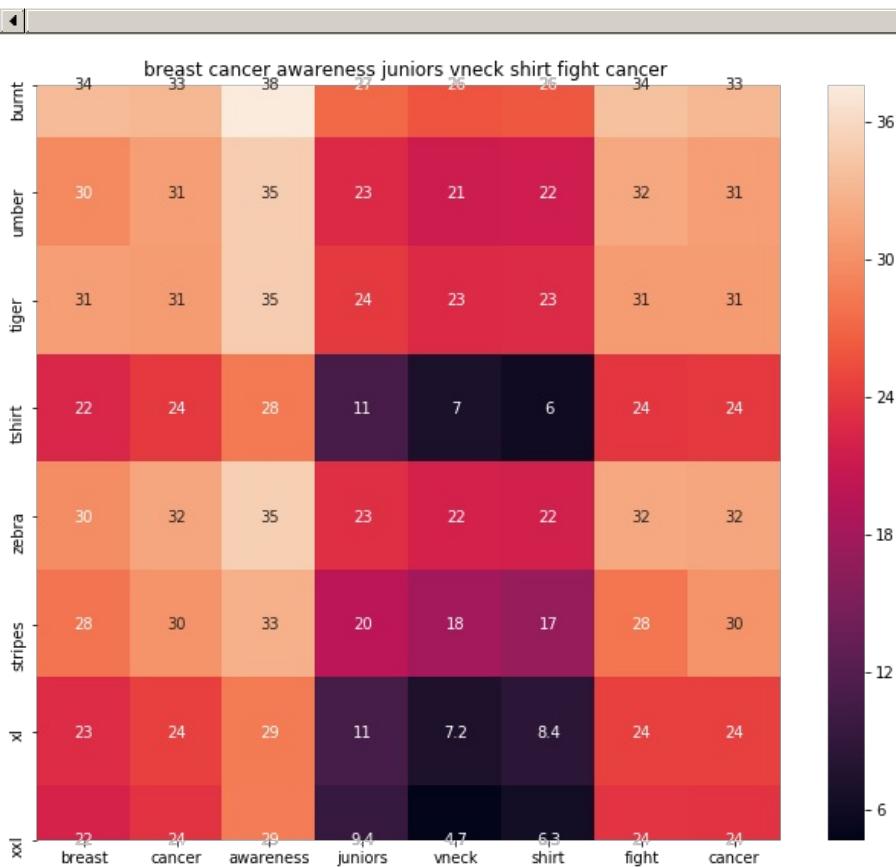




ASIN : B01L7ROZNC

Brand : Bila

euclidean distance from input : 23.21636631774441

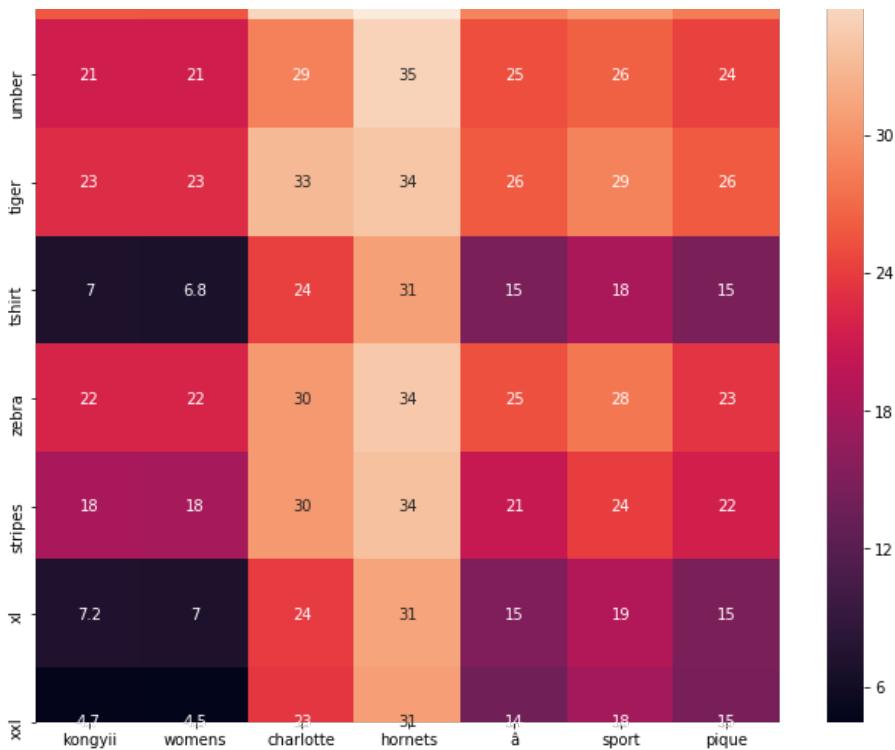


ASIN : B016CU40IY

Brand : Juiceclouds

euclidean distance from input : 23.376720678533967





ASIN : B01FJVZST2

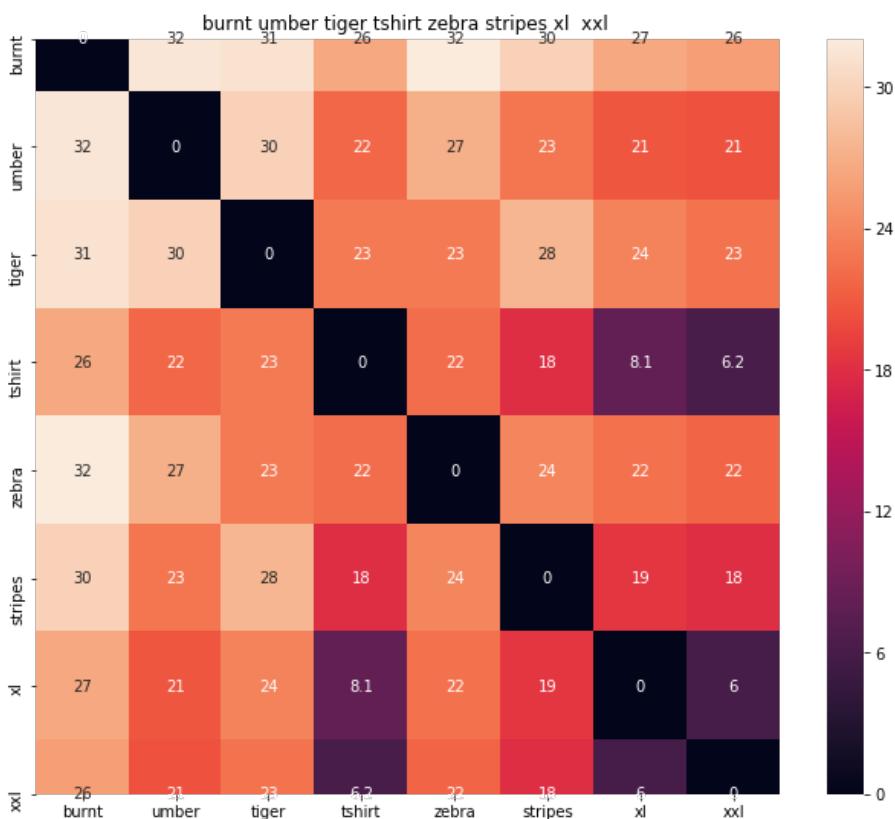
Brand : KONGYII

euclidean distance from input : 23.62043851848154

[52]:

```
function(12566,10,15,100,150,20)
```

(16042, 1) (16042, 1) (16042, 1) (16042, 1)



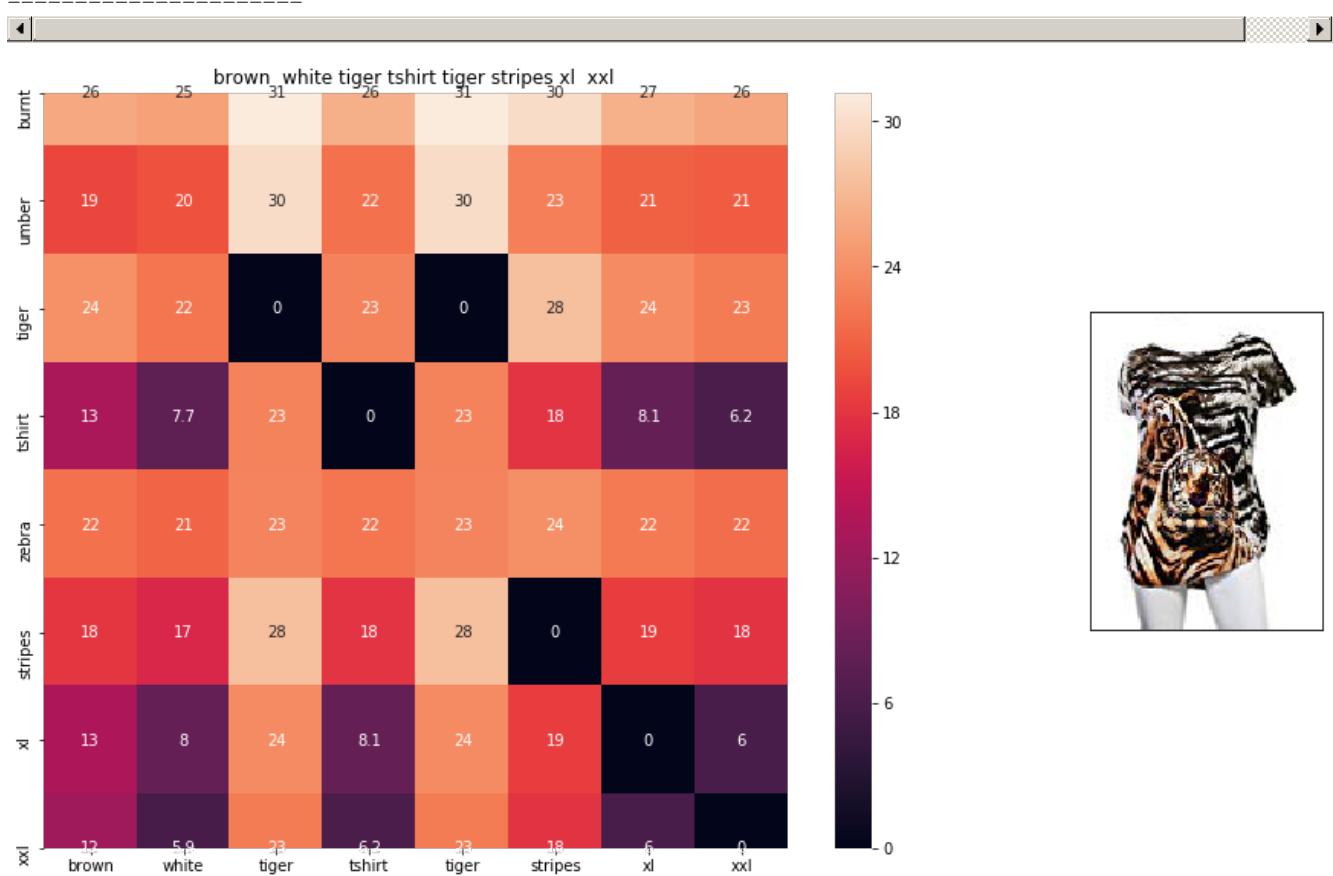
ASIN : B00JXQB5FQ

Brand : Si Row

=====

euclidean distance from input : 8.873020736868248e-09

=====

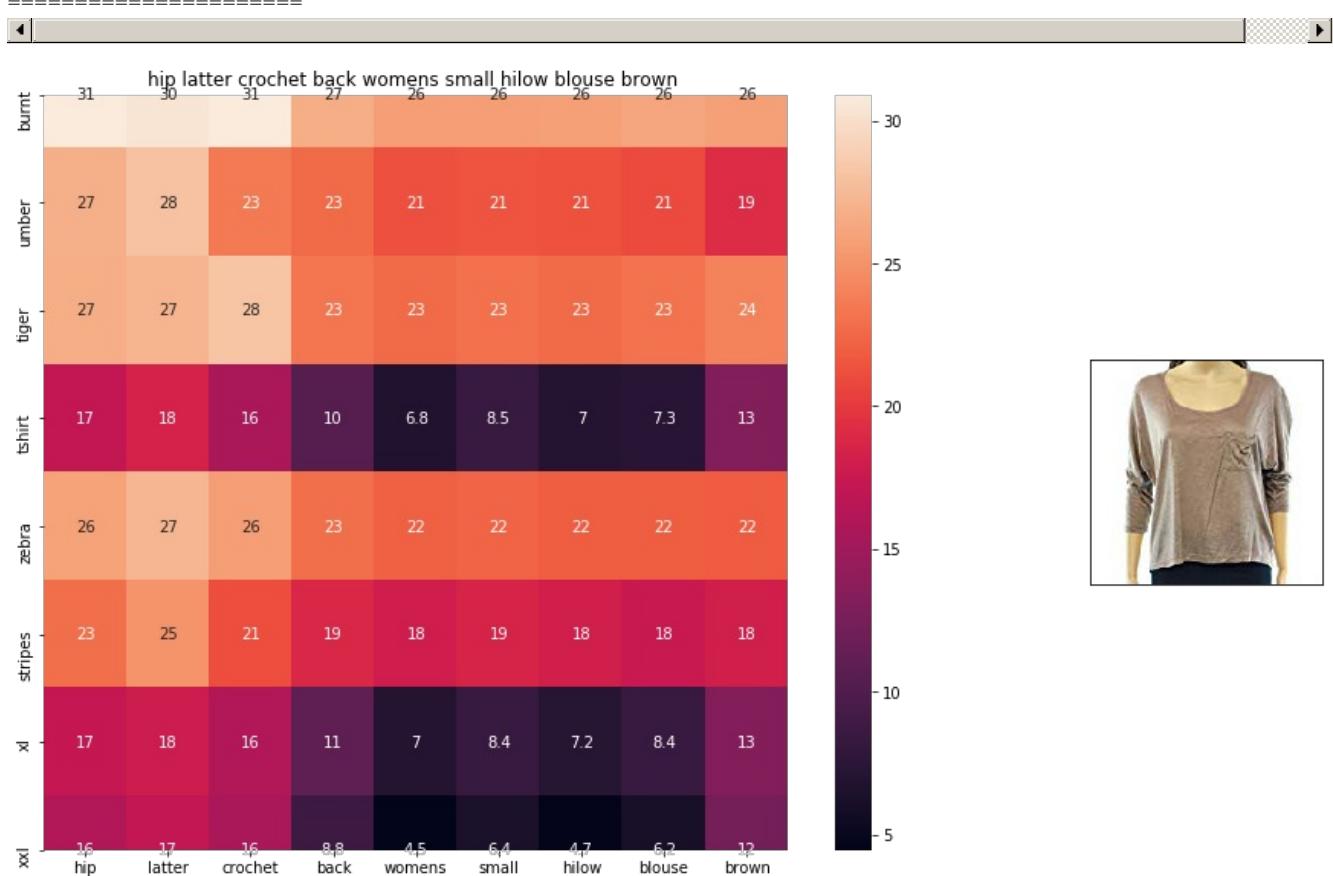


ASIN : B00JXQCWTO

Brand : Si Row

euclidean distance from input : 4.149196584601151

=====



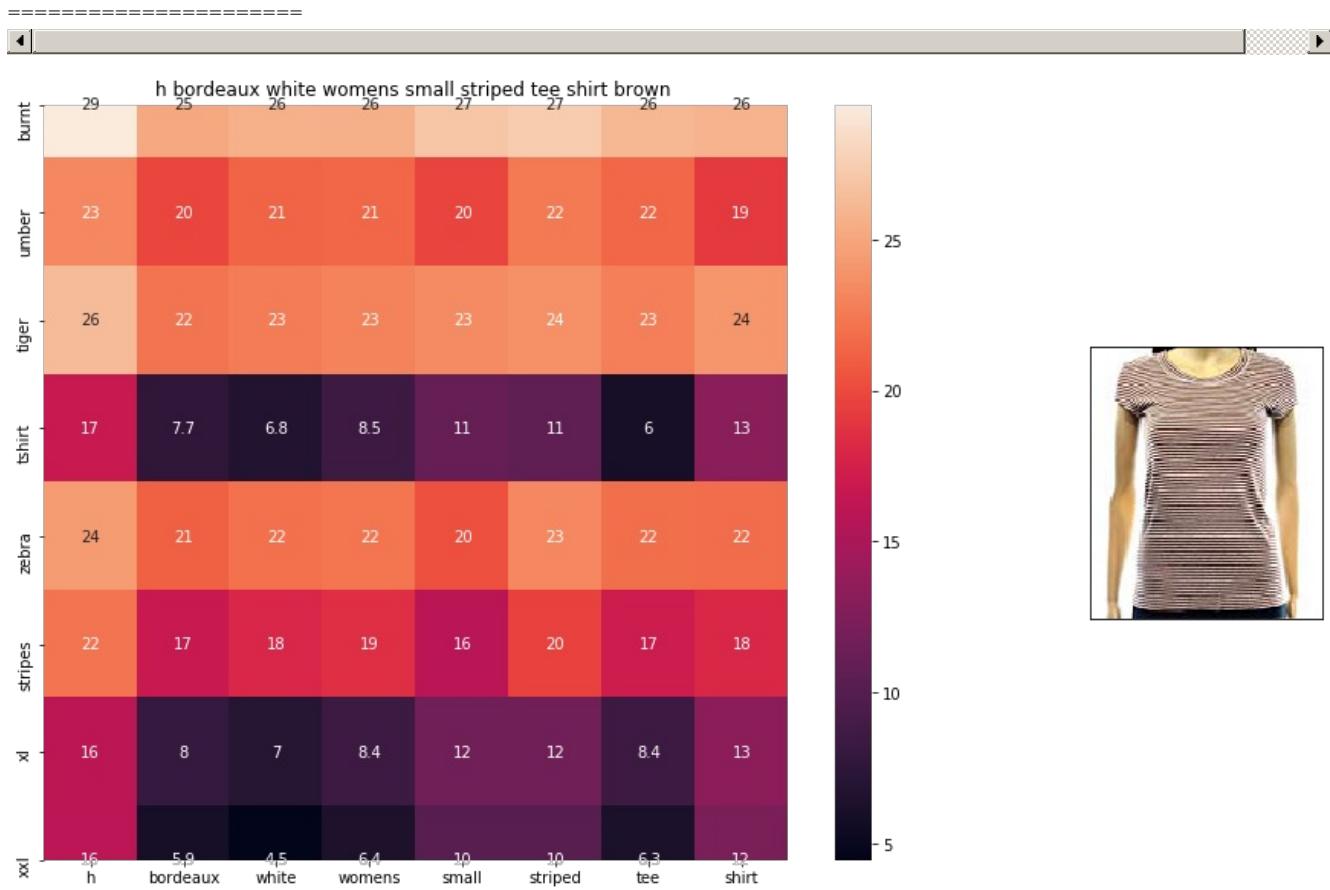
ASIN : B074MJN1K9

Brand : Hin

Brand : H By

euclidean distance from input : 4.239173041761077

=====

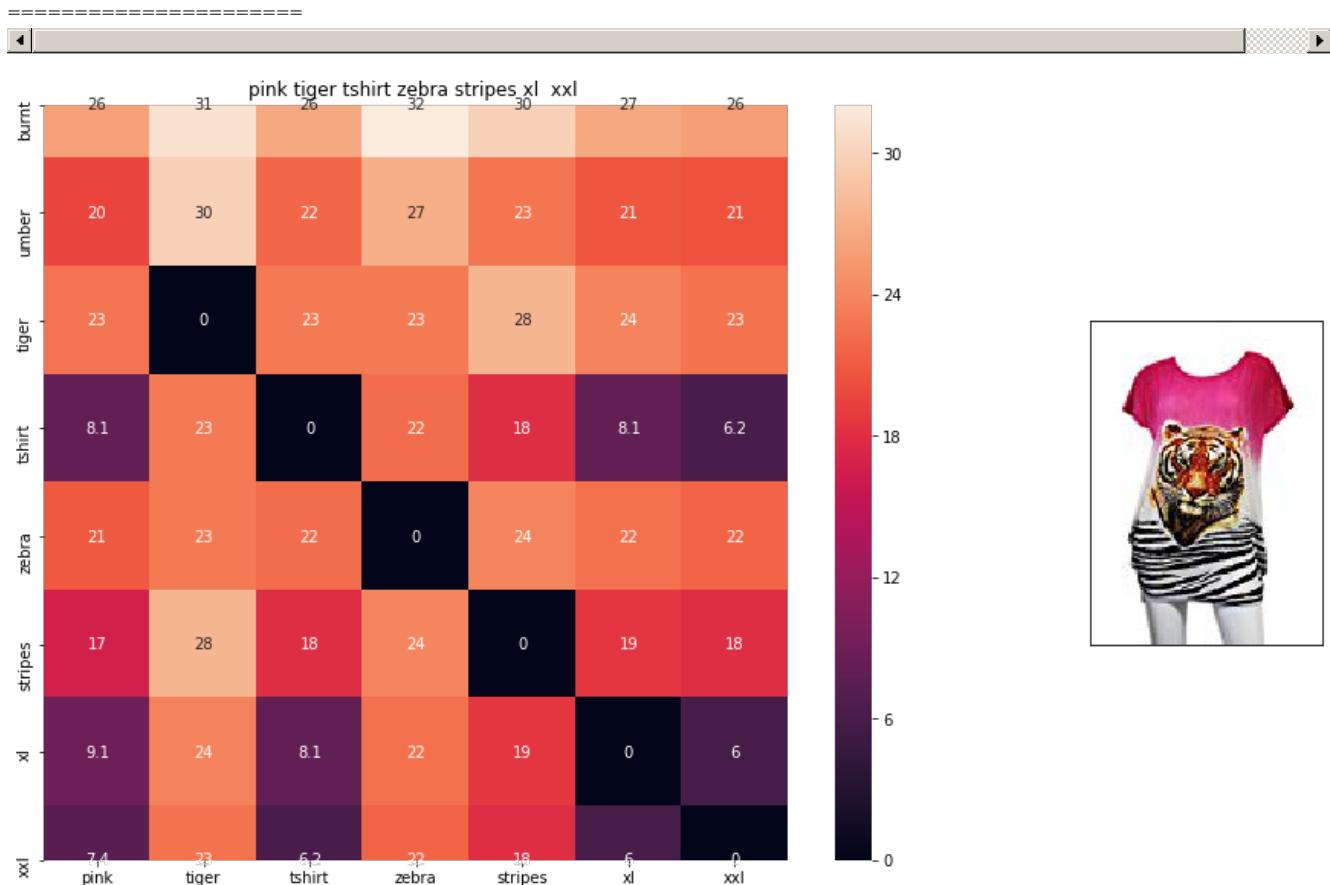


ASIN : B072BVB47Z

Brand : H By Bordeaux

euclidean distance from input : 4.283738574647067

=====



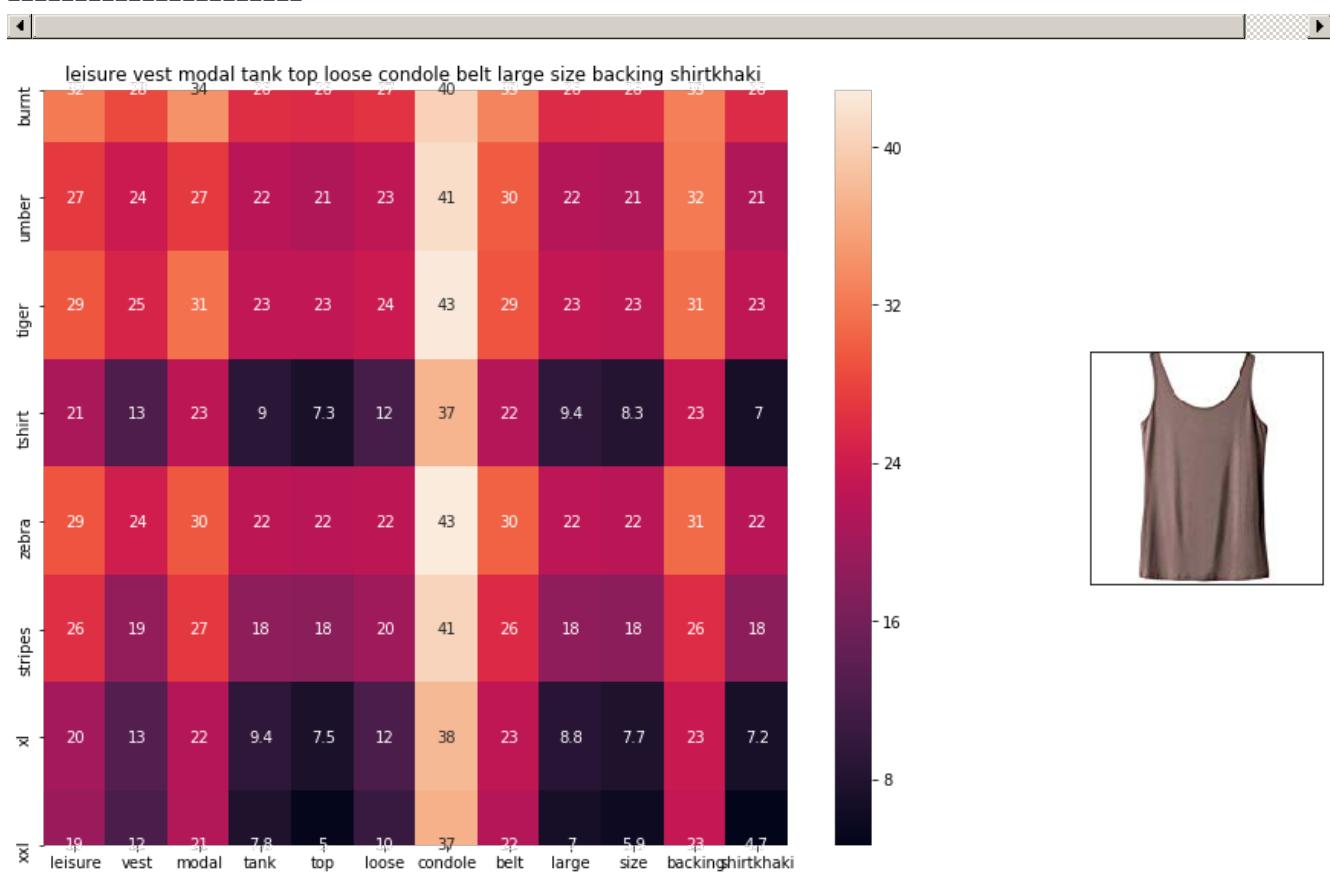
ASIN : B00JXQASS6

Brand : Si Day

Brand : SI ROW

euclidean distance from input : 4.36240898285193

=====

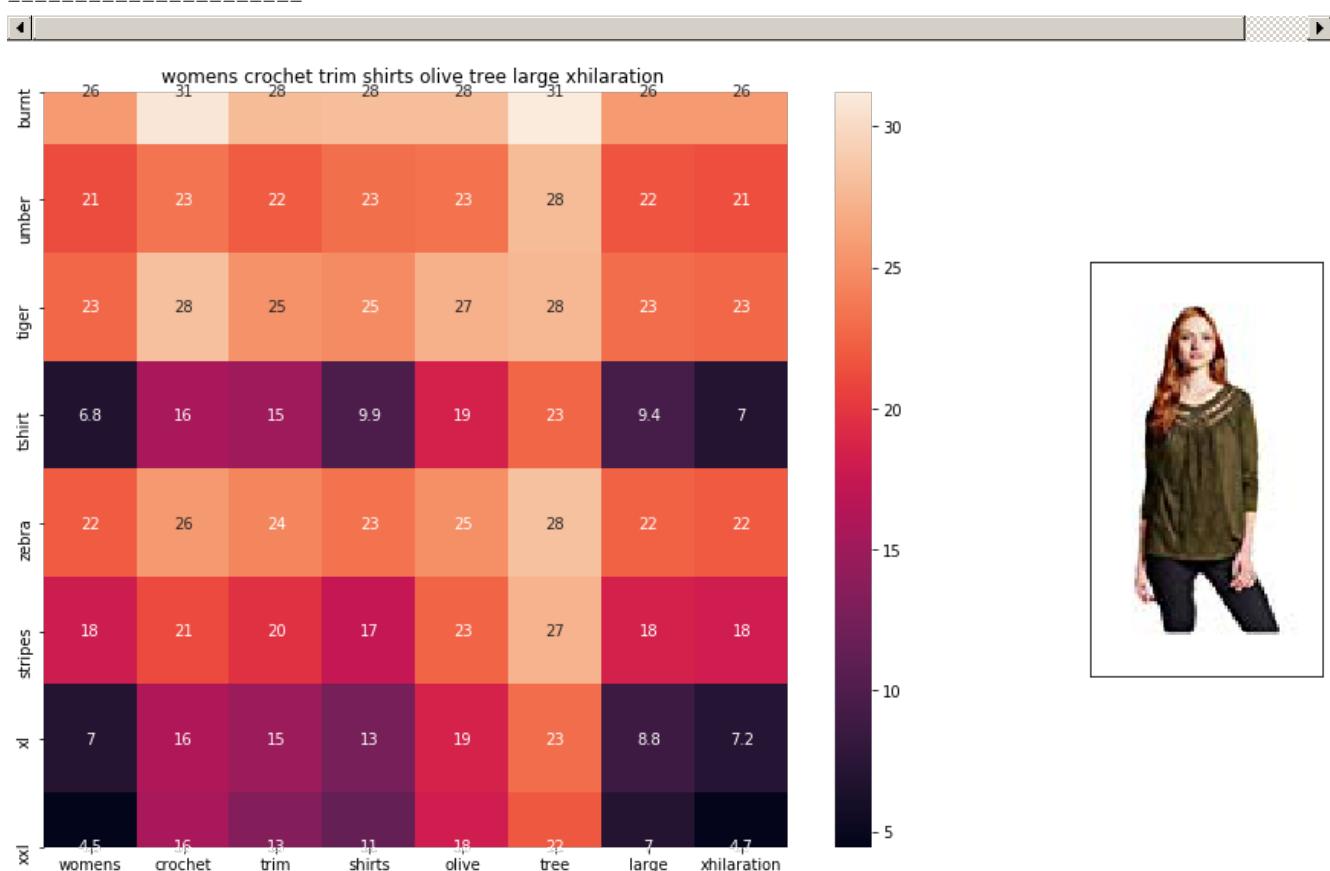


ASIN : B014OUHUZY

Brand : Black Temptation

euclidean distance from input : 4.502417902360882

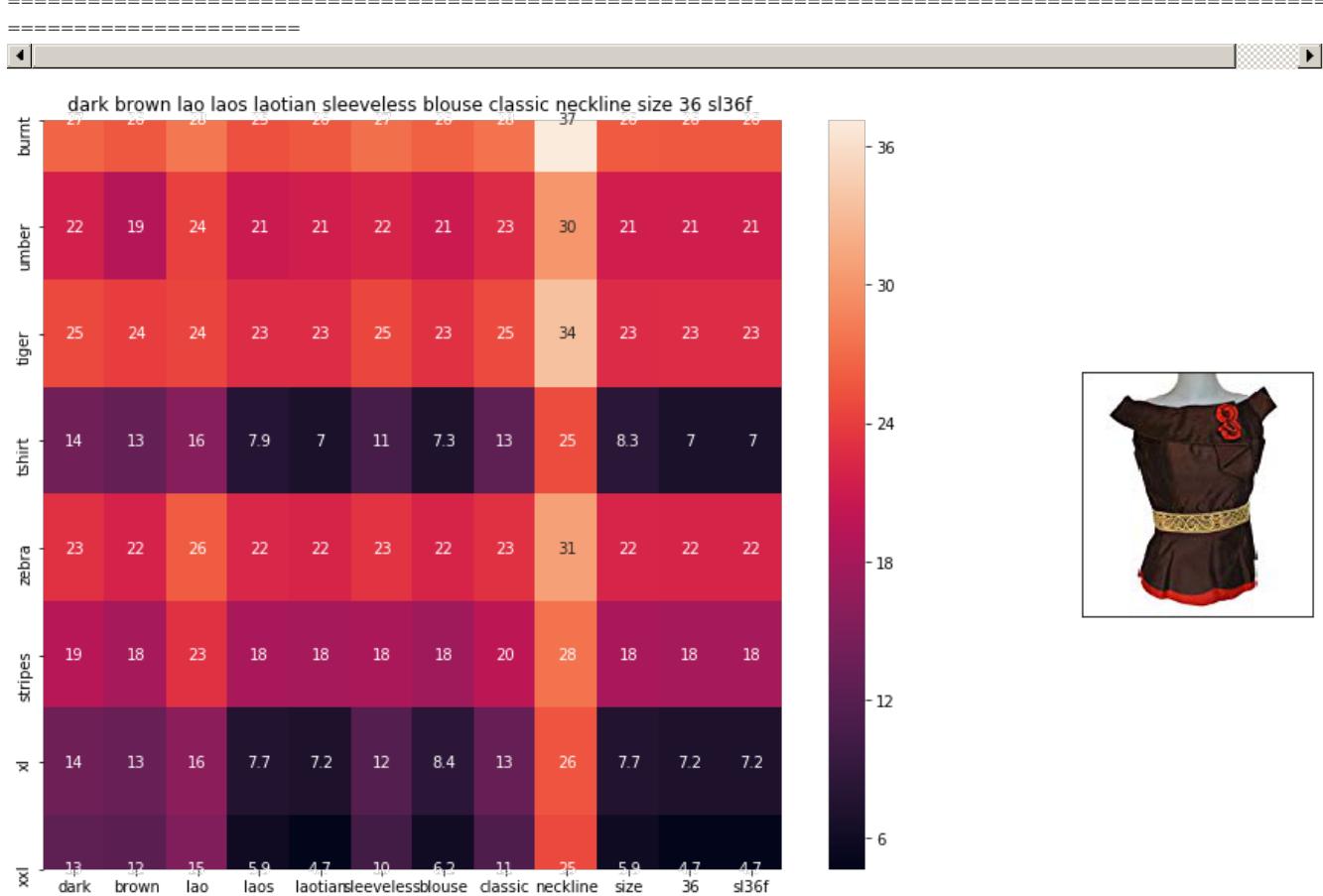
=====



ASIN : B06XBHNM7J

Brand : Xhilaration

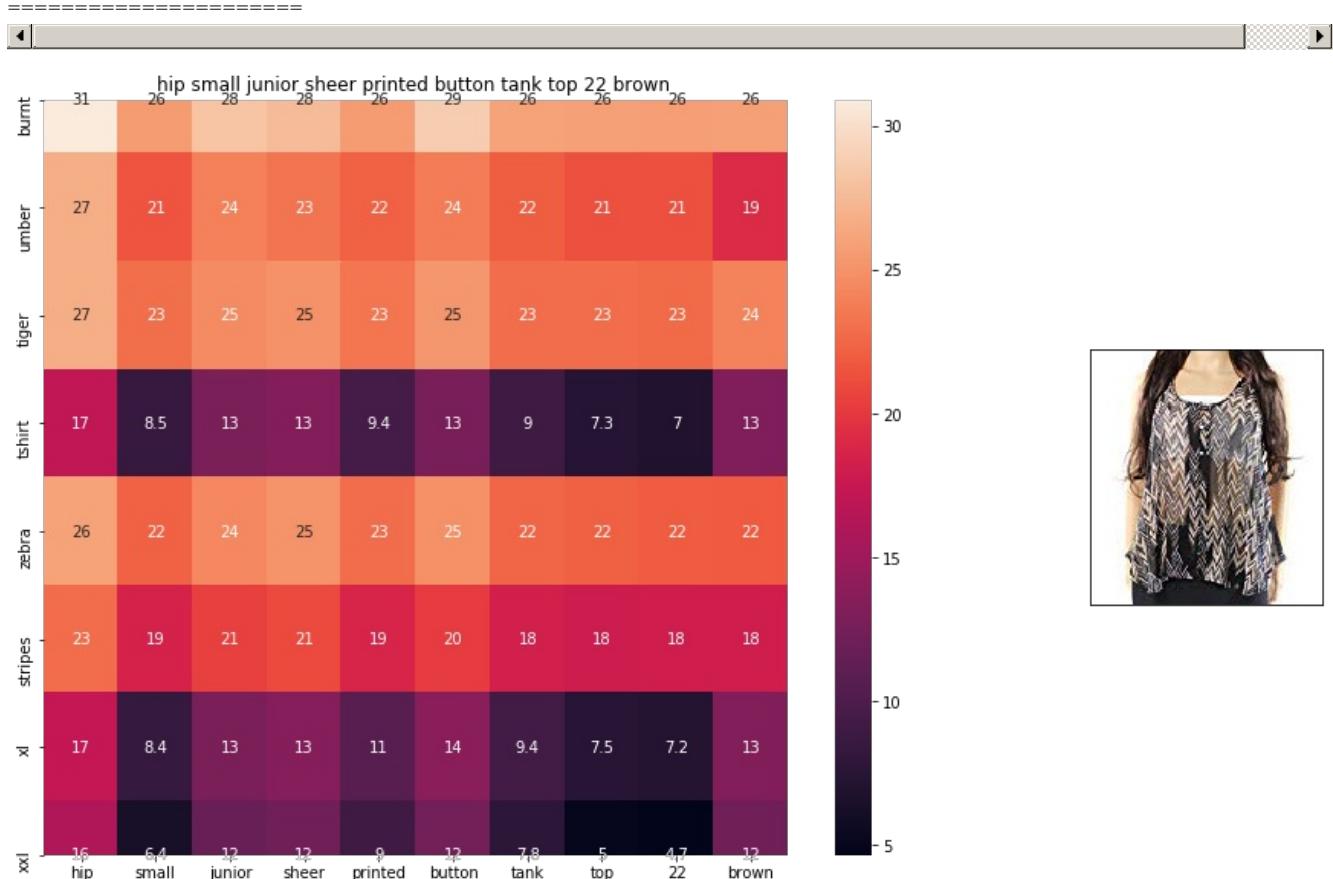
euclidean distance from input : 4.546460680039174



ASIN : B074J7BCYM

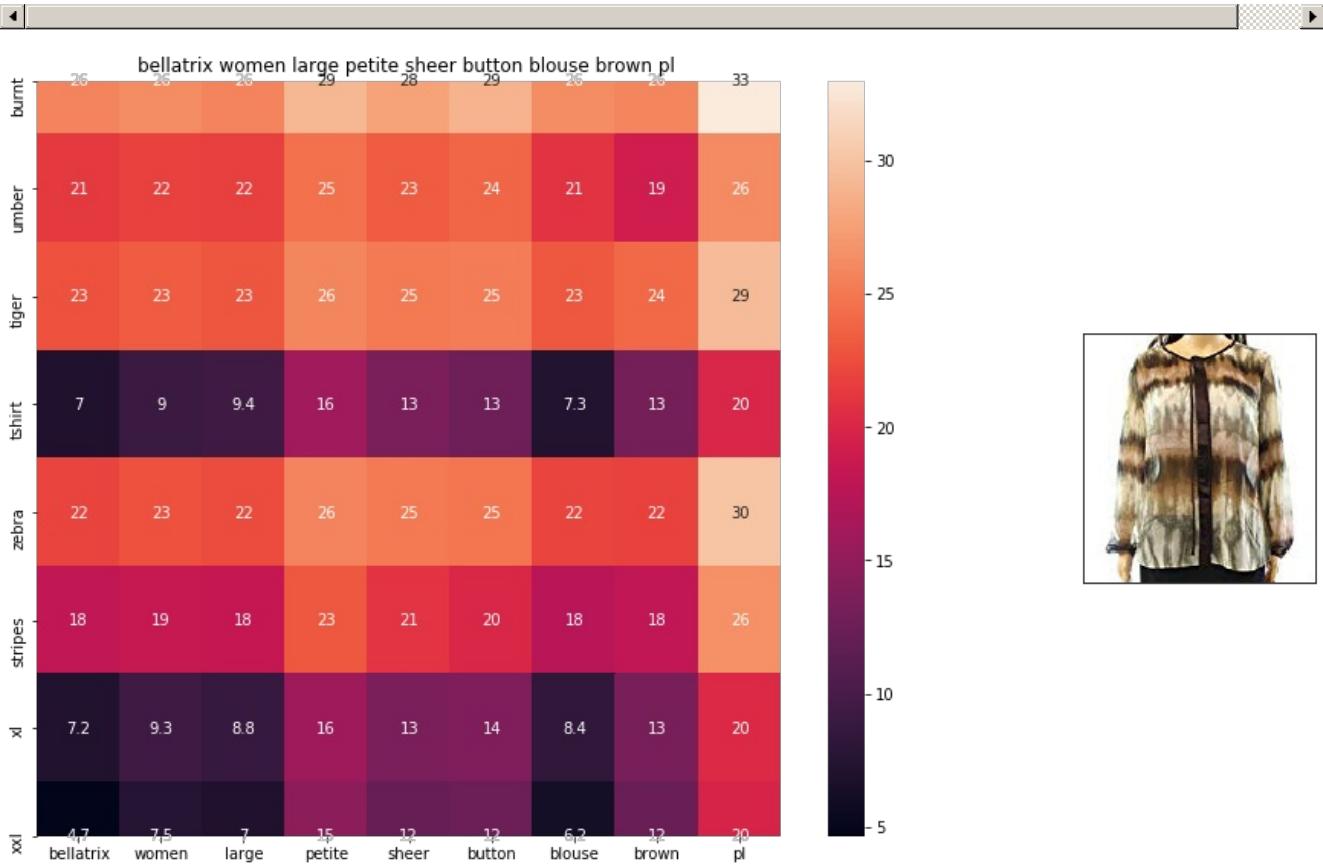
Brand : Nanon

euclidean distance from input : 4.552023913600099



ASIN : B071LDTQ1F

Brand : Hip
euclidean distance from input : 4.564540939196316



ASIN : B074QVMXSQ

Brand : bellatrix

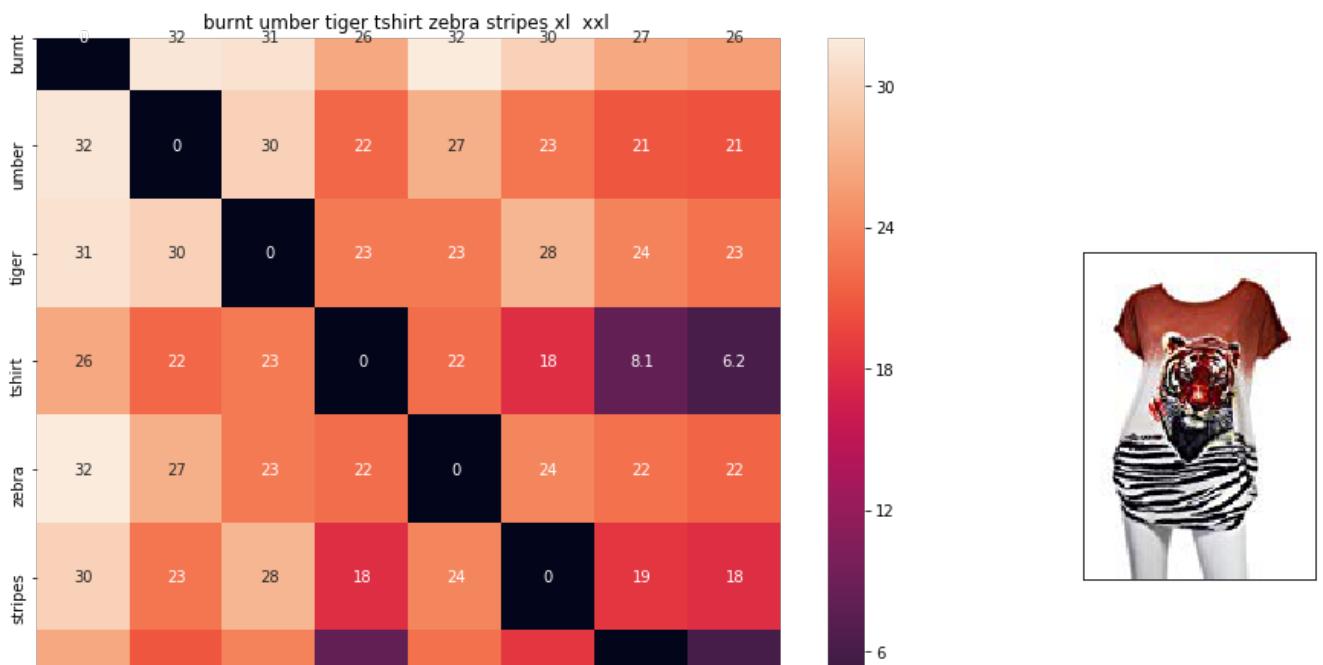
euclidean distance from input : 4.575776735137585



In [54]:

```
function(12566,10,2,100,150,1)
```

(16042, 1) (16042, 1) (16042, 1) (16042, 1)
(16042, 1)

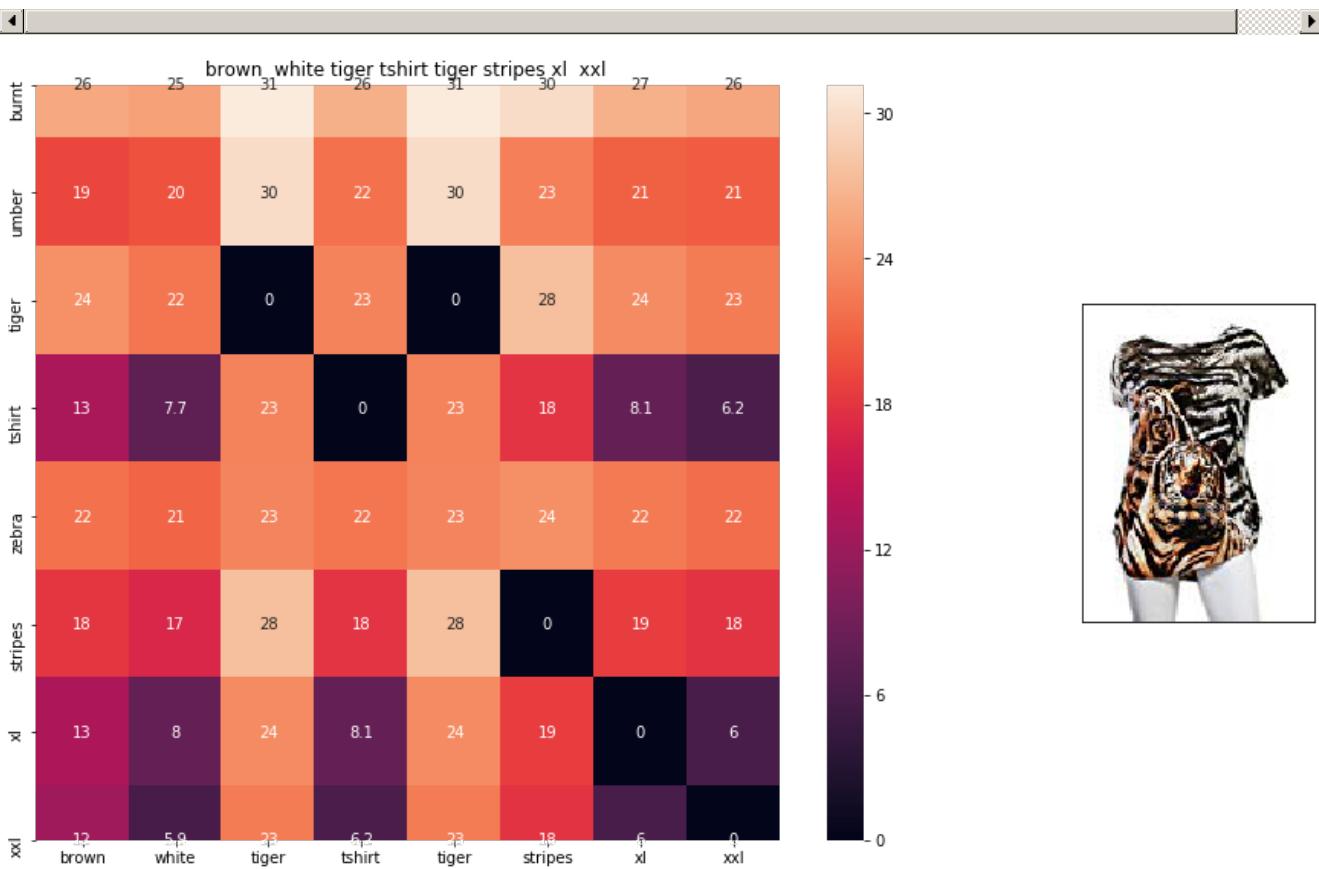




ASIN : B00JXQB5FQ

Brand : Si Row

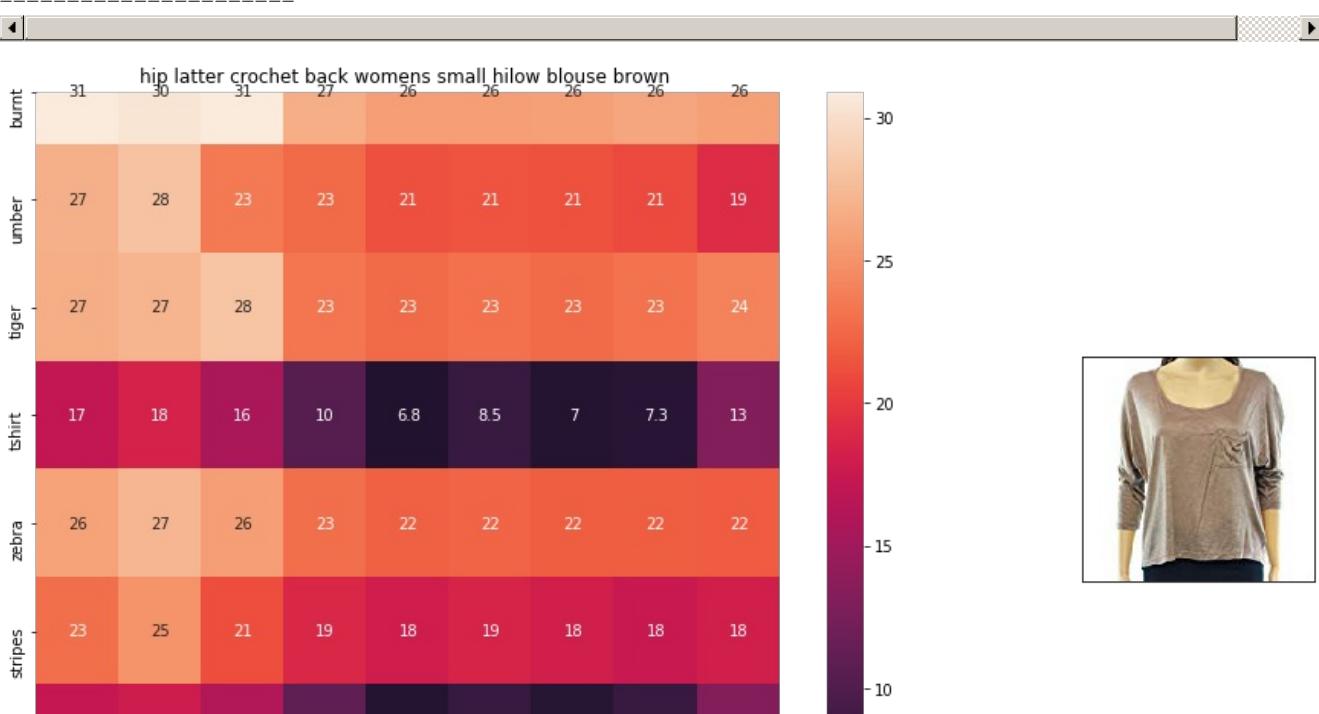
euclidean distance from input : 1.332706649472931e-09



ASIN : B00JXQCWTO

Brand : Si Row

euclidean distance from input : 0.257271661117614





ASIN : B074MJN1K9

Brand : Hip

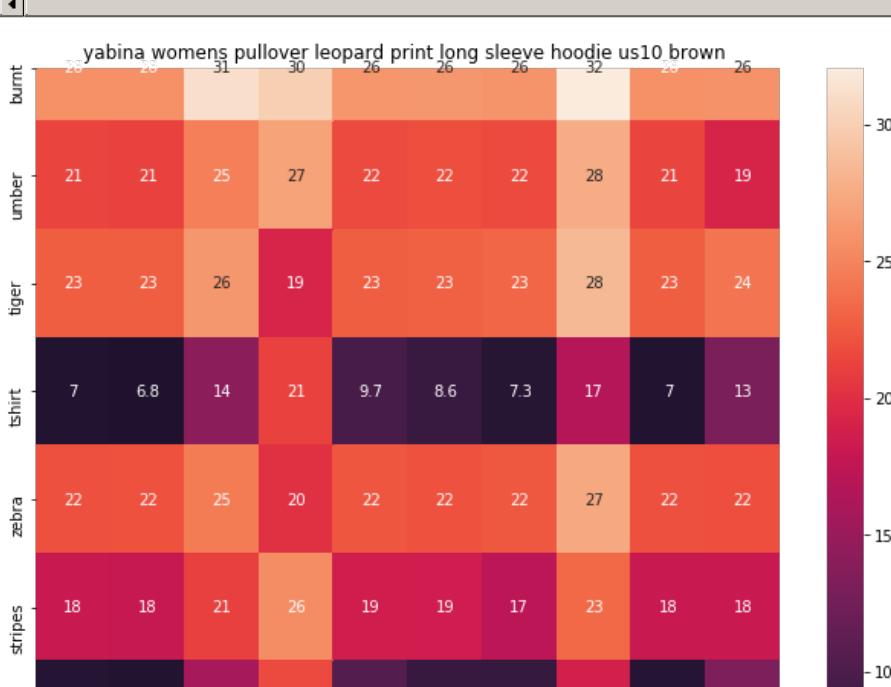
euclidean distance from input : 0.9305942247403348



ASIN : B074J7BCYM

Brand : Nanon

euclidean distance from input : 0.9440671172904753

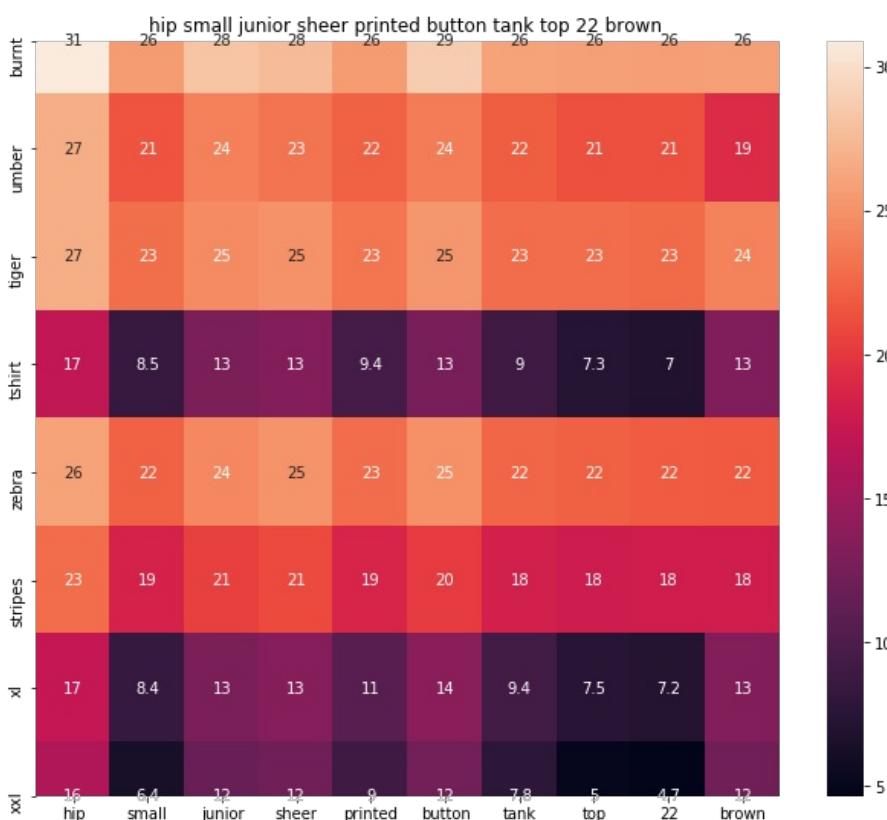




ASIN : B01KJUM6JI

Brand : YABINA

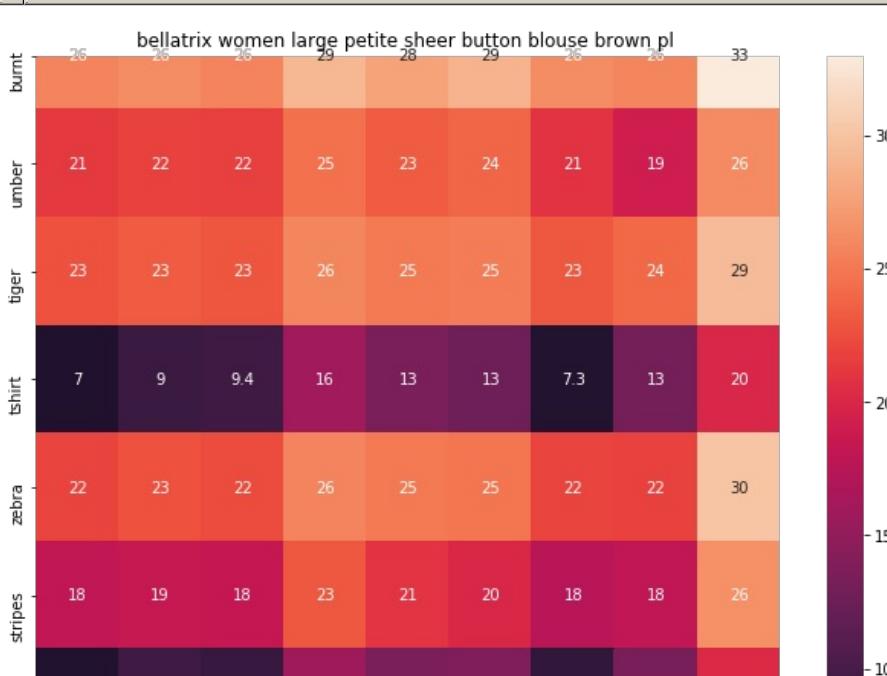
euclidean distance from input : 0.9453613701912134

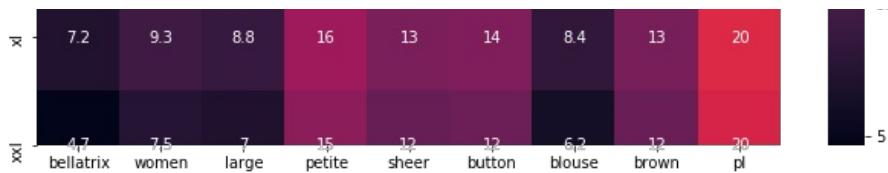


ASIN : B071LDTQ1F

Brand : Hip

euclidean distance from input : 0.9474402369613584

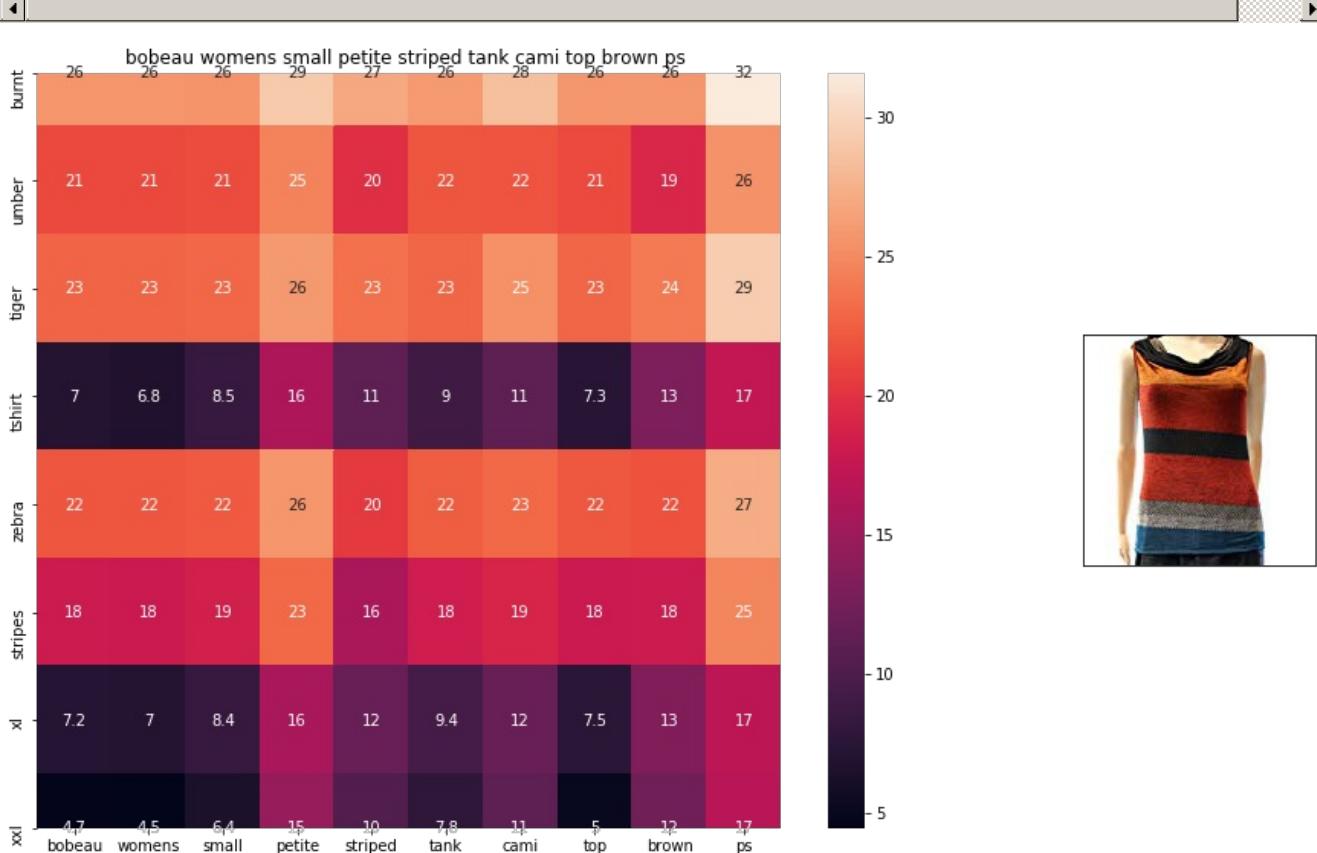




ASIN : B074QVMXSQ

Brand : bellatrix

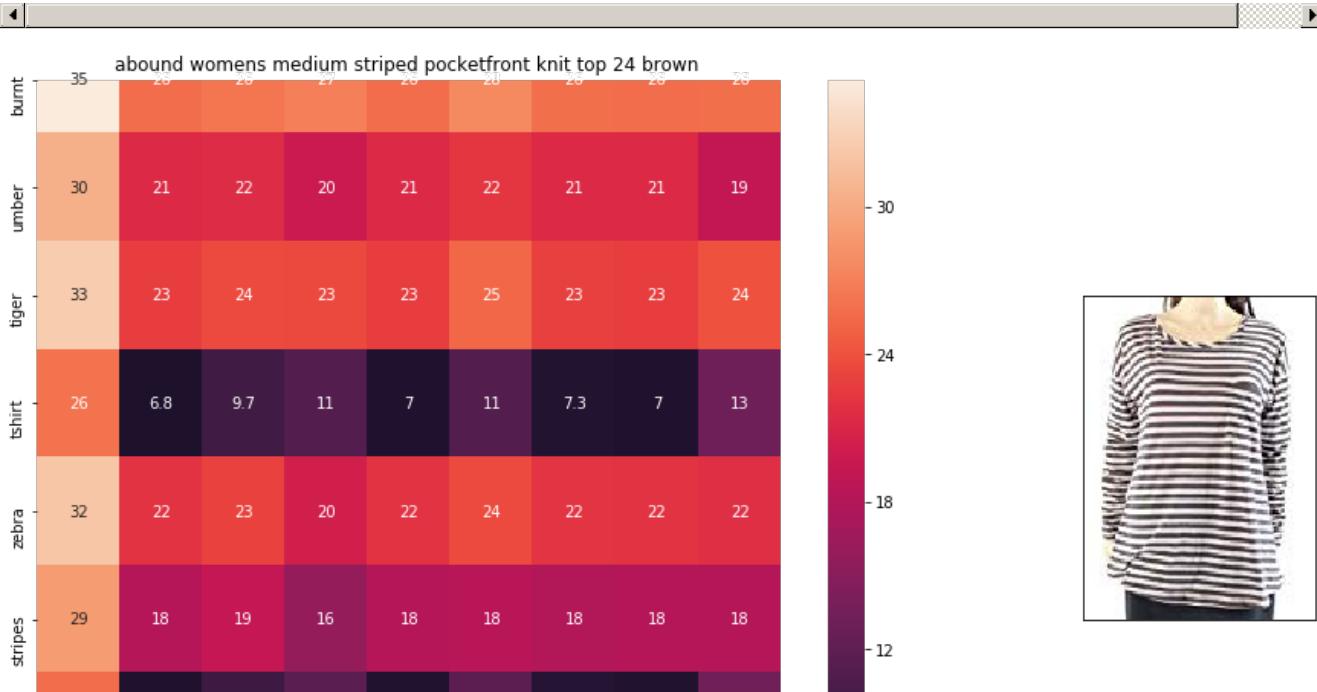
euclidean distance from input : 0.948711496133519



ASIN : B074P8YWV4

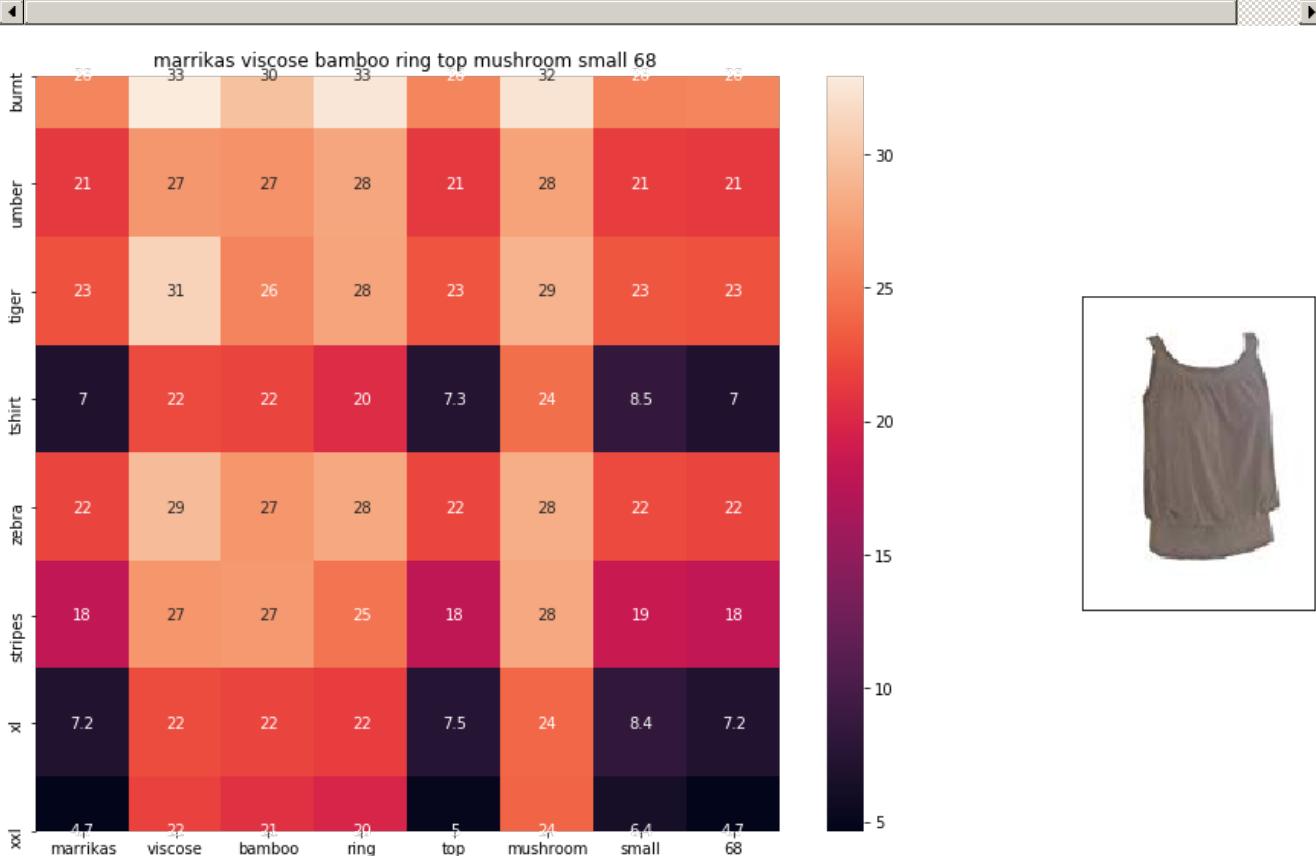
Brand : Bobeau

euclidean distance from input : 0.9495037280890661





ASIN : B072M4ZF89
 Brand : Abound
 euclidean distance from input : 0.9495496703126565

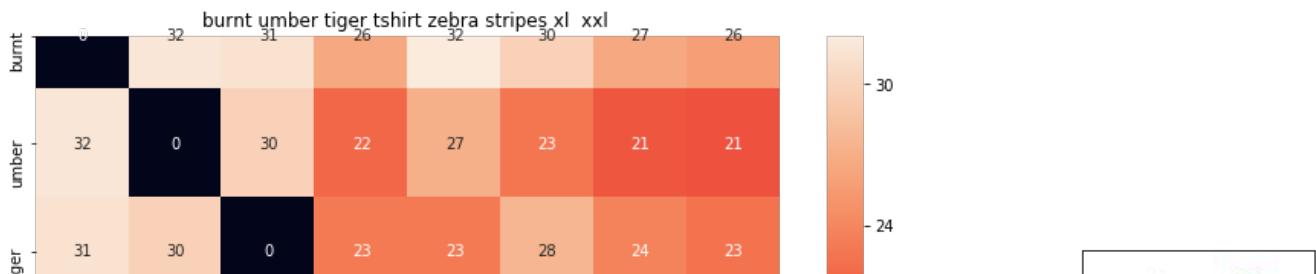


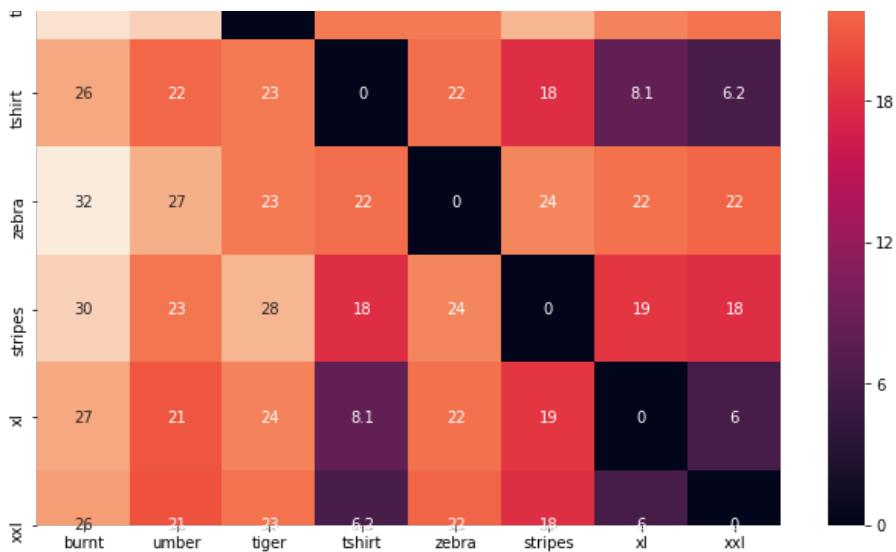
ASIN : B003SPYNAM
 Brand : Marrikas
 euclidean distance from input : 0.9500819298956584

In [57]:

```
function(12566,10,3,100,15,1)

(16042, 1) (16042, 1) (16042, 1) (16042, 1)
(16042, 1)
[12566 14072 14089 14790 10569 7330 7482 7105 13041 12934]
Int64Index([142055, 160670, 160819, 169492, 116737, 72149, 74241, 68773,
           147862, 146397],
           dtype='int64')
[142055, 160670, 160819, 169492, 116737, 72149, 74241, 68773, 147862, 146397]
```

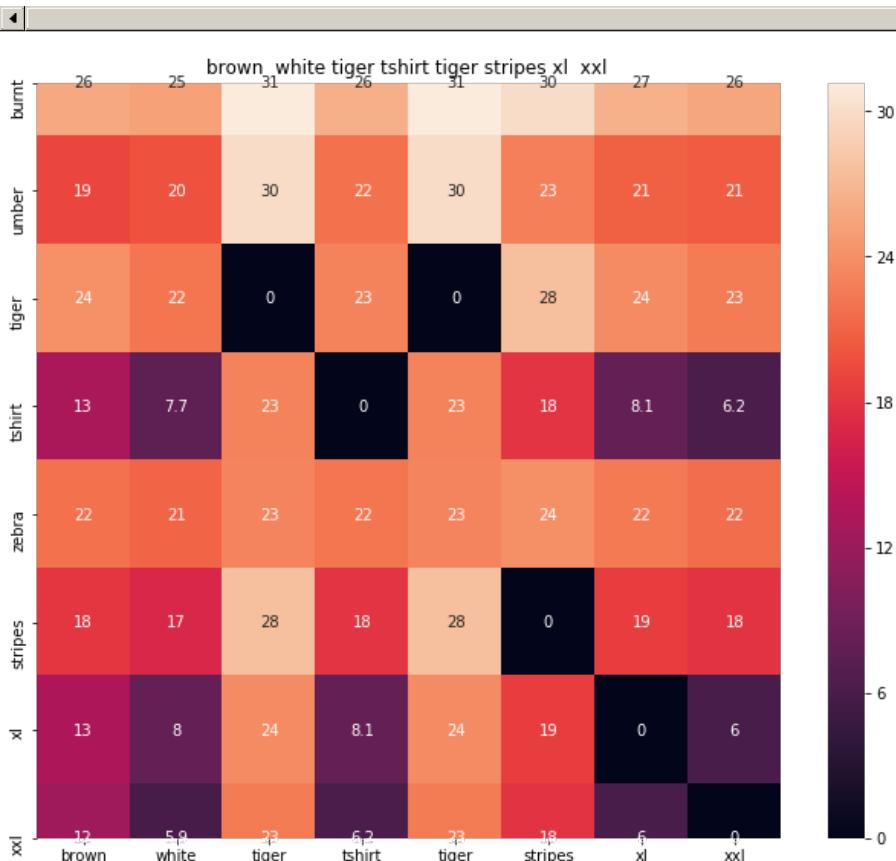




ASIN : B00JXQB5FQ

Brand : Si Row

euclidean distance from input : 4.2501021786901056e-09

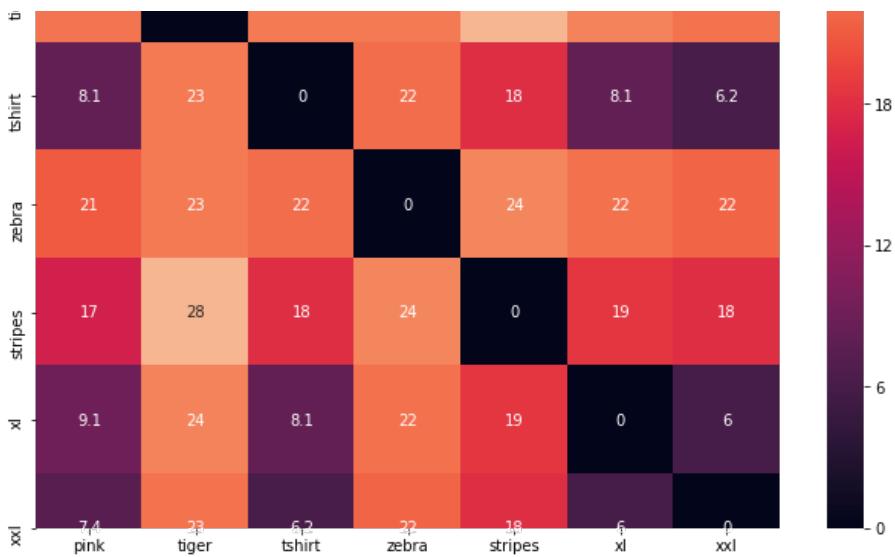


ASIN : B00JXQCWTO

Brand : Si Row

euclidean distance from input : 0.5870644705636161

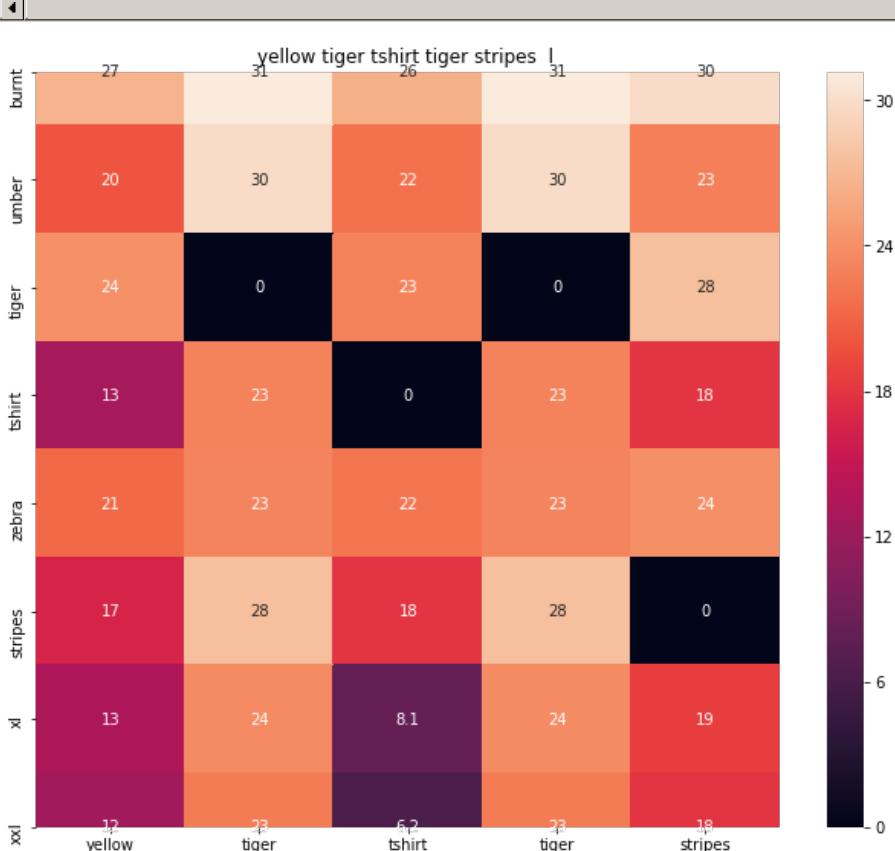




ASIN : B00JXQASS6

Brand : Si Row

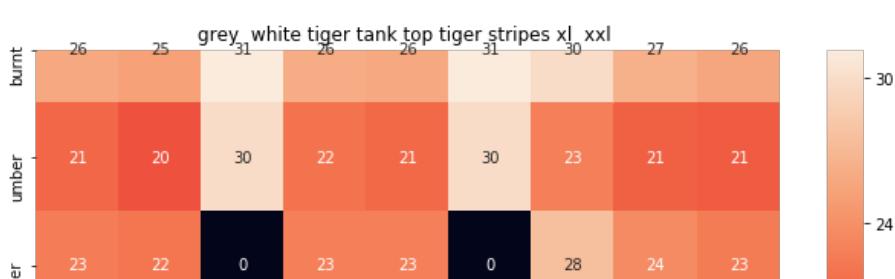
euclidean distance from input : 0.6883586114649729

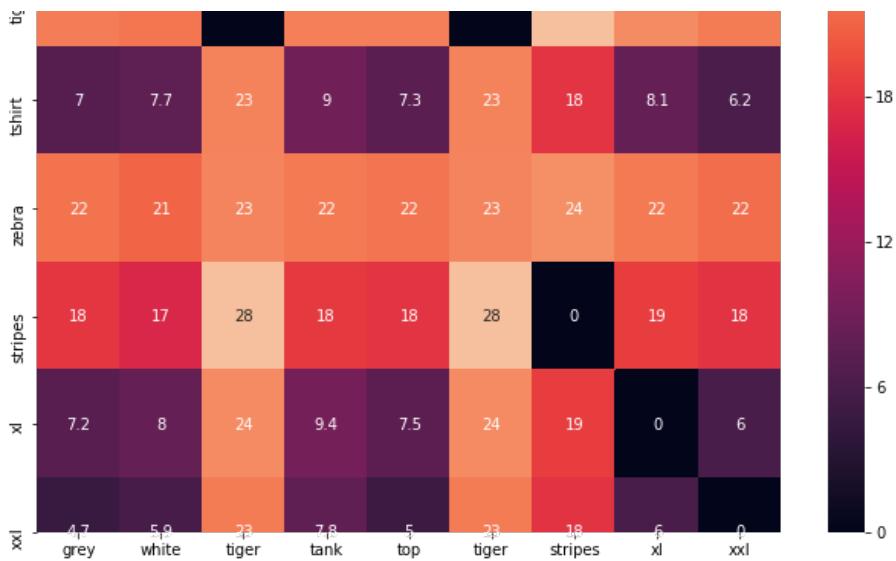


ASIN : B00JXQCUIC

Brand : Si Row

euclidean distance from input : 0.765299099879577

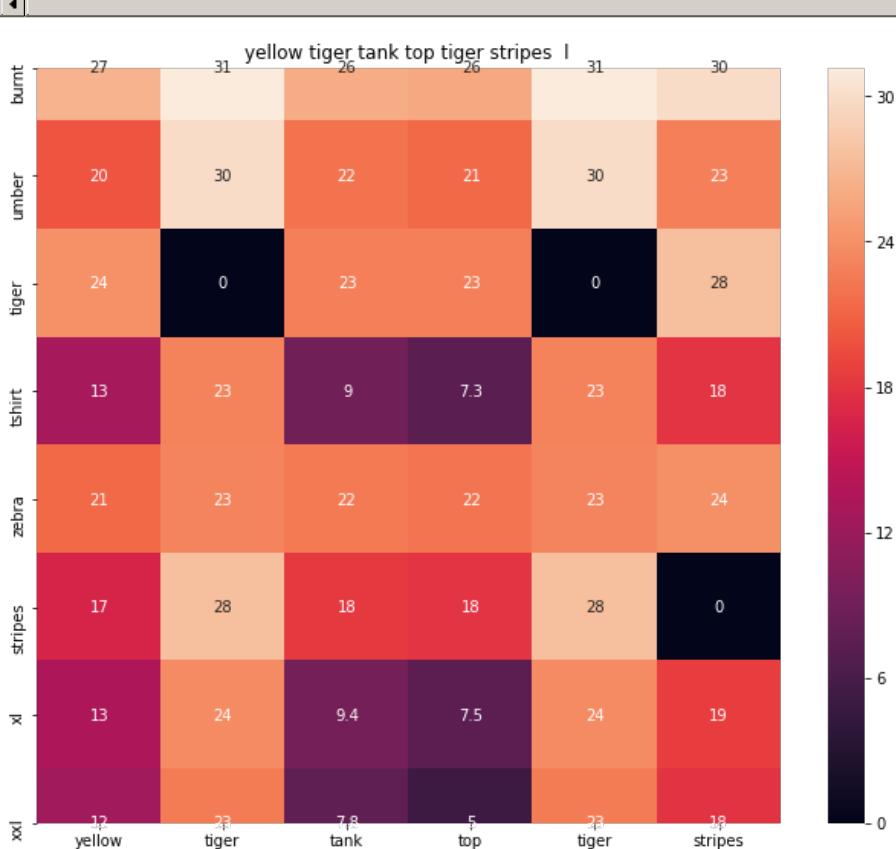




ASIN : B00JXQAFZ2

Brand : Si Row

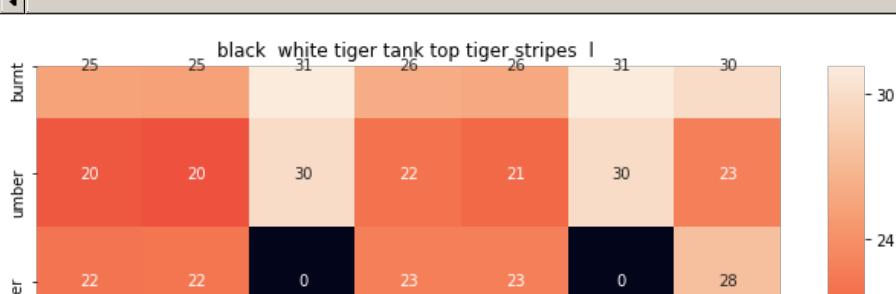
euclidean distance from input : 0.7665571165139652

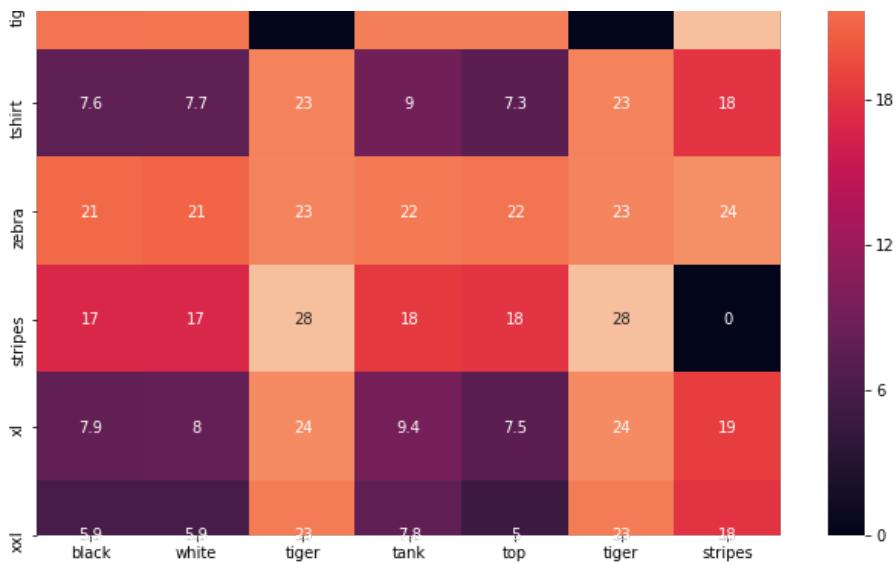


ASIN : B00JXQAUWA

Brand : Si Row

euclidean distance from input : 0.7710980327644986

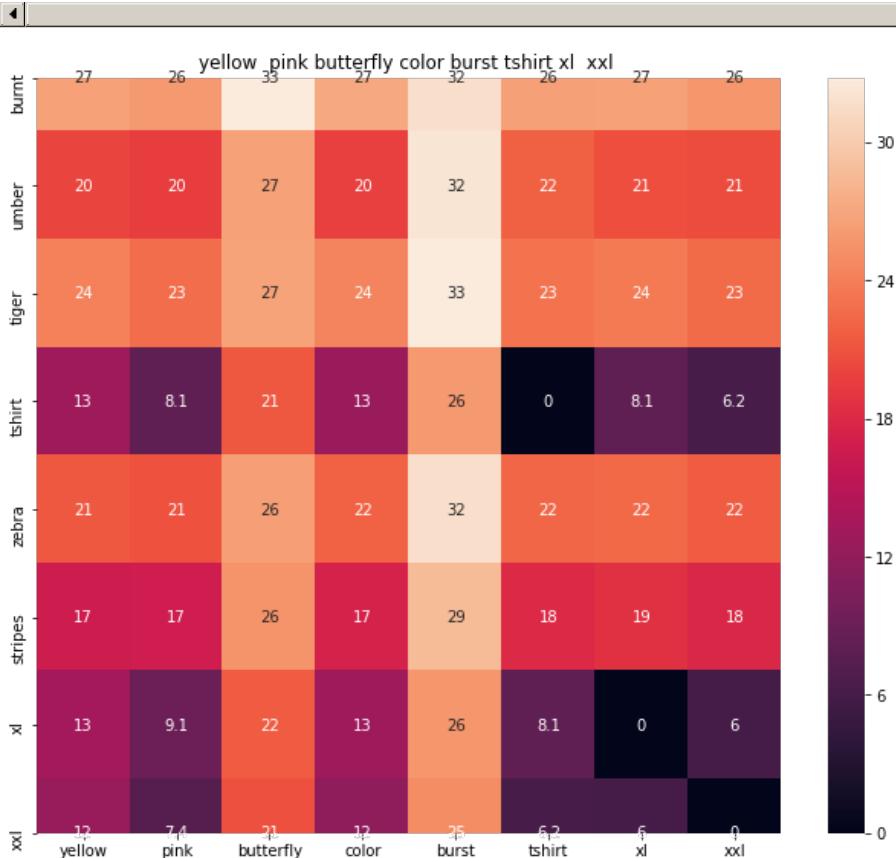




ASIN : B00JXQAO94

Brand : Si Row

euclidean distance from input : 0.7858956281332125

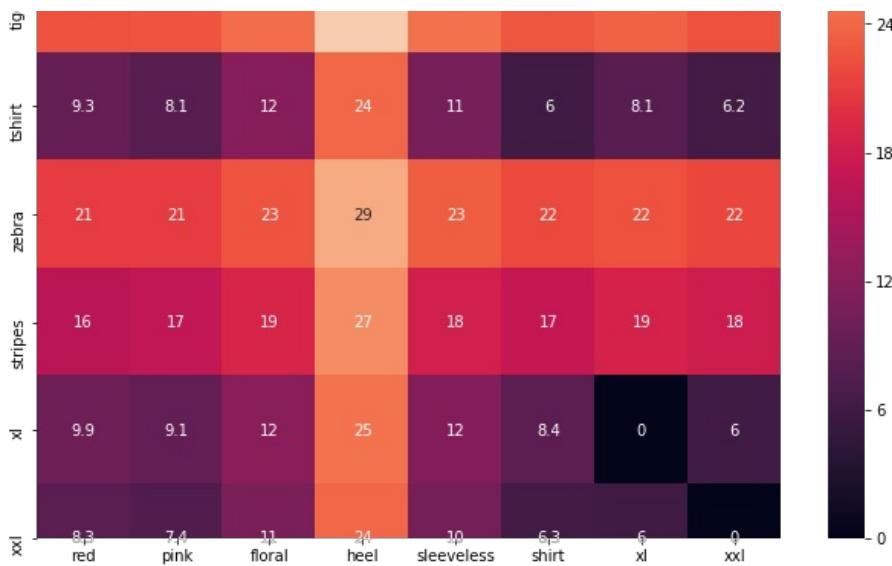


ASIN : B00JXQBBMI

Brand : Si Row

euclidean distance from input : 0.8360235390998453

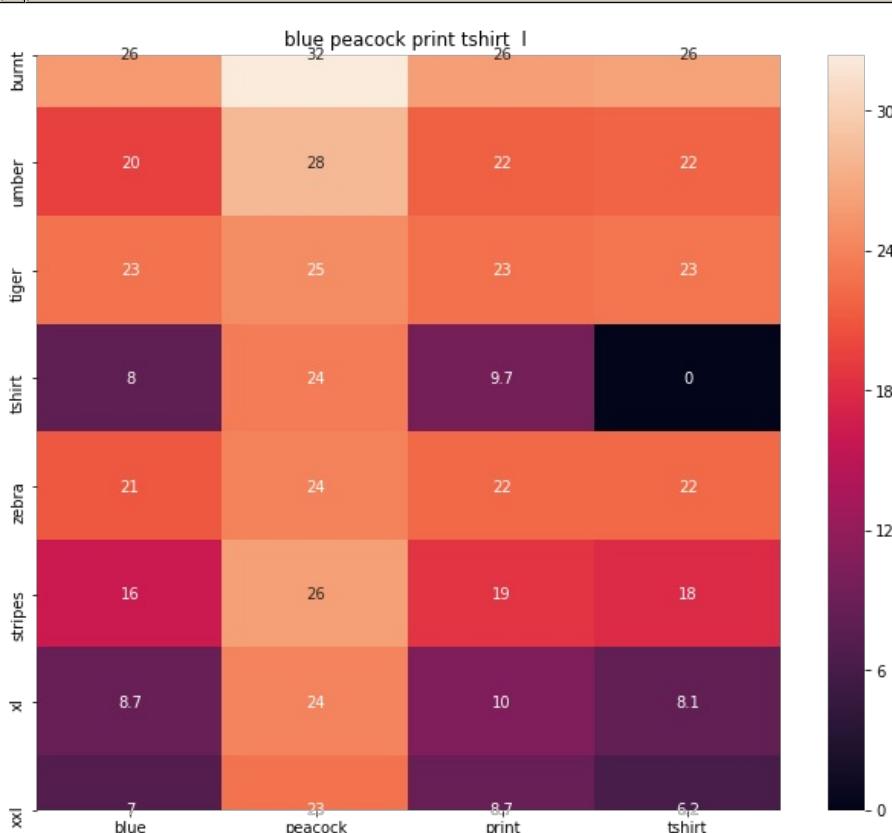




ASIN : B00JV63QQE

Brand : Si Row

euclidean distance from input : 0.8372073454470912



ASIN : B00JXQC8L6

Brand : Si Row

euclidean distance from input : 0.8375937838969828

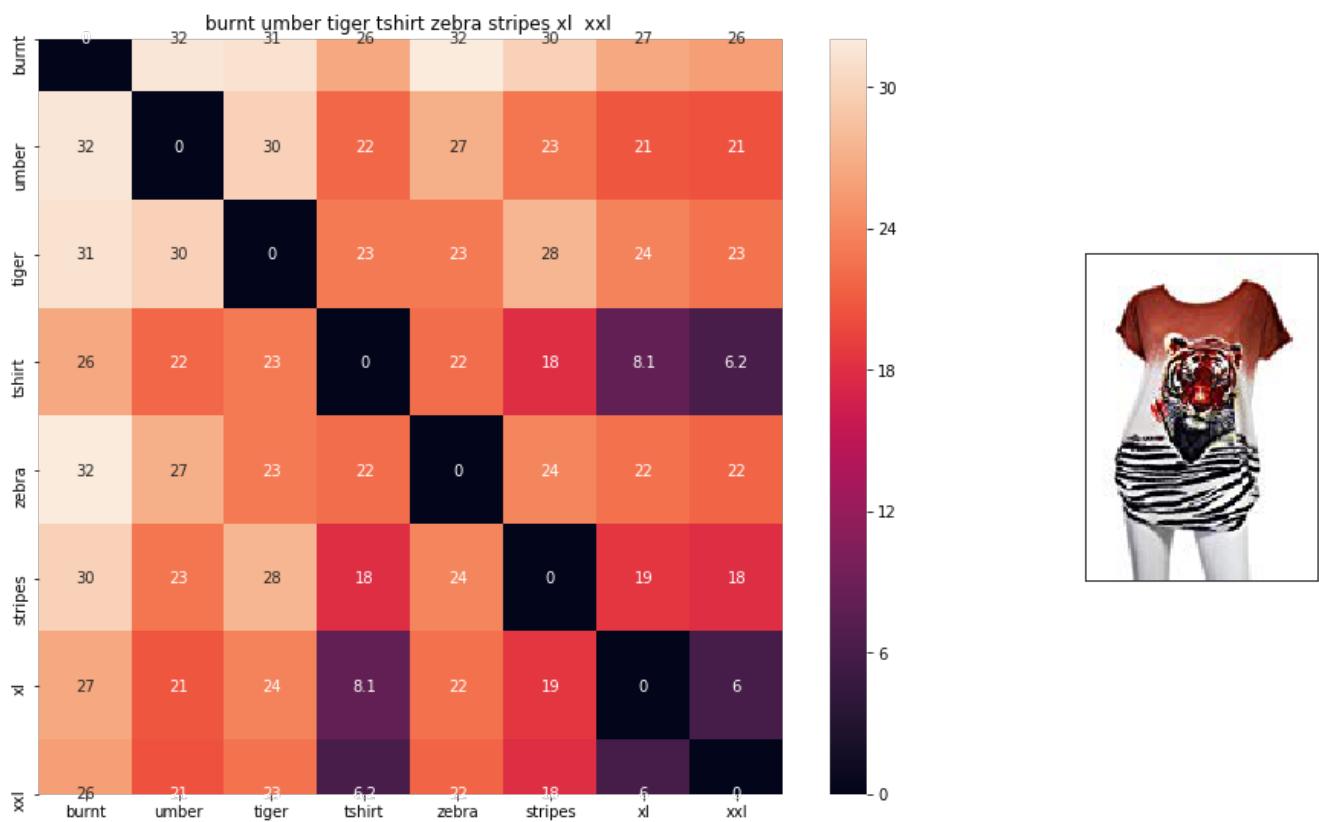
In [59]:

```
function(12566,20,3,100,15,1)
```

```
(16042, 1) (16042, 1) (16042, 1) (16042, 1)
(16042, 1)
[12566 14072 14089 14790 10569 7330 7482 7105 13041 12934 9337 14002
 12540 12065 12107 7726 1564 4028 12234 4067]
Tnt64Tndex/112055 160670 160810 160102 116737 72140 71211 68773
```

```
[147862, 146397, 99739, 159912, 141660, 135891, 136502, 77322,
18437, 41446, 138085, 41824],
dtype='int64')
```

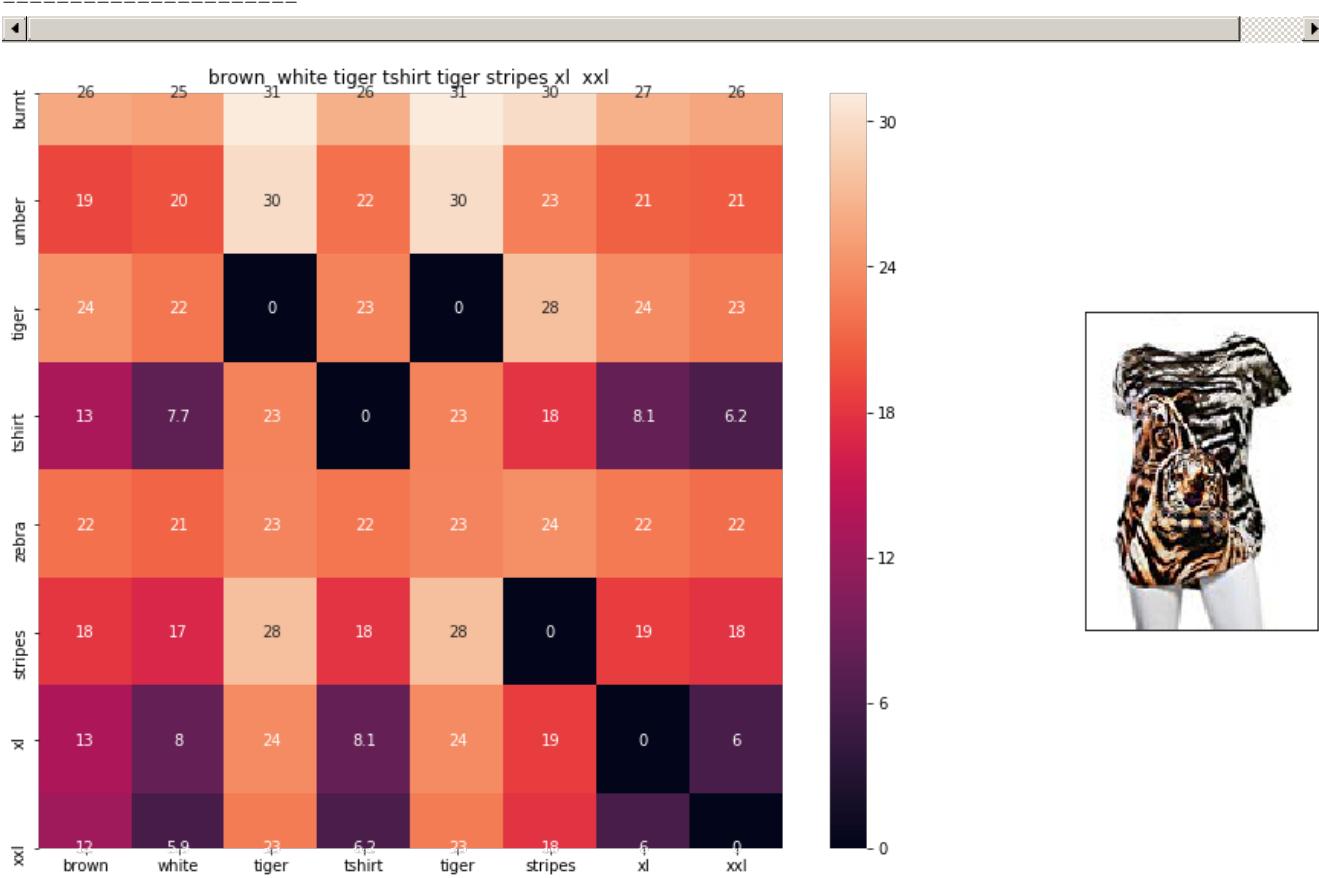
```
[142055, 160670, 160819, 169492, 116737, 72149, 74241, 68773, 147862, 146397, 99739, 159912,
141660, 135891, 136502, 77322, 18437, 41446, 138085, 41824]
```



ASIN : B00JXQB5FQ

Brand : Si Row

euclidean distance from input : 4.2501021786901056e-09



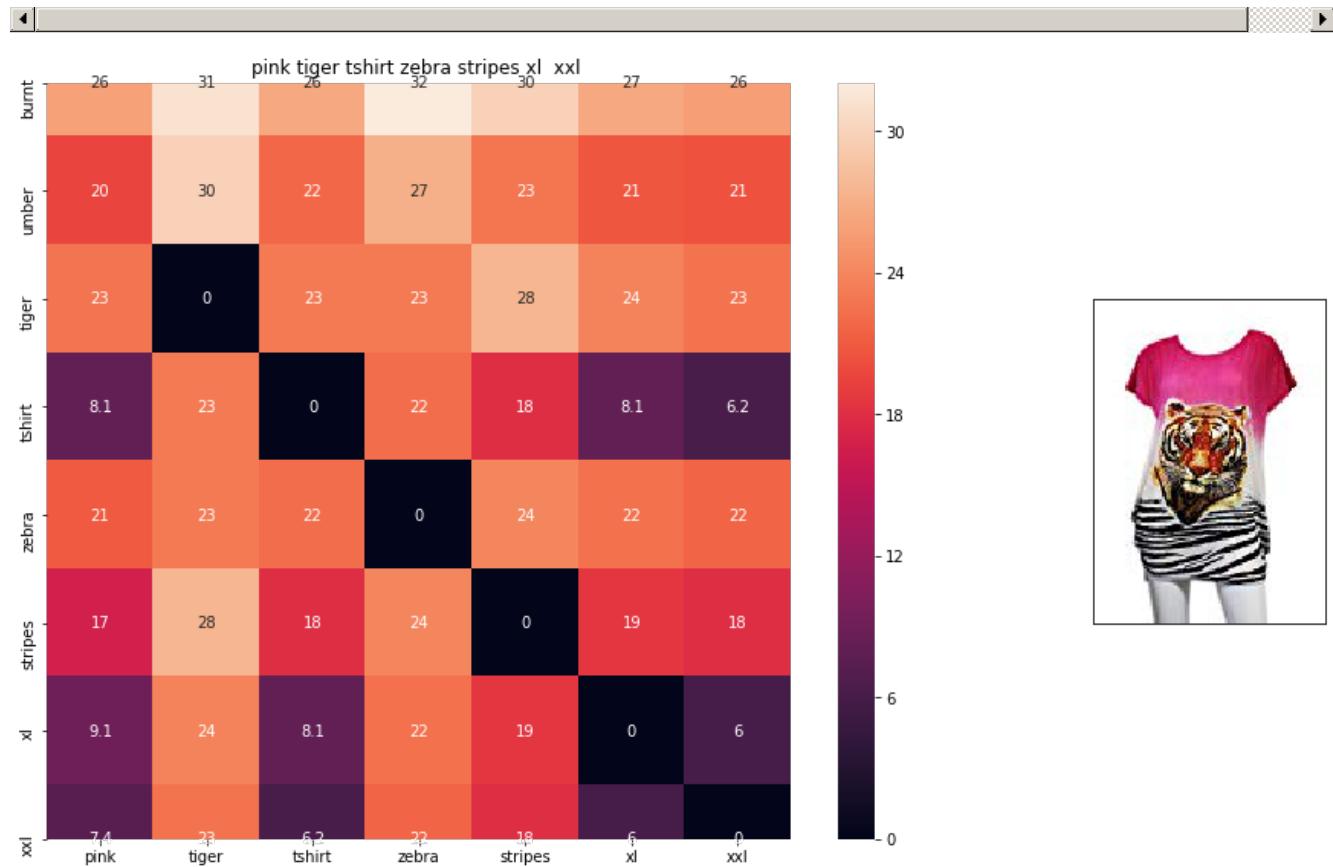
ASIN : B00JXQCWT0

Brand : Si Row

euclidean distance from input : 0.5870644705636161

=====

=====



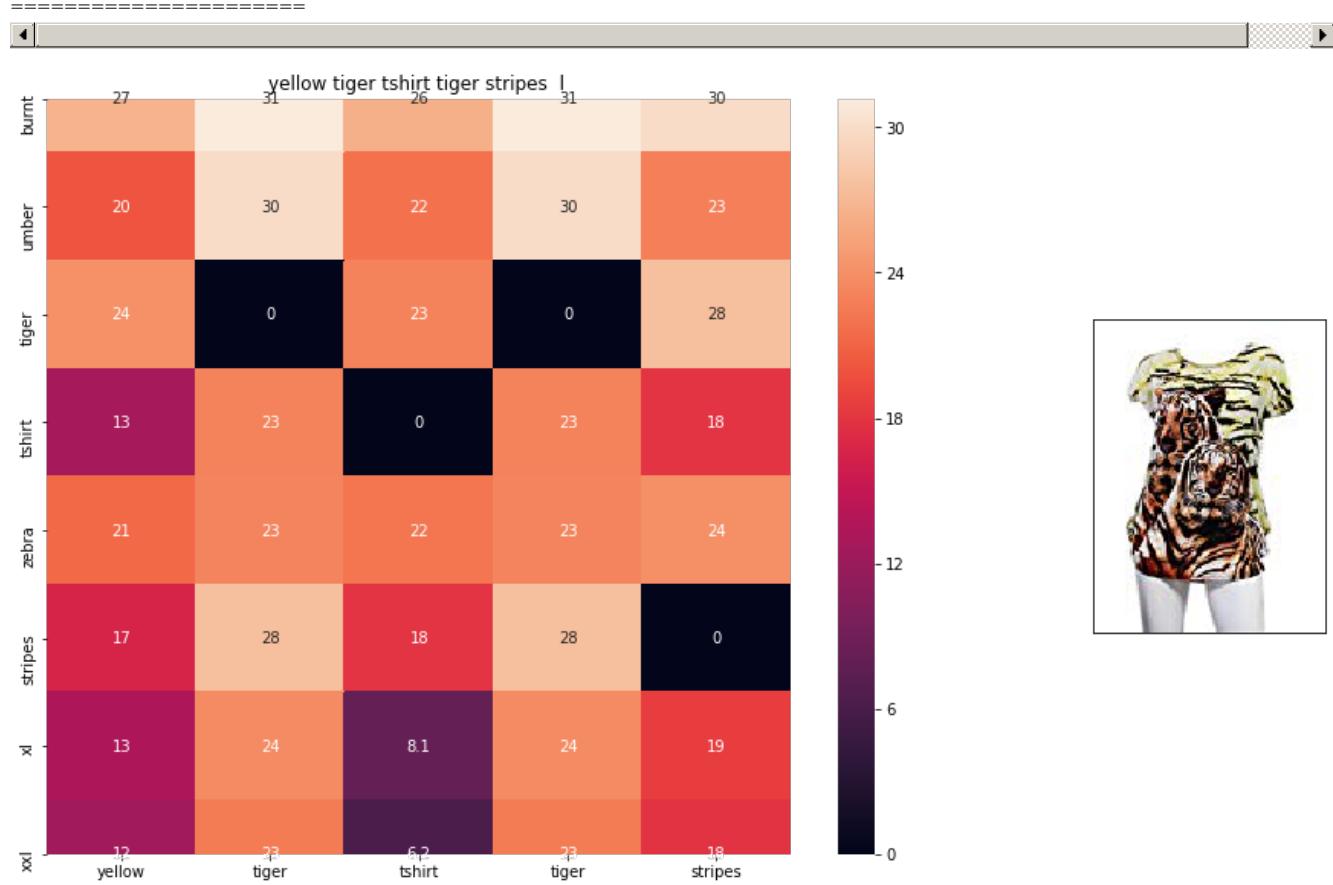
ASIN : B00JXQASS6

Brand : Si Row

euclidean distance from input : 0.6883586114649729

=====

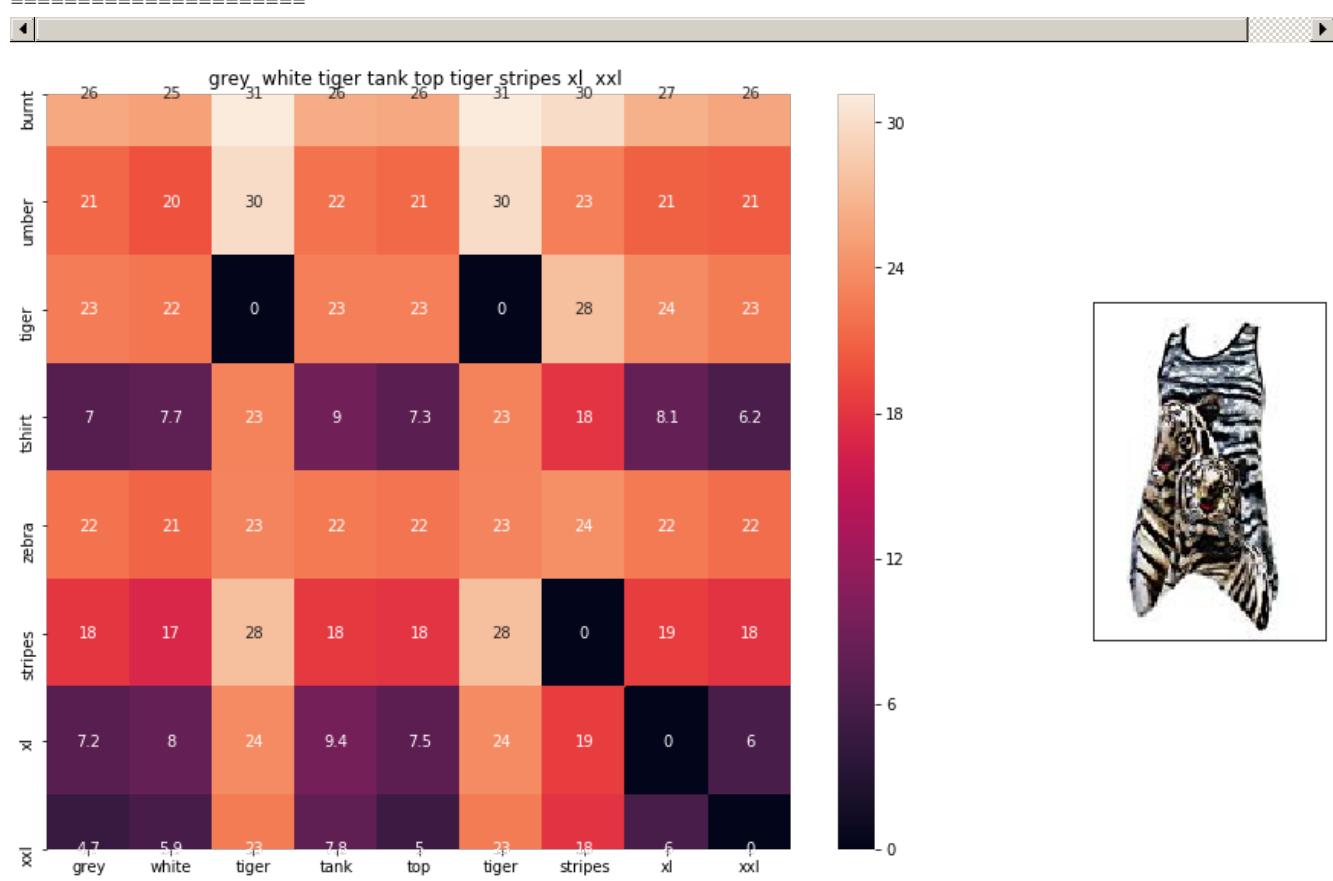
=====



ASIN : B00JXQCUIC

Brand : Si Row

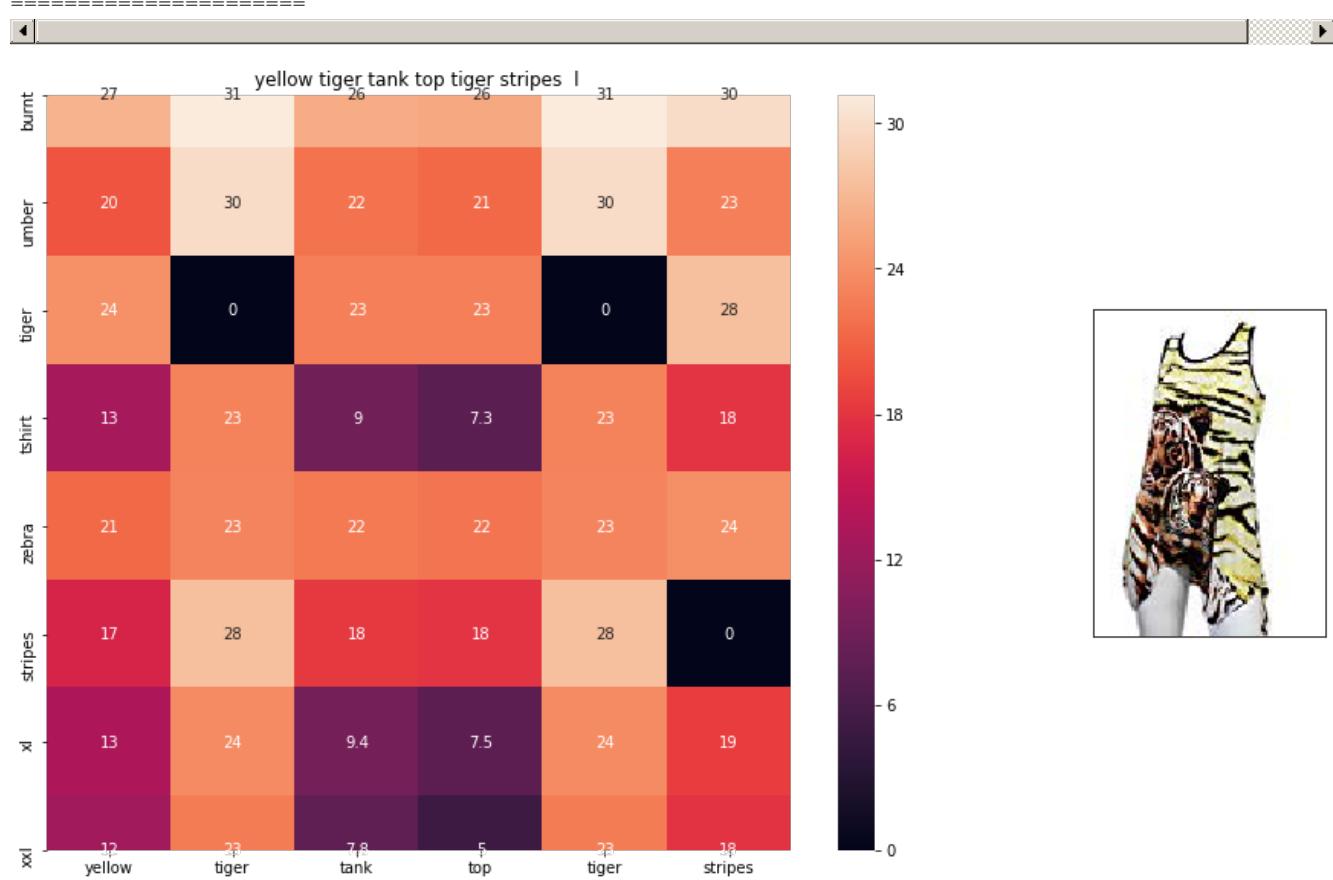
euclidean distance from input : 0.765299099879577



ASIN : B00JXQAFZ2

Brand : Si Row

euclidean distance from input : 0.7665571165139652

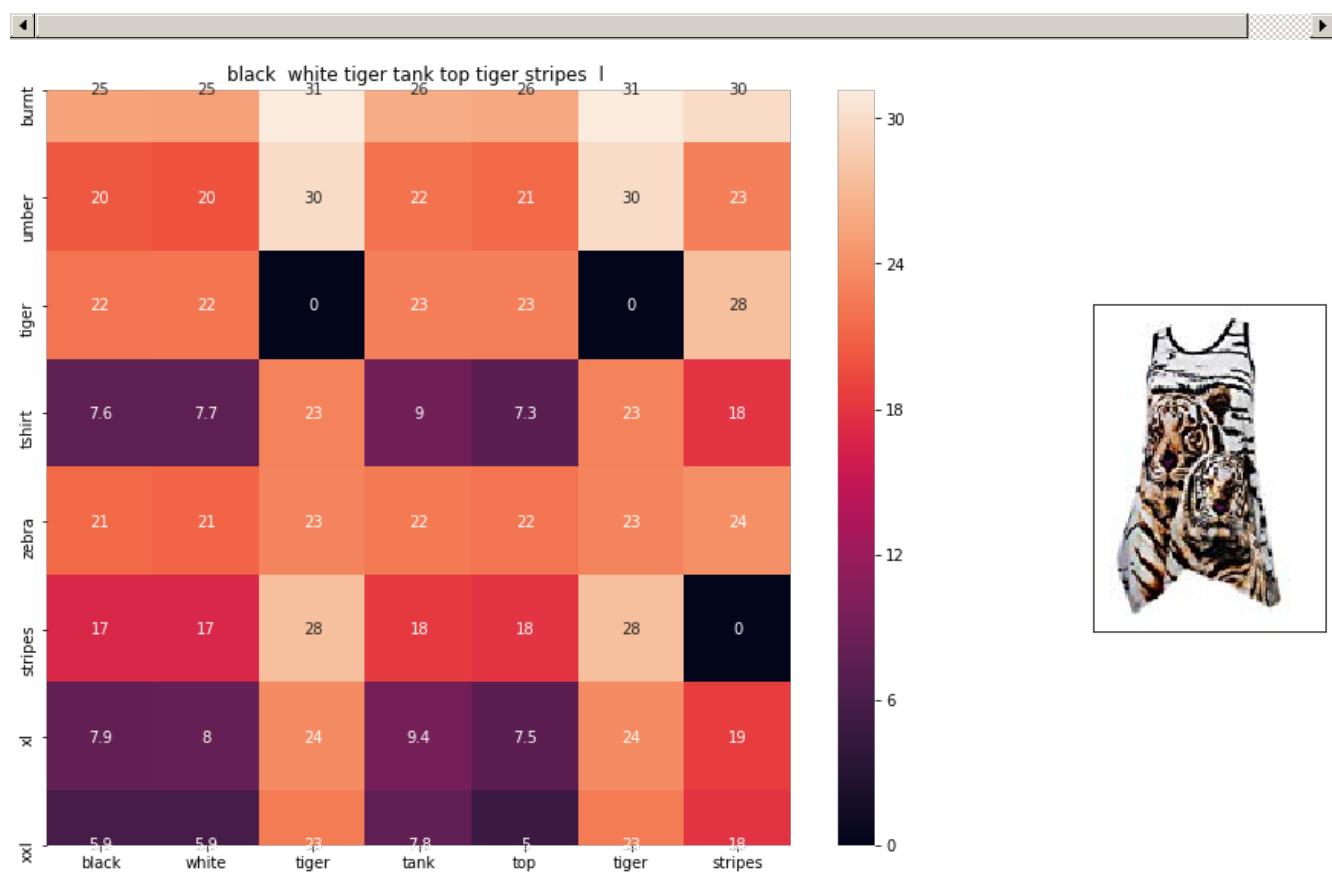


ASIN : B00JXQAUWA

Brand : Si Row

euclidean distance from input : 0.7710980327644986

=====

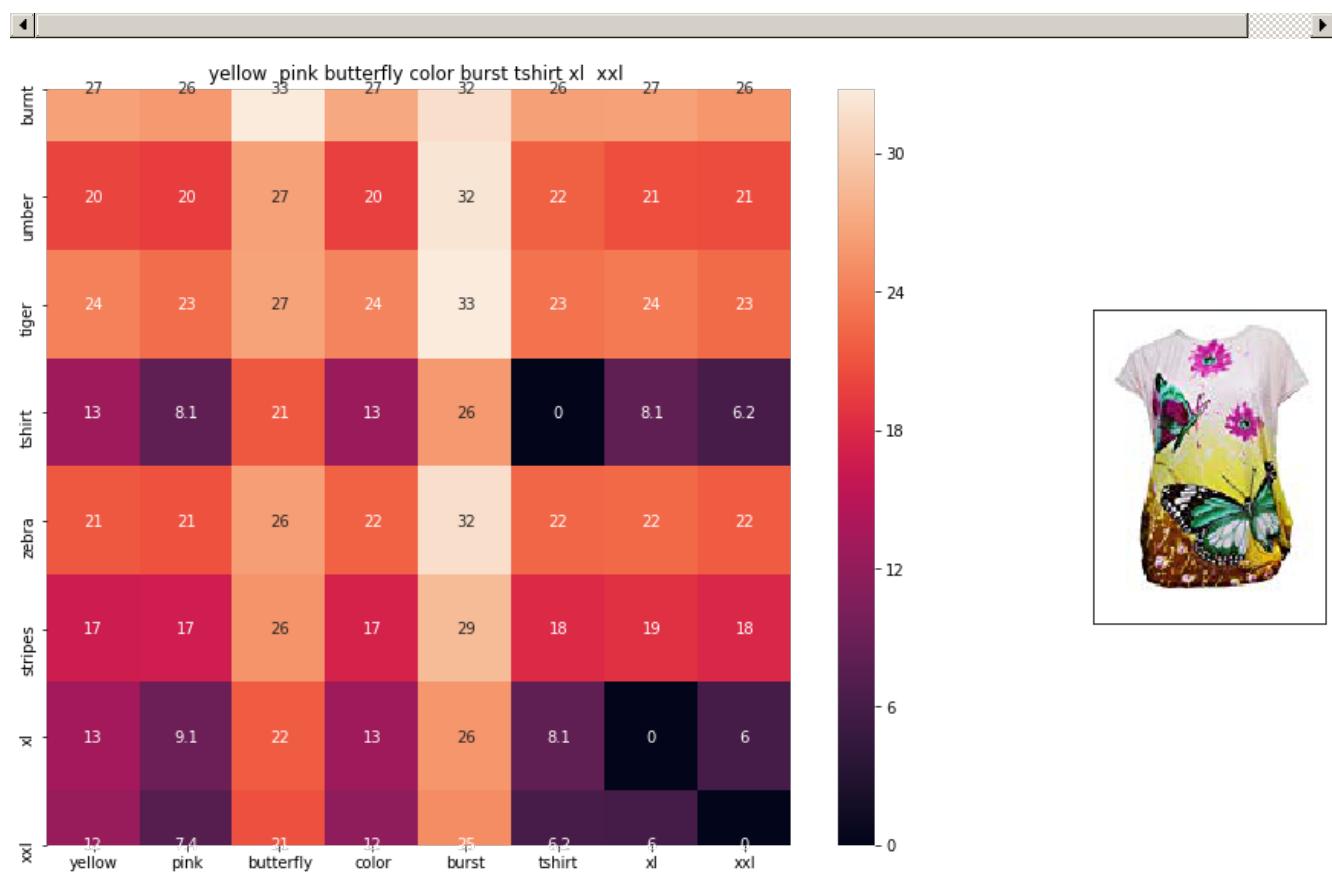


ASIN : B00JXQA094

Brand : Si Row

euclidean distance from input : 0.7858956281332125

=====

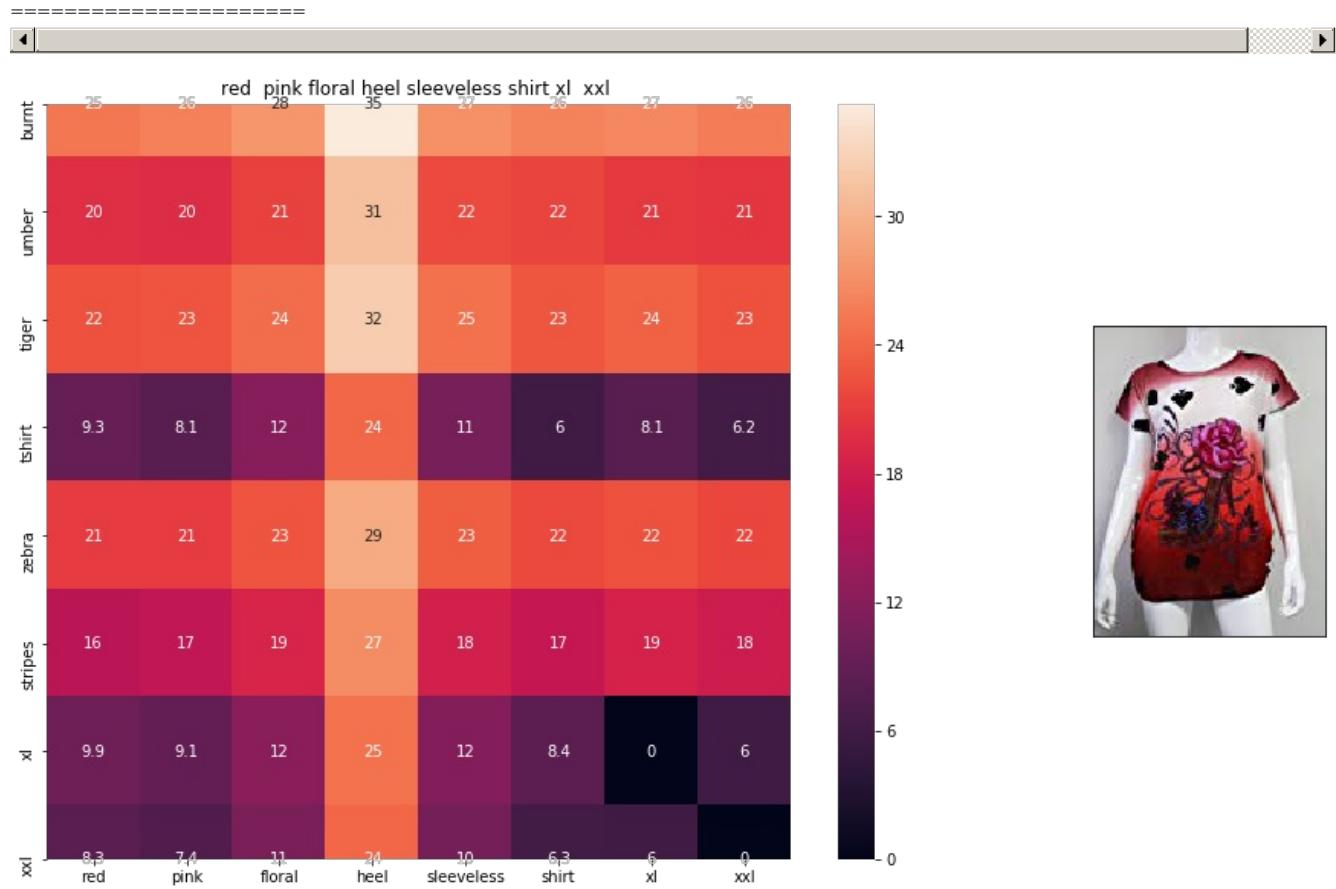


ASIN : B00JXQBBM1

Brand : Si Row

euclidean distance from input : 0.8360235390998453

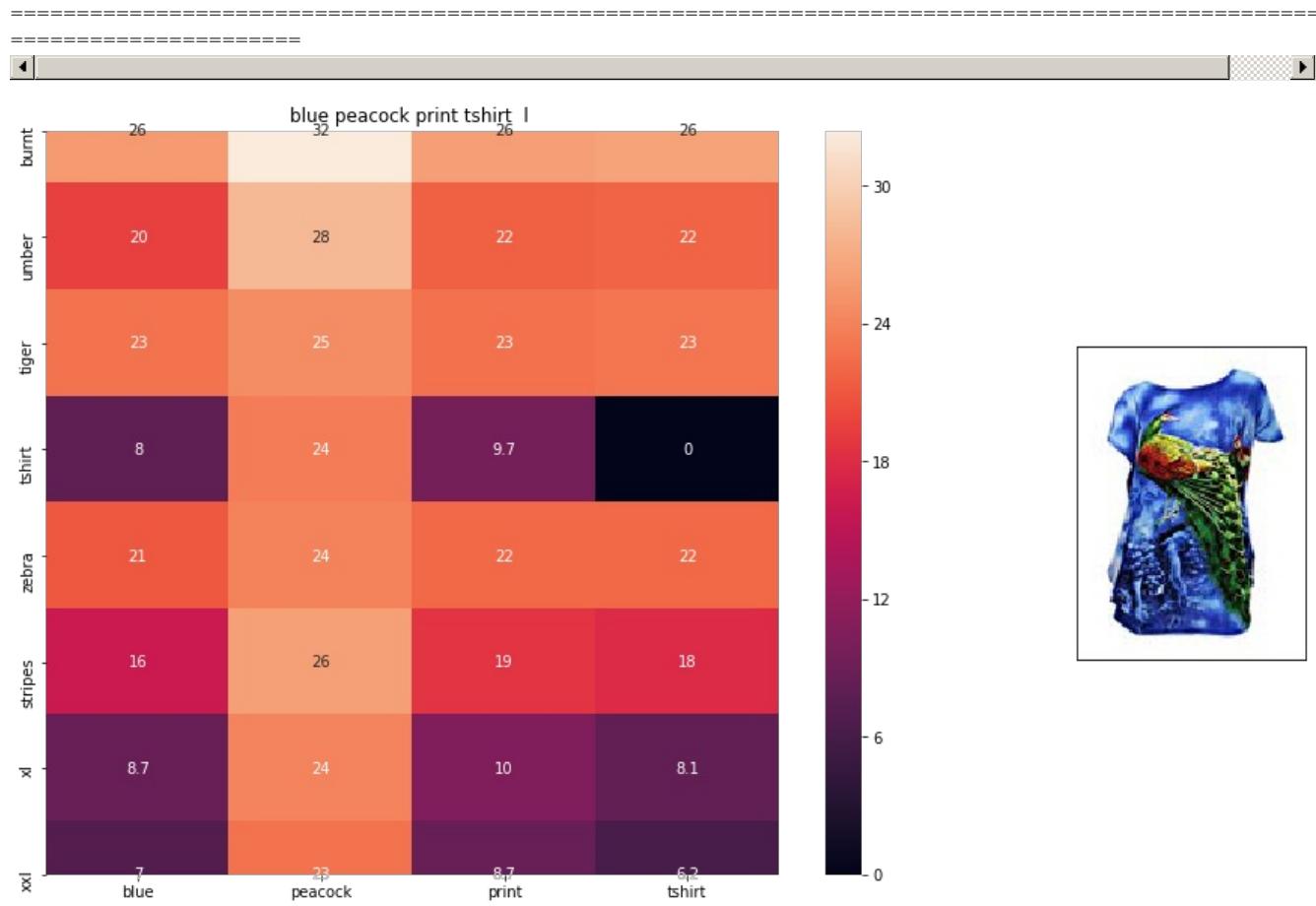
=====



ASIN : B00JV63QQE

Brand : Si Row

euclidean distance from input : 0.8372073454470912

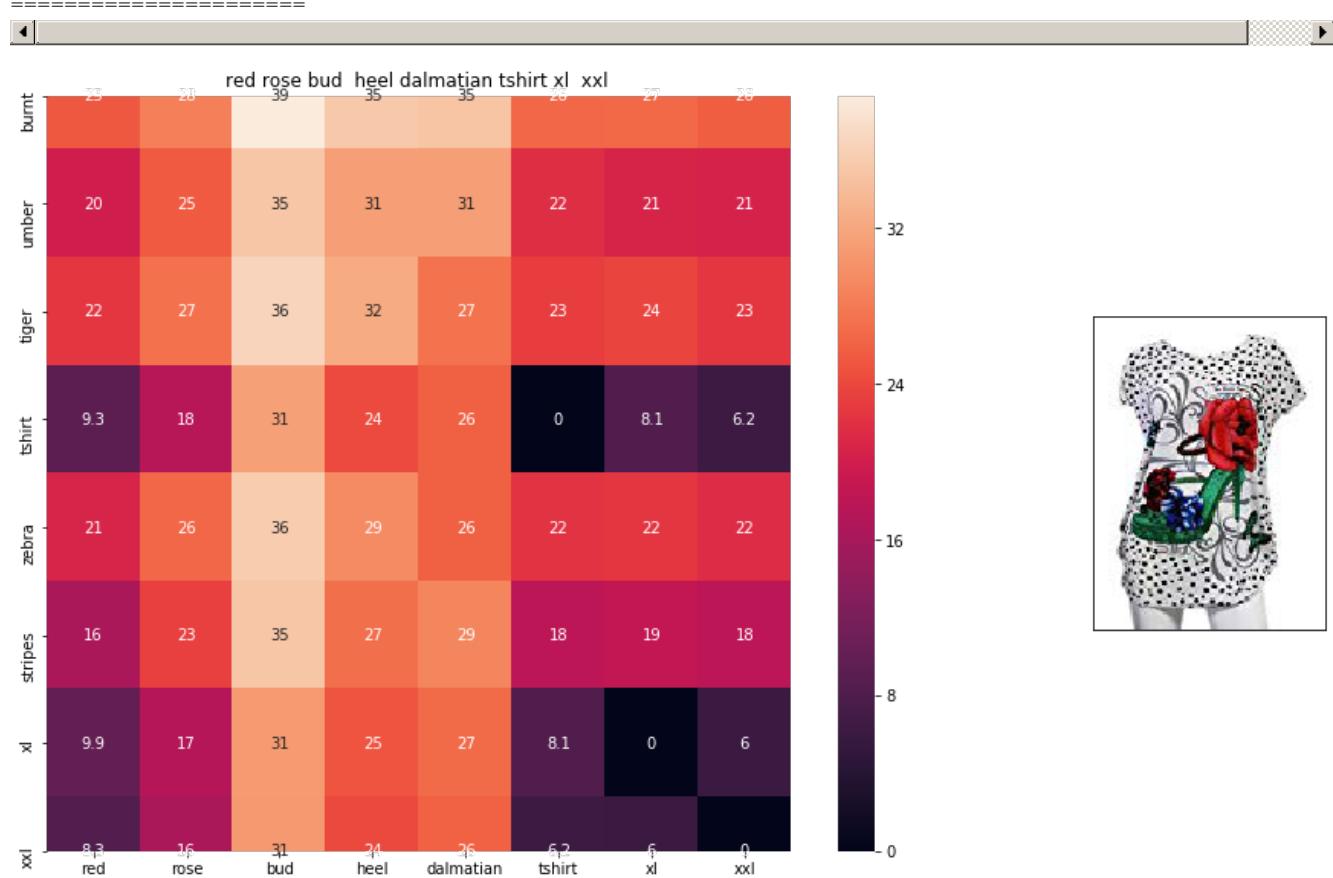


ASIN : B00JXQC8L6

Brand : Si Row

euclidean distance from input : 0.8375937838969828

=====

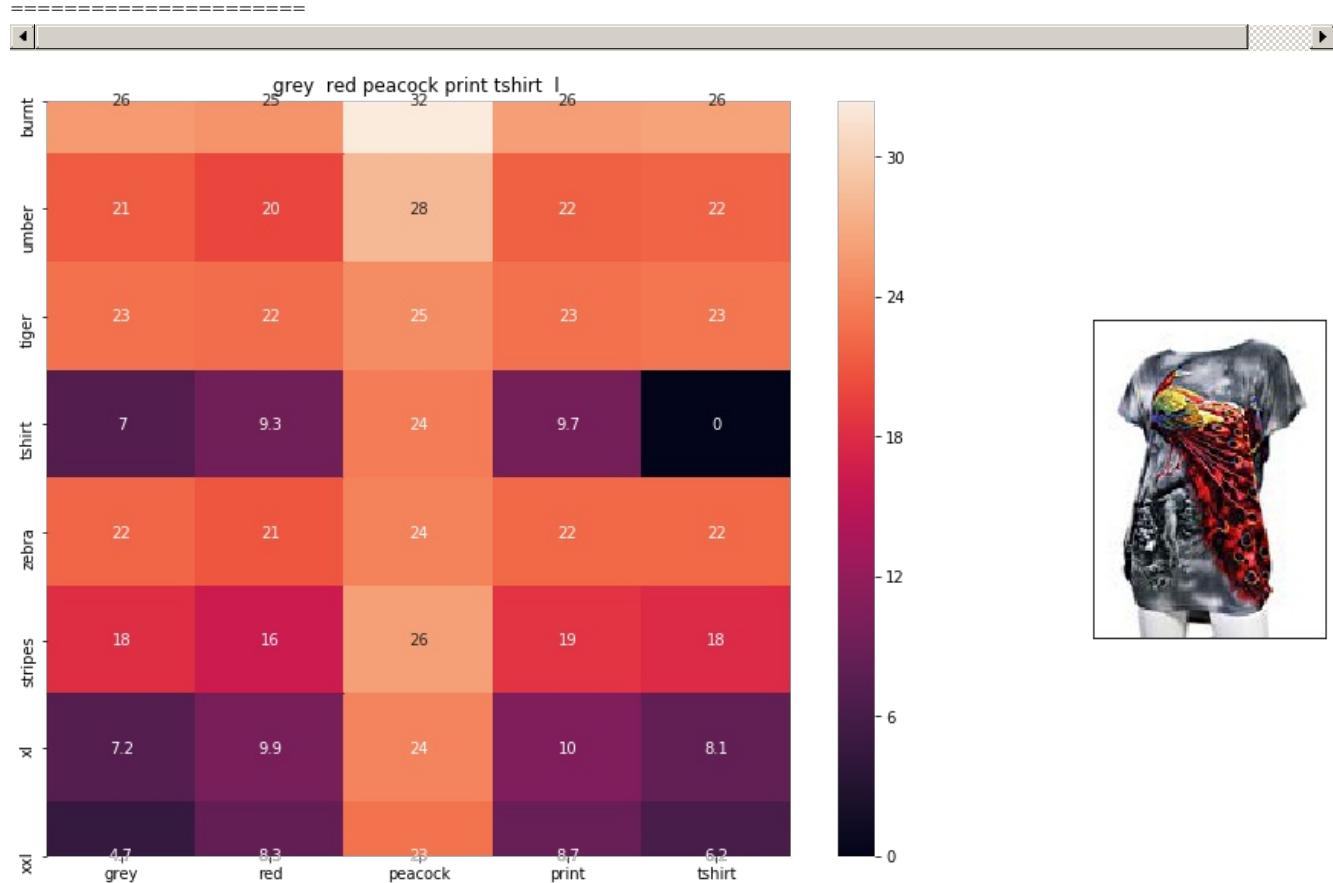


ASIN : B00JXQABBO

Brand : Si Row

euclidean distance from input : 0.8416741956637404

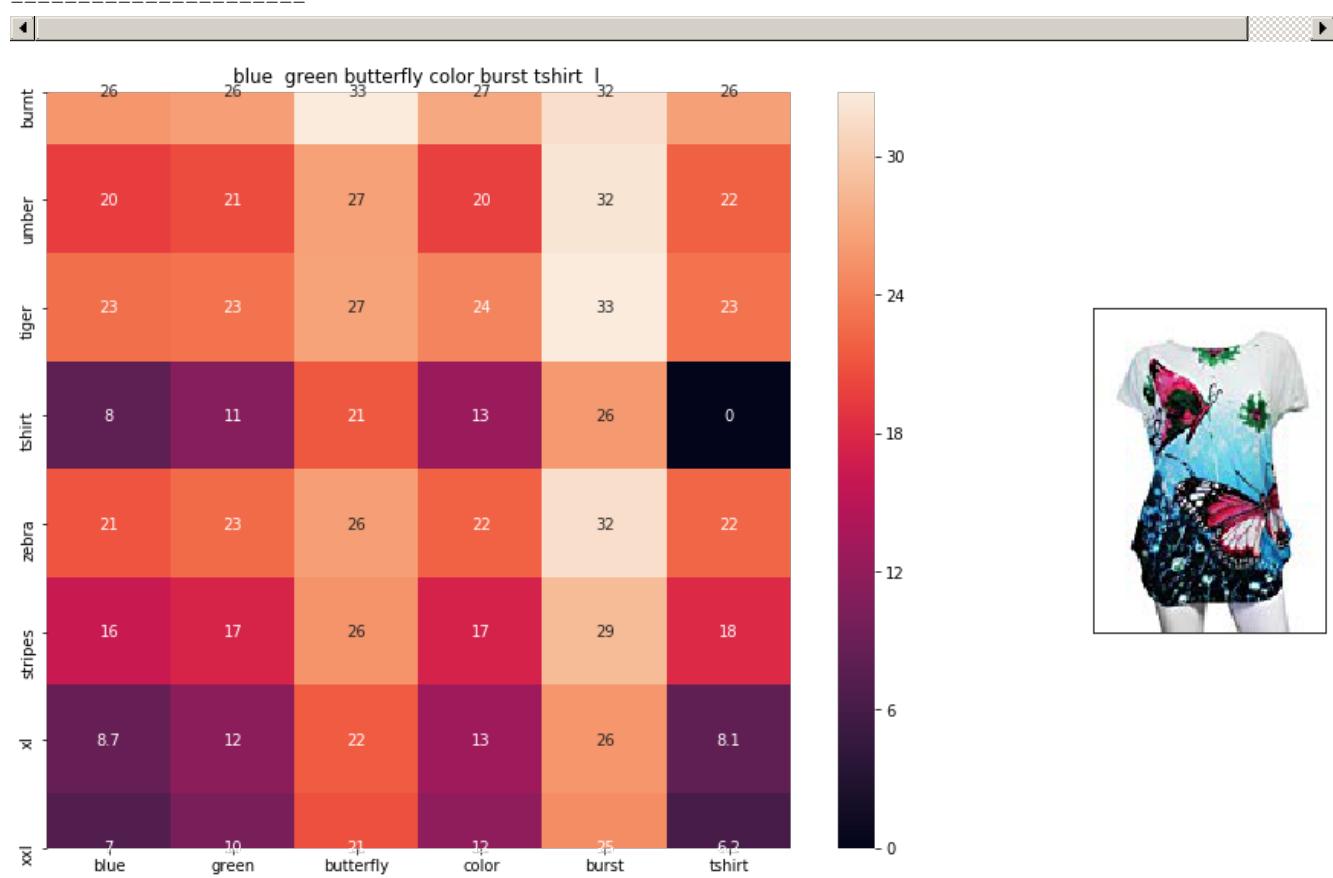
=====



ASIN : B00JXQCFRS

Brand : Si Row

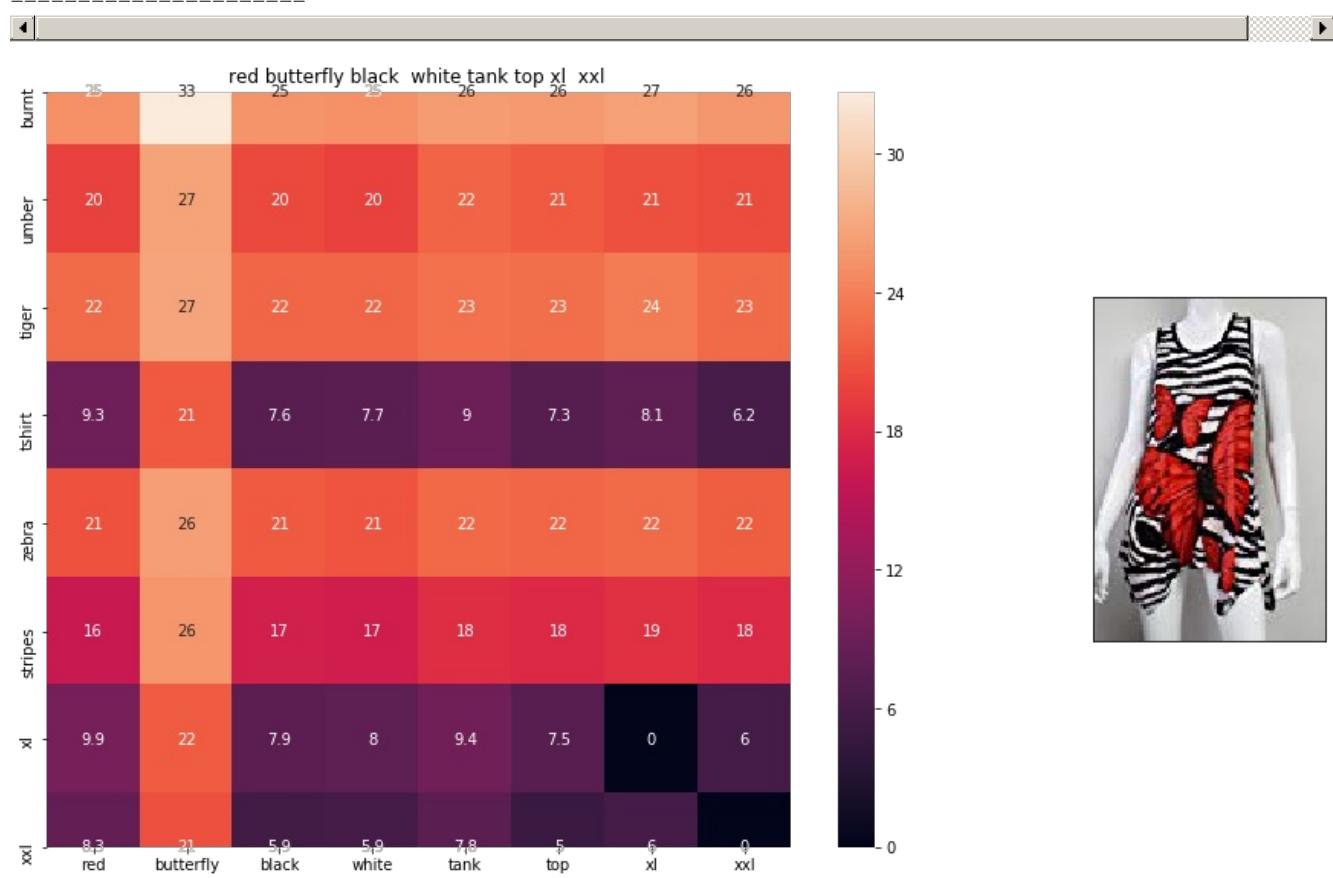
euclidean distance from input : 0.8442540850413663



ASIN : B00JXQC0C8

Brand : Si Row

euclidean distance from input : 0.8554128919784478



ASIN : B00JV63CW2

Brand : Si Row

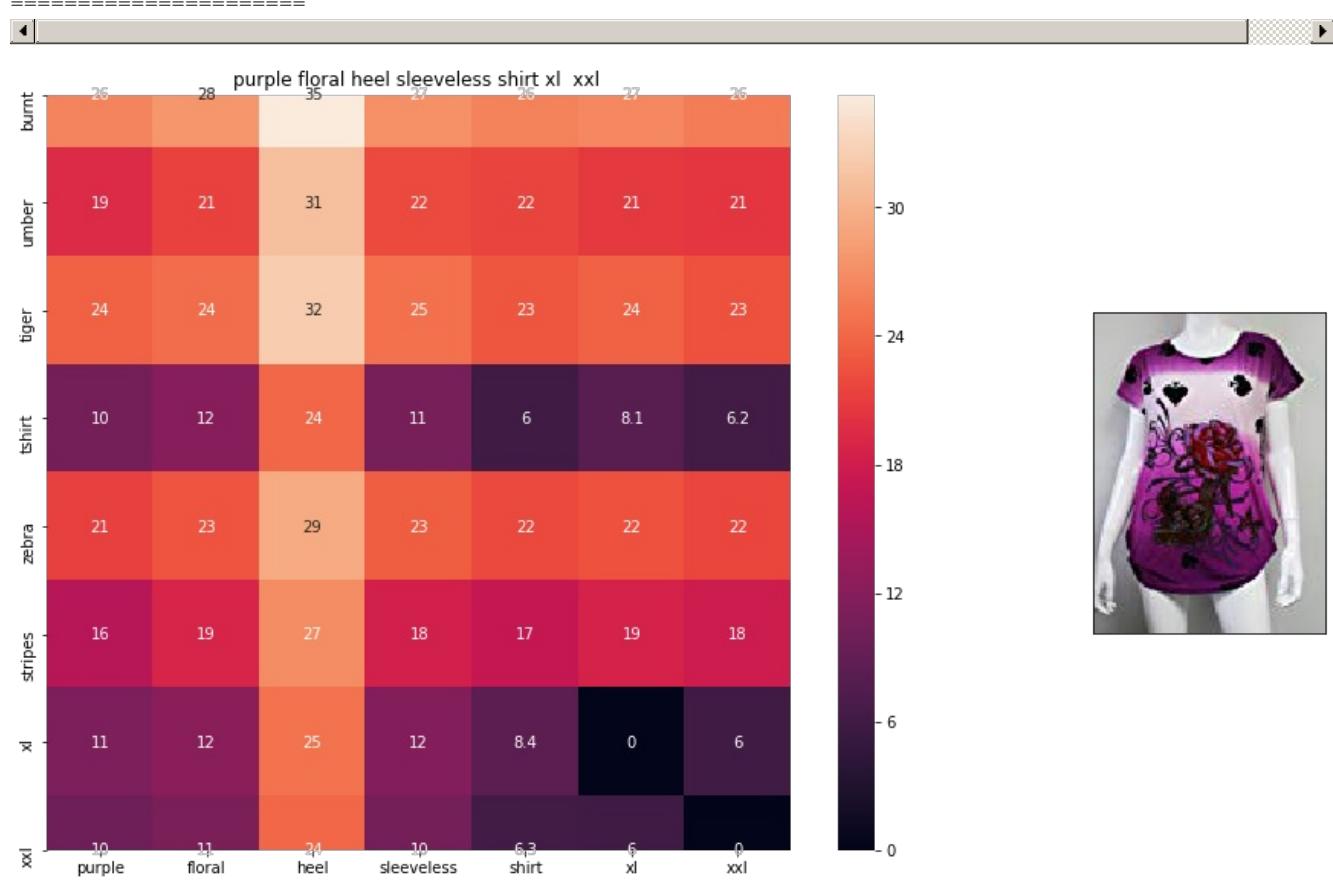
euclidean distance from input : 0.8558759609645872



ASIN : B00JXQAX2C

Brand : Si Row

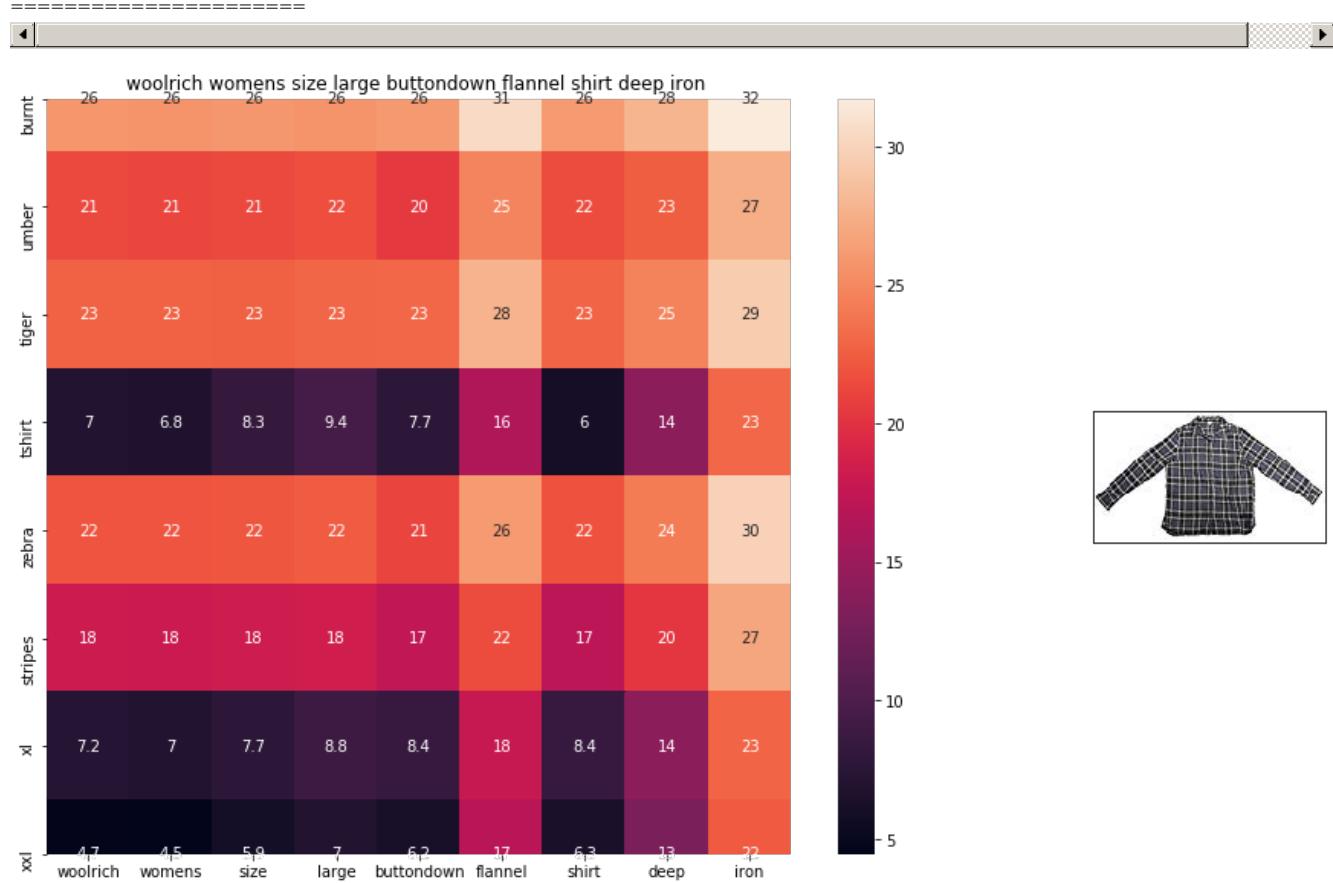
euclidean distance from input : 0.876505298820335



ASIN : B00JV63VC8

Brand : Si Row

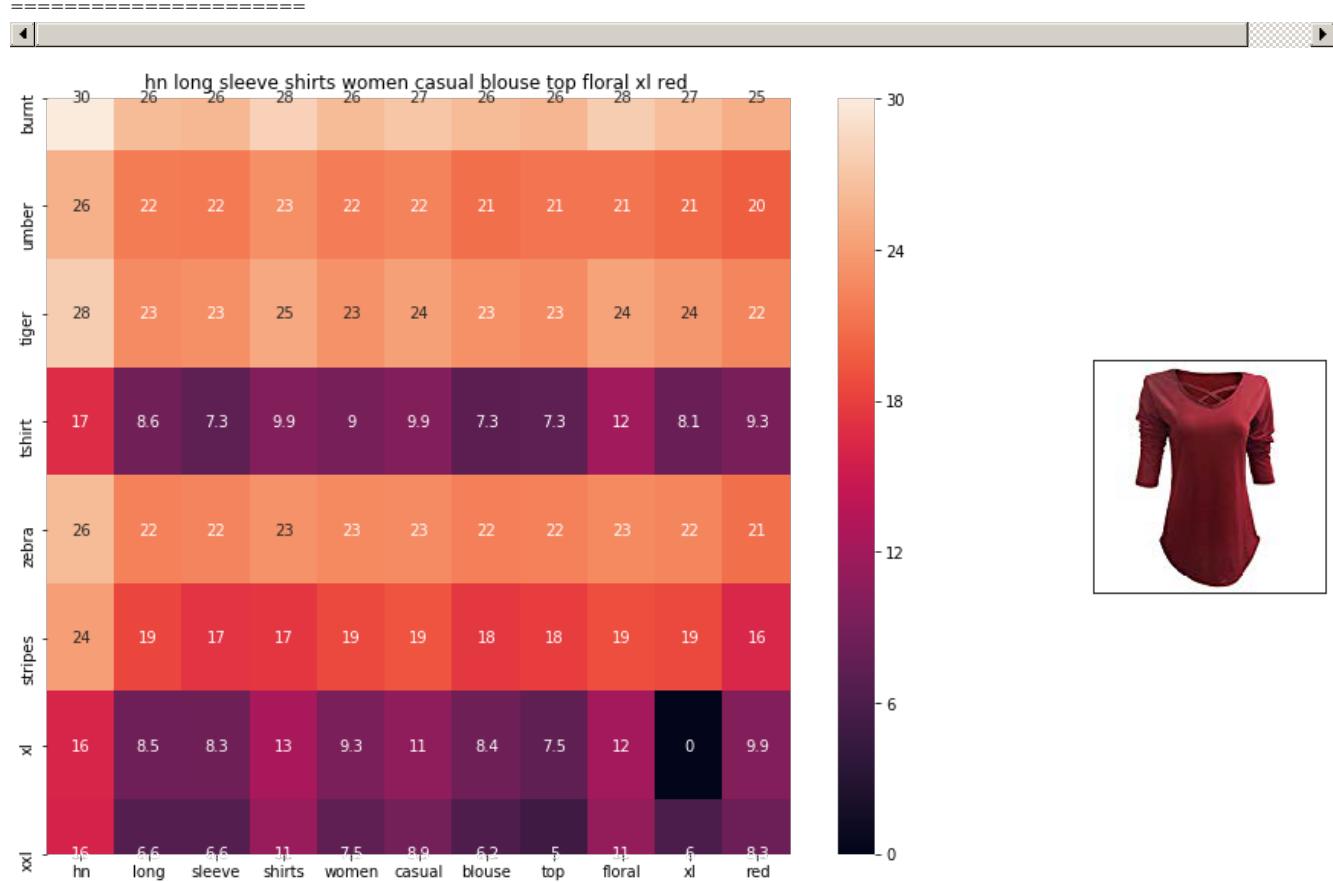
euclidean distance from input : 0.8785631196293227



ASIN : B06XJYVM9Q

Brand : W

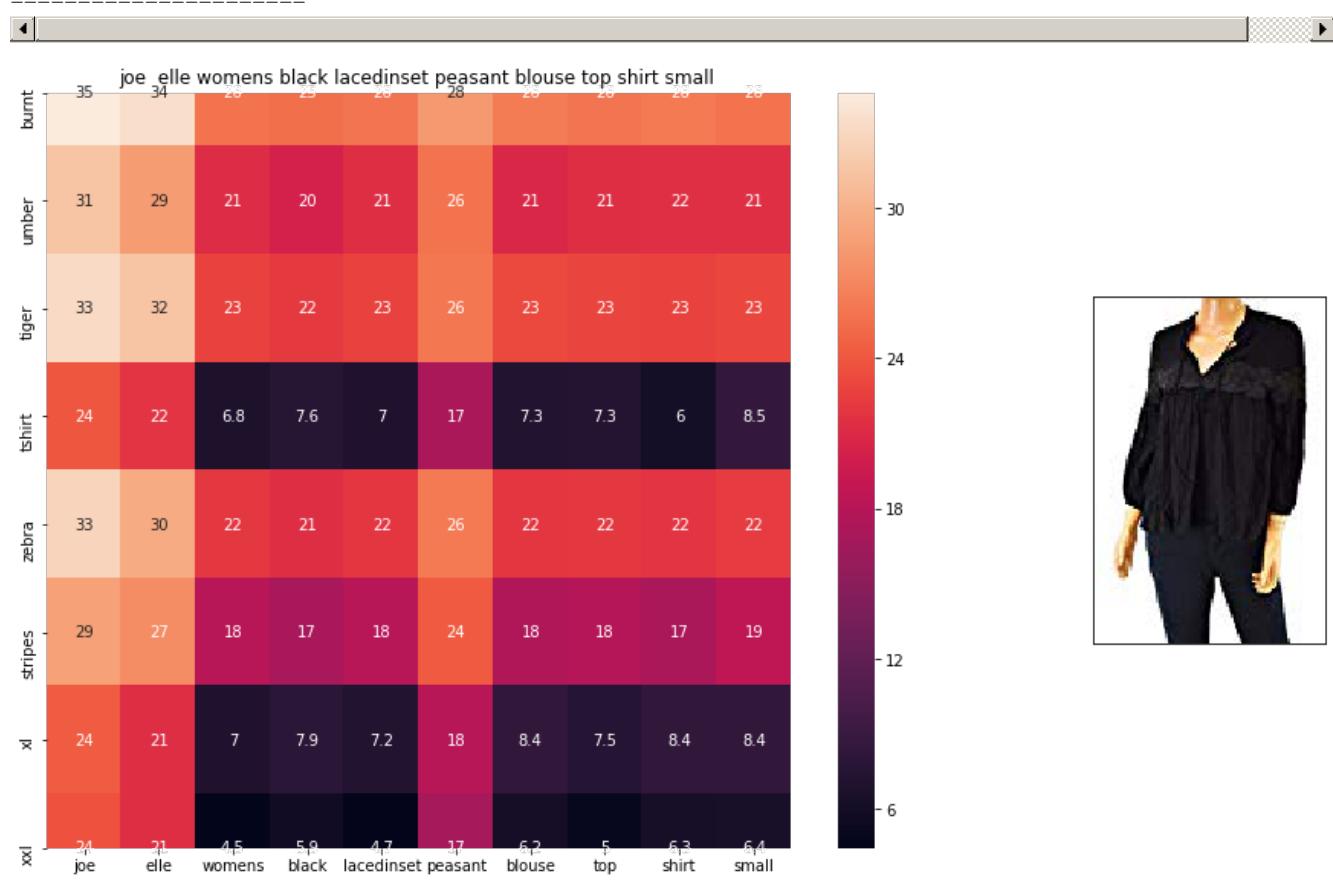
euclidean distance from input : 1.989050181187327



ASIN : B0711KH9DX

Brand : H&N

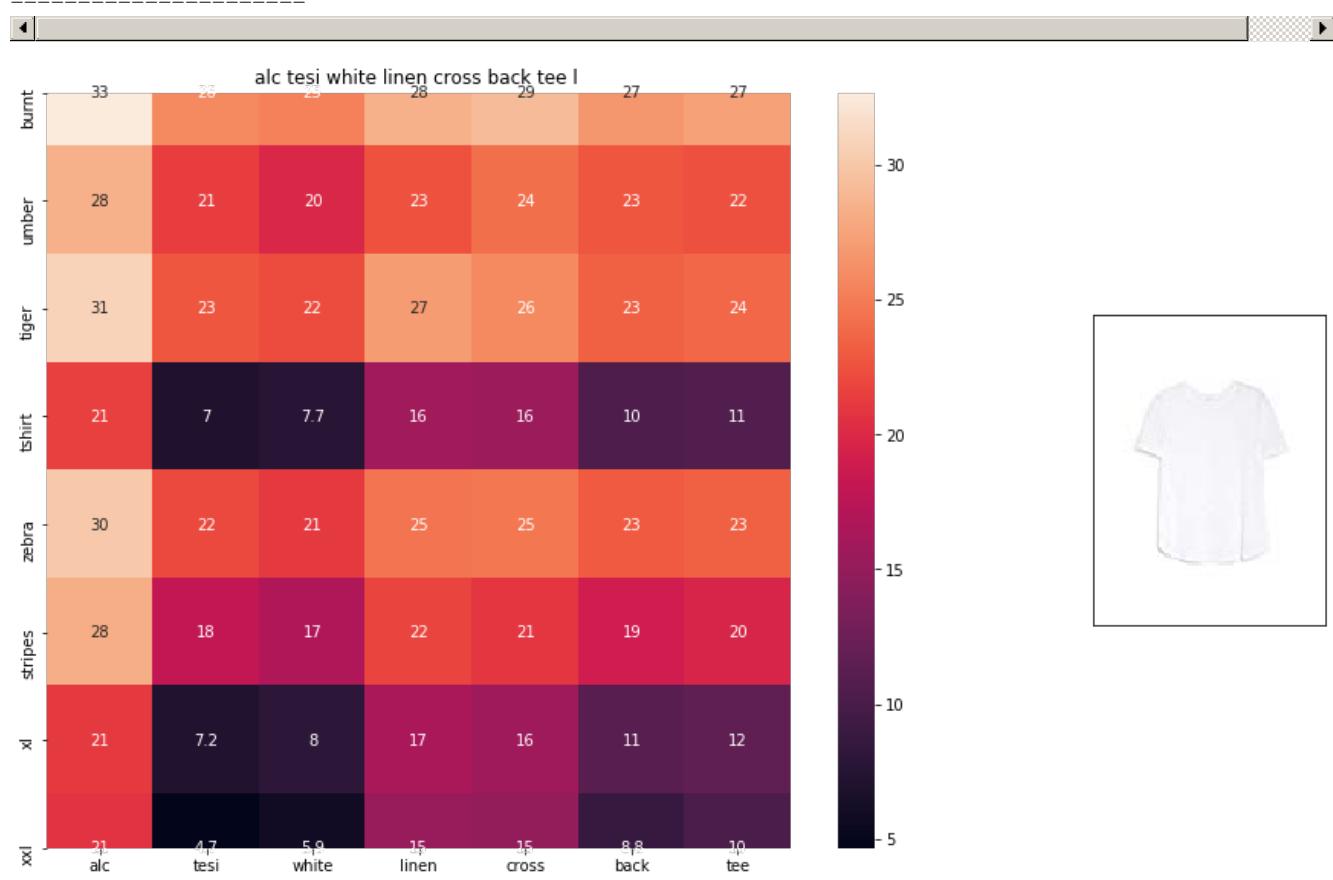
euclidean distance from input : 1.9917851299284046



ASIN : B073GDVFT2

Brand : J&E

euclidean distance from input : 2.0014043699278714



ASIN : B06ZY8FFRW

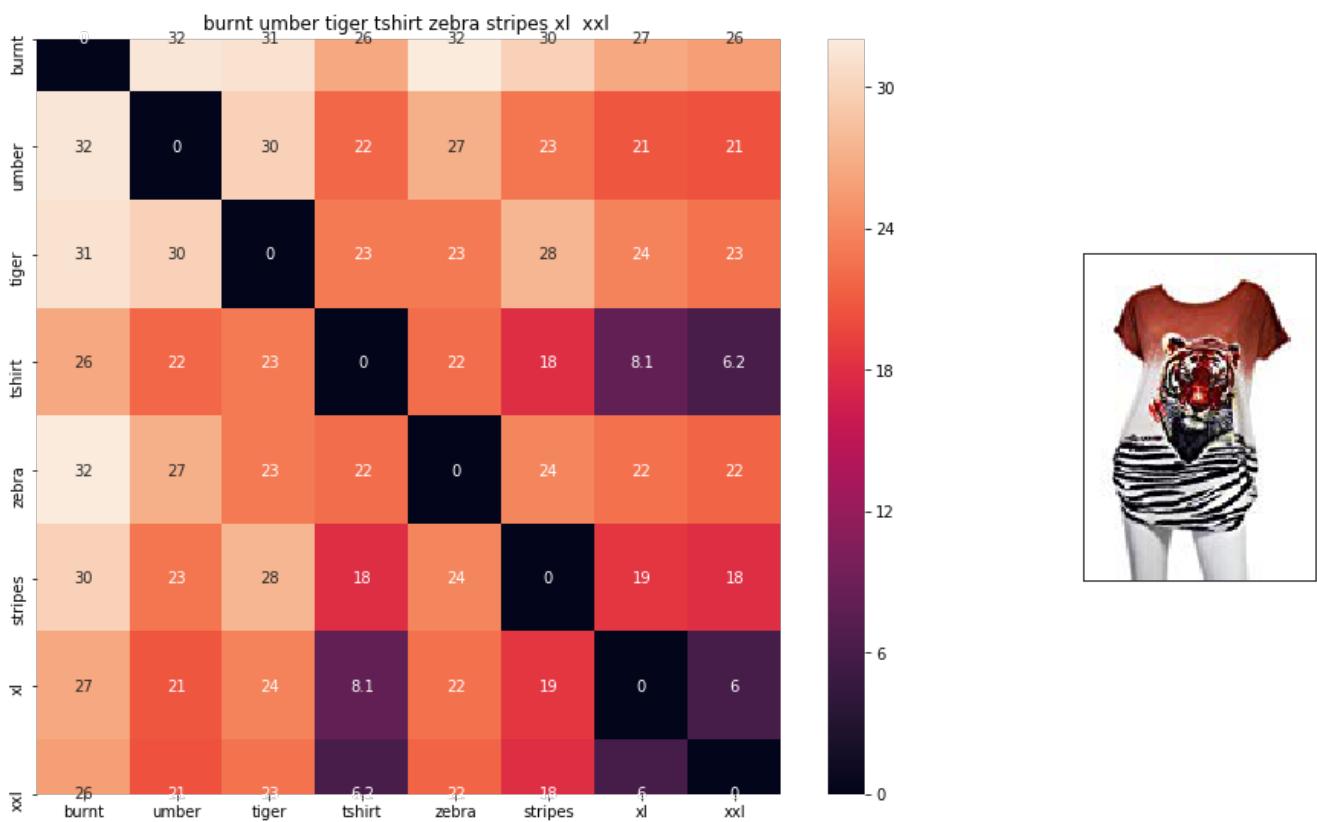
Brand : A.L.C.

euclidean distance from input : 2.0090413778255125

In [60]:

```
function(12566,20,3,15,100,1)
```

```
(16042, 1) (16042, 1) (16042, 1) (16042, 1)  
(16042, 1)  
[12566 14072 11526 2046 8581 5231 13905 15712 2097 2126 5707 4181  
 7124 10062 7074 9821 6417 770 8782 6812]  
Int64Index([142055, 160670, 129167, 23088, 89098, 52307, 158633, 178390,  
 23587, 23914, 56343, 42826, 69078, 109727, 68285, 106428,  
 62219, 9176, 91611, 65428],  
 dtype='int64')  
[142055, 160670, 129167, 23088, 89098, 52307, 158633, 178390, 23587, 23914, 56343, 42826, 69078, 1  
09727, 68285, 106428, 62219, 9176, 91611, 65428]
```

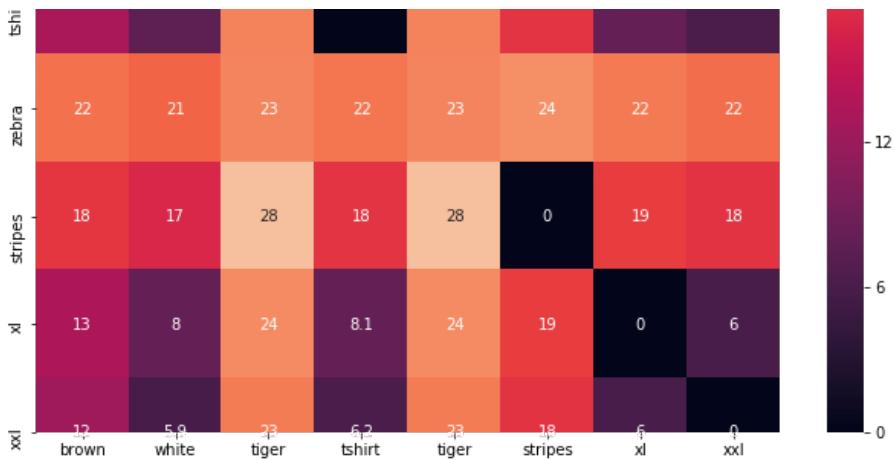


ASIN : B00JXQB5FQ

Brand : Si Row

euclidean distance from input : 4.2501021786901056e-09

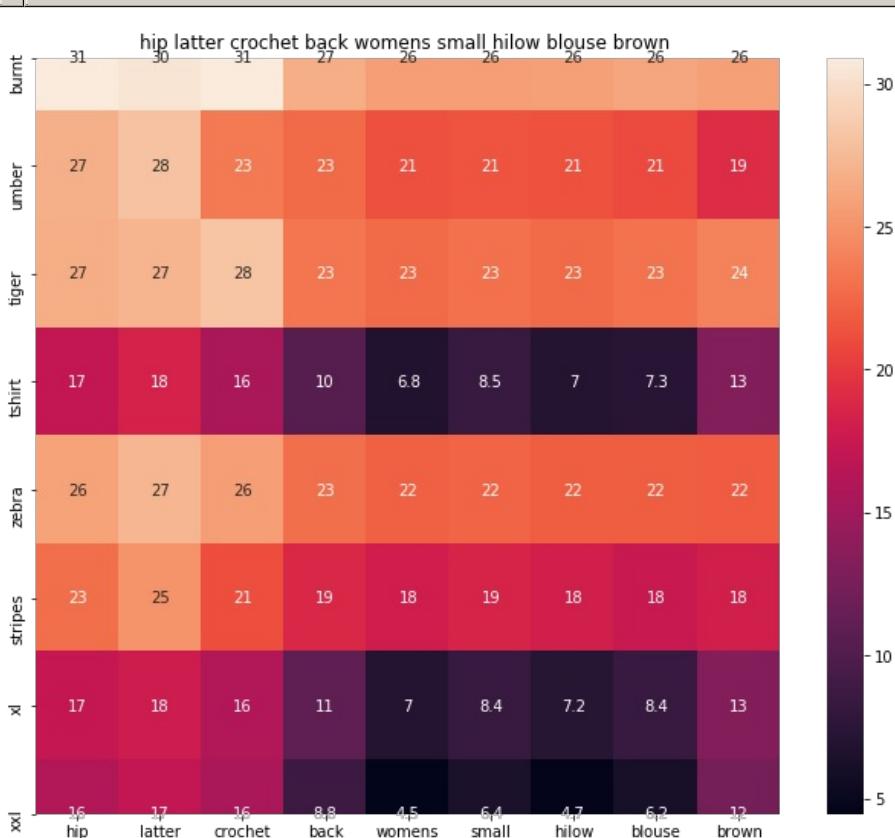




ASIN : B00JXQCWTO

Brand : Si Row

euclidean distance from input : 0.5870644705636161

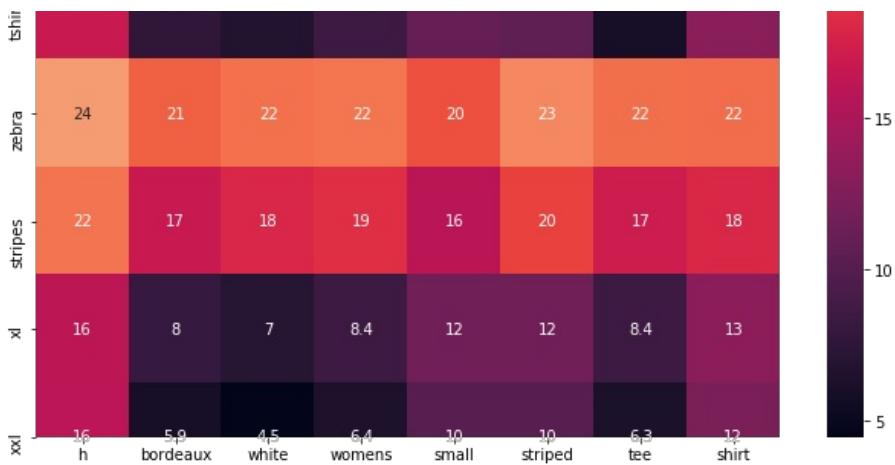


ASIN : B074MJN1K9

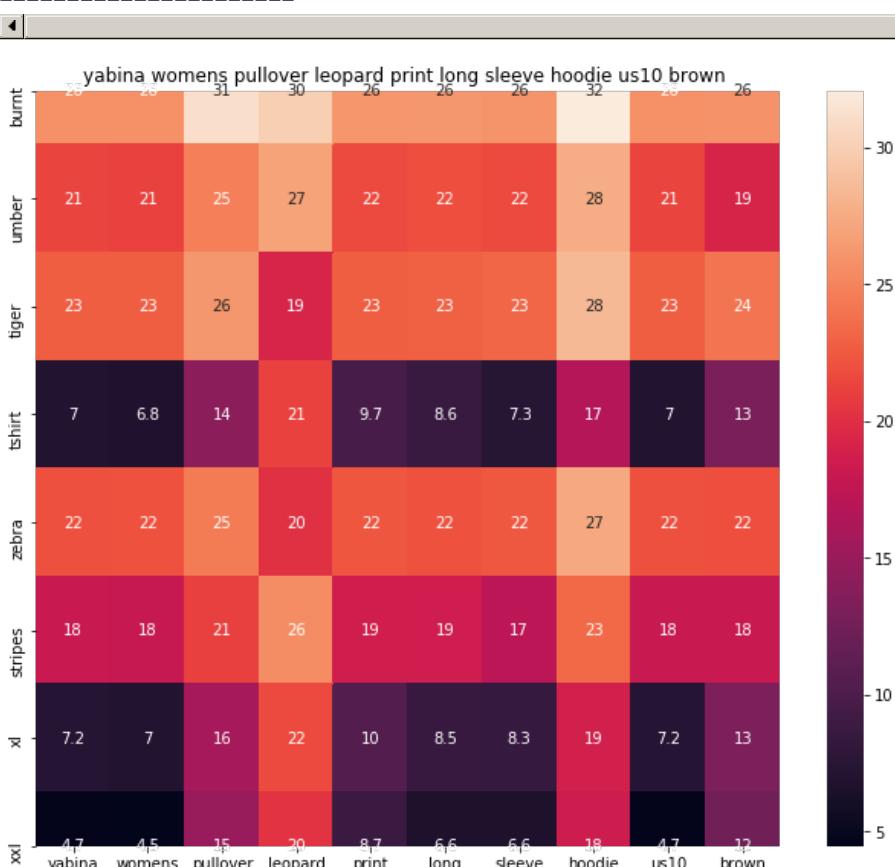
Brand : Hip

euclidean distance from input : 0.8118139284114689



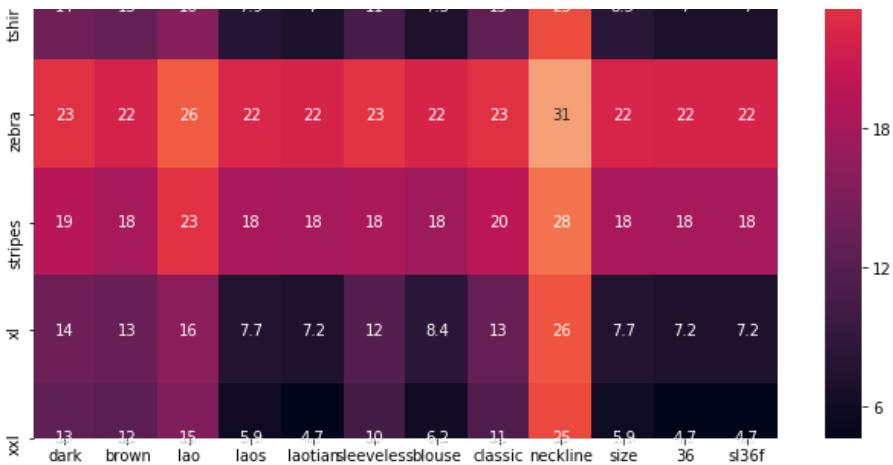


ASIN : B072BVB47Z
 Brand : H By Bordeaux
 euclidean distance from input : 0.8159045211407316



ASIN : B01KJUM6JI
 Brand : YABINA
 euclidean distance from input : 0.8330726592597653

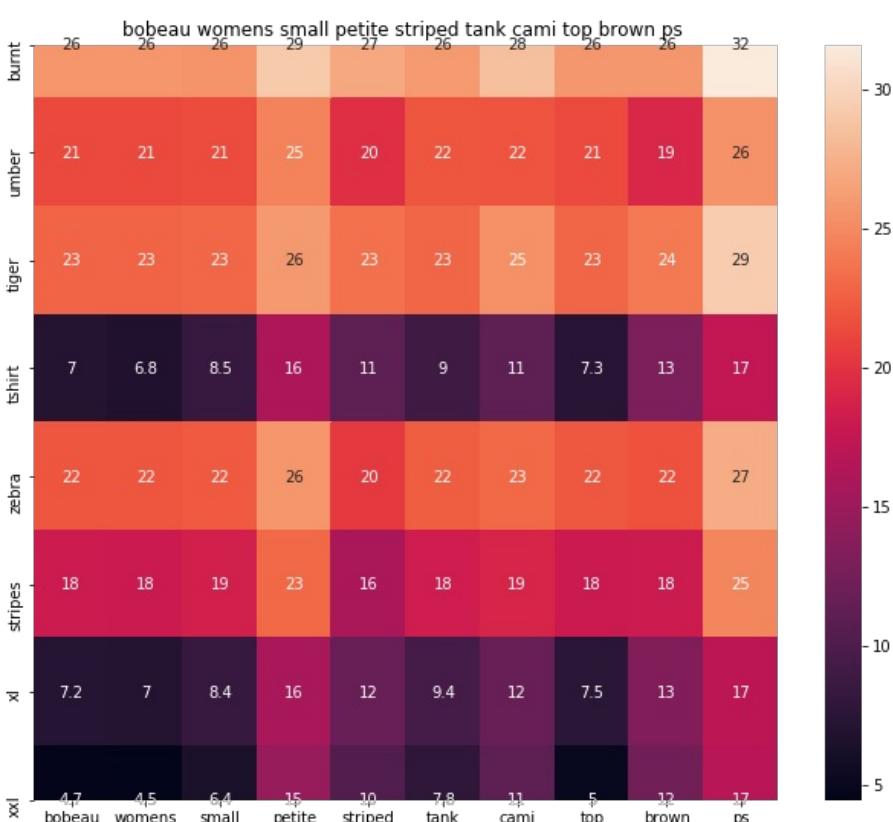




ASIN : B074J7BCYM

Brand : Nanon

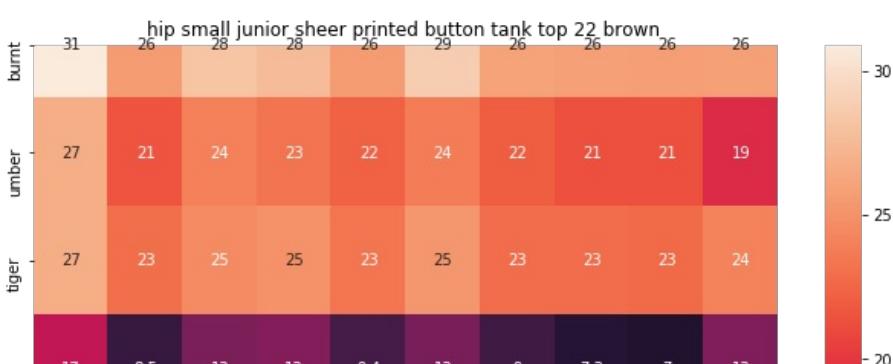
euclidean distance from input : 0.833402614545102

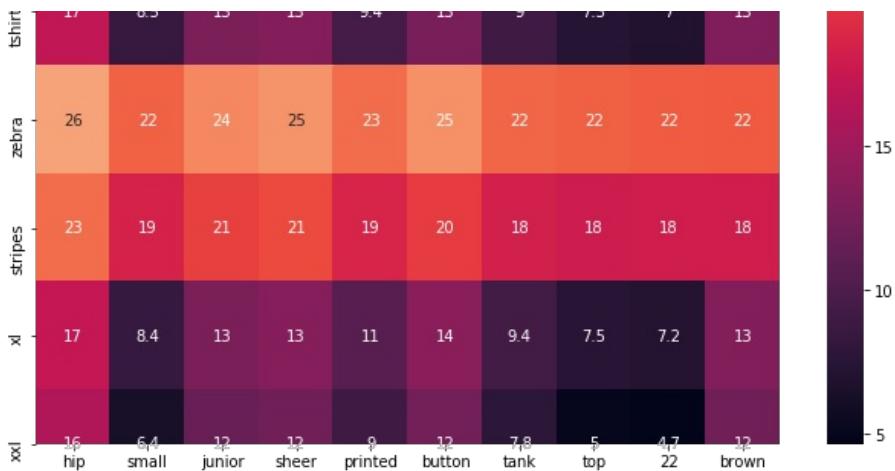


ASIN : B074P8YWV4

Brand : Bobeau

euclidean distance from input : 0.8450697218787582

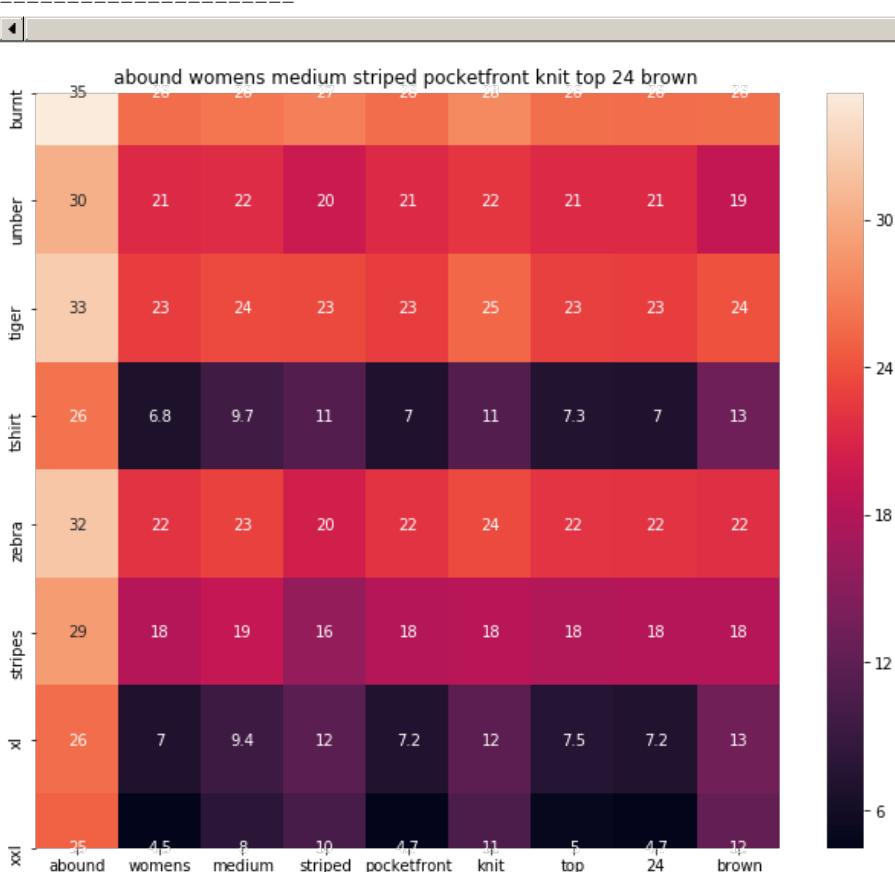




ASIN : B071LDTQ1F

Brand : Hip

euclidean distance from input : 0.8451120682240346

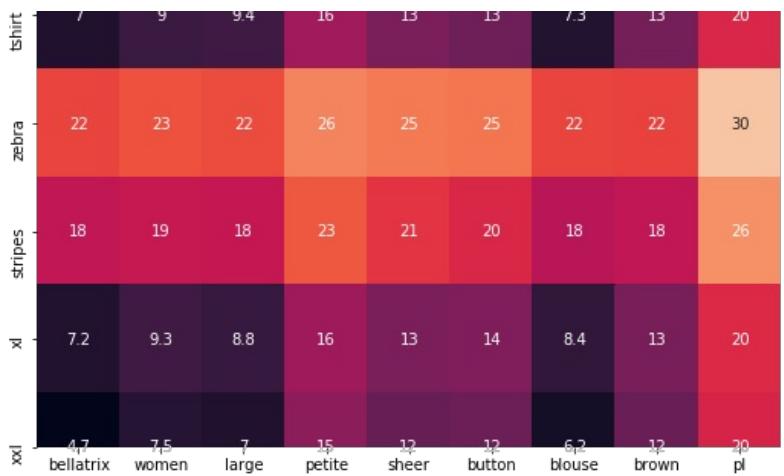


ASIN : B072M4ZF89

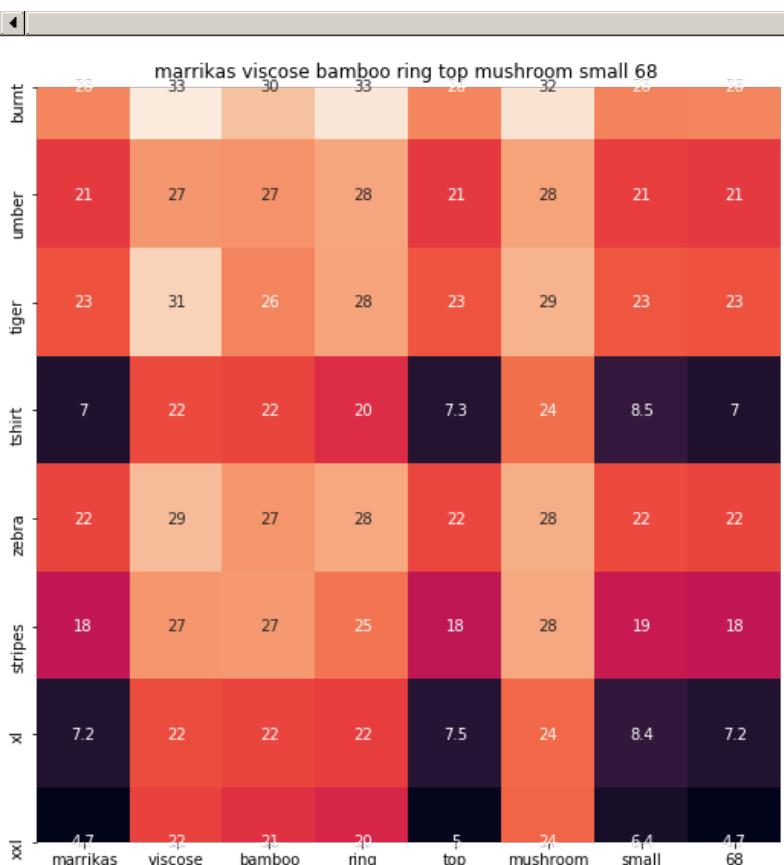
Brand : Abound

euclidean distance from input : 0.8469636229023023

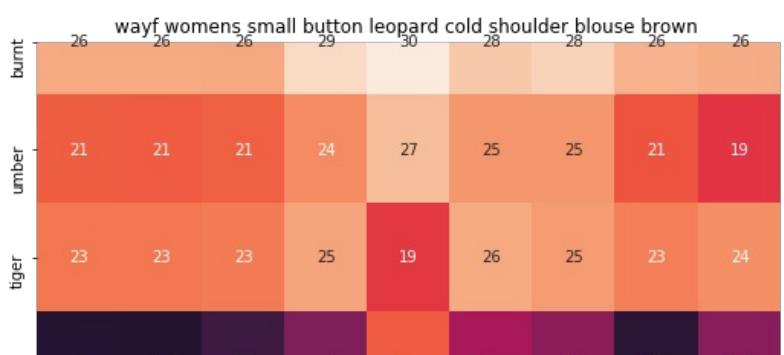




ASIN : B074QVMXSQ
 Brand : bellatrix
 euclidean distance from input : 0.8489006716781116



ASIN : B003SPYNAM
 Brand : Marrikas
 euclidean distance from input : 0.8519644185391606

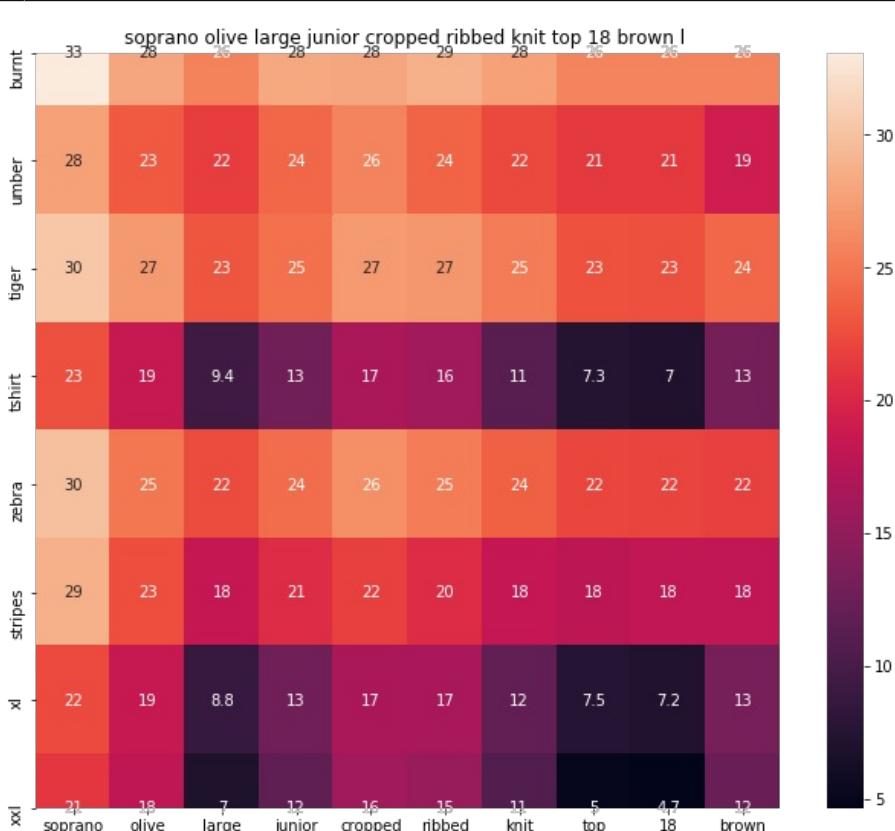




ASIN : B01M06V4X1

Brand : WAYF

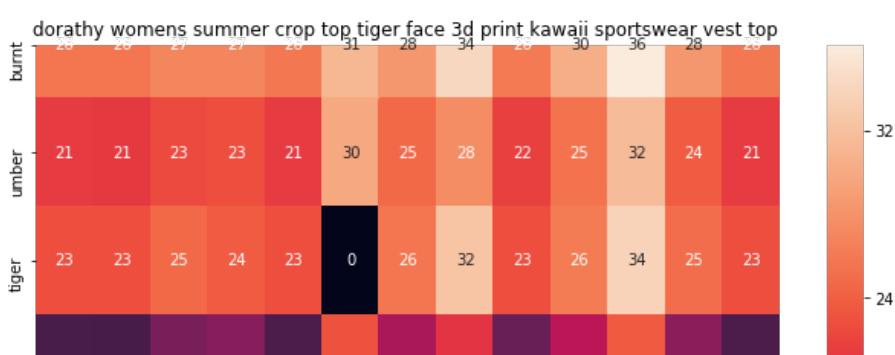
euclidean distance from input : 0.8532075610598407

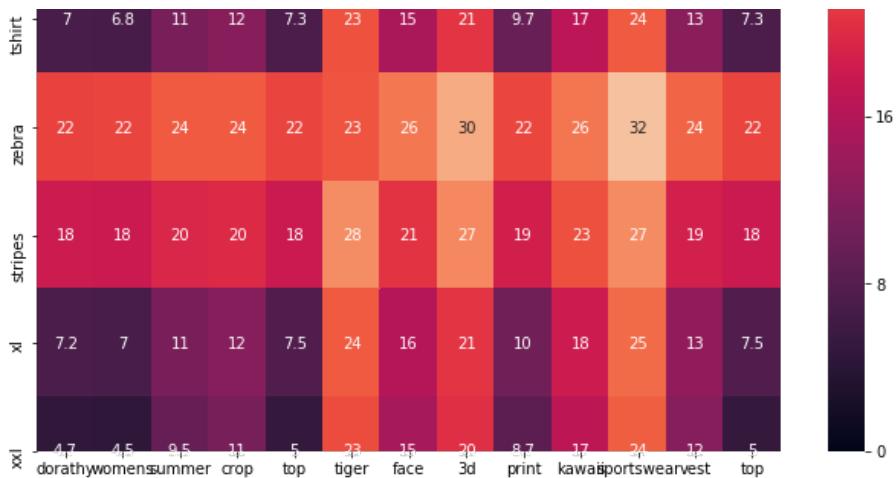


ASIN : B07288KFHF

Brand : Soprano

euclidean distance from input : 0.8558777850332192

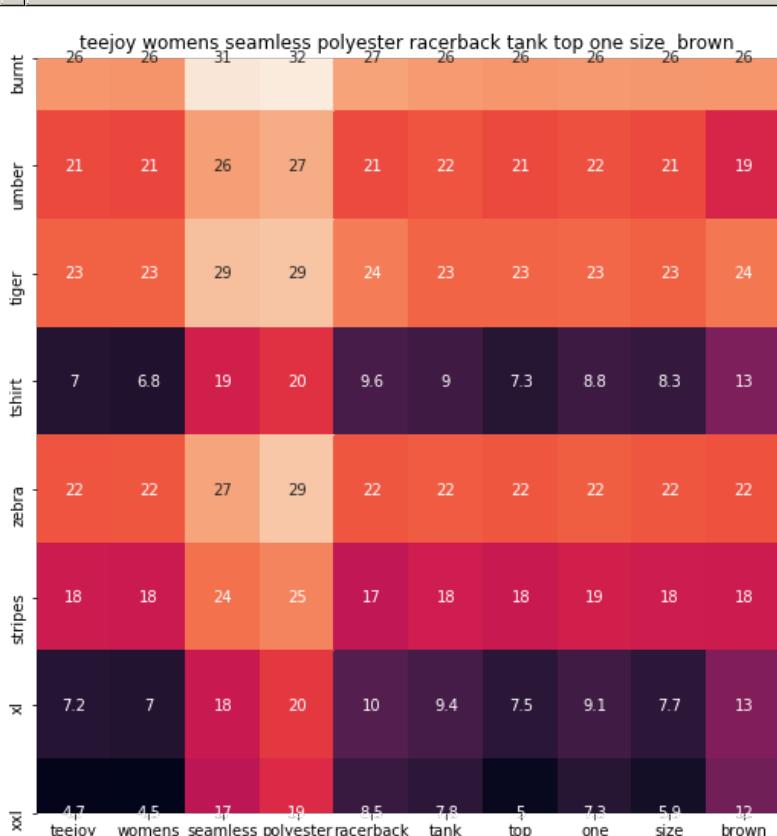




ASIN : B01HDKPMVQ

Brand : Dorothy

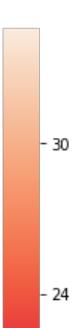
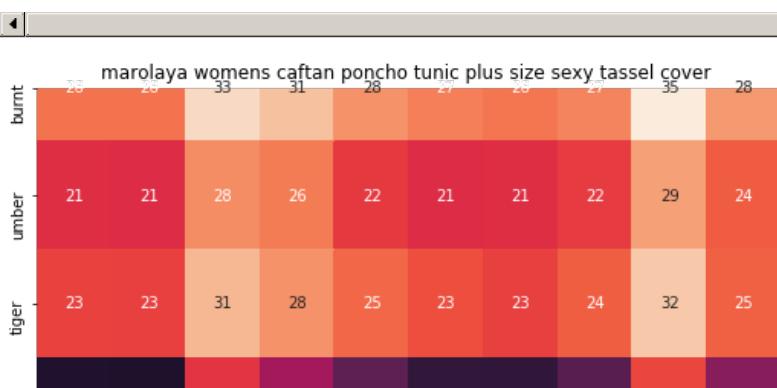
euclidean distance from input : 0.8573767687209325

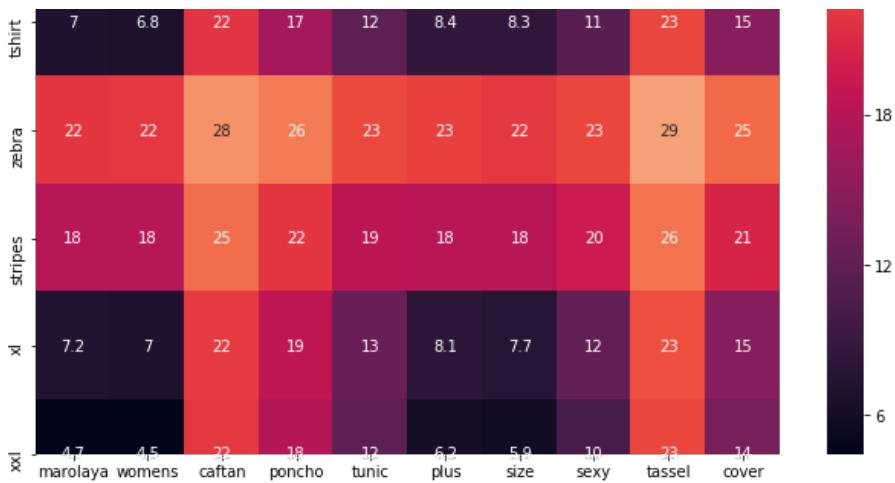


ASIN : B00JPOYCCO

Brand : Sofra

euclidean distance from input : 0.8622936891023664

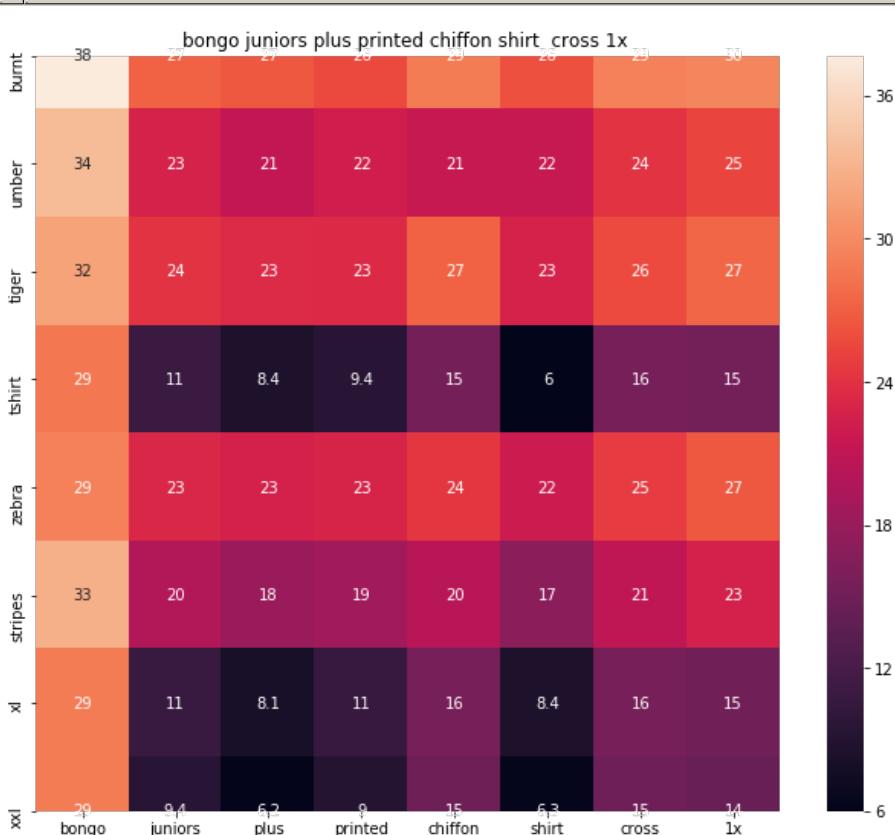




ASIN : B01CE40W16

Brand : Marolaya

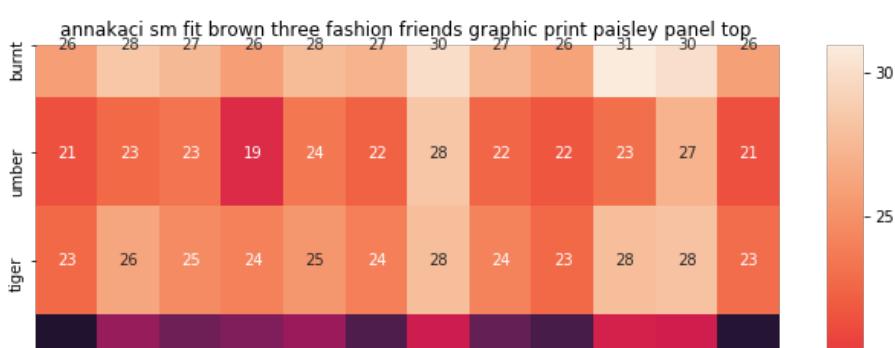
euclidean distance from input : 0.8626542100502474

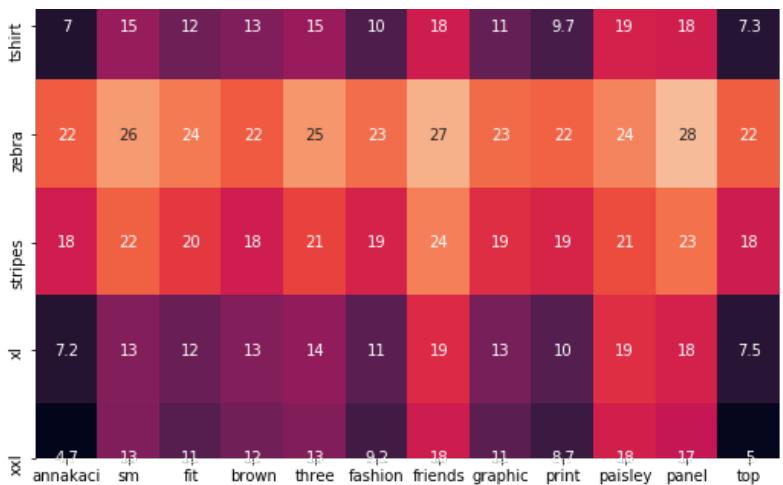


ASIN : B00K02DWJO

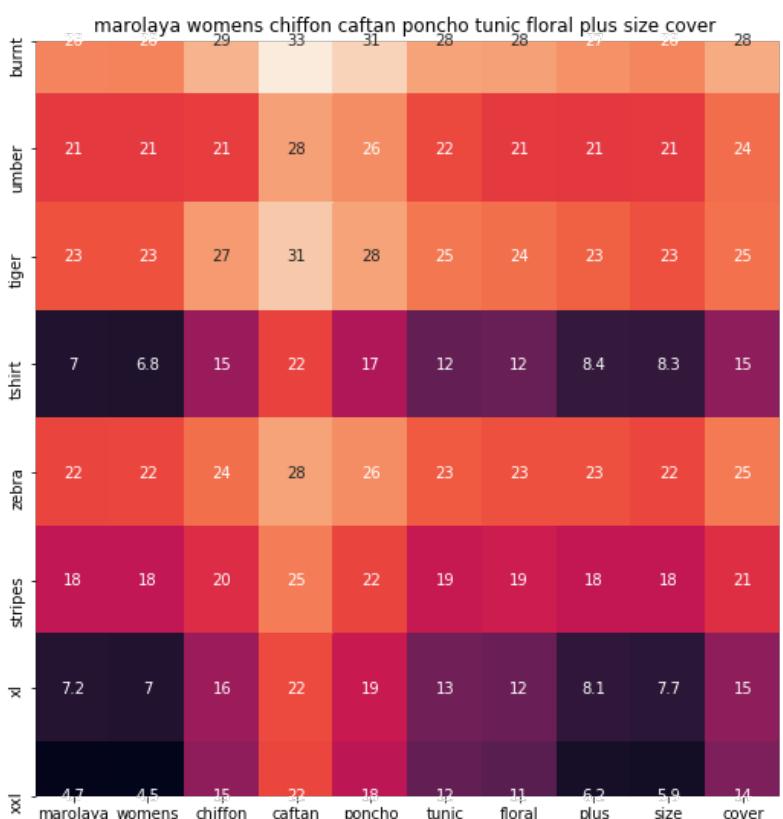
Brand : Bongo

euclidean distance from input : 0.8664538248363475



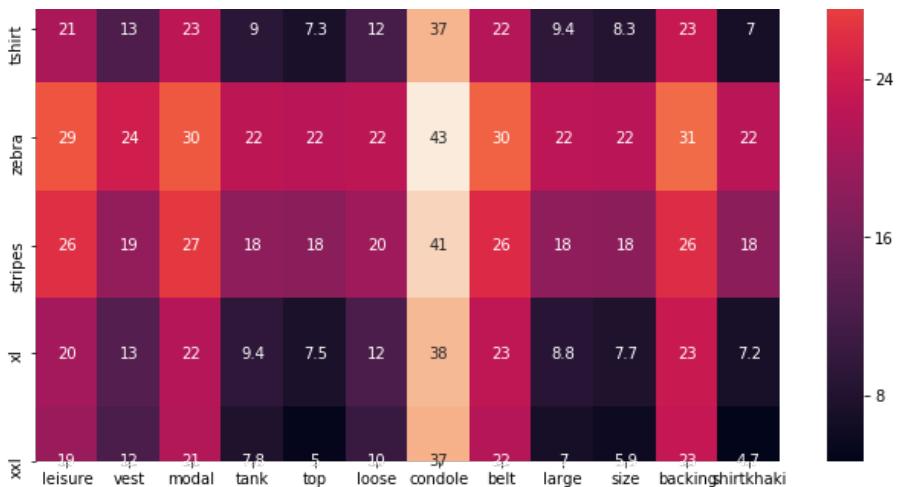


ASIN : B00BTJKAQ0
 Brand : Anna-Kaci
 euclidean distance from input : 0.8668007089310333



ASIN : B01CE40VX0
 Brand : Marolaya
 euclidean distance from input : 0.872739772748227





ASIN : B014OUHUZY

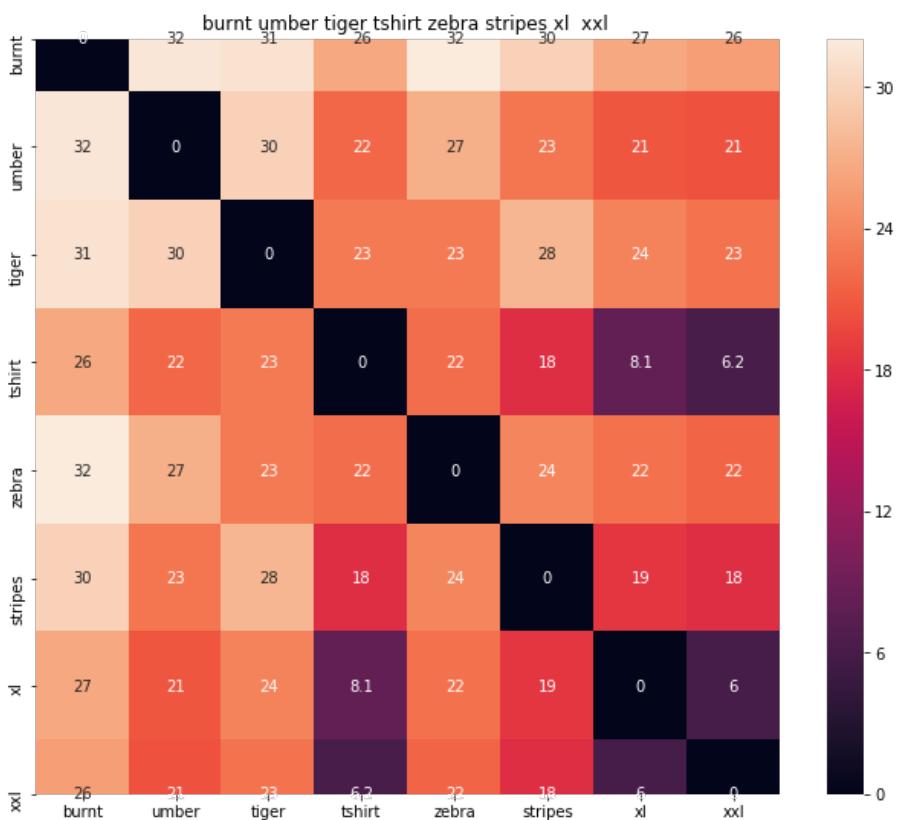
Brand : Black Temptation

euclidean distance from input : 0.8728023657277852

In [61]:

```
function(12566,20,3,10,10,1)
```

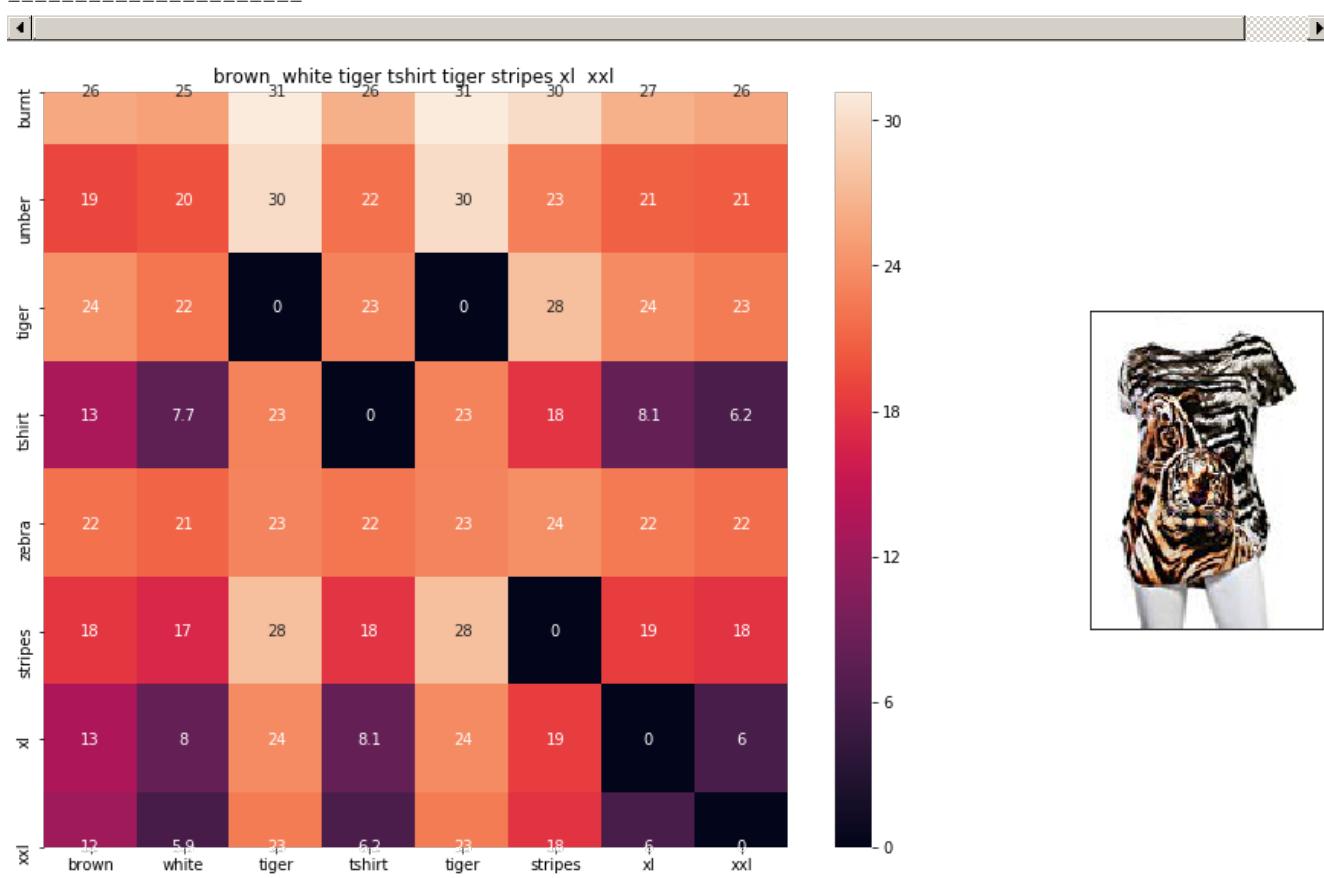
```
(16042, 1) (16042, 1) (16042, 1) (16042, 1)
(16042, 1)
[12566 14072 14089 14790 10569 7330 7482 2046 11526 8581 5231 13905
 15712 2097 2126 7105 13041 12934 5707 4181]
Int64Index([142055, 160670, 160819, 169492, 116737, 72149, 74241, 23088,
 129167, 89098, 52307, 158633, 178390, 23587, 23914, 68773,
 147862, 146397, 56343, 42826],
 dtype='int64')
[142055, 160670, 160819, 169492, 116737, 72149, 74241, 23088, 129167, 89098, 52307, 158633,
 178390, 23587, 23914, 68773, 147862, 146397, 56343, 42826]
```



ASIN : B00JXQB5FQ

Brand : Si Row

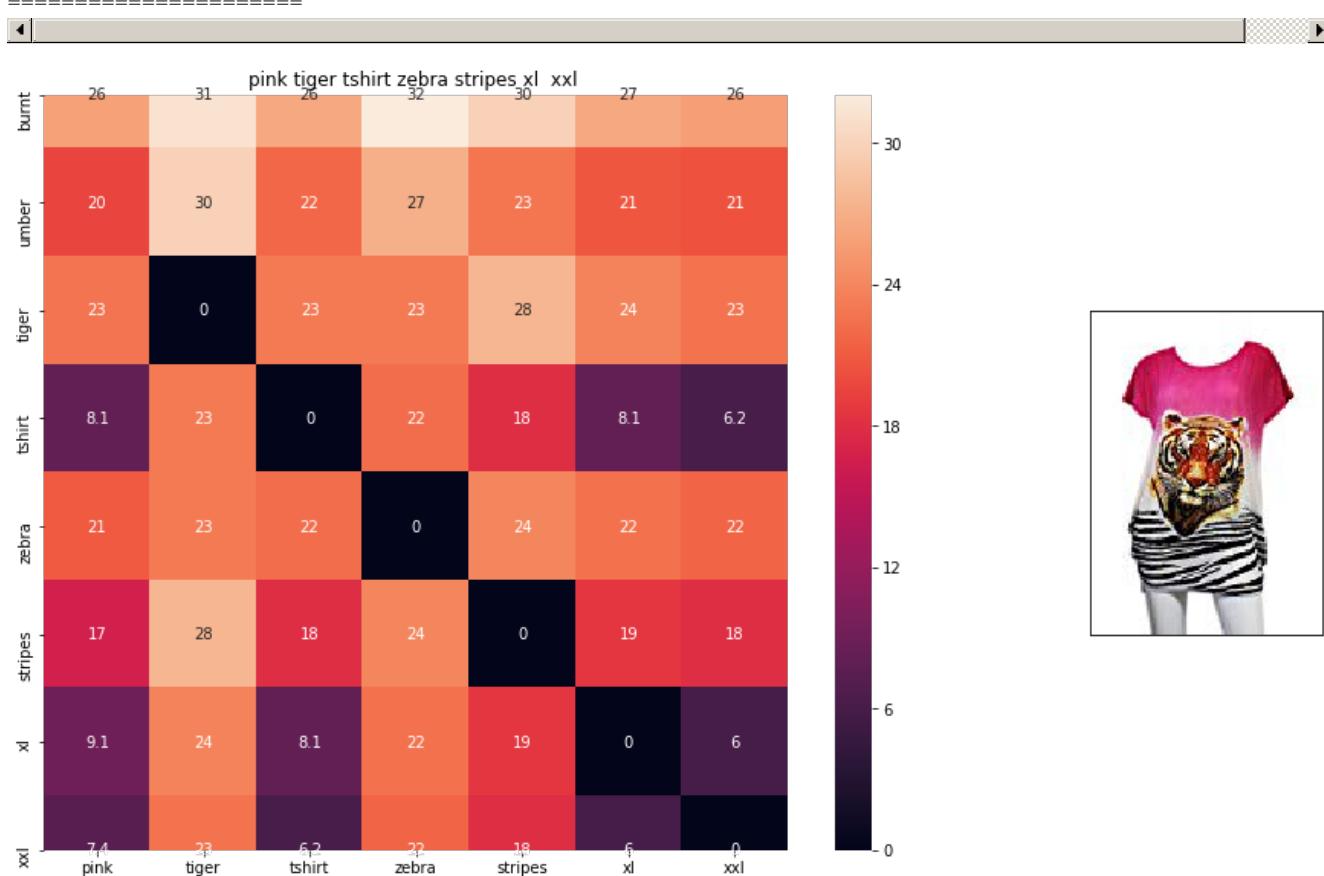
euclidean distance from input : 2.1073423302671774e-08



ASIN : B00JXQCWTO

Brand : Si Row

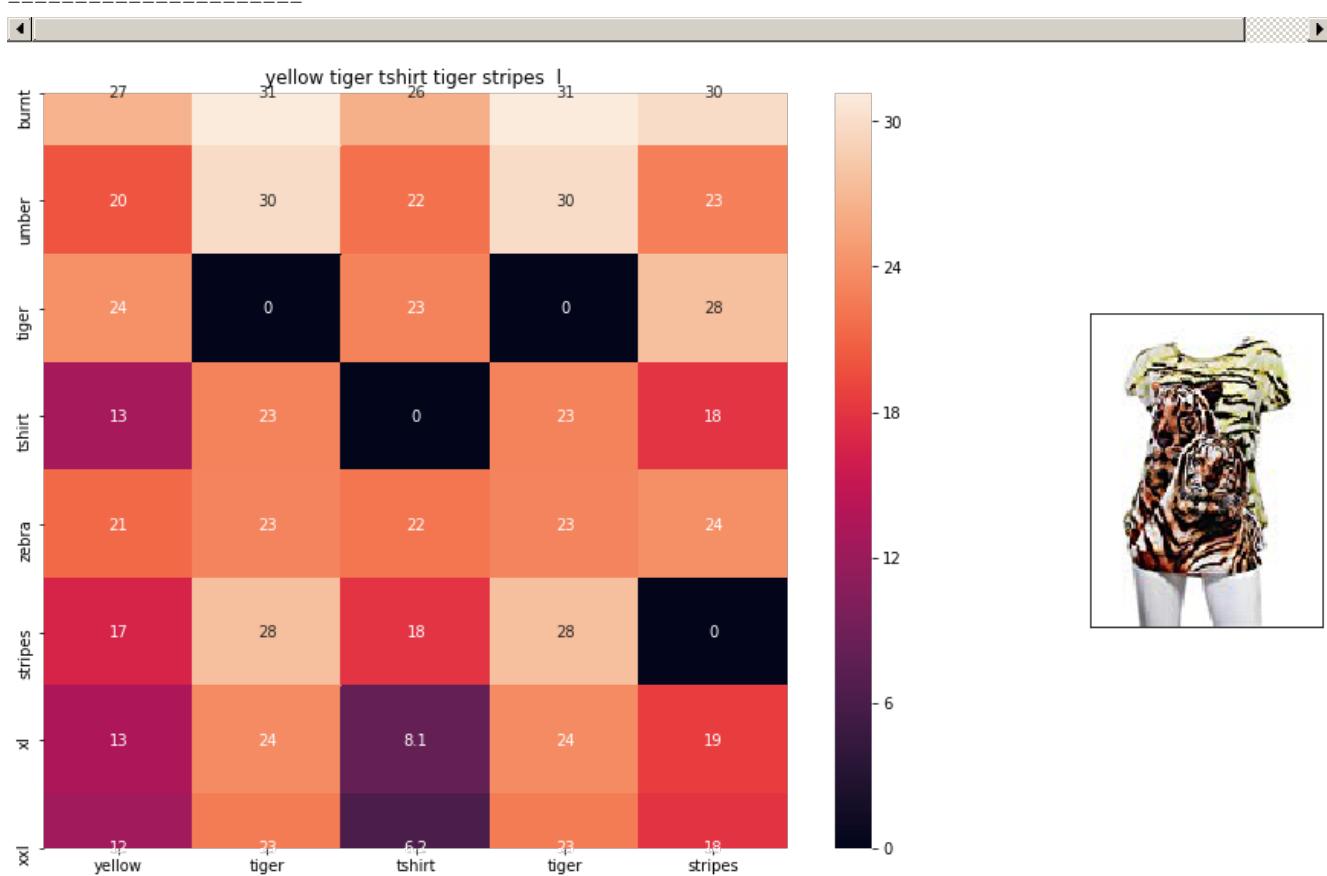
euclidean distance from input : 2.910861333211263



ASIN : B00JXQASS6

Brand : Si Row

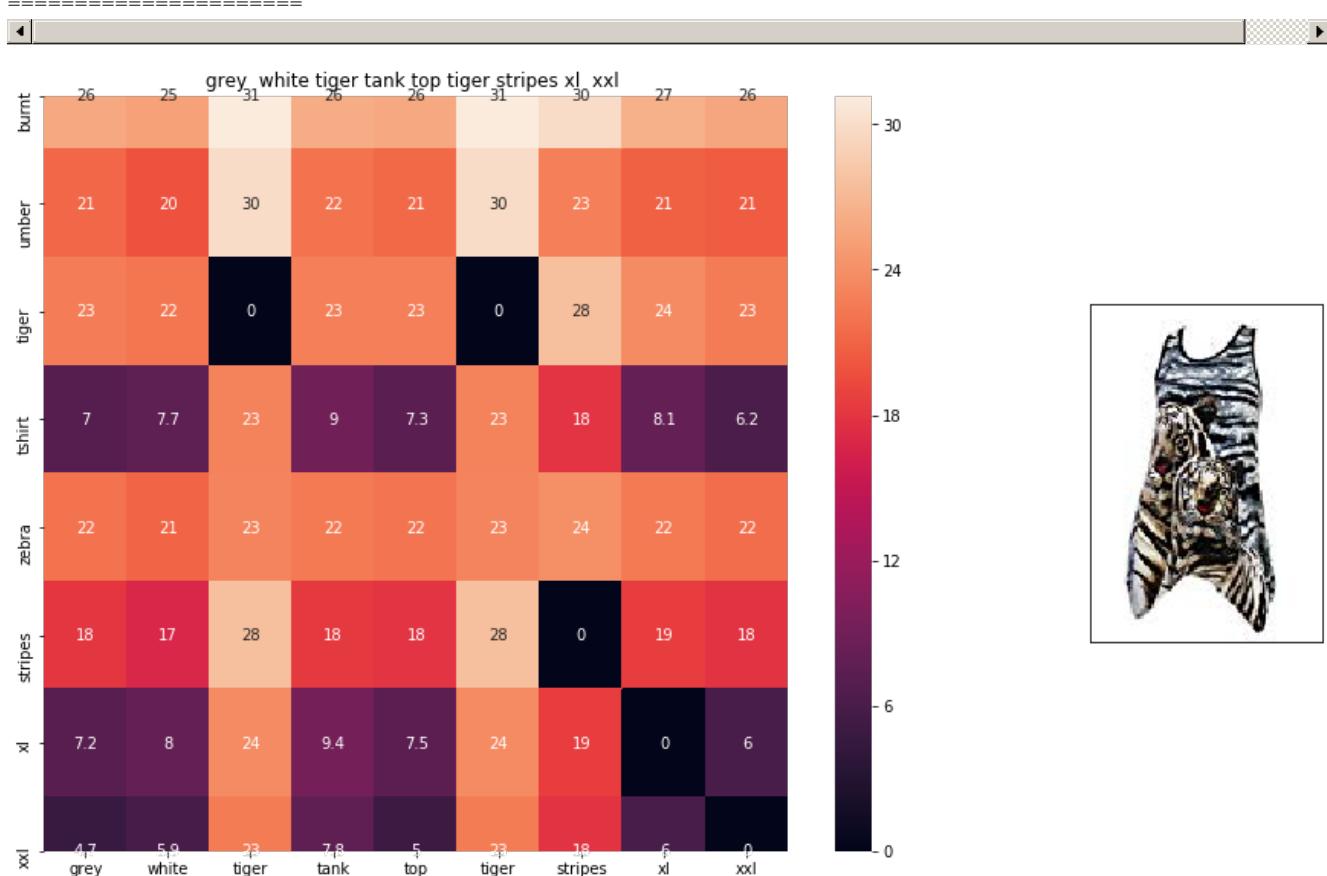
euclidean distance from input : 3.1184836230194293



ASIN : B00JXQCUIC

Brand : Si Row

euclidean distance from input : 3.499980211408508

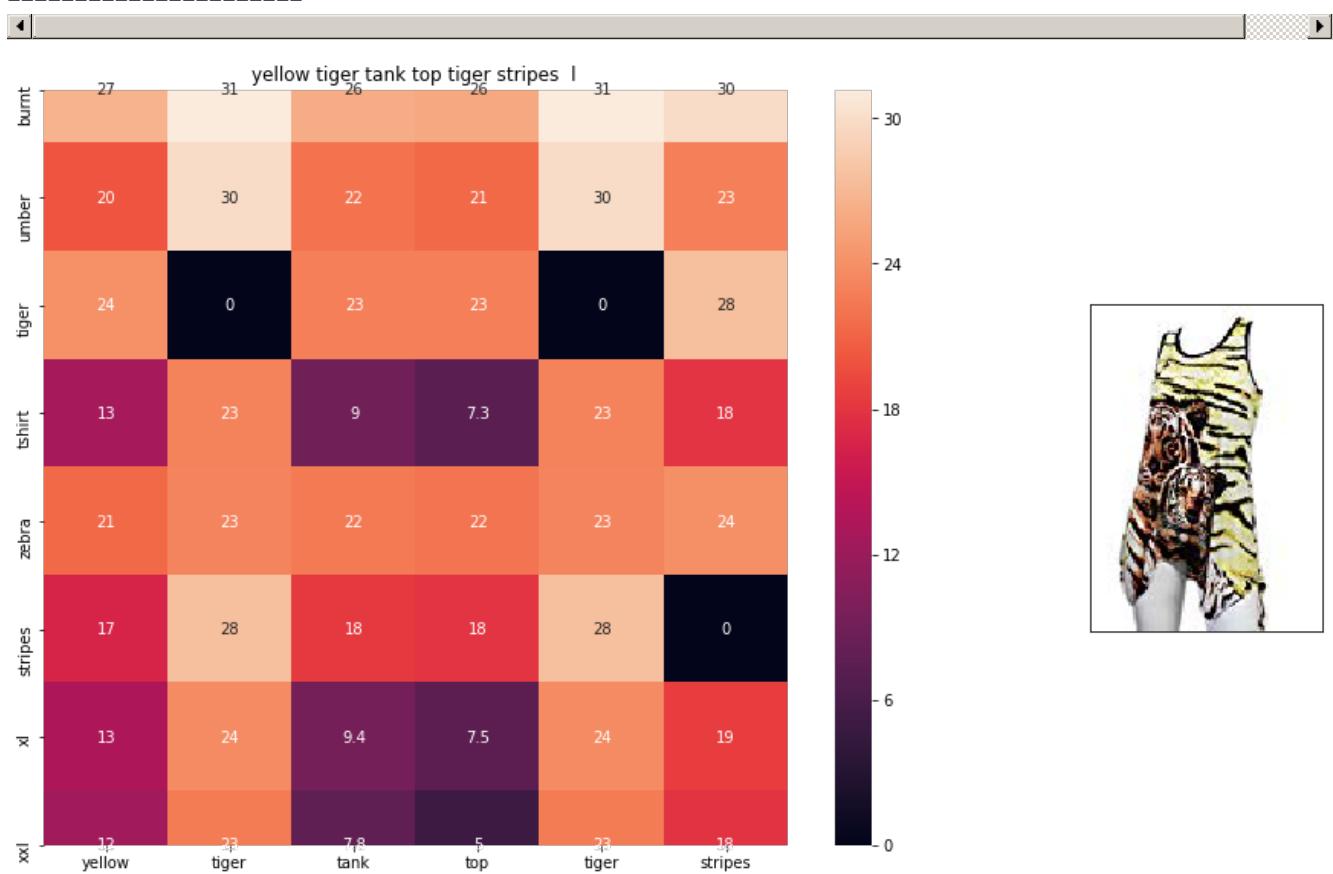


ASIN : B00JXQAFZ2

Brand : Si Row

euclidean distance from input : 3.5062178772206827

=====

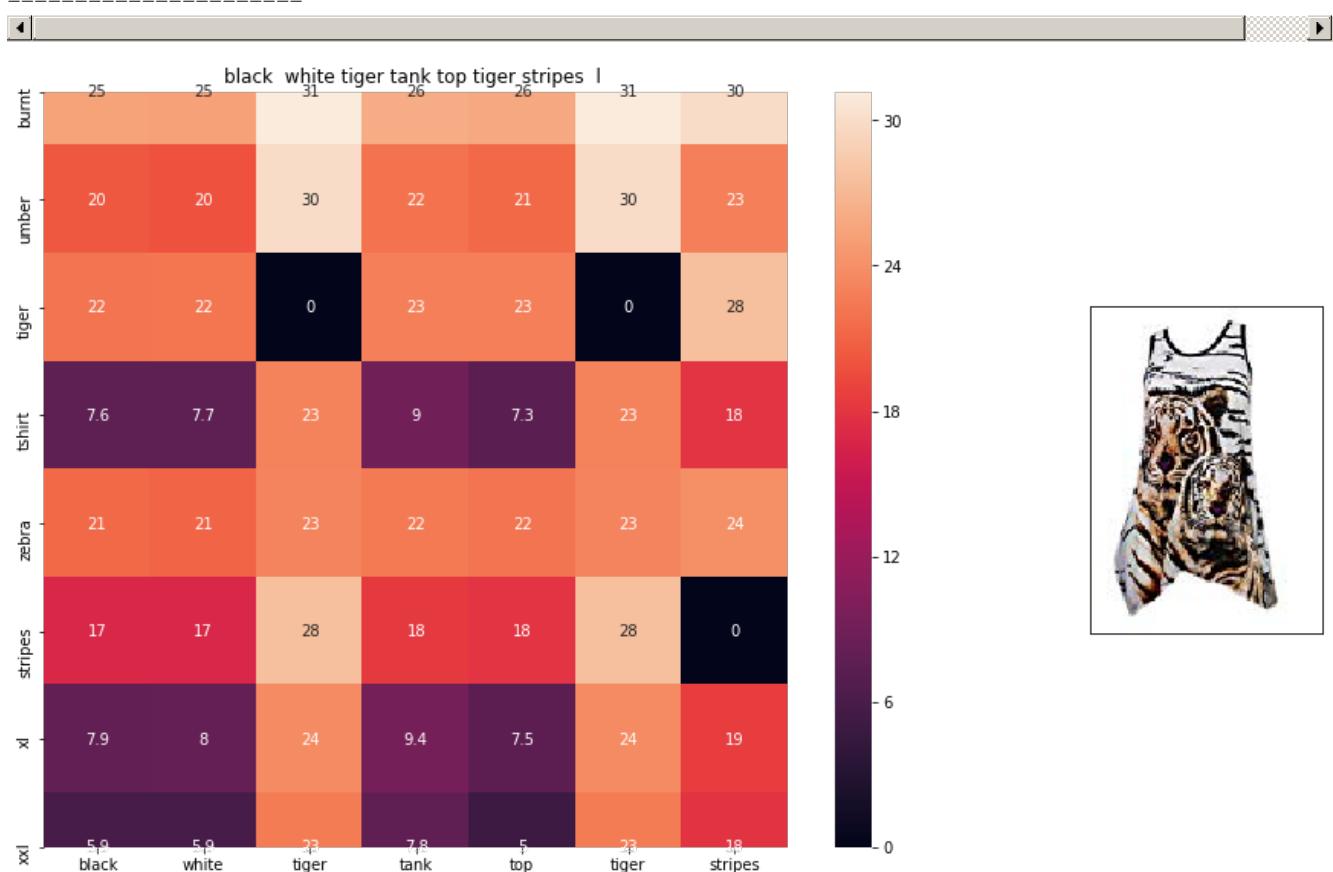


ASIN : B00JXQAUWA

Brand : Si Row

euclidean distance from input : 3.5287332536295772

=====

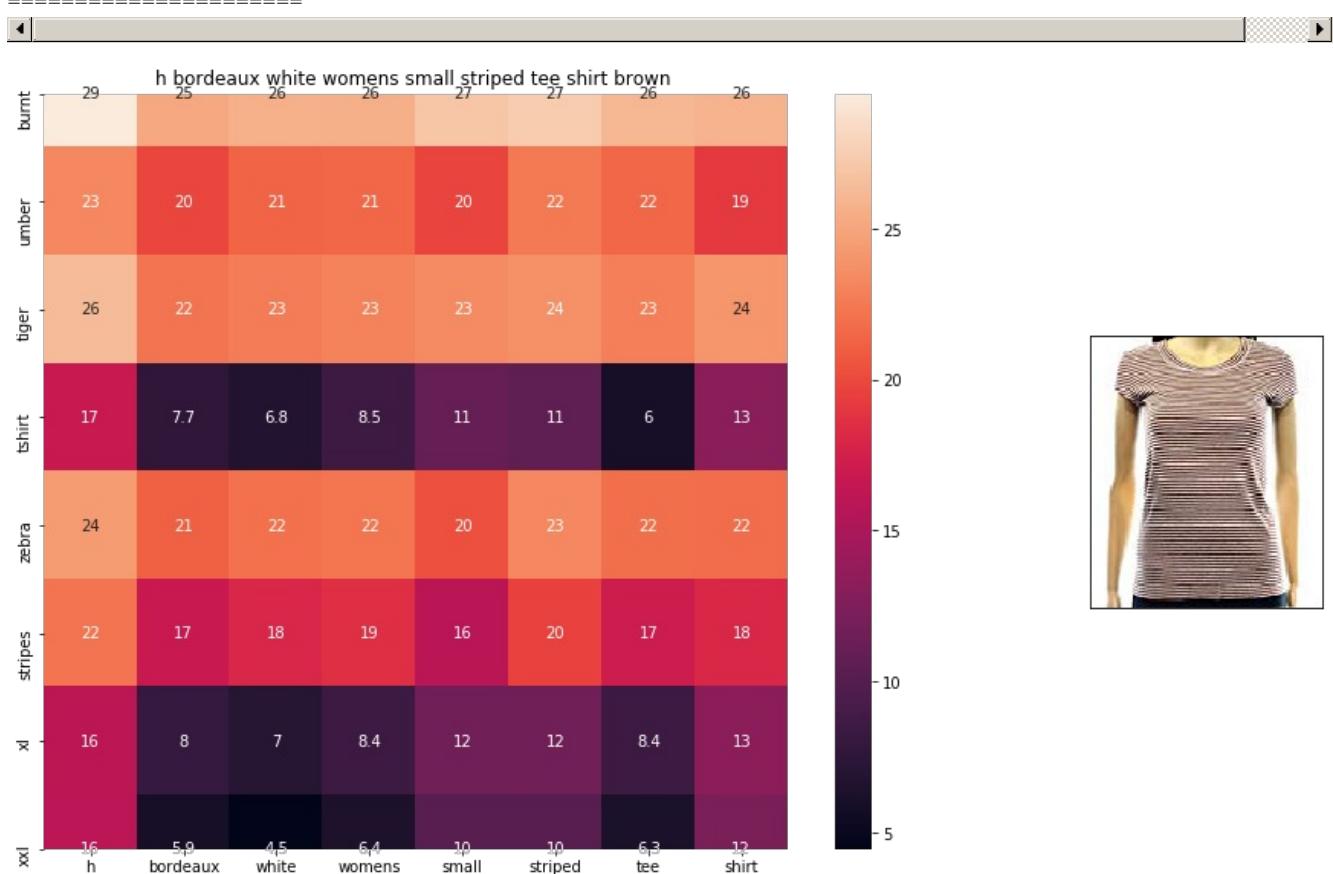


ASIN : B00JXQA094

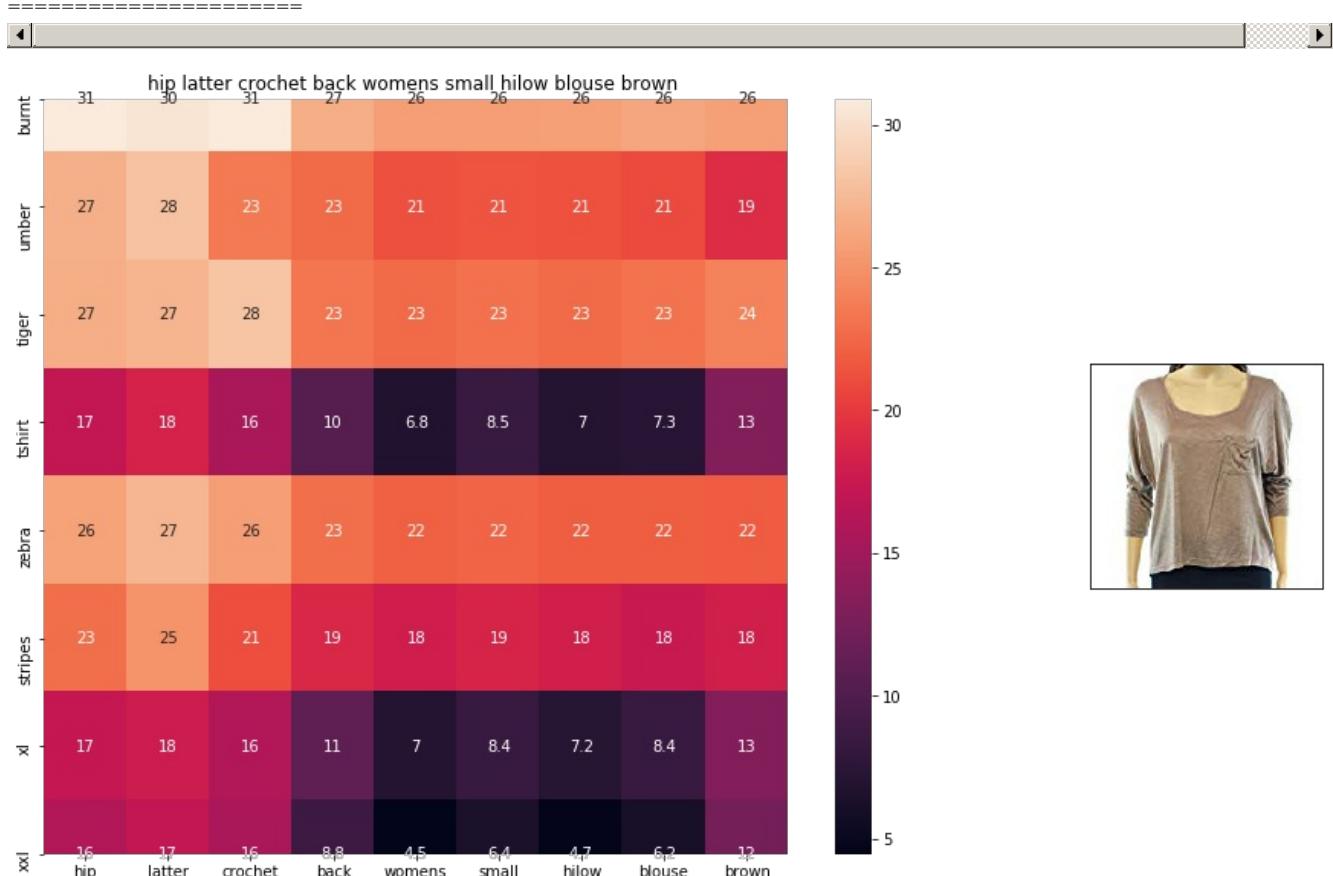
Brand : Si Row

euclidean distance from input : 3.6021046639994503

=====



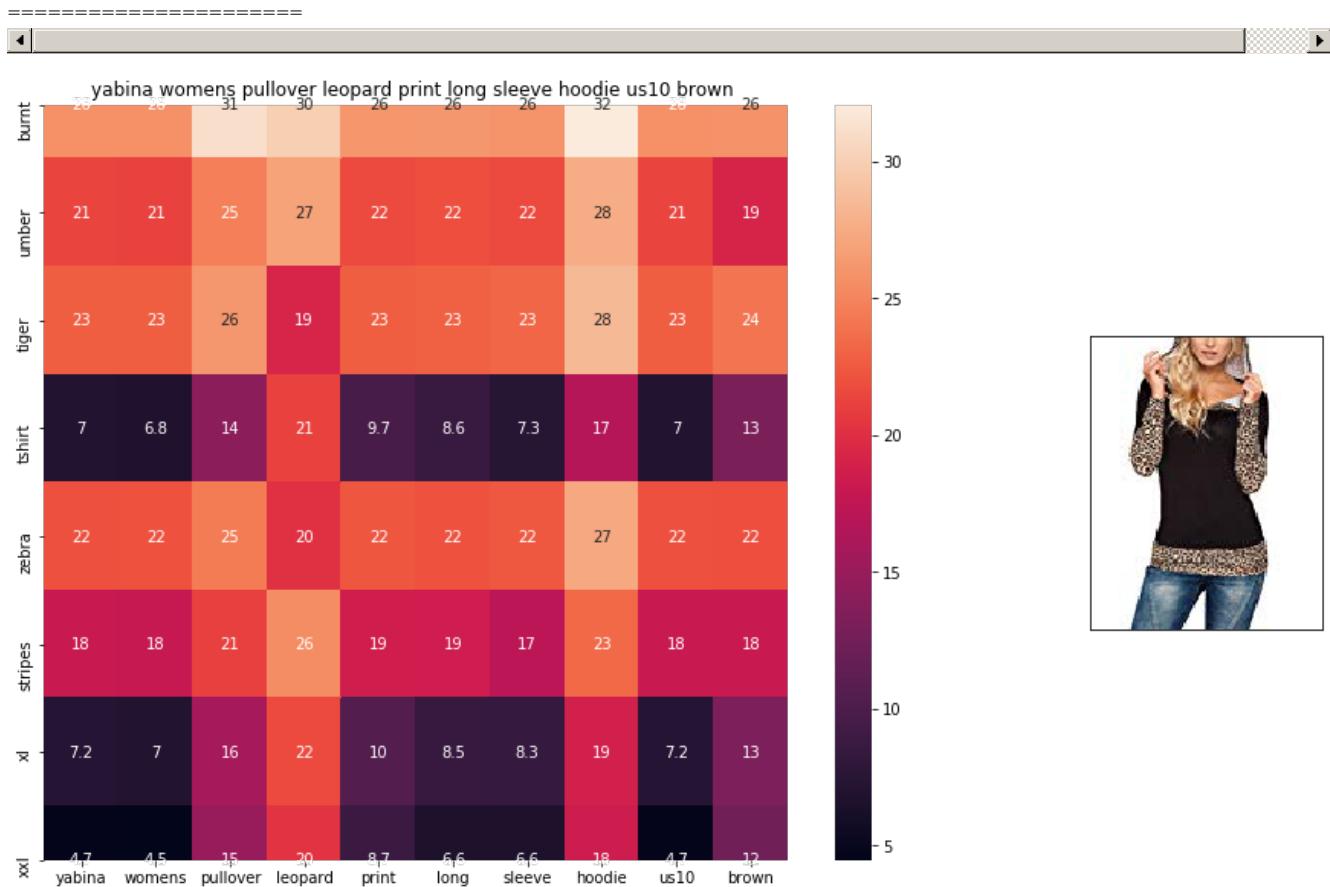
=====



ASIN : B074MJN1K9

Brand : Hip

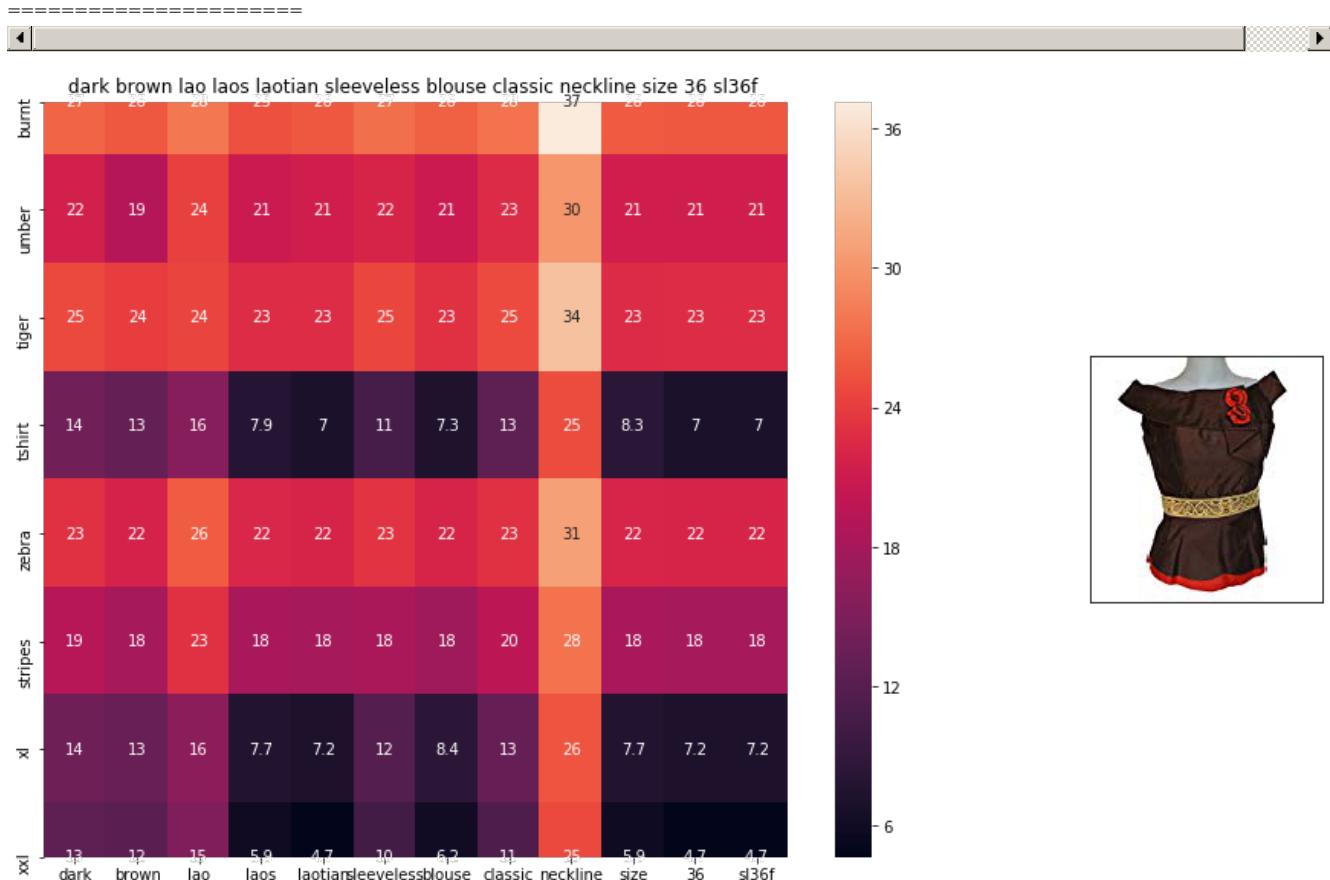
euclidean distance from input : 3.66440014346335



ASIN : B01KJUM6JI

Brand : YABINA

euclidean distance from input : 3.7698080172528194

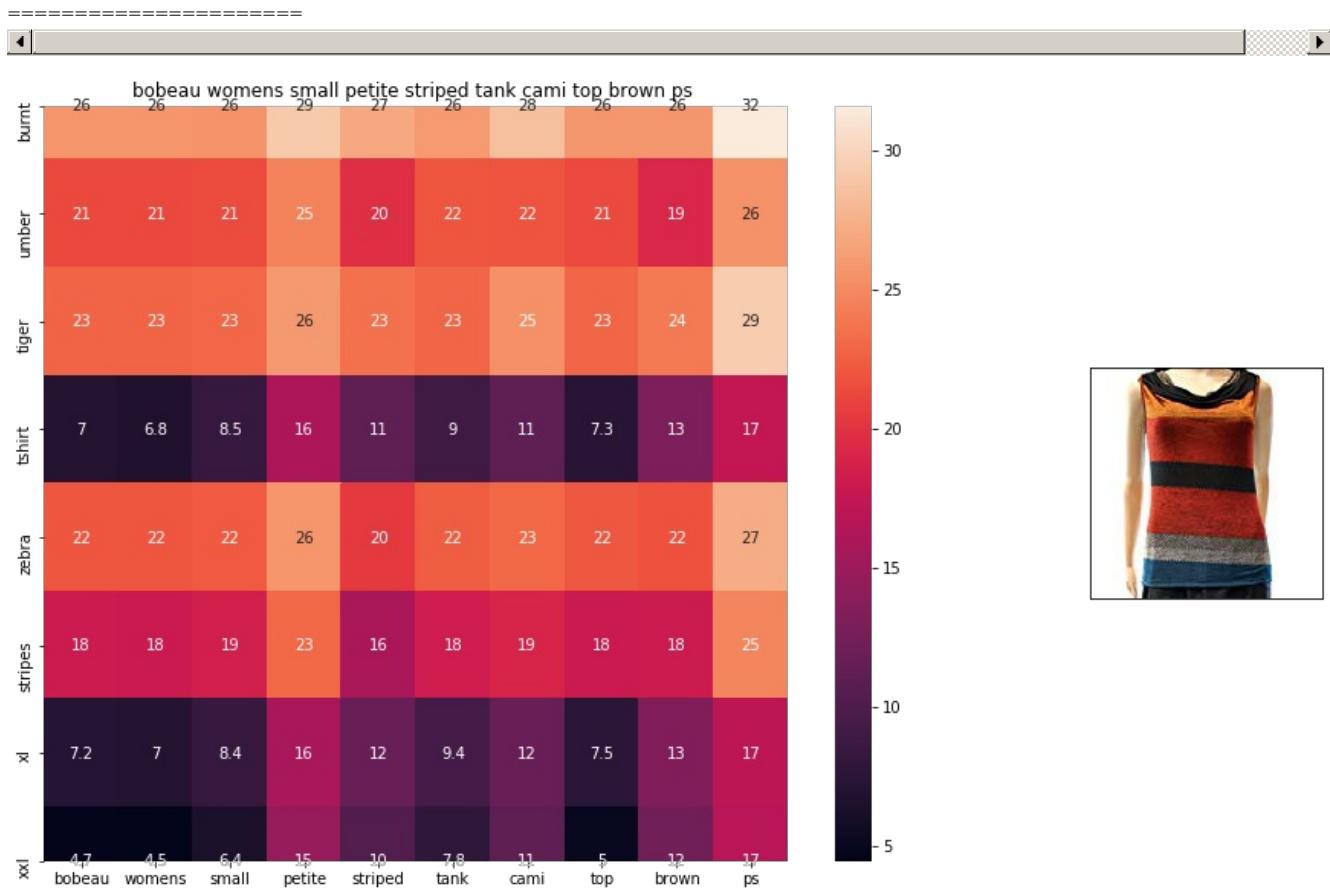


ASIN : B074J7BCYM

Brand : Nanon

euclidean distance from input : 3.7714440455426144

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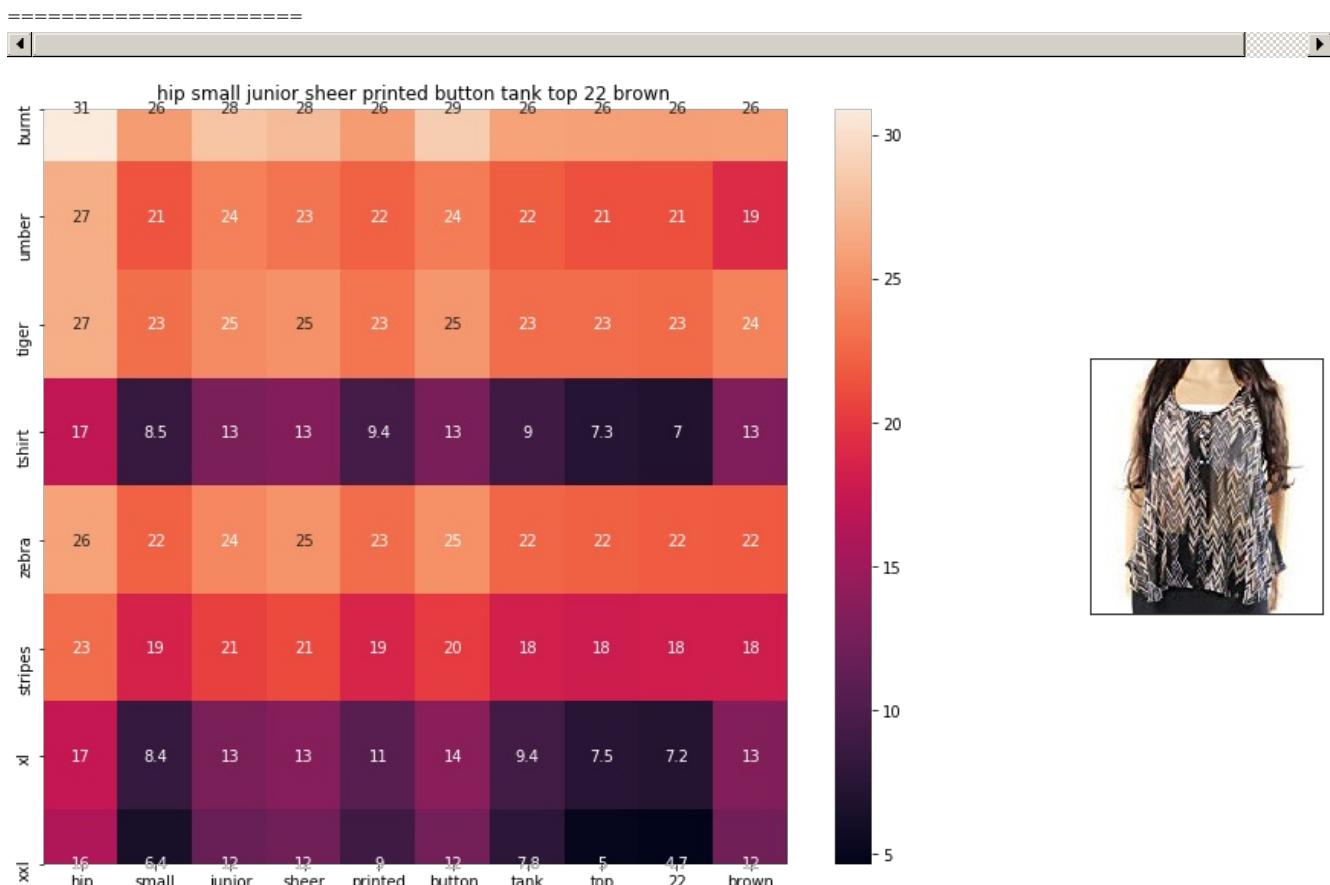


ASIN : B074P8YWV4

Brand : Bobeau

euclidean distance from input : 3.8292934527386593

=====

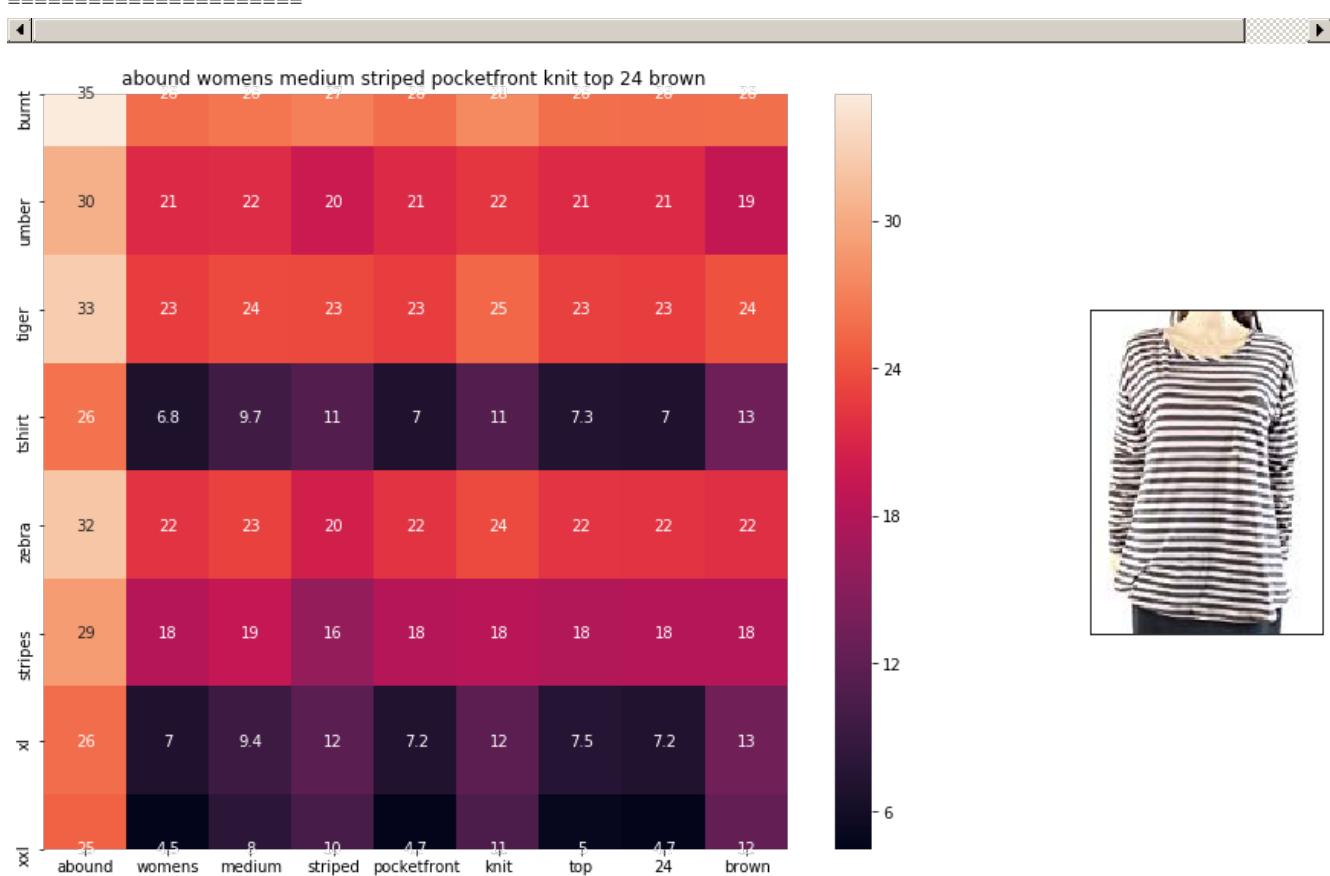


ASIN : B071LDTQ1F

Brand : Hin

Brand : nnp

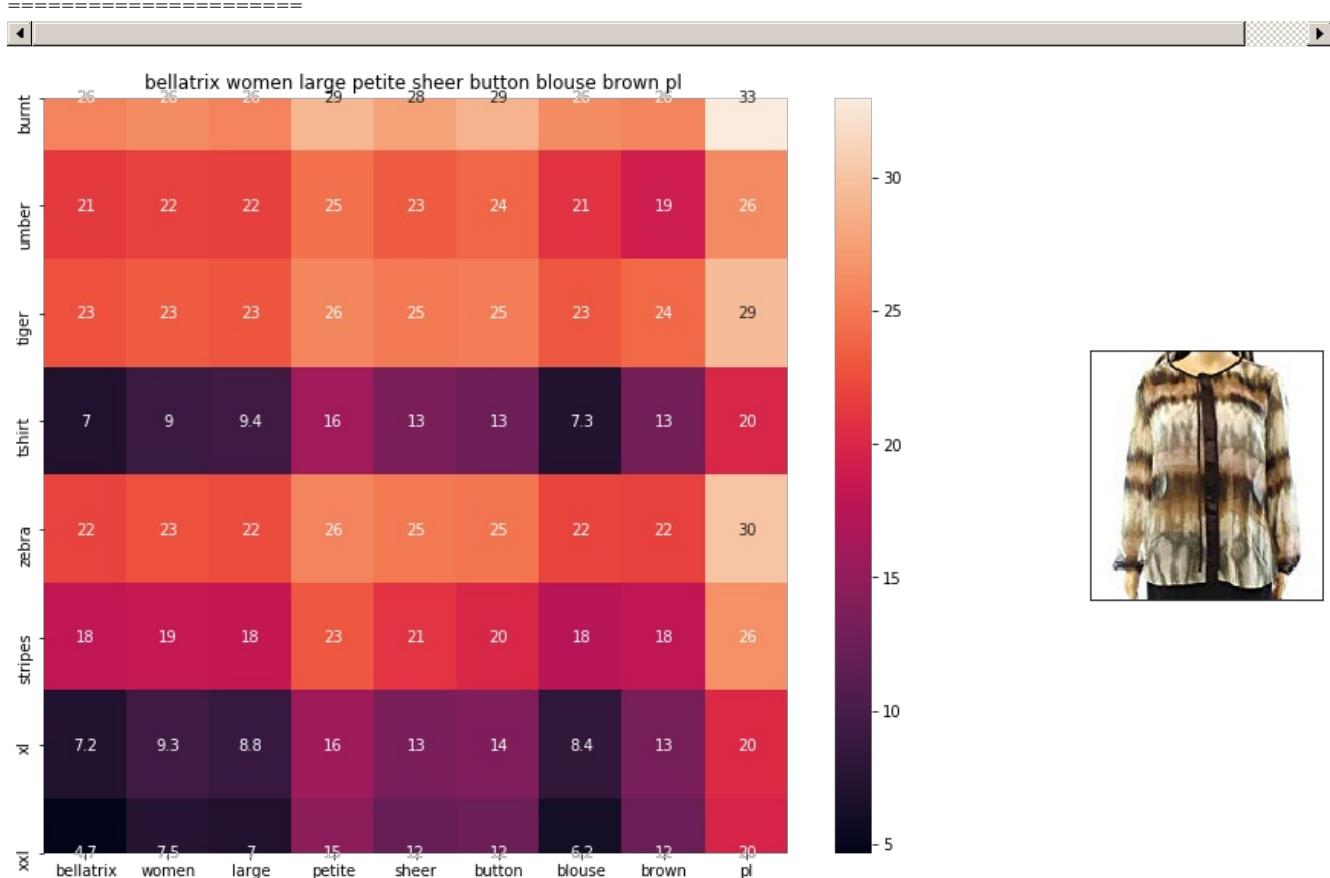
euclidean distance from input : 3.8295034200339884



ASIN : B072M4ZF89

Brand : Abound

euclidean distance from input : 3.8386840453137325



ASIN : B074QVMXSQ

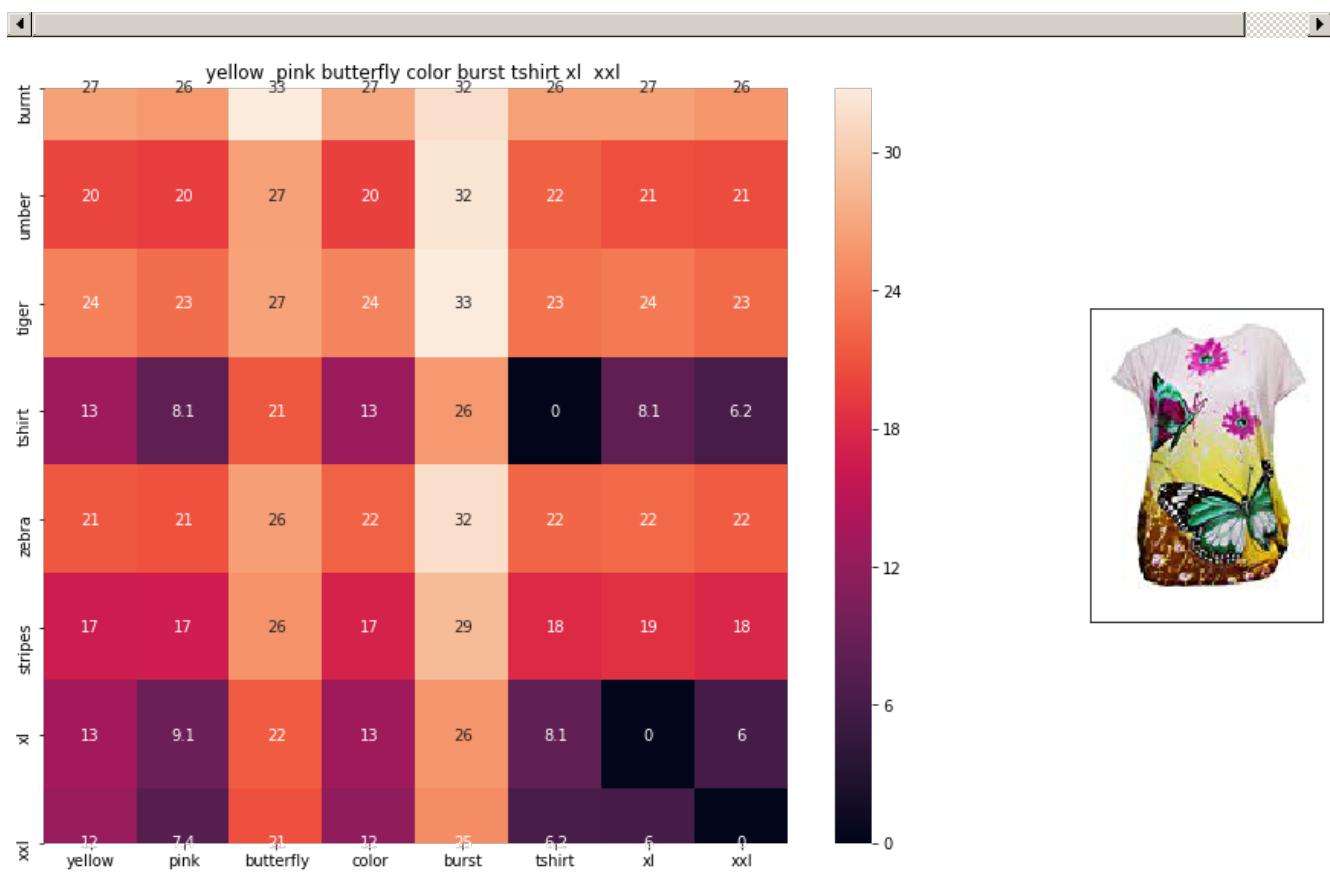
Brand : bellatrix

Brand : versatix

euclidean distance from input : 3.8482885788271197

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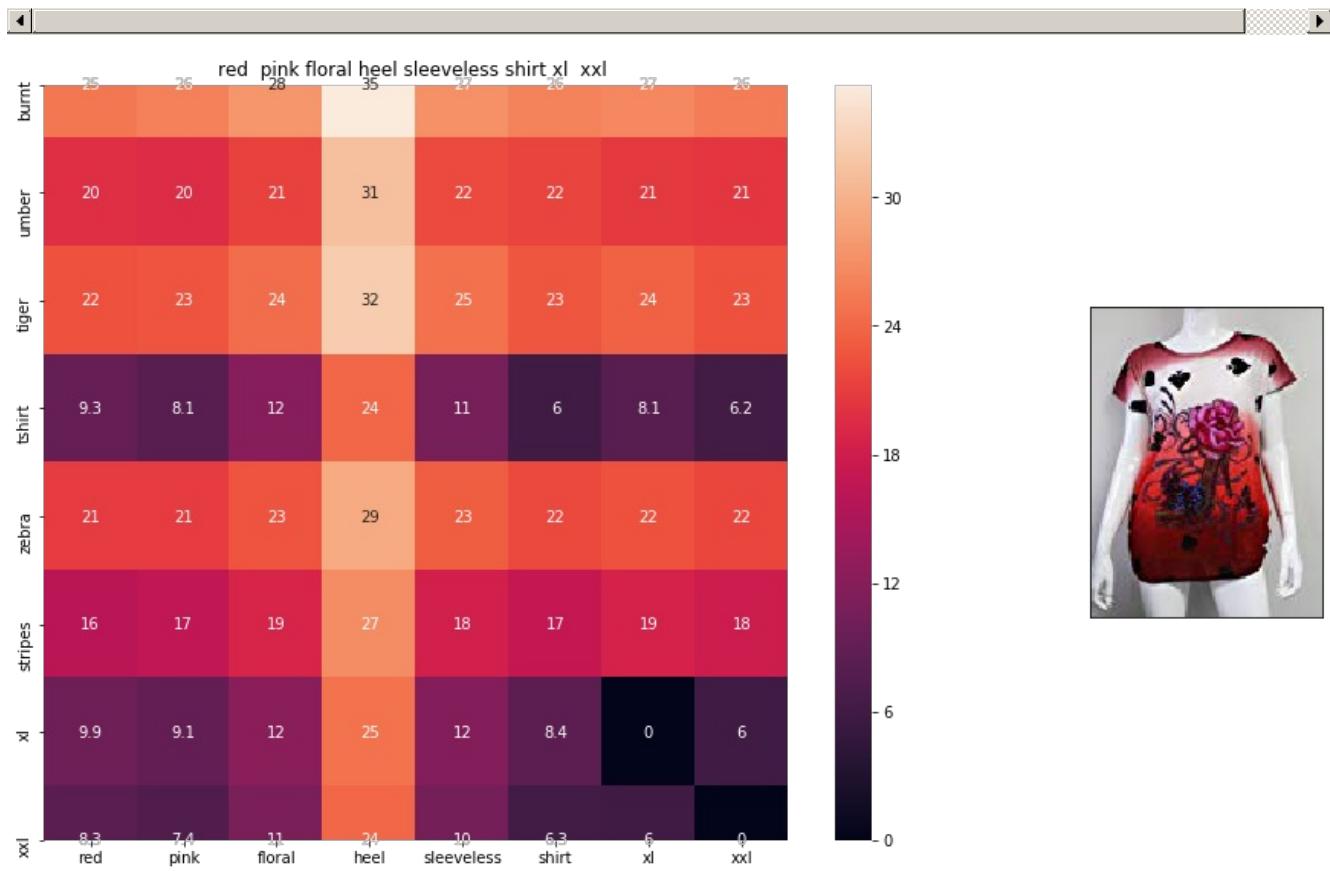
ASIN : B00JXQBBM1

Brand : Si Row

euclidean distance from input : 3.850655555875671

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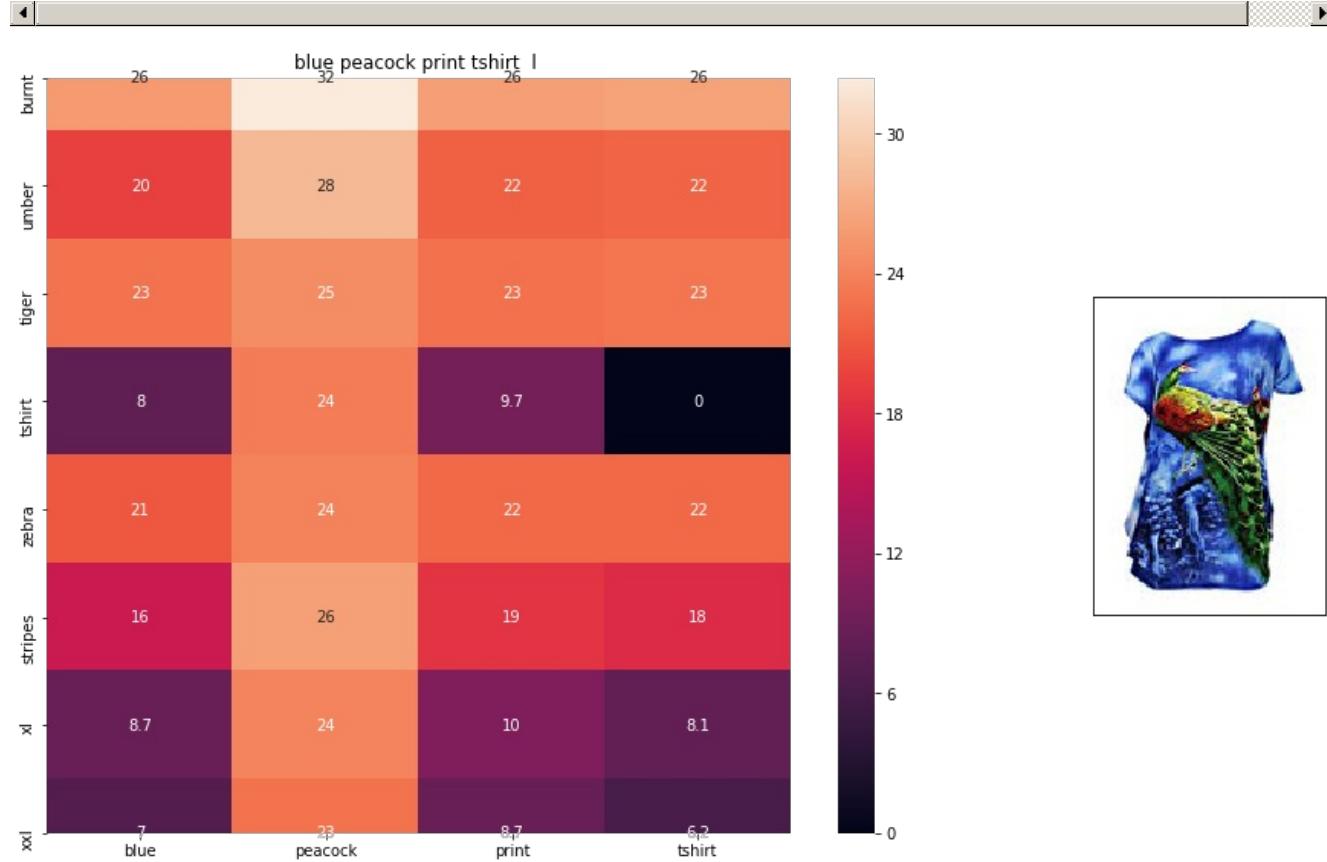


ASIN : B00JV63QOE

Brand : Si Row

euclidean distance from input : 3.856525262347432

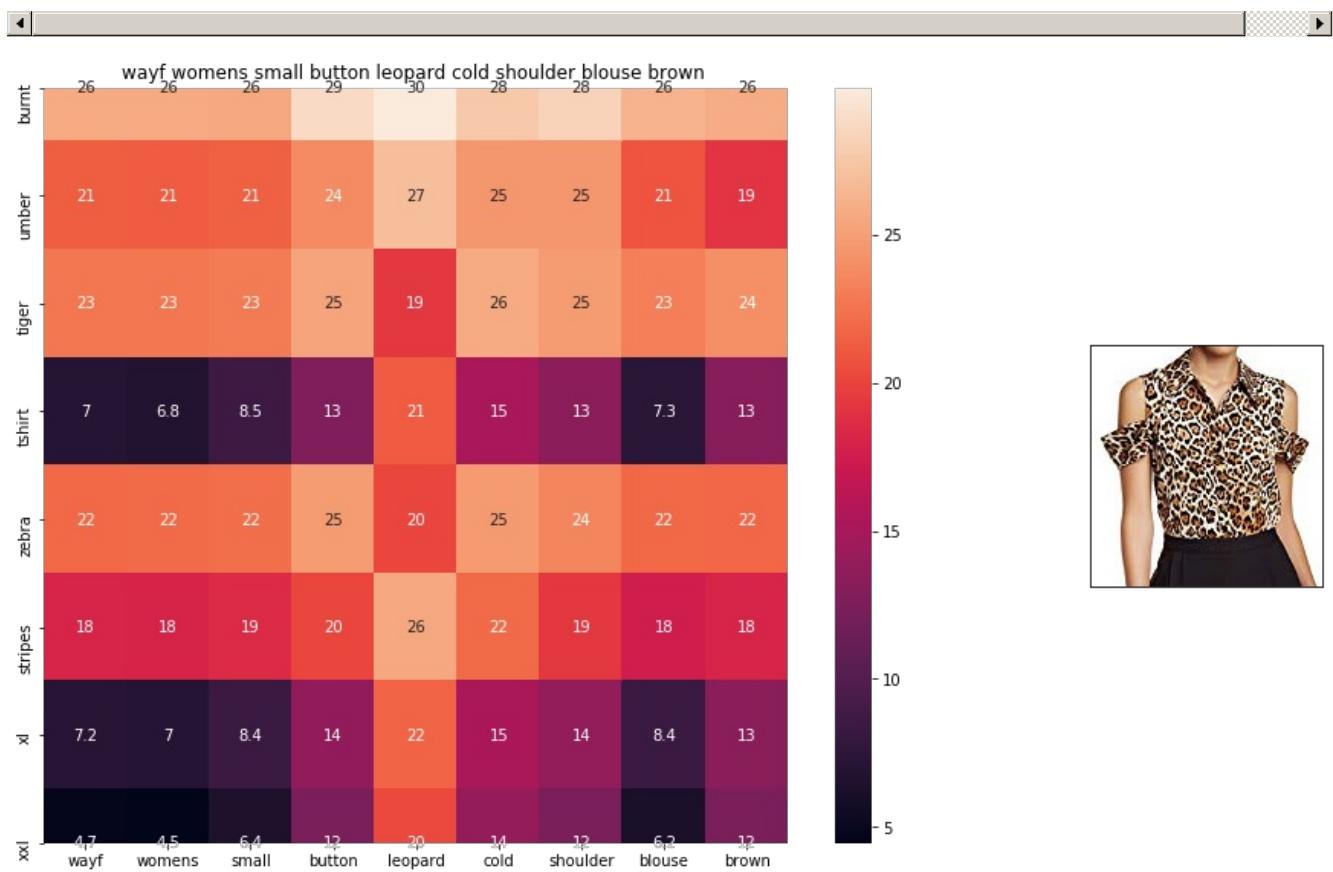
=====



Brand : Marrikas

euclidean distance from input : 3.863479657013155

=====



ASIN : B01M06V4X1

Brand : WAYF

euclidean distance from input : 3.8696435720115274

=====



Algorithm

1. compute the similarity distances between each of the features.
2. combine them based on some weights.
3. Now sort them.
4. Note that higher weight would be less importance here since we choose least distances.
5. Get the index of corresponding products and map them in data set indices and show those top n items.

Observations.

1. Recomendation were giving better results without combining any of the features.
2. Results were quite interesting in last 5 trials for different weights.
3. Note results were better by giving less weight to titles(W2V + image).

In []:

