Fashion Recommendation

Team Members:-

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Introduction

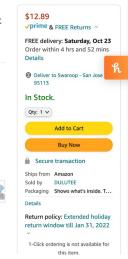
- Recommendation engines are the core of the businesses that are customer focused.
- They are responsible for bringing in a major chunk of revenue to the company.
- Amazon's 35% revenue is from product recommendations. So we believe by implementing Fashion Recommendation, we can learn more about the in-demand use case of recommendation engines.
- The recommendation engine we build in this project is content based recommendation as collaborative filtering needs item/user information which might be tough to find because of privacy concerns.

Introduction

Product







>

Recommendations

Products related to this item Page 2 of 305 Start over Sponsored @ YunJey short sleeve round neck triple colo block stripe T-shirt casual blouse.Red.XX-< SAMPEEL Women T-Amoretu Womens V Biucly Women's Summer Hount Tees for Women **Buvears Women Summer** YunJey short sleeve imesrun Womens V Neck Shirts Short Sleeve V **Neck Short Sleeve** Casual Sexy Short Sleeve Summer Short Sleeve T Shirts Short Sleeve **Tshirts Short Sleeve** round neck triple color Summer T Shirts Curved Crewneck Color Block Casual Tops Round Neck Rounded V Neck Pocket Loose Casual Summer **Neck Tunic Tops Loose** block stripe T-shirt Fitting Clothing Burgun... Hem Tee Tops(Yellow,S) Tee Loose Tops ... Loose T Shirts Coral S Tee Tops Pink X-Large casual blouse, Red, XX... Tops with Pokcet... ★★★☆ 15,611 ★★★☆ 15,356 **★★★☆☆ 7,510** ★★★★☆ 84 ★★★★☆ 113 **** 261 **★★★★** 42,464 \$21.99 \rime \$18.99 yprime \$22.99 yprime \$18.99 yprime \$19.99 yprime \$16.99 yprime \$18.98 yprime

System Design & Implementation Details

- We have built 2 types of recommendation engines:
 - Text based recommendation engines
 - BoW based
 - TF-IDF based
 - IDF based
 - Average Word2Vec (Text Semantics) based
 - IDF weighted Word2Vec (Text Semantics) based
 - Weighted similarity using brand and color
 - Visual features based
- We have built these 2 types of recommendation engines because in some cases text based engines might give better results (for books) and in some cases visual features based engines might give better results (for apparel).

- We have considered the basic techniques of text pre-processing for Text based recommendation engines with the exception of Word2Vec, we did this because our main goal was to build a robust recommendation engine and not complicate things by applying more advanced techniques.
- We have used libraries like sklearn and gensim.models for text preprocessing and Word2Vec.
- We have also used keras library to extract features from the product images.

Dataset

- The dataset consists of a total 183,138 number of records of different products. Each record is defined using 19 features, some of which are ASIN which stands for Amazon Standard Identification Number, the Brand of the product, the Type of the product, Availability of the product, Reviews given by users for the products, 3 different sizes of the product image and Manufacturer of the product and many more.
- Of these 19 features, we used only 7 features:
 - 1. ASIN (Amazon standard identification number)
 - 2. Brand (brand to which the product belongs)
 - 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/T-SHIRT)
 - 5. medium_image_url (URL of the image)
 - 6. title (title of the product.)
 - 7. formatted_price (the price of the product)

Data Preprocessing

• We have 2325 products that have the same title but different colors. These shirts are exactly the same except that they are different in size.



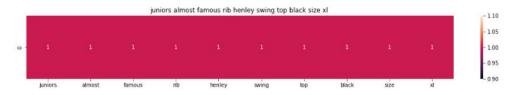


- These shirts are exactly the same except in color.
- We removed the duplicates from the dataset.

Algorithms

1. Bag of words on product titles

• We separated words from all the product titles and formed vectors of each title. We used these vectors to find similarity between other product title vectors and recommended the ones with the most similar titles.



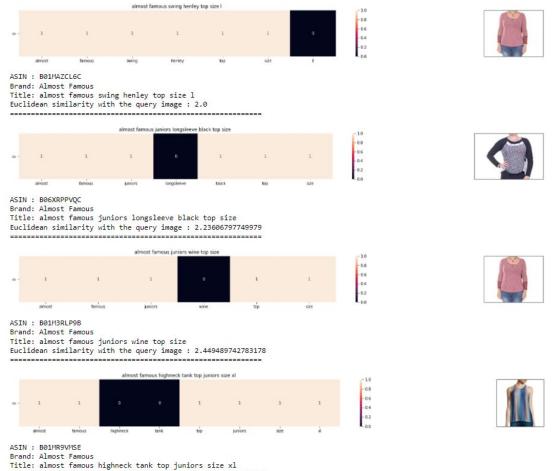


ASIN : B072M2W79L Brand: Almost Famous

Title: juniors almost famous rib henley swing top black size xl

Euclidean similarity with the query image : 0.0

Fig. Input Product for Bag of Words Model

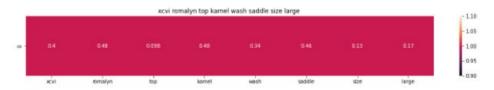


Title: almost famous highneck tank top juniors size xl Euclidean similarity with the query image : 2.449489742783178

Fig. Recommended Products

2. TF-IDF based product similarity

Term frequency times Inverse document frequency is used to scale down the impact of words that occur more frequently and do not contribute in distinguishing a title.





ASIN : B074NC9MJM BRAND : XCVI

Eucliden distance from the given image: 0.0

Fig. Input for TF-IDF Model



BRAND : XCVI

Eucliden distance from the given image: 1.22960337478935

Fig. Recommended Products

3. IDF based product similarity

Inverse document frequency is used to find similarity between the vectors to recommend products.



Fig. Input for IDF Model



euclidean distance from the given image: 19.69345982566793

Fig. Recommended Products

4. Average Word2Vec product similarity

- Used Word2Vec to generate a vector for the product titles.
- Used these vectors to recommend similar products.

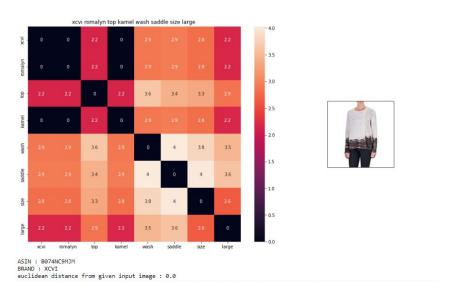


Fig. Input for Average Word2Vec Model



Fig. Recommended Products

5. IDF Weighted Word2Vec for product similarity

- Used Word2Vec to generate a vector for the product titles.
- While generating vector from Word2Vec multiply the vector provided by Word2Vec with the inverse document frequency of the title.
- Used these vectors to recommend similar products.

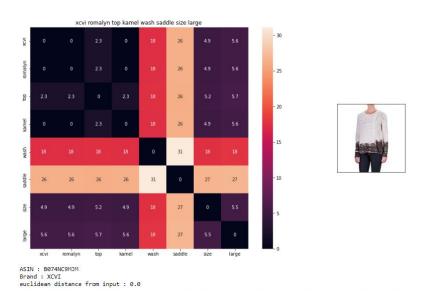


Fig. Input for IDF Weighted Word2Vec Model

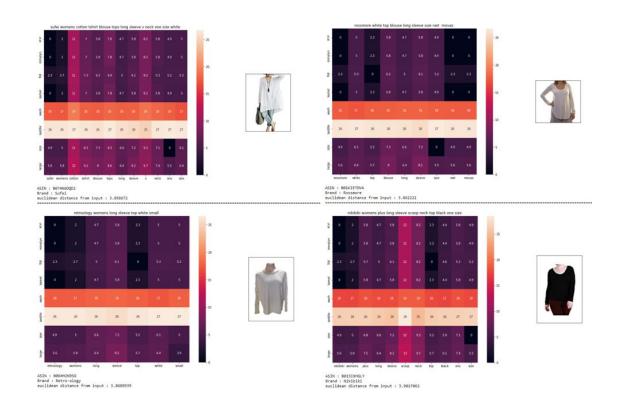
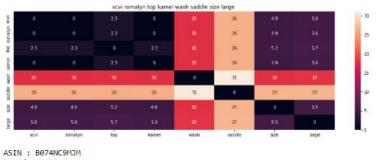


Fig. Recommended Products

6. Weighted similarity using brand and color

Made use of brand and color feature to recommend products.

Used weighted importance of brand and color while calculating similarity of two products based on the title.





Brand : XCVI euclidean distance from input : 0.0

Fig. Input for Weighted Similarity using brand and color



Fig. Recommended Products

Visual features based product similarity



Used images to recommend similar products.



Product Title: xcvi romalyn top kamel wash saddle size large Euclidean Distance from input image: 6.180516e-06 Amazon Url: www.amzon.com/dp/8074MC9MJM



Product Title: cable gauge large dark blush peasant top Euclidean Distance from input image: 35.34564 Amazon Url: www.amzon.com/dp/B074XPJWZT



Product Title: style co womens metallic scoopneck pullover top gray Euclidean Distance from input image: 36.92667 Amazon Url: www.amzon.com/do/B081M5K6P4E



Product Title: ideology raglan spaceddyed longsleeve top size xs Euclidean Distance from input image: 37.799213 Amazon Url: www.amzon.com/dp/80HFONJZ8

Analysis of Results

- We used A/B testing to find out which recommendation engine gave the best results.
- We took 10 test subjects (consisting of friends and family) and we divided them into 2 groups, namely, Group A and Group B.
- Group A was shown the results from the "Text based recommendation engines" and Group B was shown the results from the "Visual based recommendation engines".
- Since we can measure the "goodness" through user (test subjects feedback), we took the feedback from 2 groups.

Group A (shown the results from Text based recommendation engines)

Person	Feedback
Person 1	"Not so relevant results"
Person 2	"Decent, not so great"
Person 4	"I searched for shirt, recommended tank top"
Person 7	"I liked the recommendations"
Person 8	"Somewhat relevant"

Group B (shown the results from Visual based recommendation engines)

Person	Feedback
Person 3	"brilliant"
Person 5	"Relevant to my product"
Person 6	"This is quite relevant and helped me to understand my options"
Person 9	"Its okay"
Person 10	"I like the fact that my recommendations were relevant"

Conclusion

- Conclusion was that the Visual based recommendation engines performed better than the Text based recommendation engines and it makes sense because in case of apparel the "Visual" part is the core of recommendations as its the same way a human would recommend an apparel.
- In case of books, we feel **Text based recommendation engines** will perform better because there is no "Visual" part required to recommend books, book titles, genres would be the core of book recommendation as its the same way a human would recommend a book.
- Our engines are built on simple yet very effective techniques which give very good recommendations.

Decisions made/Things that worked:

 We decided to make multiple recommendation engines so that our outputs (recommendations) are more robust and this helped us a lot.

Difficulties faced/Things that didn't work well

- We tried stemming and it gave us really bad recommendations, out test subjects did not find the recommendations relevant.
- The data for the recommendation was taken from an archived page of amazon since the Amazon API was giving technical issues.

Link to Dataset and Code

Dataset Link

https://drive.google.com/file/d/1w3j8M4Oh7Ogt3sPNUmSwlmG4umYkXSxo/view?usp=sharing

Code Link

https://github.com/ankit-dhokariya/CMPE-256-Recommender-System-Project