



Leveraging Deep Learning for Product Similarity Analysis

A case study for PhD job position assessment
at Technical University of Denmark

Presented by:
Nastaran Moradzadeh



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Introduction



Recommender systems

A software tool
Suggesting personalized recommendations based on user behavior or input



Our Problem

Outfit recommender system
Traditional systems: based on user history (content-based) and users similarities (collaborative-based)
Here: based on image uploaded by user



Data Resources



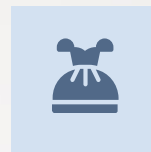
Product Images

Includes 295 images



Product Labels

Includes 9 categorical columns



User Input Image

New clothing photo uploaded by the user



id	gender	masterCategory	subCategory	articleType	baseColour	season	year	usage	productDisplayName
1185	Men	Apparel	Topwear	Tshirts	Blue	Summer	2011.0	Sports	Nike Sahara Team India Farneser Round Neck Jersey
1186	Men	Apparel	Topwear	Tshirts	Blue	Winter	2010.0	Sports	Nike Men Blue 720 Indian Cricket Jersey
1185	Men	Apparel	Topwear	Tshirts	Blue	Summer	2013.0	Sports	Nike Men Team India Cricket Jersey
1026	Unisex	Accessories	Bags	Backpacks	Navy Blue	Fall	2010.0	Casual	Puma Deck Navy Blue Backpack
1026	Unisex	Accessories	Bags	Backpacks	Black	Fall	2010.0	Sports	Puma Big Cat Backpack Black
1028	Men	Apparel	Topwear	Jackets	Black	Fall	2010.0	Sports	Puma Men Ferrari Black Fleece Jacket
1029	Men	Apparel	Topwear	Tshirts	Red	Fall	2010.0	Casual	Ferrari Tee
1030	Men	Apparel	Topwear	Jackets	Red	Fall	2010.0	Sports	Puma Men Ferrari Track Jacket
1031	Men	Apparel	Topwear	Tshirts	Grey	Fall	2010.0	Casual	Puma Men Grey Solid Round Neck T-Shirt
1032	Men	Apparel	Topwear	Tshirts	Grey	Fall	2010.0	Casual	Puma Men Grey Leaping Cat T-shirt
1033	Men	Apparel	Topwear	Tshirts	Red	Fall	2010.0	Casual	Puma Men Cat Red T-shirt
1034	Men	Apparel	Topwear	Tshirts	Black	Fall	2010.0	Casual	Puma Men Black Leaping Cat T-shirt
1035	Unisex	Accessories	Headwear	Caps	Black	Fall	2010.0	Sports	Puma Unisex Logo Cap
1036	Men	Apparel	Topwear	Tshirts	Black	Fall	2010.0	Sports	Puma Men Black Net Jersey
1037	Men	Apparel	Topwear	Tshirts	Red	Fall	2010.0	Sports	Puma Men Red Net Jersey
1038	Men	Apparel	Topwear	Tshirts	Blue	Fall	2010.0	Casual	Puma Men Ferrari Black T-shirt
1039	Men	Apparel	Topwear	Tshirts	Grey	Fall	2010.0	Casual	Puma Men Ferrari Grey T-shirt
1040	Men	Apparel	Topwear	Tshirts	Blue	Fall	2010.0	Casual	Puma Men Ferrari Vintage Black Polo T-shirt
1041	Men	Footwear	Shoes	Sports Shoes	White	Fall	2010.0	Sports	Puma Men's Ballistic Spike White Green Shoe
1042	Men	Footwear	Shoes	Sports Shoes	White	Fall	2010.0	Sports	Puma Men's Ballistic Rubber Shoe
1043	Men	Footwear	Shoes	Casual Shoes	Black	Fall	2010.0	Casual	Puma Men Basket-62 Sneaker
1044	Men	Footwear	Shoes	Casual Shoes	White	Fall	2010.0	Casual	Puma Men's Basket Bump Sneaker
1045	Men	Footwear	Shoes	Casual Shoes	White	Fall	2010.0	Casual	Puma Men's Speed Cat Shoe
1046	Men	Footwear	Shoes	Casual Shoes	White	Fall	2010.0	Casual	Puma Men's Future White Shoe
1047	Men	Footwear	Shoes	Casual Shoes	Black	Fall	2010.0	Casual	Puma Men's Tunnel Sneaker

gender

Men 201
Unisex 51
Women 43

masterCategory

Apparel 182
Footwear 61
Accessories 45
Sporting Goods 7

subCategory

Topwear 154
Shoes 59
Bags 34
Bottomwear 28
Sports Equipment 7
Headwear 5
Water Bottle 5
Sandal 2
Accessories 1

articleType

Tshirts 139
Sports Shoes 43
Backpacks 28
Casual Shoes 16
Shorts 13
Track Pants 11
Jackets 10
Footballs 6
Water Bottle 6

Caps 5
Handbags 5
Swimwear 4
Sweatshirts 4
Sandals 2
Basketballs 1
Duffel Bag 1
Tops 1

baseColour

Black 64
White 63
Blue 49
Grey 29
Red 26
Green 15
Navy Blue 11
Brown 10

Pink 7
Purple 6
Yellow 6
Orange 3
Silver 2
Beige 2
Maroon 1
Cream 1

season

Fall 187
Summer 69
Winter 31
Spring 8

year

2010.0 216
2011.0 70
2012.0 5
2015.0 2
2013.0 1
2016.0 1

usage

Sports 170
Casual 117
Travel 8

Data Analysis

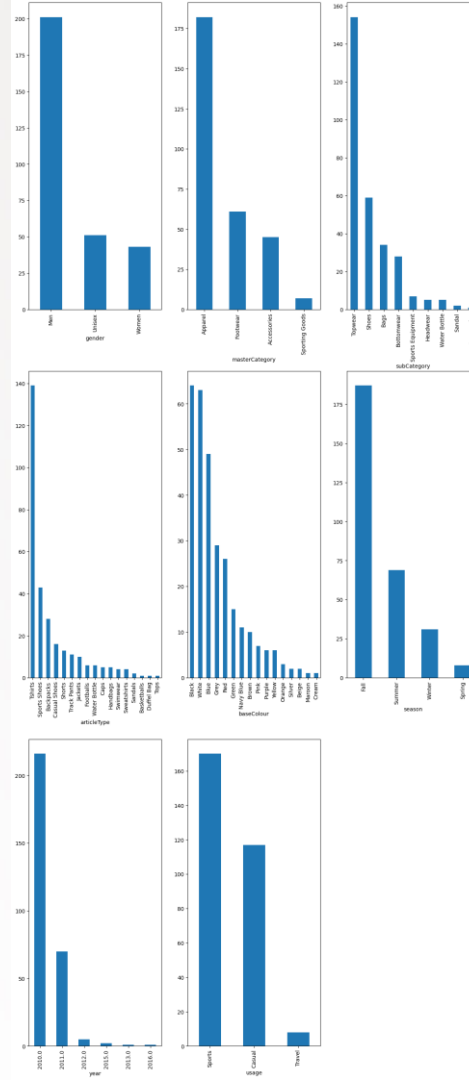
Histograms

Labels histograms are drawn
Data is not balanced across almost all
labels

Clustering

Product images are clustered
Based on various feature embedding
extractors

This method is used to evaluate feature
embeddings



Data Processing (Product images)

Step 1

Resize input
to 256*256

Step 3

Conversion to tensor

Step 2

Crop center of resized
image in 224 * 224

Step 4

Normalization



Data Processing (User Input Image)

Step 0

Preprocess such as:
clothes detection
alignment
background removal

```
53 # Define data preprocessing
54 preprocess = transforms.Compose(
55     [
56         transforms.Resize(256),
57         transforms.CenterCrop(224),
58         transforms.ToTensor(),
59         transforms.Normalize(mean=[0.485, 0.456, 0.406],
60                               std=[0.229, 0.224, 0.225]),
61     ]
62 )
```

Step 1-4

Same as product images
processing



Proposed Solution

Step 1

Extract feature embeddings from product images using deep learning.

Save the extracted features for subsequent utilization.

Step 2

A web application is implemented to facilitate the upload of new images and display similar products.

Step 3

The uploaded image is processed, and its feature embedding is extracted.

Using a similarity metric, similar products are retrieved.

Step 4

To enhance the results, the labels of products are utilized to identify less probable items.

Label voting is conducted among selected items, those with a label different from the computed label are marked accordingly.

Results are presented to the user

Main Challenges

1

User input photo preprocessing such as clothes detection, alignment and background removal

2

Feature extractor model -> dedicated model in outfit area will have better performance

Simplifying assumptions!

The system input photos are well-formed, featuring a single outfit and devoid of complex backgrounds.

Additionally, pretrained neural networks trained on the ImageNet dataset are employed as feature extractors.

Deep Learning Models

ResNet50 & ResNet18

Both resnet50 and resnet18 models performed well as feature extractors

Evaluation method

Subjective tests based on personal opinion.
Clustering results

VGG16

VGG16 model did not perform well as a feature extractor *

so is not utilized

Final Model

To further improve the results both ResNet50 and ResNet18 feature vectors are concatenated *

* In the notebook file, the results are presented.



Implementation Details

Utilized Tools

PyTorch: Deep learning models

Dash – Plotly: Web app implementation

Docker: Containerization for deployment and scalability

Uploaded files

Are saved!

Feature Embeddings

Have been saved!

To prevent re-computation upon each restart.

Sample input images

Are prepared and uploaded in the github

Similarity metric

Cosine similarity

Further results enhancement

The 'articleType' label is utilized.

Overview of the Web App

Outfit Recommender System

Recode Case Study

Drag and Drop or [Select Files](#)

Number Of Similar Items:

FIND

The most similar Items

Results




Outfit Recommender System

Recode Case Study

Drag and Drop or [Select Files](#)




0a7e5fe0-d592-40e6-b9b8-75aac9a2d685.jpg
2021-01-17T13:06:08



Number Of Similar Items:
3

FIND

The most similar Items


Items	Scores	description
	0.7608	
	0.7574	
	0.7383	

Outfit Recommender System

Recode Case Study

Drag and Drop or [Select Files](#)











2a9c05ae-88ae-42a7-9bd7-761c55805e67.jpg
2021-01-17T13:06:08



Number Of Similar Items:
10

FIND

The most similar Items

Items	Scores	description
	0.808	
	0.7916	
	0.7865	
	0.7862	not predicted
	0.785	
	0.7848	
	0.7847	
	0.7479	
	0.7386	not predicted
	0.725	not predicted

Conclusion & more Ideas

Conclusion

If the input is sufficiently well-formed and similar products are present in the dataset, the final recommended items are desirable.

More ideas

- 1- Capturing user's interested labels and categories can enhance performance.
- 2- Recommending items from related categories can be beneficial. For instance, if a user uploads a blue T-shirt, recommending a blue cap or navy jeans can complement their outfit.
- 3- Suggest complementary accessories, such as belts, bags, or sunglasses, based on the uploaded item.
- 4- Offer variations of the uploaded item, such as different colors, patterns, or styles, to provide options for the user.
- 5- Include options for user feedback or customization preferences to further refine recommendations and improve user experience.

RESOURCES

[1] <https://pytorch.org/vision/0.8/models.html>

[2] <https://github.com/alexeygrigorev/clothing-dataset>





THANKS!

any questions?