Market Basket Analysis & Recommendation System

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Overview

The H&M Group is a brand and business family with 53 online markets and approximately 4,850 stores. H&M's online store provides customers with a wide range of products to choose from.

Product recommendations are essential for improving the shopping experience. More importantly, assisting customers in making the best decisions has a positive impact on sustainability because it reduces returns and thus reduces transportation emissions.

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DATA ARTICLES

	0	1
article_id	108775015	108775044
product_code	108775	108775
prod_name	Strap top	Strap top
product_type_no	253	253
product_type_name	Vest top	Vest top
product_group_name	Garment Upper body	Garment Upper body
graphical_appearance_no	1010016	1010016
graphical_appearance_name	Solid	Solid
colour_group_code	9	10
colour_group_name	Black	White
perceived_colour_value_id	4	3
perceived_colour_value_name	Dark	Light
perceived_colour_master_id	5	9
perceived_colour_master_name	Black	White
department_no	1676	1676
department_name	Jersey Basic	Jersey Basic
eboo_xebni	Α	Α
index_name	Ladieswear	Ladieswear
index_group_no	1	1
index_group_name	Ladieswear	Ladieswear
section_no	16	16
section_name	Womens Everyday Basics	Womens Everyday Basics
garment_group_no	1002	1002
garment_group_name	Jersey Basic	Jersey Basic
detail_desc	Jersey top with narrow shoulder straps.	Jersey top with narrow shoulder straps.

CUSTOMERS

	customer_ld	FN	Active	club_member_status	fashlon_news_frequency	age	postal_code
0	00000dbacae5abe5e23885899a1fa44253a17956c6d1c3	NaN	NaN	ACTIVE	NONE	49.0	52043ee2162cf5aa7ee79974281641c6f11a68d276429a
1	0000423b00ade91418cceaf3b26c6af3dd342b51fd051e	NaN	NaN	ACTIVE	NONE	25.0	2973abc54daa8a5f8ccfe9362140c63247c5eee03f1d93
2	000058a12d5b43e67d225668fa1f8d618c13dc232df0ca	NaN	NaN	ACTIVE	NONE	24.0	64f17e6a330a85798e4998f62d0930d14db8db1c054af6
3	00005ca1c9ed5f5146b52ac8639a40ca9d57aeff4d1bd2	NaN	NaN	ACTIVE	NONE	54.0	5d36574f52495e81f019b680c843c443bd343d5ca5b1c2

TRANSACTION_TRAIN

	t_dat	customer_ld	article_id	price	sales_channel_ld
0	2018-09-20	000058a12d5b43e67d225668fa1f8d618c13dc232df0ca	663713001	0.050831	2
1	2018-09-20	000058a12d5b43e67d225668fa1f8d618c13dc232df0ca	541518023	0.030492	2
2	2018-09-20	00007d2de826758b65a93dd24ce629ed66842531df6699	505221004	0.015237	2
3	2018-09-20	00007d2de826758b65a93dd24ce629ed66842531df6699	685687003	0.016932	2

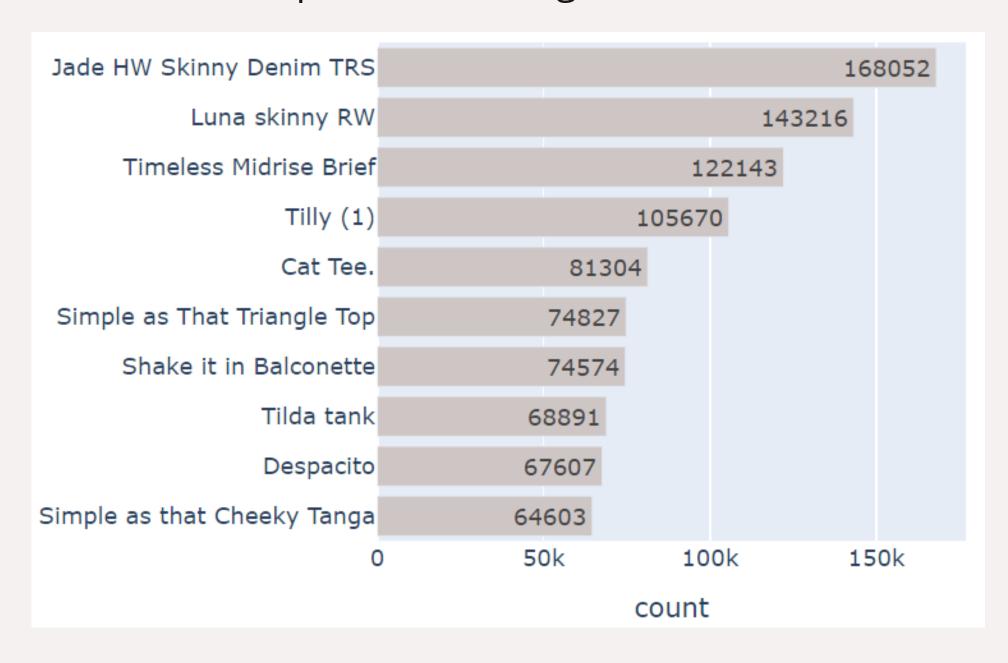
ARTICLES = 1054 ROWS

CUSTOMERS = 1371980 ROWS

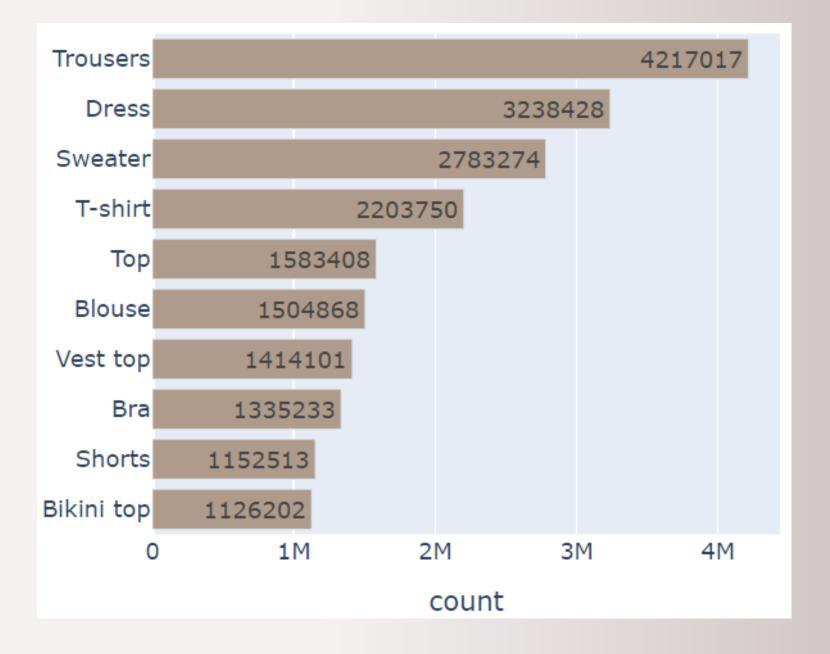
TRANSACTION_TRAIN = 31788324 ROWS

EXPLORATORY DATA ANALYSIS (EDA)

Top 10 Bestselling Products

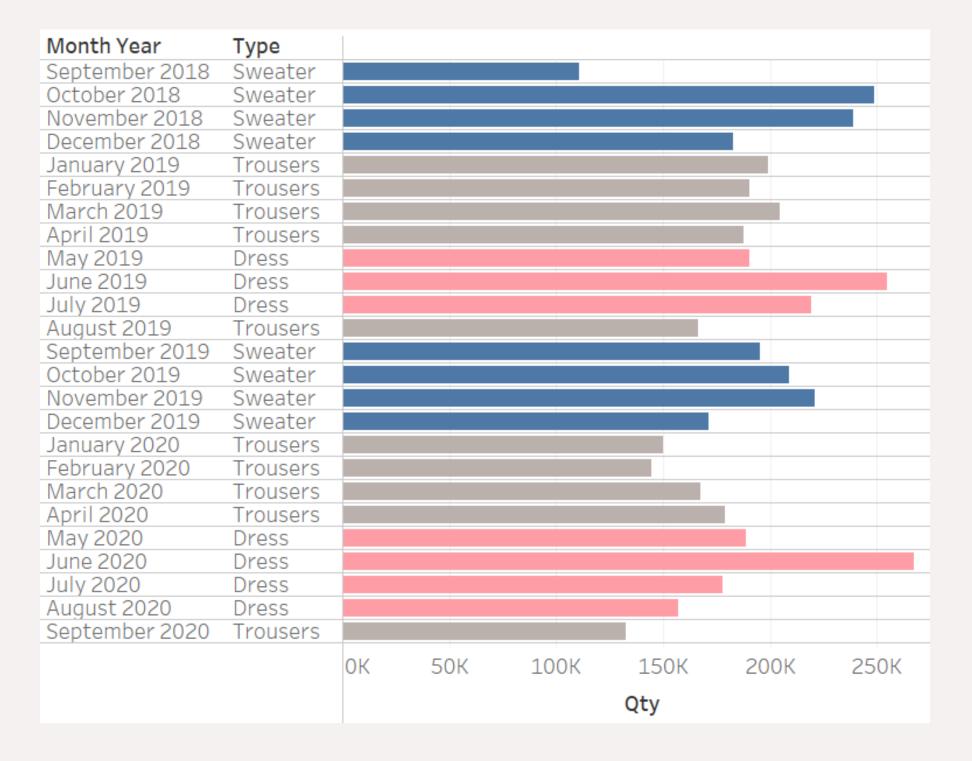


Top 10 Bestselling Product Types



EXPLORATORY DATA ANALYSIS (EDA)

Bestselling Product Types Each Month

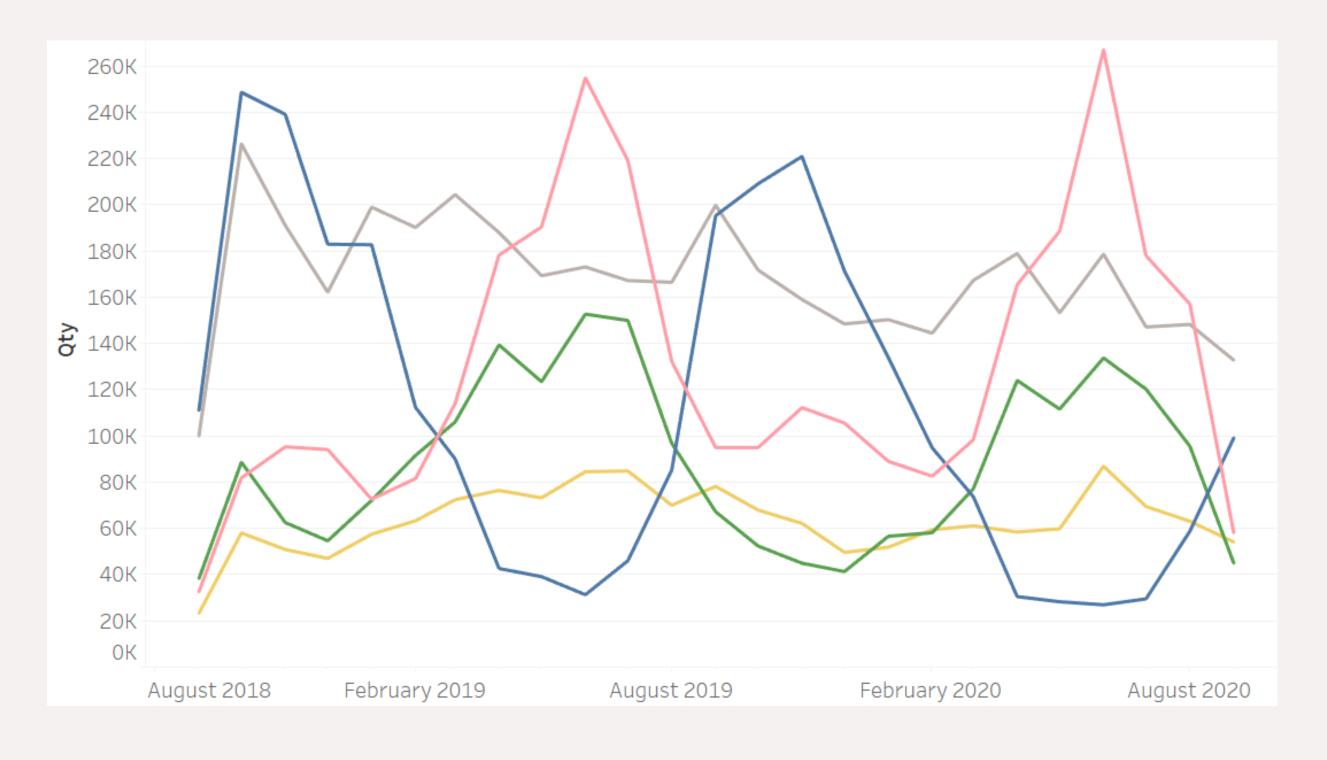


Observations

Sweater, trousers, or dress are the most popular products each month from 2018 to 2020.

EXPLORATORY DATA ANALYSIS (EDA)

Top 5 Best Selling Product Types from August 2018 - August 2020



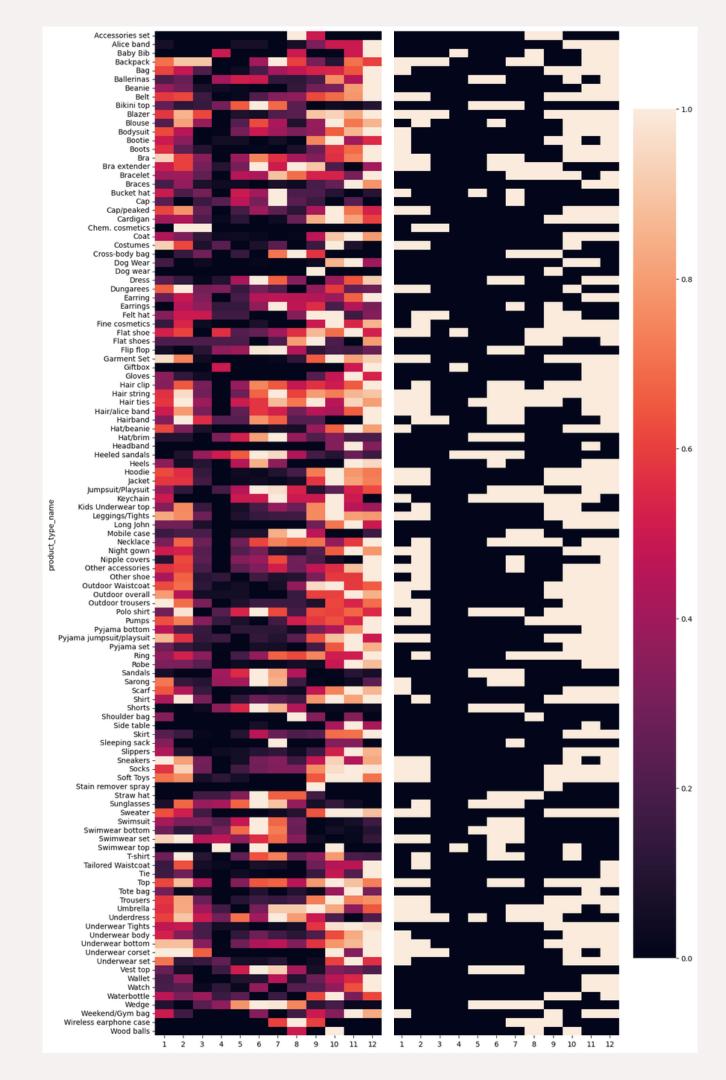
Observations

The top five products of all time have different sales patterns.

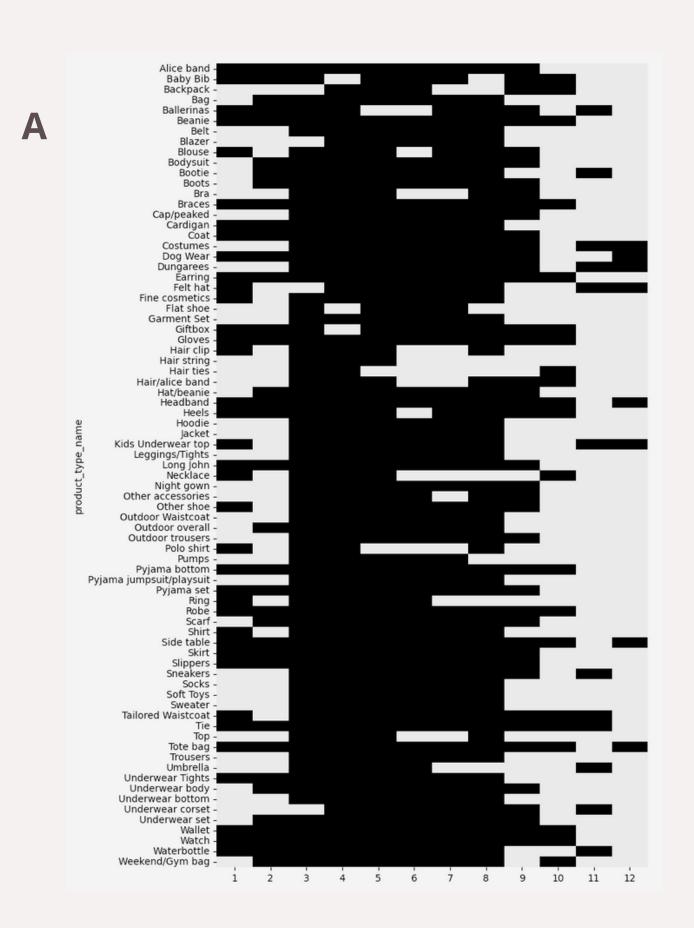
SALES PATTERN OF PRODUCT TYPES FROM JANUARY TO DECEMBER

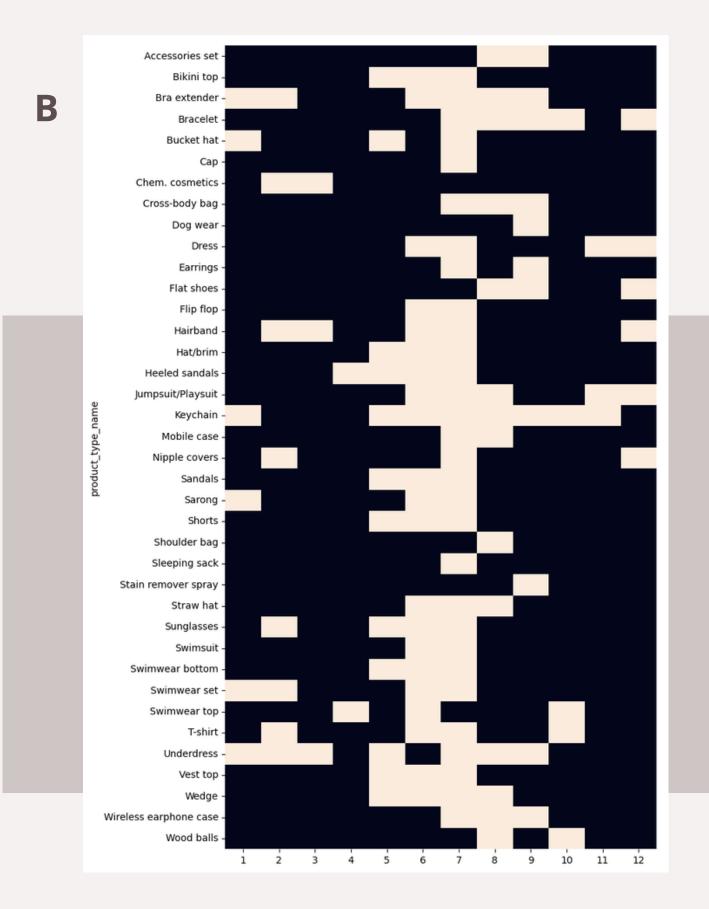
Additional information:

H&M Group is a global fashion brand with 53 online markets. The postal codes are anonymized. Since the postal codes might be from different countries or climate zones, the sales pattern of the same type of product may differ.



GROUPING TYPE OF PRODUCTS





Observations

- 1. Products in group A mostly sold well in September February and products in group B mostly have high sales in around May August
- 2. Customers bought more products in autumn-winter than in spring-summer.
- 3. There are type of products that sold well nearly all the time (e.g. hair ties, hair string, and keychain).

Recommendation

To recommend products to customers, **consider new products** and when certain type of products will be in **high demand**. It is critical to generate recommendations **using the most appropriate data.**

Market basket analysis (MBA) is a data mining techinique. The primary goal of MBA is to provide information to retailers in order to help them better understand their customers' purchasing habits, allowing them to make better-informed decisions.

SUPPORT

Probability a product appears in transaction data

CONFIDENCE

Probability customers purchase product B, given they have purchased product A.

Support(A&B) / Support(B)

LIFT

Support(A&B)/
Support(B)*Support(A)

	FP Growth	Apriori
Speed	Faster	Slower
Candidates	X	V
Memory	Small	Large

- 1. PURCHASED TOGETHER PRODUCTS
- 2. PRODUCT PLACEMENT
- 3. PRODUCT BUNDLES
- 4. OFFER SPECIAL DEALS
- 5. RECOMMENDATION
 PRODUCT BOUGHT
 TOGETHER

MARCH-MAY 2020 DATA

	Involce_no	prod_name
0	106601135	[Dixie]
1	106601171	[Shirtdress Lou]
2	106601220	[Kendrick Como, Kendrick tie thong]
3	106601222	[Mona, Bob v-neck 2-pack, Pingu spring, Twenty
4	106601267	[TEA HOOD S.1]
•••		***
11923	107560416	[Eleonor button dress, Olivia long linen shirt]
11924	107560424	[Sadie Shirt]
11925	107560631	[Lazer Razer High Rise Hipster, Lazer Razer Pa
11926	107560915	[Olivia woven Shorts, Dessie HW Belted Woven S
11927	107560962	[Sporty Spice PRICE Swimsuit]
11928 r	ows × 2 col	umns

MAY 2020 DATA

	Involce_no	prod_name				
0	107198512	[V-neck Strap Top.]				
1	107198783	[Charlotte bralette laguna opt2, Coachella Pus				
2	107198842	[TILDA HIP BELT, Alicia wrap skirt]				
3	107198952	[PEACH STRAP]				
4	107199508	[Flipflop solid]				
	•••	•••				
5417	107560416	[Eleonor button dress, Olivia long linen shirt]				
5418	107560424	[Sadie Shirt]				
5419	107560631	[Lazer Razer High Rise Hipster, Lazer Razer Pa				
5420	107560915	[Olivia woven Shorts, Dessie HW Belted Woven S				
5421	107560962	[Sporty Spice PRICE Swimsuit]				
5422 r	5422 rows × 2 columns					

MARCH-MAY 2020 DATA PREDICT JUNE 2020

RULES= 215

	antecedent	consequent	confidence	lift	support
0	{Baby Love Cheeky Brief}	(Baby Love Triangle Top)	0.727273	1084.363636	0.000671
1	(Baby Love Triangle Top)	{Baby Love Cheeky Brief}	1.000000	1084.363636	0.000671
2	(Ginger Top)	(Ginger Highwaist)	0.545455	813.272727	0.000503
3	(Ginger Highwaist)	{Ginger Top}	0.750000	813.272727	0.000503
4	(Ursula Cheeky V Brief)	(Ursula Top)	0.571429	757.333333	0.000671
	•••				

ESTIMATED ADDITIONAL REVENUE

16.56

REVENUE

>654.2

ESTIMATED INCREASE

>2.53%

PREDICTION

	Involce_no	prod_name_x	prediction	llft	prod_name_y	price	total_price	price_est
8376	107993163	[Susan tank dress, Eleonor button dress]	0	0.000000	0	0.000000	0.053356	0.00000
7196	107931720	[Shake it in Balconette]	Swish HR wide side Brief	32.237838	Swish HR wide side Brief	0.006763	0.025407	0.00218
3316	107715147	[PIXI PINNAFORE, Price TEE TVP]	Money tee TVP	28.673077	Money tee TVP	0.004220	0.014373	0.00121
8278	107987321	[Gemini skirt.]	0	0.000000	0	0.000000	0.020322	0.00000
3789	107736191	[Mademoiselle dress, Chiba Dress, Dragonfly dr	0	0.000000	0	0.000000	0.067339	0.00000

MAY 2020 DATA PREDICT JUNE 2020

RULES= 487

	antecedent	consequent	confidence	llft	support
0	{Tropicana Brazilian CON}	(Tropicana Top)	1.000000	1355.500000	0.000553
1	{Tropicana Top}	(Tropicana Brazilian CON)	0.750000	1355.500000	0.000553
2	{Tropic Like Its Hot Brief}	(Tropic Like Its Hot Top)	1.000000	1084.400000	0.000553
3	{Tropic Like Its Hot Top}	(Tropic Like Its Hot Brief)	0.600000	1084.400000	0.000553
4	(Bondi Rescue Tie Brief)	(Bondi Rescue Bandeau)	0.600000	1084.400000	0.000553
		•			

ESTIMATED ADDITIONAL REVENUE

24.777

REVENUE

>654.2

ESTIMATED INCREASE

>3.787%

PREDICTION

	Invoice_no	prod_name_x	prediction	lift	prod_name_y	price	total_price	price_est
5198	107815294	[Desert Top, Class Omar ring pk, Simple as tha	Timeless Cheeky Brief	37.136986	Timeless Cheeky Brief	0.004051	0.033508	0.001504
4177	107757168	[Liza coral padded wire]	0	0.000000	0	0.000000	0.016932	0.000000
6177	107869665	[SF Michelle dress, Mia]	0	0.000000	0	0.000000	0.067763	0.000000
3100	107706784	[New Girl Cheeky Tanga, Timeless Midrise Brief	Superwoman Superpush	28.574440	Superwoman Superpush	0.010661	0.111780	0.003046
1223	107624181	[ED Jafar, ED House]	0	0.000000	0	0.000000	0.037424	0.000000

DATA USED TO GENERATE RECOMMENDATIONS IN JUNE 2020

MARCH-MAY 2020	MAY 2020
LARGE	SMALL
SLOW	FAST
LESS	MORE
	LARGE SLOW

RECOMMENDATION:

Use previous month's data to generate rules for cross-selling or recommending products to customers for the next month.

Product Recommendation

CONTENT-BASED RECOMMENDATION

NEW ITEMS

POPULAR ITEMS

COLLABORATIVE FILTERING

Item-based and user-based recommendation

HYBRID FILTERING

REAL-WORLD
RECOMMENDATION
SYSTEM

1. THE RETRIEVAL
STAGE

2. THE RANKING STAGE

A Simple Recommendation System

- REPURCHASEDITEMS
 - PREVIOUSLY PURCHASED ITEMS

2 USER-ITEM
RECOMMENDATION
(TENSORFLOW
RECOMMENDER)

LAST WEEK'S POPULAR
ITEMS

Evaluation Metric

MEAN AVERAGE PRECISION @ 12 (MAP@12)

$$MAP@12 = \frac{1}{U} \sum_{u=1}^{U} \sum_{k=1}^{\min(n,12)} P(k) \times rel(k)$$

P(K) = Precision at K

rel(K) = Rel at K (Indicator if the item at rank K is relevant)

Result

	customer_ld	prediction
0	00000dbacae5abe5e23885899a1fa44253a17956c6d1c3	0568601043 0858856005 0568601023 0927225001 09
1	0000423b00ade91418cceaf3b26c6af3dd342b51fd051e	0797988005 0832331003 0878190005 0809238001 07
2	000058a12d5b43e67d225668fa1f8d618c13dc232df0ca	0794321007 0805000008 0805000007 0794321011 07
3	00005ca1c9ed5f5146b52ac8639a40ca9d57aeff4d1bd2	0919365008 0677930066 0864216003 0736923009 09
4	00006413d8573cd20ed7128e53b7b13819fe5cfc2d801f	0791587015 0896152002 0730683050 0896152001 09
•••		
1371975	ffffbbf78b6eaac697a8a5dfbfd2bfa8113ee5b403e474	0557599022 0713997002 0720125039 0740922009 07
1371976	ffffcd5046a6143d29a04fb8c424ce494a76e5cdf4fab5	0818754002 0762846031 0762846026 0889460003 09
1371977	ffffcf35913a0bee60e8741cb2b4e78b8a98ee5ff2e6a1	0762846027 0689365050 0794819001 0884081001 07
1371978	ffffd7744cebcf3aca44ae7049d2a94b87074c3d4ffe38	0714790020 0906612001 0928088001 0714790028 09
1371979	ffffd9ac14e89946416d80e791d064701994755c3ab686	0685813018 0685813043 0685813042 0784926002 08
1371980 ı	rows × 2 columns	

MAP@12: 0.0225

Recommendations

- In some months from March August, the sales of products in group B didn't reach 50% of the highest sales of the products.
- To recommend products to customers, consider new products and when certain type of products will be in high demand. It is critical to generate recommendations using the most appropriate data.
- Offer special deals, product bundles based on the rules generated from the market basket analysis using previous month's data.
- Personalized product recommendations using hybrid recommendation system

References

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- 2. www.techtarget.com/searchcustomerexperience/definition/market-basket-analysis (What is market basket analysis? Definition from Whatls.com)
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- 5. www.tensorflow.org/recommenders/examples/basic_retrieval (Recommending movies: retrieval | TensorFlow Recommenders)