



# MARKET BASKET ANALYSIS AND 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.

# 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
index_code	A	A
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_Id	FN	Active	club_member_status	fashion_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_Id	article_Id	price	sales_channel_Id
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

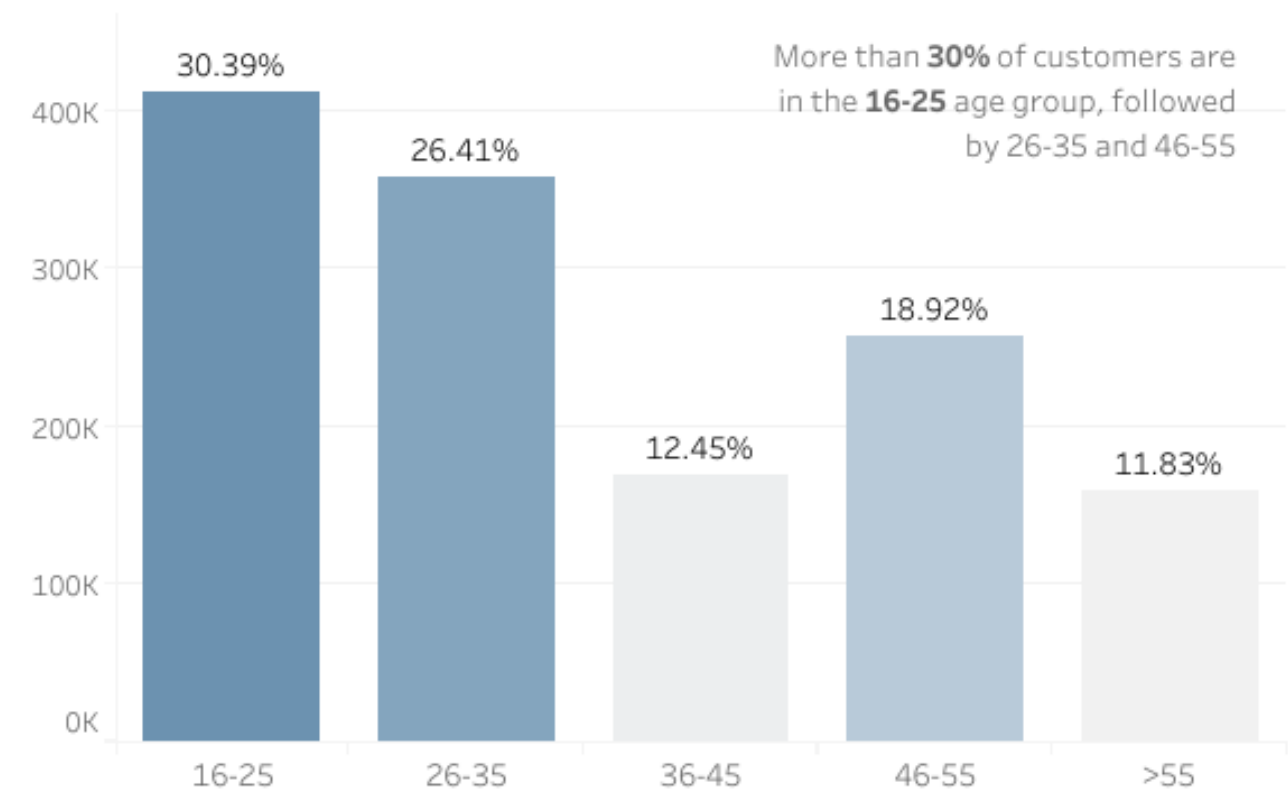
ARTICLES = 1054 ROWS

CUSTOMERS = 1371980 ROWS

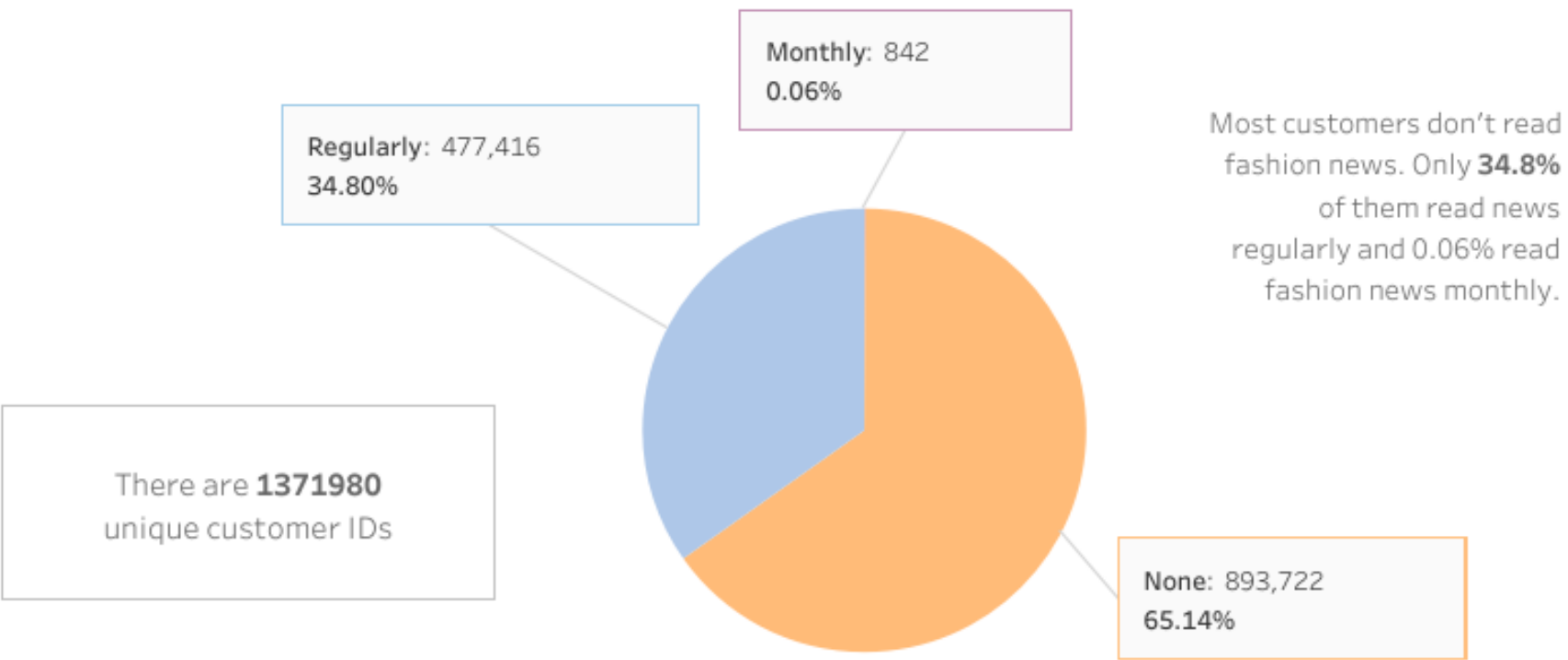
TRANSACTION\_TRAIN = 31788324 ROWS

# EXPLORATORY DATA ANALYSIS (EDA)

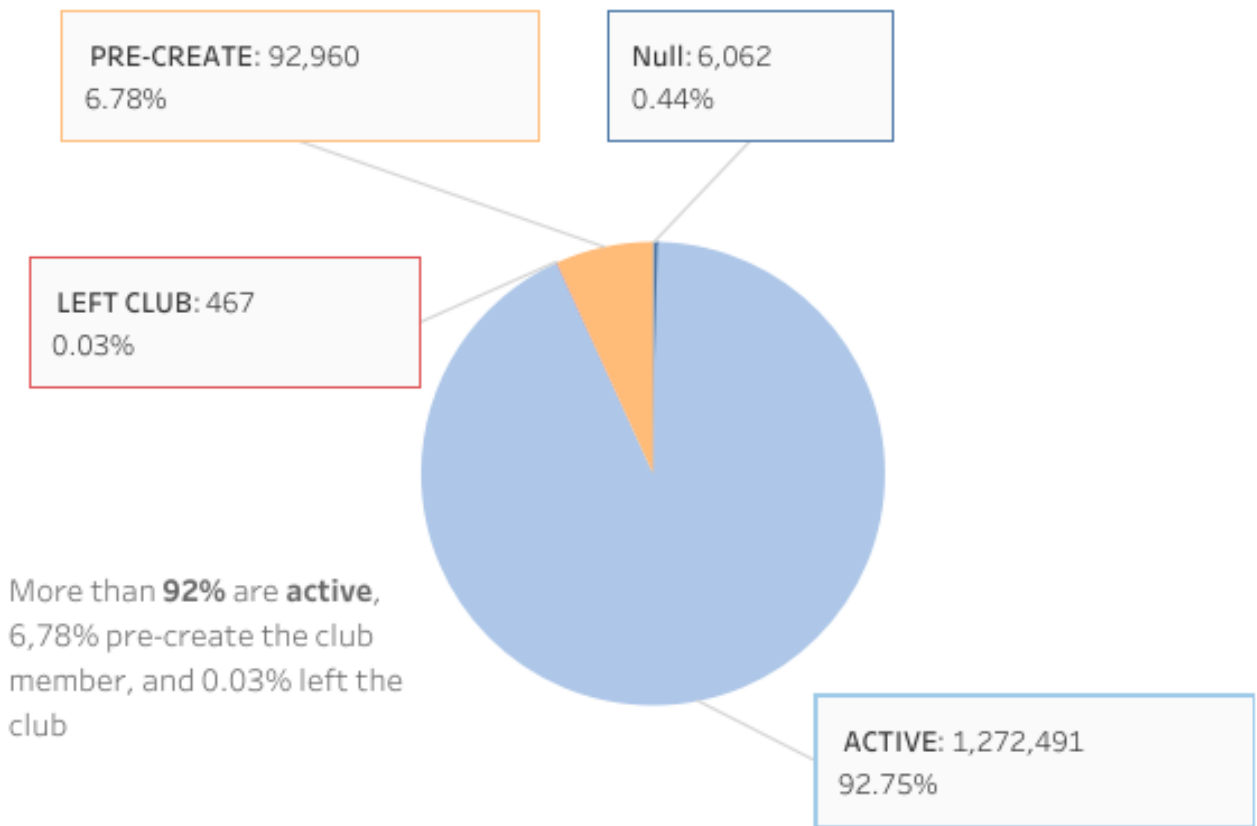
Age Groups



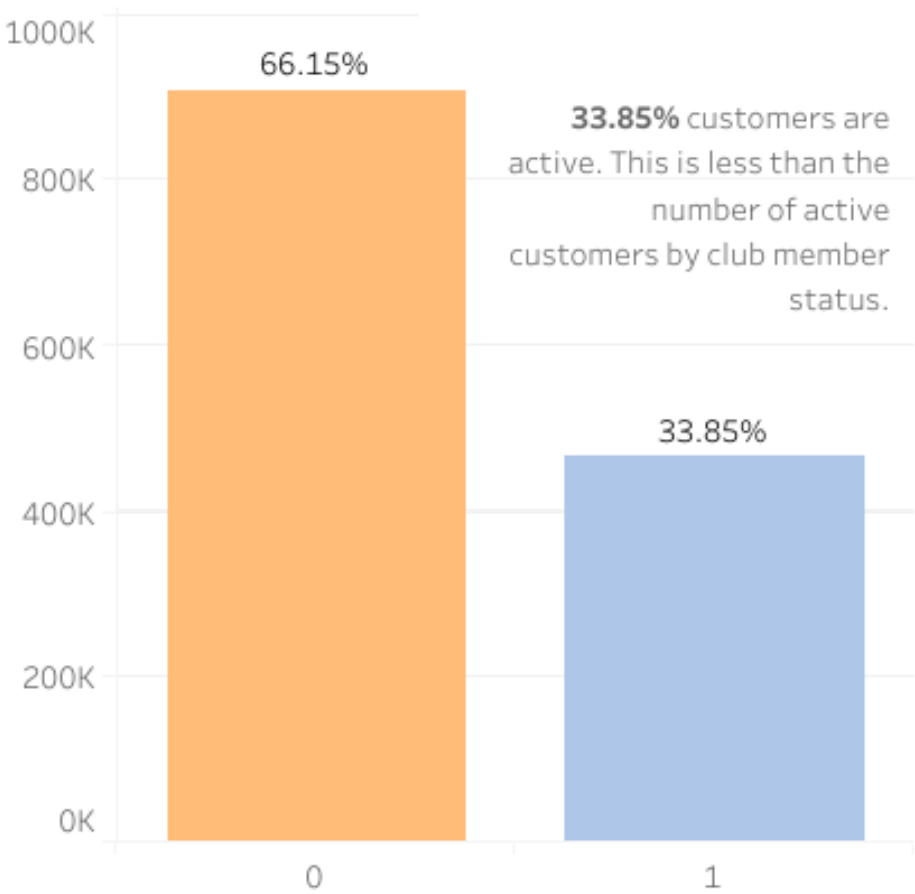
Fashion News Frequency



Club Member Status



Active Customers

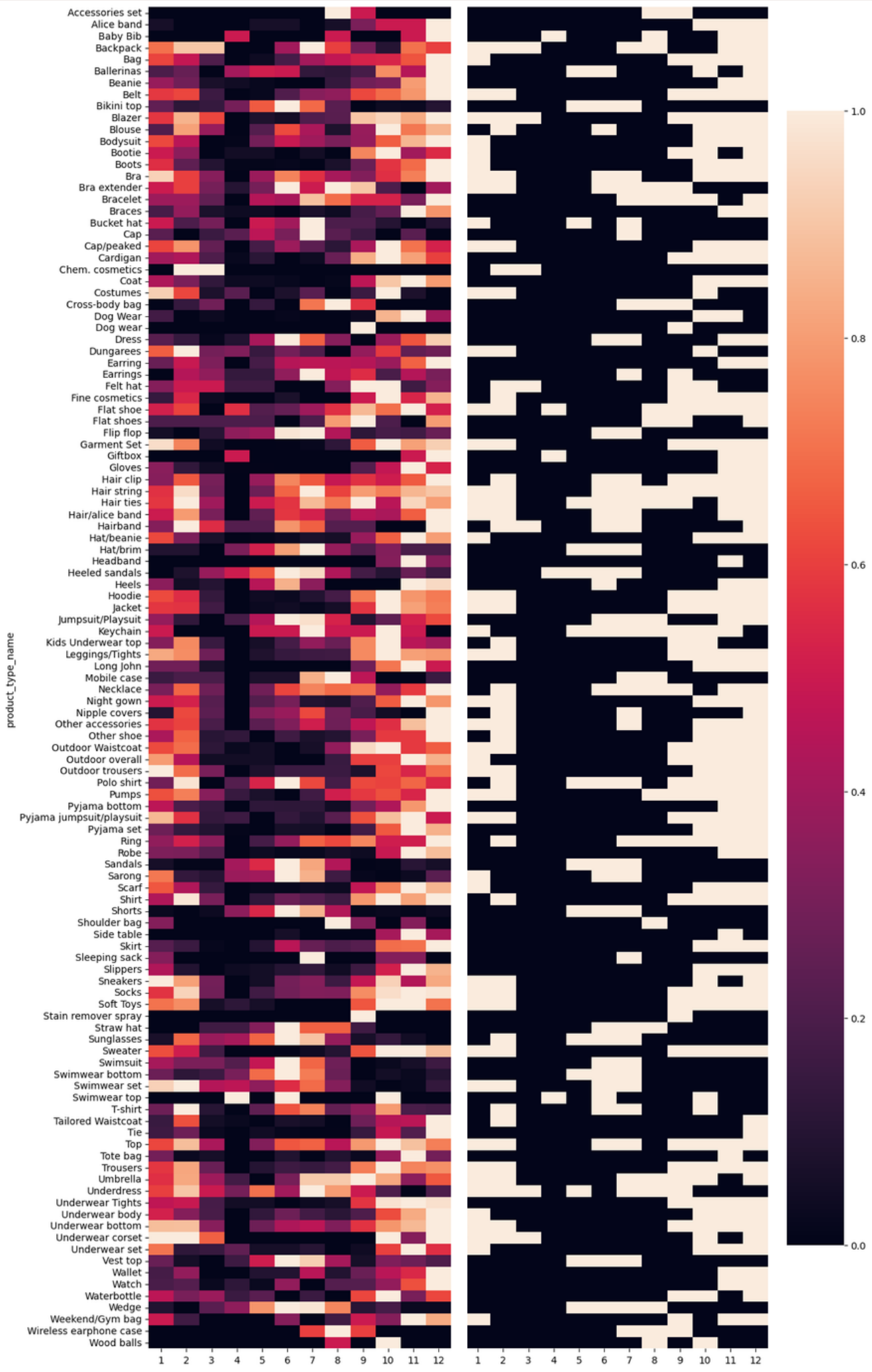


Postal code

2c29ae653a9282cce4151..	120,303
cc4ed85e30f4977dae476..	261
714976379549eb90aae4..	159
7c1fa3b0ec1d37ce2c3f34..	157
5b7eb31eabebd3277de6..	156
Others	1,250,944

There are 352,899 unique postal codes. The highest number of customers by postal code is 120,303 or 8.77% of all customers, meanwhile the second highest is only 261.

# SALES PATTERN BY TYPE OF PRODUCTS FROM JANUARY TO DECEMBER

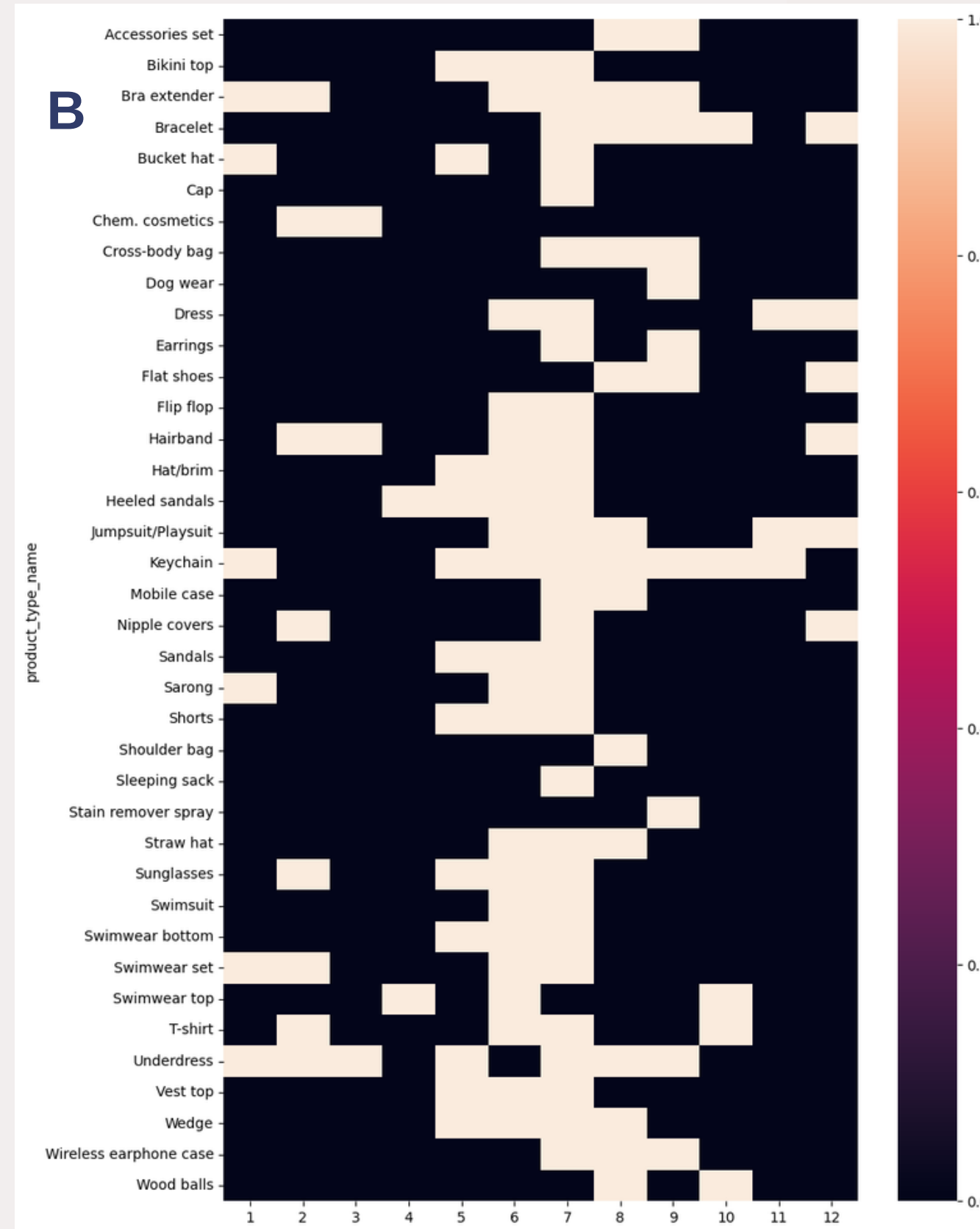
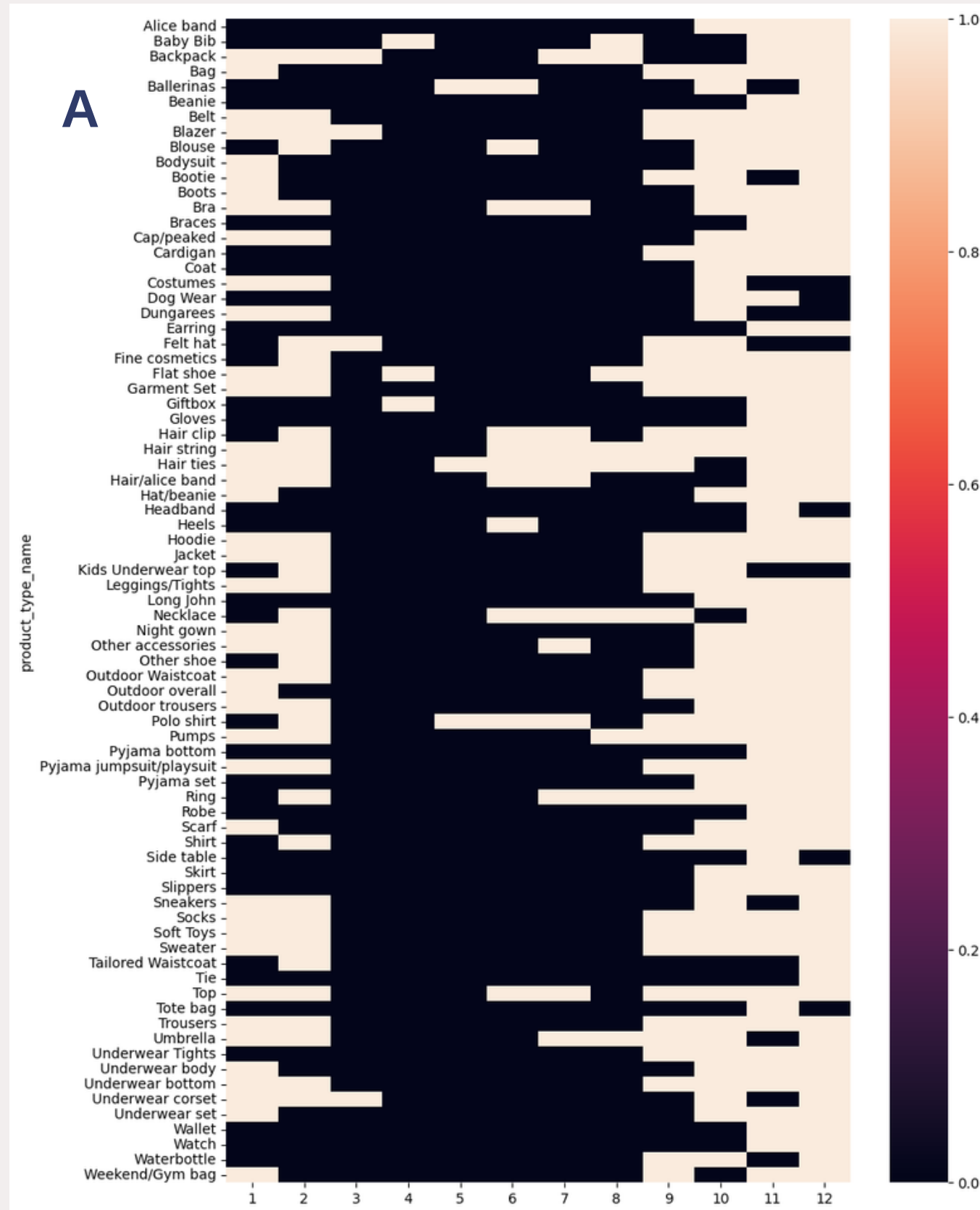




# GROUPING TYPE OF PRODUCTS

## OBSERVATIONS

1. Products in group A mostly sold well in September - February
2. Products in group B mostly have high sales in around May - August
3. There are winter and autumn clothing in group A, but not all of them are associated with winter and autumn
4. Customers bought products mostly in autumn and winter



# DATA EXPLORATION II

MARCH 2020 - AUGUST 2020

TOP 2 POSTAL CODE BY THE NUMBER OF TRANSACTION

	postal_code	transaction
0	2c29ae653a9282cce4151bd87643c907644e09541abc28...	199523
1	1f5bd429acc88fbbf24de844a59e438704aa8761bc7b99...	1439

THE NUMBER OF SALES BY PRODUCT NAME FROM MARCH-AUGUST

Product Name	March	April	May	June	July	August	Grand T..
Tilly (1)	76	45	124	213	252	256	966
Tilda tank	59	65	121	207	148	123	723
Cat Tee.	41	26	66	175	167	145	620
Nora T-shirt	33	13	101	155	120	134	556
Jade HW Skinny ..	77	50	77	67	133	151	555
Timeless Midrise..	20	40	97	162	145	83	547
Simple as That T..	31	23	126	204	118	30	532
Luna skinny RW	76	22	87	100	122	122	529
Pluto RW slacks ..	81	12	83	118	91	117	502
Charlie Top	82	3	88	115	117	96	501
Jen tee	65	17	86	167	75	77	487
Melrose	47	16	49	115	124	111	462
Skinny Ankle R...	117	2	73	87	126	54	459
Primo slacks	61	39	71	93	100	95	459
Simple as that C..	34	11	113	171	85	44	458
Brit Baby Tee	27	14	67	70	125	133	436
7p Basic Shaftless	46	12	56	81	127	108	430
SUPREME RW ti..	83	16	44	58	123	104	428
Sorrel	10	5	71	120	128	94	428
Tequila denim			40	121	163	73	397
Mariette Blazer	83		39	62	72	111	367
Eleven top	5	1	28	111	142	77	364
Therese tee	37	31	32	82	102	66	350
Lemon HW			11	120	144	73	348
Thyme ton	4	22	64	138	79	40	347



# MARKET BASKET ANALYSIS

Market basket analysis (MBA) is a data mining technique.

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.  
 $\text{Support}(A \& B) / \text{Support}(B)$

## **LIFT**

$\text{Support}(A \& B) /$   
 $\text{Support}(B) * \text{Support}(A)$

# MARKET BASKET ANALYSIS

	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

# MARKET BASKET ANALYSIS

## DATA MARCH-MAY 2020

	Invoice_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 rows x 2 columns

## DATA MAY 2020

	Invoice_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 rows x 2 columns

# MARKET BASKET ANALYSIS

DATA MARCH-MAY 2020

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

	Invoice_no	prod_name_x	prediction	lift	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

# MARKET BASKET ANALYSIS

DATA MAY 2020  
PREDICT JUNE 2020  
RULES= 487

	antecedent	consequent	confidence	lift	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

# PRODUCT RECOMMENDATION

**POPULAR ITEMS**

**CONTENT-BASED RECOMMENDATION**

**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

1

- REPURCHASED ITEMS
- PREVIOUSLY PURCHASED ITEMS

2

USER-ITEM  
RECOMMENDATION  
(TENSORFLOW  
RECOMMENDER)

3

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_id	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 rows x 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.
- OFFER SPECIAL DEALS, PRODUCT BUNDLES BASED ON RULES GENERATED FROM MARKET BASKET ANALYSIS USING PREVIOUS MONTH'S DATA.
- PERSONALIZED PRODUCT RECOMMENDATIONS USING HYBRID RECOMMENDATION SYSTEM

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