Background

H&M is the world's second-largest clothing retailer

- Operates in more than 74 countries with more than 5,000 stores worldwide
- Plan to close 5% of stores in 2021 because of COVID

Shift from traditional retail settings to eCommerce

Problem

Lots of competition within the online retail space

H&M needs to differentiate itself

Different consumer expectations in eCommerce that H&M needs to adapt to

Create the best online shopping experience for shoppers

Our Proposed Solution

Create an organized database to help H&M more easily access consumer data that can be used to understand their customers and ultimately develop product recommendation systems

Original Data Description



• **Data Source**: Kaggle

• **Data Name**: H&M Personalized Fashion

Recommendation datasets

- Article csv file
- Customer csv file
- Transaction csv file
- Data Provider: H&M Group
- Data Access Link:

https://www.kaggle.com/competitions/h-and-m-perso

nalized-fashion-recommendations/data

Key data information:

Article file

Article id, product name, product type,
 graphical appearance, color, department etc.

Customer file

 Customer id, customer active status, customer membership status, age.

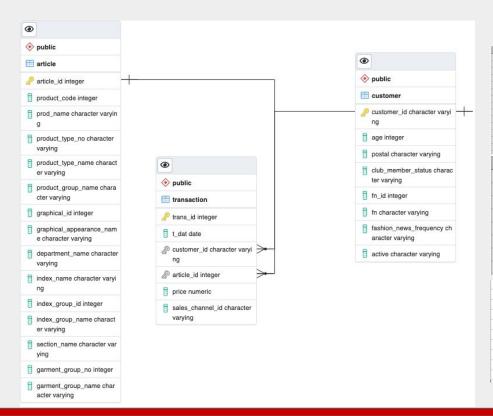
Transaction file

Transaction date, customer id, article article
 (products purchased), price, sales channel

H&M Personalized Fashion Recommendations

Provide product recommendations based on previous purchases

Normalization - 1NF

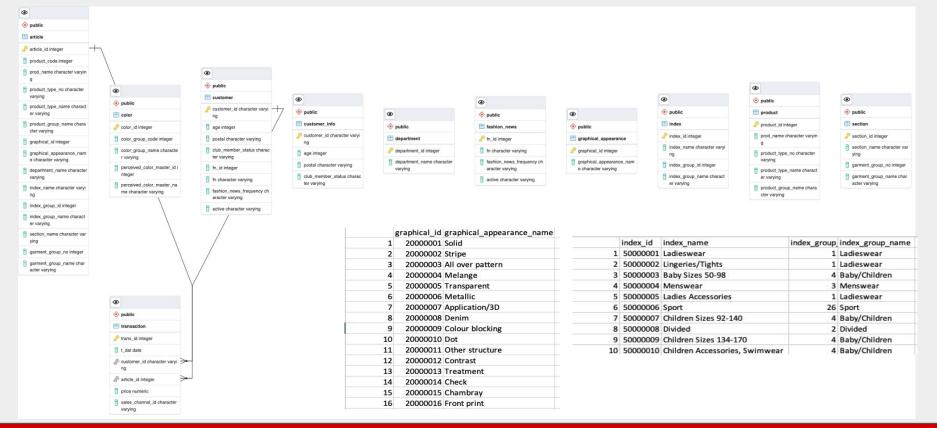


- Each table cell should contain a single value.
- Each record needs to be unique.

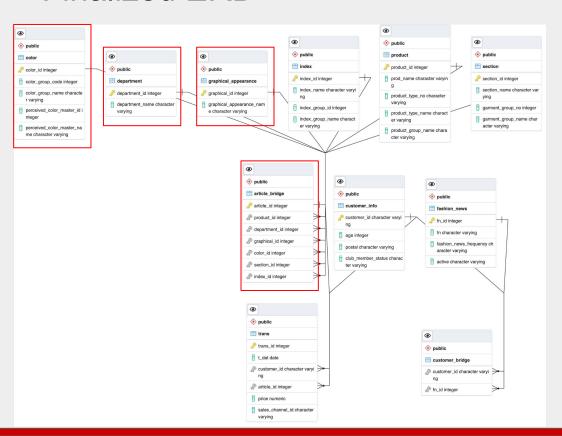
article_id	product_code	prod_name	product_type_no	product_type_name	product_group_name	graphical_appearan	on_ec	graphical_appe	earance_name	colour_group_code col	our_group_name perce	eived_colour_value_id	perceiv	ed_color
108775015	108775	Strap top	253	Vest top	Garment Upper body	10	10016	Solid		9 Bla	ok	4	Dark	
108775044	108775	Strap top	253	Vest top	Garment Upper body	10	10016	Solid		10 Wh	ite	3	Light	
108775051		Strap top (1)		Vest top	Garment Upper body		10017			11 Off			Dusty Li	ight
110065001		OP T-shirt (ldro)		Bra	Underwear		10016			9 Bla			Dark	
110065002		OP T-shirt (Idro) OP T-shirt (Idro)	_	Bra Bra	Underwear		10016			10 Wh			Light Dusty Li	
111565001		20 den 1p Stockings		Underwear Tights	Socks & Tights		10016			12 Ug			Dark	ignt
111565003		20 den 1p Stockings		Socks	Socks & Tights		10016			13 Bei			Medium	Dusty
11586001		Shape Up 30 den 1p Tights	273	Leggings/Tights	Garment Lower body	10	10016	Solid		9 Bia	ok		Dark	
111593001	111593	Support 40 den 1p Tights	304	Underwear Tights	Socks & Tights	10	10016	Solid		9 Bla	ok	4	Dark	
11609001	111609	200 den 1p Tights	304	Underwear Tights	Socks & Tights	10	10016	Solid		9 Blo	ok	4	Dark	
12679048		SWEATSHIRT OC		Sweater	Garment Upper body			All over pattern		7 Gre			Dusty Li	
12679052		SWEATSHIRT OC		Sweater	Garment Upper body			All over pattern		71 Ligi			Dusty Li	ight
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000058a12d5b43e67d225668fa1f8d618c13dc232df0cad8ffe7ad4a1091e318						91e318			ACTIVE		NONE			24
00005ca1c9ed5f5146b52ac8639a40ca9d57aeff4d1bd2c5feb1ca5dff07c43e						07c43e	1		ACTIVE		NONE			54
00006413d8573cd20ed7128e53b7b13819fe5cfc2d801fe7fc0f26dd8d65a85a						165a85a	1.0	1.0	ACTIVE		Regularly			52
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00007	'd2de826	758b65a93dd2	4ce629ed66	842531df669	9338c5570910	0a014cc2	1.0	1.0	ACTIVE		Regularly			32
	t_dat customer_id									article_id price sales_ch		nanne	el_id	
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	2 9/20/18 000058a12d5b43e67d225668fa1f8d618c13dc232df0cad8ffe7						7ad	4a1091e318 541518023 0.030491525				- 2		
	3 9/20/18 00007d2de826758b65a93dd24ce629ed66842531df6699338d						557	70910a01	14cc2	505221004	05221004 0.015237288			- 2
4 9/20/18 00007d2de826758b65a93dd24ce629ed66842531df6699338c						557	70910a01	14cc2	2 685687003 0.016932203			- 2		
5 9/20/18 00007d2de826758b65a93dd24ce629ed66842531df6699338						557	70910a01	14cc2	685687004	7004 0.016932203			- 2	
6 9/20/18 00007d2de826758b65a93dd24ce629ed66842531df6699338c						557	70910a01	14cc2	685687001	0.0169322	2203		- 2	
7 9/20/18 00007d2de826758b65a93dd24ce629ed66842531df6699338c					:557	70910a01	14cc2	505221001	0.0203220	34		- 2		
	8 9/20	18 00083cda04	41544b2fbb	0e0d2905ad	17da7cf10079	26fb4c7323	35dd	cbbc132	280	688873012	0.030491525			

Normalization - 2NF

- The table must be already in 1 NF and all non-key columns of the tables must depend on the PRIMARY KEY
- The partial dependencies are removed and placed in a separate table



Normalization - 3NF Finalized ERD



- Non-Primary key columns shouldn't depend on the other non-Primary key columns
- There is no transitive functional dependency

article_id	co	lour_id	department_	graphical_id	index_id	product_i	d section_id	articlebridge	
1087750	015	41000001	31000001	2000000	50000001	100000	001 60000001	108775015	
1087750	044	41000002	31000001	2000000	50000001	100000	001 60000001	108775044	
1087750	051	41000003	31000001	20000002	50000001	100000	002 60000001	108775051	
1100650	001	41000001	31000002	2000000	50000002	100000	003 60000002	110065001	
1100650	002	41000002	31000002	2000000	50000002	100000	003 60000002	110065002	
1100650	011	41000004	31000002	2000000	50000002	100000	003 60000002	110065011	
1115650	001	41000001	31000003	2000000	50000002	100000	004 60000003	111565001	
1115860	001	41000001	31000003	2000000	50000002	100000	006 60000003	111586001	
1115930	001	41000001	31000003	2000000	50000002	100000	007 60000003	111593001	
1116090	001	41000001	31000003	2000000	50000002	100000	008 60000003	111609001	
1163790	047	41000009	31000001	2000000	50000001	100000	011 60000001	116379047	
1184580	004	41000011	31000001	20000004	50000004	100000	012 60000005	118458004	
1184580	029	41000010	31000001	20000004	50000004	50000004 1000001		118458029	
1184580	038	41000013	31000001	20000004	50000004	100000	012 60000005	118458038	
color_id	color_	group_code	color_group	_name perc	eived_color_m	aster_id p	perceived_color_r	master_name	
41000001 9		Black			5 E	lack			
41000002		10 White			9 White				
41000003		1:	1 Off White		9 White				

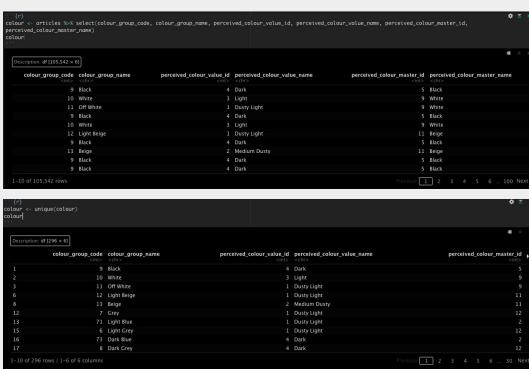
color_id	color_group_code	color_group_name	perceived_color_master_id	perceived_color_master_name
41000001	9	Black	5	Black
41000002	10	White	9	White
41000003	11	Off White	9	White
41000004	12	Light Beige	11	Beige
41000005	13	Beige	11	Beige
41000006	7	Grey	12	Grey
41000007	71	Light Blue	2	Blue
41000008	6	Light Grey	12	Grey
41000009	73	Dark Blue	2	Blue
41000010	8	Dark Grey	12	Grey
41000011	7	Grev	12	Grev

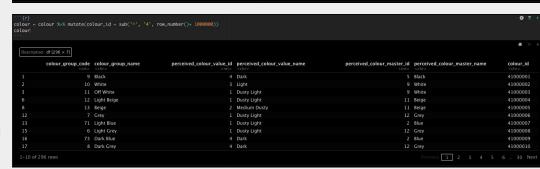
department_id	department_name	graphical id	graphical_appearance_name
31000001	Jersey Basic		
31000002	Clean Lingerie	20000001	
		20000002	Stripe
	Tights basic	20000003	All over pattern
31000004	Baby basics	2000004	
31000005	Casual Lingerie		
31000007	Jersev		Transparent
	EQ & Special Collections	20000006	Metallic
		20000007	Application/3D
	Hair Accessories	20000008	Denim
31000010	Other items		
31000011	Rahy Nightwear	20000009	Colour blocking

ETL Process

- 1. Create color table by cleaning and manipulating the data to only the attributes we want for the table. As you can see there are 105,542 rows.
- Find only the distinct rows.
 We eliminated the duplicate rows. Now we have 296 distinct color combination rows.

3. Create a primary key (ID) for the color table.





Analytical Procedure - Why

Why do we interact with data?

- Data is not always in the format as wish
- Employees who have different job responsibilities have different requirements for data analysis results

Customer Needs

- Fast, easy and convenient
- Getting a better understanding of the customers profile and demand for marketing campaign
- Different requirements for data

Results and Benefits

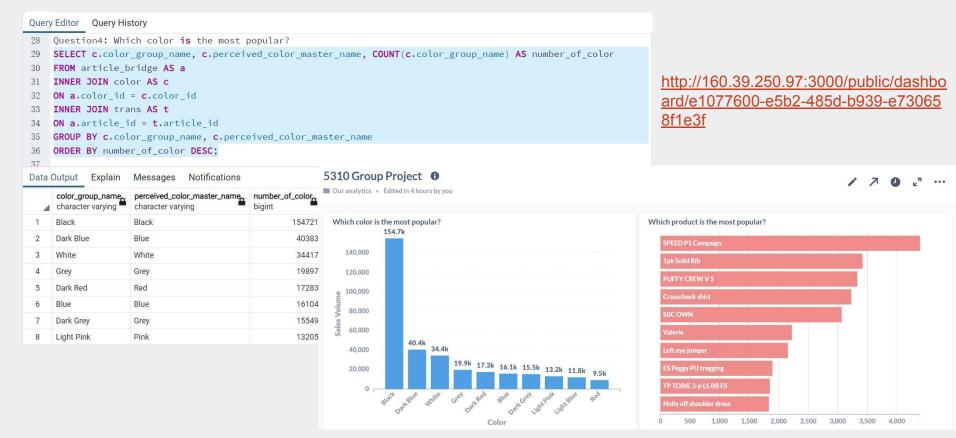
- Ops team: To know better about the fashion and trends in the market and make appropriate designing and manufacturing decisions
- C-level: To obtain business insights and support decision-making procedure through data analysis.

Analytical Procedure - How

We have multiple front-end and back-end interaction tools for our SQL database, such as Python, PGAdmin, and Metabase. Each tool has its own feature.

- Python is a front end programming language to help with data visualization, ETL process before data ingestion, and data analytics. Psycopg2 is the library we used for connection to our SQL server. We can use Hadoop for HDFS and Spark for ML analytics.
 - **pgAdmin** is the platform for our SQL server. Our clients or our product manager can directly interact with our database using query language on pgAdmin.
 - Metabase is the automated data visualization tool for our SQL database. Our data security is ensured
 with limited views for different levels of analyst and manager. Different positions within the company can
 view different visualizations and it is always updated.

Database Interaction Demo



Conclusion



- Gain deeper business insights from the results generated by data analytics
 - o more targeted marketing strategy and campaign based on the data analytical results.

- Gain better insight of customer behaviors, and helps H&M decision making process
 - o eg. most popular color and graphical print among customers

 Provide information and experience for ETL procedure implementation in other department (e.g accounting, HR, public relations), improving company operation efficiency.