MRA Project Part-B

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Problem Statement:

A grocery store shared the transactional data with you. Your job is to conduct a thorough analysis of Point of Sale (POS) data, identify the most commonly occurring sets of items in the customer orders, and provide recommendations through which a grocery store can increase its revenue by popular combo offers & discounts for customers.

Dataset:

Grocery Store Data: dataset group.csv

Top few records to get a feel of the data structure are shown below:

	Date	Order_id	Product	-	Date	Order_id	Product
0	01-01-2018	1	yogurt	20636	25-02-2020	1138	soda
1	01-01-2018	1	pork	20637	25-02-2020	1138	paper towels
2	01-01-2018	1	sandwich bags	20638	26-02-2020	1139	soda
3	01-01-2018	1	lunch meat	20639	26-02-2020	1139	laundry detergent
4	01-01-2018	1	all- purpose	20640	26-02-2020	1139	shampoo

Exploratory data analysis (EDA)

The dataset contains 20641 rows and 3 columns

```
df.shape
(20641, 3)
```

The info gives the information of the dataset such as Column name, Non-null count, Dtype. The dataset contains 3 columns in which 1 int and 2 object datatype

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20641 entries, 0 to 20640
Data columns (total 3 columns):
# Column Non-Null Count Dtype
    Date 20641 non-null
                             object
1 Order id 20641 non-null
                             int64
    Product 20641 non-null
                             object
dtypes: int64(1), object(2)
memory usage: 483.9+ KB
```

The five point summary of the dataset can be obtained from the describe function.

	Order_id
count	20641.000000
mean	575.986289
std	328.557078
min	1.000000
25%	292.000000
50%	581.000000
75%	862.000000
max	1139.000000

 Null values should be check before performing calculations. Checking the null values by using isnull function. We have observed that there are no null values present in the dataset.

```
Date 0
Order_id 0
Product 0
dtype: int64
```

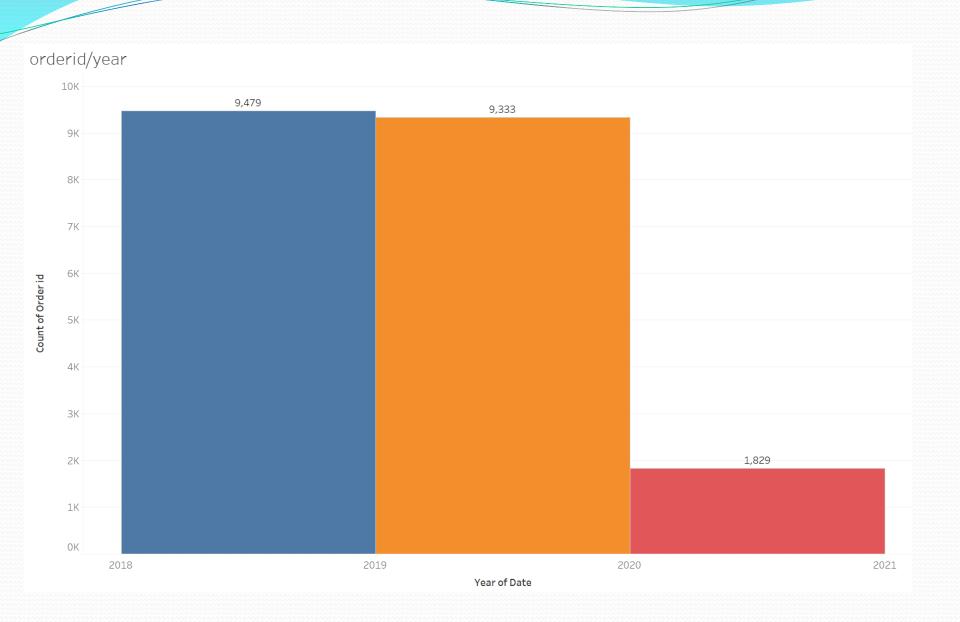
 On checking the duplicate values we have observed that there are 4730 duplicate values. Thus before calculation we will remove the duplicate values.

```
df.duplicated().sum()
4730
```

Drop function is used to remove the duplicate values.

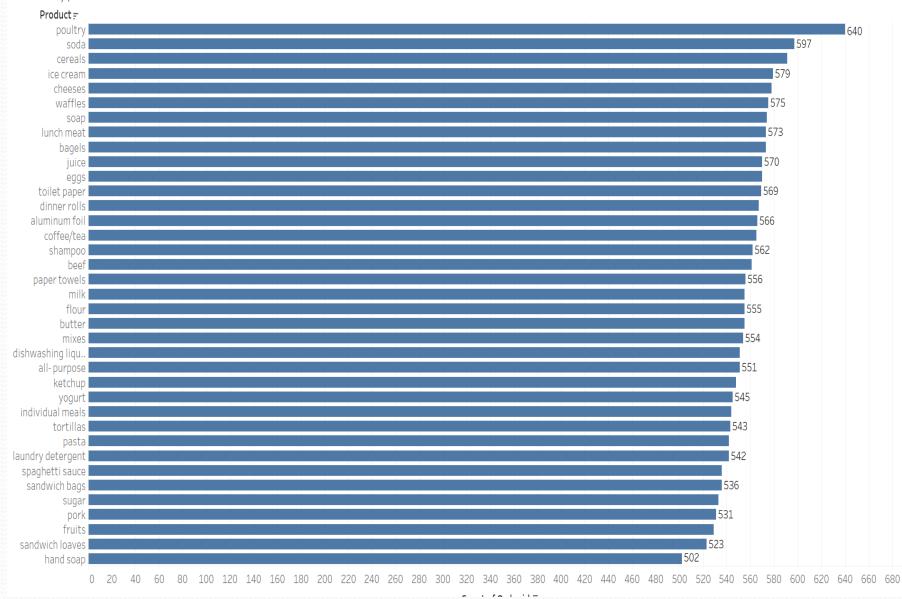
The number of duplicate values are 0

- Order-id is plotted against Year in Tableau.
- The blue color shows the sale from 2018-2019, orange color shows the sale from 2019-2020 where as red color shows the sale from 2020-2021.
- The sale from 2018-2019 is 9479\$, from 2019-2020 it is 9333\$ where as from 2020-2021 is 1829\$.
- Thus, it is clear from the graph that the sale is decreases as in 2021 lowest sale is observed.

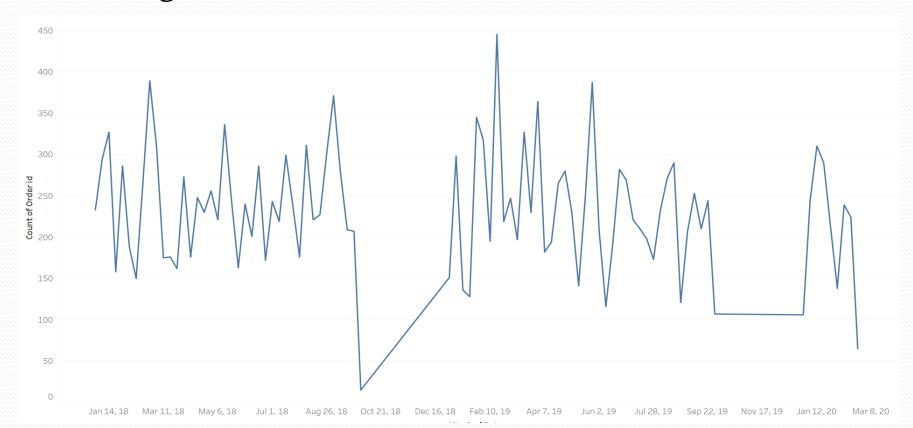


- Tableau is used to make a graph of count of order-id and products that are plotted against each other.
- It is observed from the graph that the maximum orderid is placed for poltry, soda, cereals where as minimum order-id is placed for handsoaps and sanwich loaves.
- The maximum order count for poltry is 640 where as minimum order count for hansoaps is 502

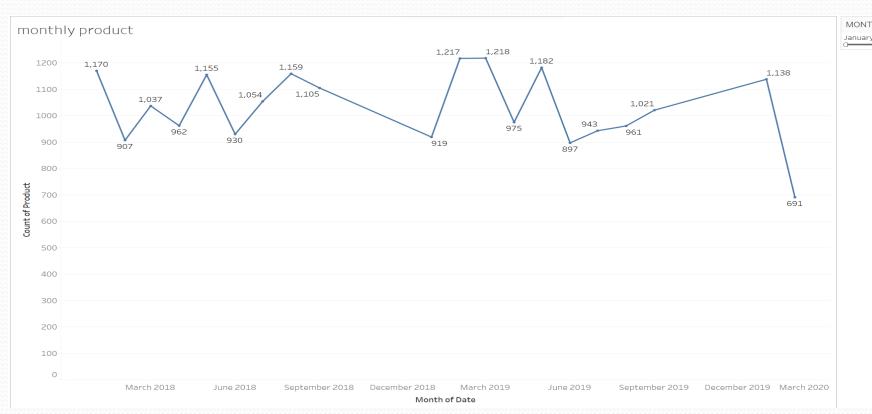
orderid/product



- The weekly order count is plotted in below graph.
- It is observed that only in second week of feb2020 sale was increase other than that the sale is continuously decreasing in 2020

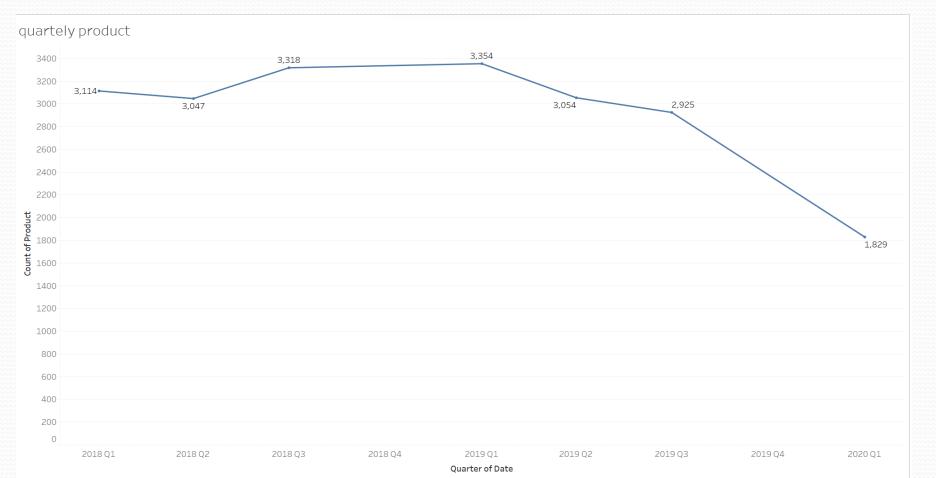


- The monthly product count is shown in below line graph.
- It is observed that product count decreases from 1138 to 691

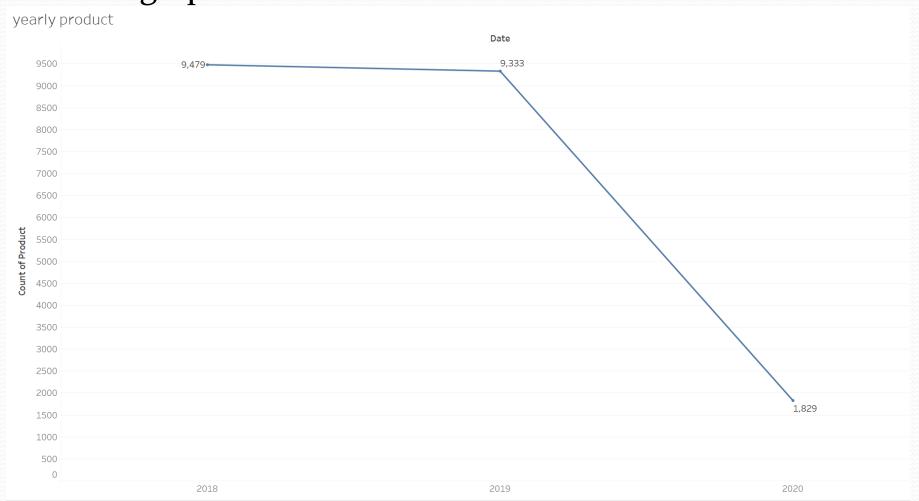


MONTH(Date)
January 2018 February 2020

- The quartely product count is shown below which shows.
- It shows that from 2019 Q3 to 2020Q1 there is huge loss in product count



• The yearly decrease in product count in shown in below line graph.

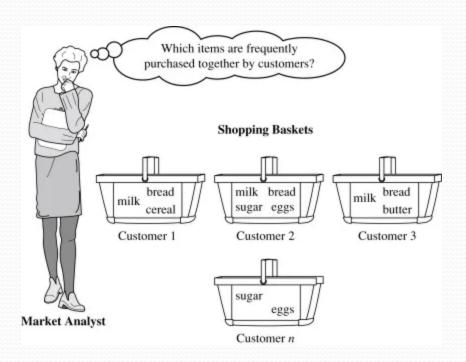


INFERENCES

- Since sale is deceasing in 2020 from 2nd Quarter onwards thus more offers should be provided on combos and update customers about the offers
- Sales can also increase by providing different options to customers such as buy using buy online and pickup in store, in house delivery, reducing wait time at counters etc.
- Sale can also increase by providing discounts at various seasons, weekend etc

Market Basket Analysis

 Market Basket Analysis is a technique used in Data Mining to increase the sales by understanding customer purchase pattern.



The three important terms which are used in Market Basket Analyses.

Lift =
$$((A+B)/A)/(B/Total)$$

The KNIME tool is used to do the Market Basket Analysis

The data is imported in the KNIME using the CSV node.

Row	s: 20641	Colu	ımns: 3					C
	#	RowID	Date String	~	Order_id Number (integer)	~	Product String	
	1	Row0	01-01-2018		1		yogurt	
	2	Row1	01-01-2018		1		pork	
	3	Row2	01-01-2018		1		sandwich bags	
	4	Row3	01-01-2018		1		lunch meat	
	5	Row4	01-01-2018		1		all- purpose	
	6	Row5	01-01-2018		1		flour	
	7	Row6	01-01-2018		1		soda	
	8	Row7	01-01-2018		1		butter	
	9	Row8	01-01-2018		1		beef	
	10	Row9	01-01-2018		1		aluminum foil	
	11	Row	01-01-2018		1		all- purpose	
	12	Row	01-01-2018		1		dinner rolls	
	13	Row	01-01-2018		1		shampoo	
	14	Row	01-01-2018		1		all- purpose	
	15	Row	01-01-2018		1		mixes	
	16	Row	01-01-2018		1		soap	
	17	Row	01-01-2018		1		laundry detergent	
	18	Row	01-01-2018		1		ice cream	
	19	Row	01-01-2018		1		dinner rolls	
	20	Row	01-01-2018		2		toilet paper	
	21	Row	01-01-2018		2		shampoo	
	22	Row	01-01-2018		2		hand soap	
	23	Row	01-01-2018		2		waffles	
П	24	Row	01-01-2018		2		cheeses	

• Group By node is used to convert the data in the form of set as shown below :-

Rows	s: 1139	Colum	nns: 2	
	#	RowID	Order_id Number (integer)	Set(Product) Set
	1	Row0	1	[yogurt,pork,sandwich bags,]
	2	Row1	2	[toilet paper,shampoo,hand soap,]
	3	Row2	3	[soda,pork,soap,]
	4	Row3	4	[cereals,juice,lunch meat,]
	5	Row4	5	[sandwich loaves,pasta,tortillas,]
	6	Row5	6	[laundry detergent,toilet paper,eggs,]
	7	Row6	7	[individual meals,paper towels,tortillas,]
	8	Row7	8	[ice cream,juice,paper towels,]
	9	Row8	9	[juice,poultry,coffee/tea,]
	10	Row9	10	[ketchup,coffee/tea,toilet paper,]
	11	Row	11	[sandwich loaves,ice cream,soda,]
	12	Row	12	[pork,tortillas,shampoo,]
	13	Row	13	[sugar,fruits,all- purpose,]
	14	Row	14	[fruits,dinner rolls,individual meals,]
	15	Row	15	[individual meals,ice cream,cereals,]
	16	Row	16	[sugar,sandwich bags,flour,]
	17	Row	17	[milk,hand soap,pasta,]
	18	Row	18	[sandwich bags,toilet paper,bagels,]
	19	Row	19	[individual meals,laundry detergent,coffee/tea,]
	20	Row	20	[shampoo,dishwashing liquid/detergent,yogurt,]
	21	Row	21	[waffles,fruits,all- purpose,]
	22	Row	22	[cheeses,cereals,sugar,]
	23	Row	23	[aluminum foil,bagels,shampoo,]
	24	Row	24	[fruits,all- purpose,pasta,]

The Association Rule Learner node is used to get the Support, Confidence, Lift, Consequent as shown in table below

Rows	s: 20272	Colu	ımns: 6										
	#	RowID	Support Number (double)	~	Confidence Number (double)	~	Lift Number (double)	~	Consequent	implies String	~	Items Set	√ √ √
	1	rule0	0.05		0.576		1.497		yogurt	<		[toilet paper,juice,	al
	2	rule1	0.05		0.594		1.544		aluminum foil	<		[yogurt,toilet pape	er,j
	3	rule2	0.05		0.538		1.421		toilet paper	<		[yogurt,juice,alum	in
	4	rule3	0.05		0.64		1.7		juice	<		[yogurt,toilet pape	er,
	5	rule4	0.05		0.523		1.36		yogurt	<		[poultry,juice,alum	nin
	6	rule5	0.05		0.564		1.468		aluminum foil	<		[yogurt,poultry,juid	ce]
	7	rule6	0.05		0.538		1.276		poultry	<		[yogurt,juice,alum	in
	8	rule7	0.05		0.62		1.645		juice	<		[yogurt,poultry,alu	ım
	9	rule8	0.05		0.588		1.528		yogurt	<		[cheeses,cereals,	co
	10	rule9	0.05		0.594		1.52		cheeses	<		[yogurt,cereals,co	ff
	11	rule10	0.05		0.582		1.469		cereals	<		[yogurt,cheeses,c	of
	12	rule11	0.05		0.613		1.616		coffee/tea	<		[yogurt,cheeses,c	er
	13	rule12	0.05		0.588		1.564		mixes	<		[dishwashing liqu	id
	14	rule13	0.05		0.528		1.395		laundry detergent	<		[dishwashing liqu	id
	15	rule14	0.05		0.576		1.484		dishwashing liquid/	<		[poultry,laundry de	et
	16	rule15	0.05		0.6		1.424		poultry	<		[dishwashing liqu	id
	17	rule16	0.051		0.563		1.464		yogurt	<		[poultry,mixes,alu	mi
	18	rule17	0.051		0.569		1.479		aluminum foil	<		[yogurt,poultry,mi	xes]
	19	rule18	0.051		0.63		1.678		mixes	<		[yogurt,poultry,alu	ım
	20	rule19	0.051		0.586		1.39		poultry	<		[yogurt,mixes,alur	mi
	21	rule20	0.051		0.611		1.66		sandwich bags	<		[cheeses,bagels,c	er
	22	rule21	0.051		0.674		1.726		cheeses	<		[bagels,cereals,sa	an

The recommended items according to basket is obtained from the KNIME tool as shown below :

Row	s: 20272	Colu	ımns: 5									0
	#	RowID	Rec_Item ↑	~	Support Number (double)	~	Confidence Number (double)	~	Lift Number (double)	~	Items_basket String	V
	1	Row0	all- purpose		0.07		0.444		1.186		aluminum foil, ketchup	
	2	Row1	all- purpose		0.073		0.464		1.237		aluminum foil, soap	
	3	Row2	all- purpose		0.157		0.409		1.09		aluminum foil	
	4	Row3	all- purpose		0.07		0.417		1.111		bagels, aluminum foil	
	5	Row4	all- purpose		0.062		0.41		1.095		bagels, coffee/tea	
	6	Row5	all- purpose		0.067		0.411		1.096		bagels, juice	
	7	Row6	all- purpose		0.063		0.414		1.104		bagels, laundry detergent	
	8	Row7	all- purpose		0.07		0.444		1.186		bagels, lunch meat	
	9	Row8	all- purpose		0.063		0.416		1.11		bagels, pasta	
	10	Row9	all- purpose		0.065		0.418		1.115		bagels, pork	
	11	Row	all- purpose		0.06		0.41		1.093		bagels, soap	
	12	Row	all- purpose		0.062		0.413		1.101		bagels, sugar	
	13	Row	all- purpose		0.067		0.409		1.09		bagels, toilet paper	
	14	Row	all- purpose		0.066		0.405		1.081		bagels, tortillas	
	15	Row	all- purpose		0.065		0.407		1.085		bagels, waffles	
	16	Row	all- purpose		0.067		0.418		1.114		beef, aluminum foil	
	17	Row	all- purpose		0.058		0.41		1.093		beef, coffee/tea	
	18	Row	all- purpose		0.066		0.424		1.13		beef, juice	
	19	Row	all- purpose		0.062		0.415		1.108		beef, pork	
	20	Row	all- purpose		0.064		0.41		1.094		beef, sandwich bags	
	21	Row	all- purpose		0.072		0.451		1.202		beef, soap	
	22	Row	all- purpose		0.062		0.401		1.07		beef, soda	
	23	Row	all- purpose		0.066		0.419		1.118		beef, toilet paper	
Ж.П.	24	Row	all- purpose		0.064		0.422		1.126		butter. bagels	
												ANNANA.

INFERENCES

 The customer will buy dinner rolls, spaghetti, sauce, bagels, cereals then he will buy poultry because its probability is approx. 70%

Rec_Item String	~	Support Number (double)	~	Confidence ↓ Number (double)	~	Lift Number (double)	~	Items_basket ∨ √
poultry		0.052		0.686		1.628		dinner rolls, spaghetti sauce,
cheeses		0.051		0.674		1.726		bagels, cereals, sandwich ba
poultry		0.054		0.656		1.556		dinner rolls, spaghetti sauce,
paper towels		0.055		0.649		1.791		eggs, ice cream, pasta
pasta		0.055		0.643		1.731		paper towels, eggs, ice cream
dinner rolls		0.054		0.642		1.651		spaghetti sauce, poultry, laun

 A lift greater than 1 indicated that the chances of buying item increases where less than 1 indicates that there are less chances of buying the item. Hence, paper towel, pasta, cheeses, juices are more with egg, icecream, cereal etc

Row	s: 20272	Colu	ımns: 5									(
	#	RowID	Rec_Item String	~	Support Number (double)	~	Confidence Number (double)	~	Lift ↓ Number (double)	~	Items_basket String	∇
	12265	Row	paper towels		0.055		0.649		1.791		eggs, ice cream, pasta	
	13006	Row	pasta		0.055		0.643		1.731		paper towels, eggs, ice cream	
	3315	Row	cheeses		0.051		0.674		1.726		bagels, cereals, sandwich ba	
	9164	Row	juice		0.05		0.64		1.7		yogurt, toilet paper, aluminu	
	12060	Row	mixes		0.051		0.63		1.678		yogurt, poultry, aluminum foil	
	14314	Row	sandwich bags		0.051		0.611		1.66		cheeses, bagels, cereals	
	5057	Row	dinner rolls		0.054		0.642		1.651		spaghetti sauce, poultry, laun	
	5055	Row	dinner rolls		0.052		0.641		1.649		spaghetti sauce, poultry, ice	
	9155	Row	juice		0.05		0.62		1.645		yogurt, poultry, aluminum foil	
	13722	Row	poultry		0.052		0.686		1.628		dinner rolls, spaghetti sauce,	
	6136	Row	eggs		0.052		0.634		1.627		paper towels, dinner rolls, pa	
	13005	Row	pasta		0.052		0.602		1.621		paper towels, eggs, dinner ro	
	5054	Row	dinner rolls		0.051		0.63		1.621		spaghetti sauce, poultry, cer	
	6142	Row	eggs		0.055		0.63		1.616		paper towels, ice cream, pasta	
	4480	Row	coffee/tea		0.05		0.613		1.616		yogurt, cheeses, cereals	
	5056	Row	dinner rolls		0.052		0.628		1.614		spaghetti sauce, poultry, juice	
	5911	Row	eggs		0.052		0.628		1.61		dinner rolls, poultry, soda	
	17089	Row	spaghetti sauce		0.054		0.598		1.603		dinner rolls, poultry, laundry	
	11335	Row	milk		0.051		0.604		1.589		poultry, laundry detergent, ce	

- From the Association Rule Learner table we can get the products which are having high possibility of purchase.
- Discounts should be given on combos.
- Shopkeeper should also put some recommended items on sale to increase the sale
- Offers should also be given on some items to increase the sales

KNIME MODEL

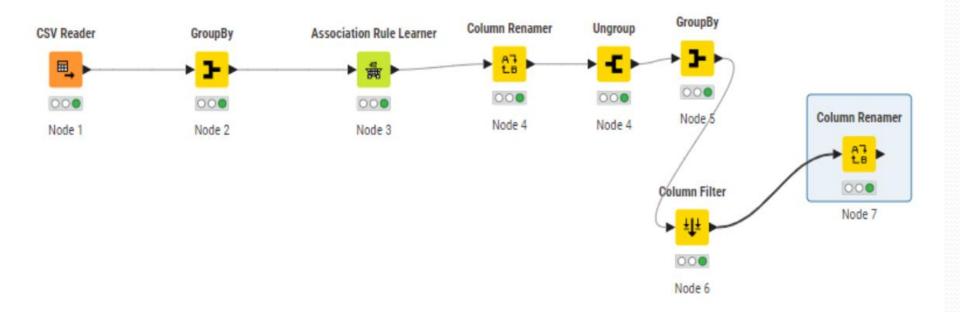


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