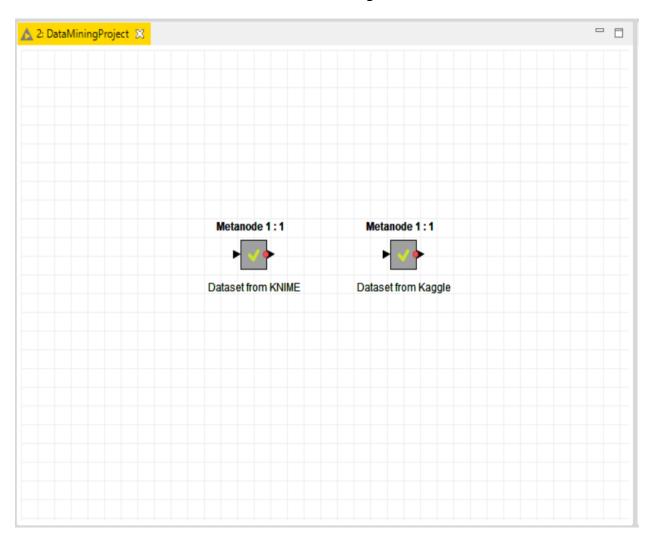
## **ABSTRACT**

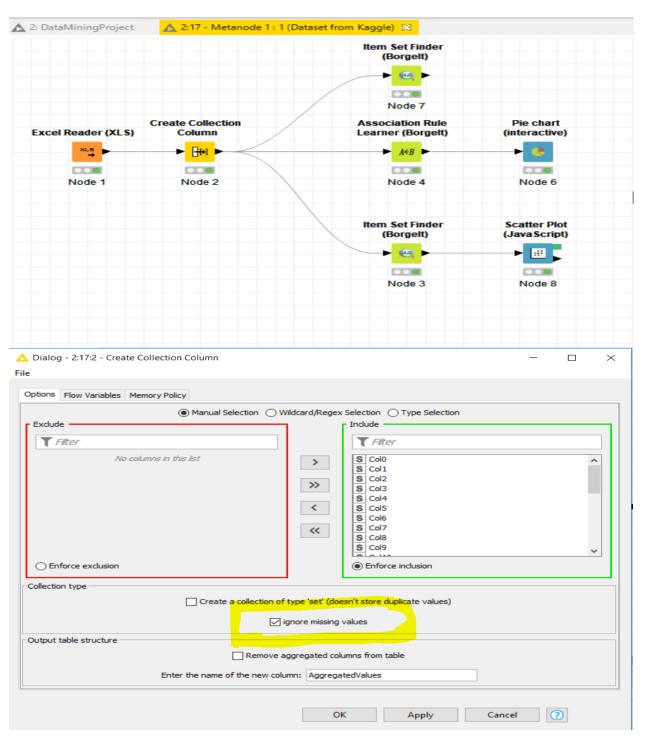
Main purpose of working on this project is analyzing data sets which contain market data to apply some algorithms such as Apriori, FPGrowth, TANIMOTO and more to find frequent item sets to show results through pie chart and scatter plot on KNIME.

## **STEPS OF THE PROJECT**

We used two different data sets to get better results. The first data set that we applied algorithms on from Kaggle (<a href="https://www.kaggle.com/apmonisha08/market-basket-analysis">https://www.kaggle.com/apmonisha08/market-basket-analysis</a>) and the second data set from KNIME's data set example for Basket Analysis. We created two different metanodes not to be distracted while working on data sets from different sources.



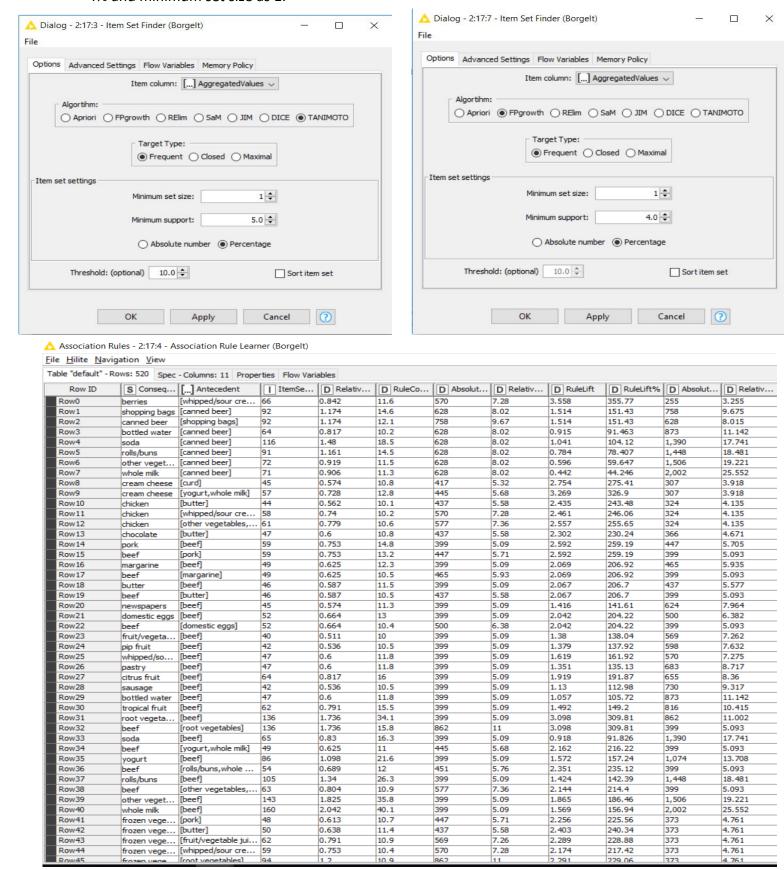
Firstly, we are going to explain our project which we used the data set from Kaggle. In the figure below, you will be able to see the steps of the project such as reading the data as an Excel file, then we created data collection (like an array) to apply algorithms on our data set through "Create Collection Column" and as a step of preprocessing and due to have a lot of missing values, we ignored missing values not to get wrong/worse results from our data analysis.



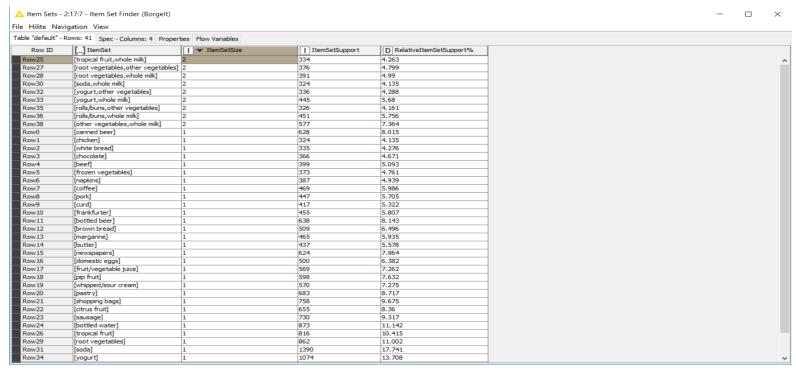
Before ignoring the missing values, our data set looked like in the figure below and it might cause have wrong results if we also add them our analysis.

ble "DataSet.:	dsx [Sheet1]" - Rows: 783	Spec - Columns: 2	9 Properties	Flow Variables														
Row ID	S Col0	S Col1	S Col2	,	\$ Col4	S Col5	S Col6	S Col7	S Col8	S Col9	S Col10	S Col11	S Col12	S Col13	S Col14	S Col15	S Col 16	S Col17
Row0	citrus fruit	semi-finished bread	margarine	ready soups	?	?	?	?	?	?	?	?	?	,	?	?	?	?
Row1	tropical fruit		coffee	?	?	?	,	?	?	?	?	?	?	?	?	?	?	,
Row2	whole milk	7	7	?	7	?	?	?	?	?	7	?	7	?	2	?	?	,
Row3	pip fruit	yogurt	cream cheese	meat spreads	?	?	,	?	?	?	7	?	?	?	?	?	?	,
Row4	other vegetables	-	condensed milk		?	?	,	?	?	?	7	?	7	?	2	?	?	,
Row5	whole milk	butter			abrasive de	7	,	7	?	7	7	2	7	7	7	?	7	,
Row6	rolls/buns	7	?	?	?	?	,	?	?	?	7	?	?	?	?	?	?	,
Row7	other vegetables	UHT-milk	rolls/buns	bottled beer	liquor (appe	?	,	7	?	7	7	2	7	?	7	?	?	,
Row8	pot plants	7	7	7	7	,	,	7	?	7	7	2	7	7	7	?	7	,
Row9	whole milk	cereals	7	,	,	,	,	7	?	,	,	2	7	7	7	7	,	,
Row10	tropical fruit		white bread	bottled water	chocolate	,	,	,	,	,	,	2	,	,	7	7	,	,
Row11	citrus fruit	tropical fruit			curd	yogurt	flour	bottled water	dishes	7	,	2	7	,	2	7	,	,
Row12	beef	7	7	)	7	)	)	7	)	,	,	,	7	,	,	,	7	,
Row13	frankfurter	rolls/buns	soda	,	,	,	,	7	7	7	,	,	7	,	,	,	7	,
Row14	chicken	tropical fruit	)	,	,	,	,	7	,	7	7	,	7	,	,	,	7	,
Row15	butter		fruit/vegeta	newenanere	)	)	)	)	)	)	)	2	2	)	2	2	)	)
Row16	fruit/vegetable juice	)	)	)	)	)	)	)	)	)	)	2	)	,	)	2	)	)
Row17	packaged fruit/veget	)	)	)	)	)	)	)	)	)	)	2	)	)	)	)	)	)
Row18	chocolate	)	)	)	)	)	)	)	;	)	)	2	)	)	2	2	)	)
Row19	specialty bar	)	1	2	2	2	2	2	1	2	r 2	2	2	2	2	2	2	1
Row20		)	2	2	2	2	,	2	2	2	2	2	5	2	2	2	2	1
Row21	other vegetables butter milk	f nachry	1	2	1	5	1	1	1	2	2	2	5	5	2	2	2	1
Row22	whole milk	pastry	1	2	1	1	1	1	<u>ا</u> د	f 2	ŗ	2	ŗ	5	2	?	f 2	1
Row23		· · · · · · · · · · · · · · · · · · ·	·	f datasant	<u>f</u>	,		1	? 1	,	,	2	,	,		?	,	,
	tropical fruit		processed c		newspapers	f 4	f	f a anlite annule	···	f anadu	f hathroom d	2	ŗ	5	2	7	r 2	1
Row24	tropical fruit		other veget	rrozen dessert	rolls/buns	flour	sweet spread	s saity snack	waffles	candy	bathroom d				!			
Row25	bottled water	canned beer	1	,	1	1		1	? 1	,	,	?	,	,		?		1
Row26	yogurt	rala horas	· · ·	f decelete				1	·	(	· ·				!			
Row27	sausage	rolls/buns	soda	chocolate			!	(	<u> </u>	/	/	/		/	/	/	/	/
Row28	other vegetables	/	f == 1	/	/	!	!	!	<u> </u>	/	/	/		/	/	/	/	/
Row29	brown bread		fruit/vegeta		newspapers	shopping bags	?	?	?	?	?	?	?	?	?	?	?	?
Row30	yogurt		bottled water		l of the	<u>/</u>	<u>/</u>	?	<u>/</u>	?	?	/	!	!	?	?	?	?
Row31	hamburger meat	other vegetables			bottled water	hygiene arti	napkins	?	?	?	?	?	?	?	?	?	?	?
Row32	root vegetables	other vegetables			sugar	?	?	?	?	?	?	?	?	?	?	?	?	?
Row33	pork		other veget	whole milk	whipped/so	artif. sweet	soda	abrasive de	?	?	?	?	!	?	?	?	?	?
Row34	beef		detergent	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row35	pastry	soda	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row36	fruit/vegetable juice	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row37	canned beer	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row38	root vegetables	other vegetables		dessert	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row39	citrus fruit		newspapers	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row40	sausage			canned beer				?	?	?	?	?	?	?	?	?	?	?
Row41	tropical fruit	_	whole milk	yogurt	domestic eggs	brown bread	pastry	sugar	cereals	coffee	soda	waffles	candy	?	?	?	?	?
Row42	berries	yogurt	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?
Row43	canned beer	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?	?

After we applied data preprocesses, then we applied FPGrowth and TANIMOTO algorithms to see the results of analysis. We have chosen minimum support values as 5% and 4% and minimum set size as 1.



In the results below, we are able to get some of information according to the result. Firstly, let's consider "ItemSetSize" and sort them descending order. If we look Row 25,Row 27,Row 28 and Row 30,we are able to say that if the market sells tropical fruit and whole milk together,4.263 percent of customers will buy and if we check "ItemSetSupport" for Row 25 is exist 334 times in all data set as a frequent item set. We can consider what we need to see on this results and show them in the results section. First photo represents the results of FPGrowth and second is TANIMOTO algorithm shows "similarity" which is used to check the model's robustness.



Row ID	[] ItemSet	ItemSetSize	ItemSetSupport	D RelativeItemSetSupport%	D AbsoluteItemCoverSimilarity	D RelativeItemCoverSimilarity%	
Row0	[whole milk,other vegetabl	2	577	7.364	0.538	53.793	
Row1	[whole milk,rolls/buns]	2	451	5.756	0.509	50.92	
Row2	[whole milk,yogurt]	2	445	5.68	0.564	56.372	
Row3	[whole milk]	1	2002	25.552	1	100	
Row4	[other vegetables]	1	1506	19.221	1	100	
Row5	[rolls/buns]	1	1448	18.481	1	100	
Row6	[yogurt]	1	1074	13.708	1	100	
Row7	[soda]	1	1390	17.741	1	100	
Row8	[root vegetables]	1	862	11.002	1	100	
Row9	[tropical fruit]	1	816	10.415	1	100	
Row10	[bottled water]	1	873	11.142	1	100	
Row11	[sausage]	1	730	9.317	1	100	
Row12	[citrus fruit]	1	655	8.36	1	100	
Row13	[shopping bags]	1	758	9.675	1	100	
Row14	[pastry]	1	683	8.717	1	100	
Row15	[whipped/sour cream]	1	570	7.275	1	100	
Row16	[pip fruit]	1	598	7.632	1	100	
Row17	[fruit/vegetable juice]	1	569	7.262	1	100	
Row 18	[domestic eggs]	1	500	6.382	1	100	
Row19	[newspapers]	1	624	7.964	1	100	
Row20	[butter]	1	437	5.578	1	100	
Row21	[margarine]	1	465	5.935	1	100	
Row22	[brown bread]	1	509	6.496	1	100	
Row23	[bottled beer]	1	638	8.143	1	100	
Row24	[frankfurter]	1	455	5.807	1	100	
Row25	[curd]	1	417	5.322	1	100	
Row26	[pork]	1	447	5.705	1	100	
Row27	[coffee]	1	469	5.986	1	100	
Row28	[beef]	1	399	5.093	1	100	

8.015

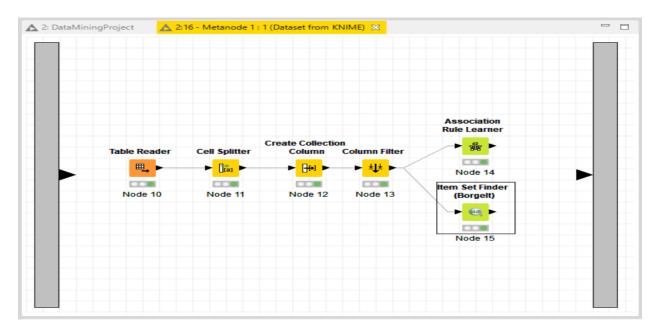
🛕 Item Sets - 0:17:3 - Item Set Finder (Borgelt)

File Hilite Navigation View

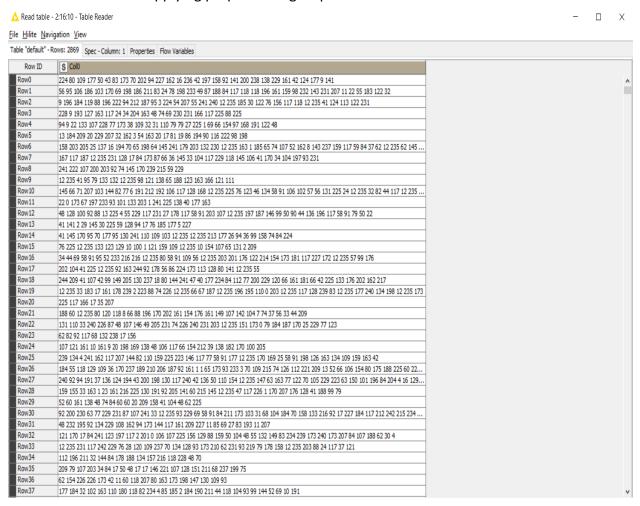
Row29

[canned beer]

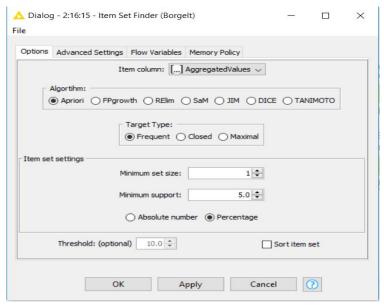
Now, we are going to explain the second part of our project and it looks like in the figure below.

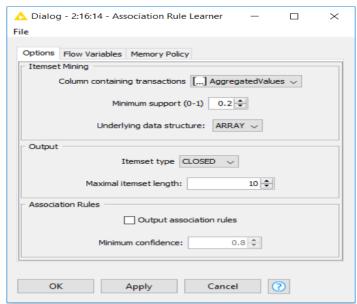


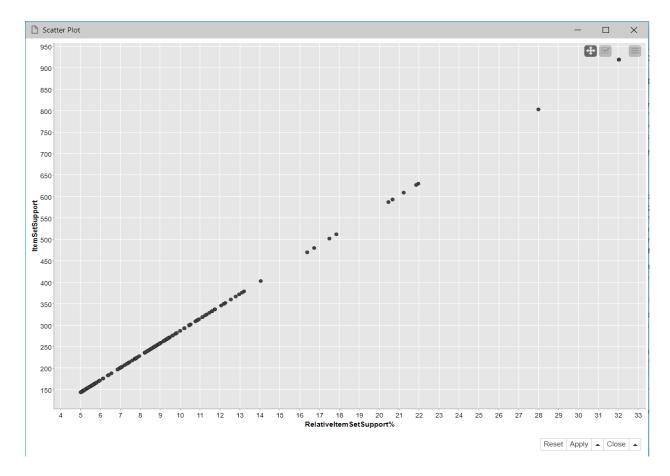
## Our data set before applying preprocessing steps look like that:

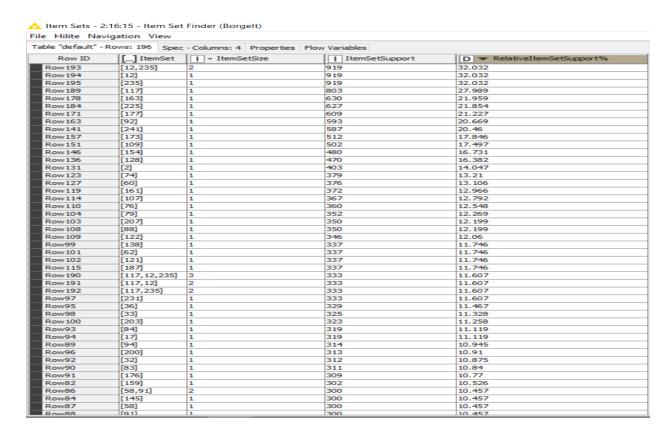


After uploading the data set on KNIME, we used "CellSplitter" to make these data which was separated by blanks into different values them and combine them to make an array to apply algorithms through "CreateCollectionColumn". As an one more step of preprocessing, we used "ColumnFilter" not to consider the first column of data set which looks like the column above. In this project, we used "AssociationRuleLearner" which learns itself frequent itemsets and association rules and "Apriori" algorithm to see the results of data analysis after all used preprocessing steps.









The results of data analysis above shows the results of Apriori algorithm. If we look "ItemSet" column,we see that some of item sets which are greater than 1 have high "RelativeItemSetSupport". It shows that if we sell items [12,235] together,32.032 percent of customers buy that or if we sell items [117,12,235] together, 11.607 percent of customers buy. If we check the results from "Association Rule Learner", we get the same results after applying Apriori algorithm to the data set.

🛕 Frequent ite	msets/Associati	on rules - 0:16:14 - Association Rule Learner
ile Hilite Navi	gation View	
Table "default" - R	Rows: 43 Spec -	- Columns: 2 Properties Flow Variables
Row ID	D ▼ Sup	[] Items
item set 42	0.32	[235, 12]
item set 41	0.28	[117]
item set 40	0.22	[163]
item set 39	0.219	[225]
item set 38	0.212	[177]
item set 37	0.207	[92]
item set 36	0.205	[241]
item set 35	0.178	[173]
item set 34	0.175	[109]
item set 33	0.167	[154]
item set 32	0.164	[128]
item set 31	0.14	[2]
item set 30	0.132	[74]
item set 29	0.131	[60]
item set 28	0.13	[161]
item set 27	0.128	[107]
item set 26	0.125	[76]
item set 25	0.123	[79]
item set 23	0.122	[207]
item set 24	0.122	[88]
item set 22	0.121	[122]
item set 18	0.117	[138]
item set 19	0.117	[187]
item set 20	0.117	[62]
item set 21	0.117	[121]
item set 16	0.116	[231]
item set 17	0.116	[117,235,12]
item set 15	0.115	[36]
item set 14	0.113	[33]
item set 13	0.113	[203]
item set 11	0.111	[84]
item set 12	0.111	[17]
item set 10	0.109	[94]
item set 9	0.109	[200]
item set 8	0.109	[32]
item set 7	0.108	[83]
item set 6	0.108	[176]
item set 5	0.105	[159]

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