APPLICATION OF DATA MINING TO DETERMINE PRODUCT ADMINISTRATION USING APRIORI ALGORITHM

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ABSTRACT

Product Arrangement is a way of arranging products to attract consumers. Product arrangement is also known as display. But in fact there are still many entrepreneurs who have not paid attention to the arrangement of their products. So that many consumers are not interested in buying products because the arrangement is less attractive which makes the company's income does not increase. Therefore researchers have a solution to solve the problems that exist in the shop. That is, with the application of data mining to determine the arrangement of products using a priori algorithm, the results can be used as a company guide to determine the display of goods and as a guide to promote goods that are less sold to participate or sell quickly. Based on the results of the trials that have been carried out, it is known that the a priori algorithm is used to help determine the product structuring solution from the sales data of Warung Sayur Segar products so that it can later be used as a consideration in determining an effective structuring and sales strategy.

Keywords: Data Mining, Apriori Algorithm, Product Arrangement

1. Introduction

Product Arrangement is a way of structuring products to attract consumer interest. Product arrangement is also called display. But in fact there are still many entrepreneurs who have not paid attention to the arrangement of their products. So that many consumers are not interested in buying products because of the less attractive arrangement that makes the company's income does not increase. Tarigan's research also explains that companies must understand what consumers are interested in determining which items are needed by consumers (Tarigan, 2017).

Based on the above problems, researchers conducted research with a concept such as a mini market called Warung Sayur Segar which sells various kinds of raw vegetable products. In addition to raw vegetable

products, Warung Sayur Segar also sells fresh seasoning products, fruits, fresh fish, fresh chicken meat, fresh beef, and others.

Data Mining is a term used to extract knowledge in a database. According to Kusrini and Lutfi (2009), Data Mining is a term used in the database. One method that exists in data mining is a priori algorithm. The a priori algorithm includes the types of association rules for data mining. Rules of association or association rule mining is the method used to find associative autran between item combinations. Examples of associative rules that we can apply in everyday life are application in supermarkets. Supermarket owners can find out how likely a buyer is to buy instant noodles along with sausages. With the information or knowledge obtained, if instant noodles and sausages are known that both are often bought together, the supermarket owner can make the placement of the two products close together so the buyer will easily take both products and the self-owner gets higher profits after applying the association rules.

Based on the description above, the researcher has a solution to solve the problems that exist in the shop. Namely with the application of data mining to determine product arrangement using a priori algorithm, the results can be used as company guidelines to determine the display of goods and as a guideline for promoting goods that are not sold well so they can be sold or sold quickly. In addition, customer satisfaction and profits obtained by the company can be maximized and the risk of loss of unsold products will decrease.

2. SYSTEM PLANNING

A. FLOWCHART

Figure 2.1 is a system flowchart that describes the flow of the a priori algorithm method system that is applied to the program.



Figure 2.1 Flowchart of the Apriori Algorithm System

The flow in Figure 2.2 explains that Product Ranking is started from entering Sales Data into the program. Then the data is imported into the database. Then the next process is carried out, namely the process of calculating an item, that is if there is data in data A four times and then there is data B (A, A, A, A, B), then the program will count items one by one and if there is the same data then the data is not raised but is put together with the same data. So the program will display A with the number three (3) and B with the number one (1). After going through the above process, the program issues the product ranking table.



Figure 2.2 Product Ranking System Flowchart

The flow in the system for obtaining a structuring solution explains the process of form rule (pattern) by calculating the value of confidence in the figure 2.1 of a priori algorithm system flowchart. The flow starts then there is the input data in the form of the last k-itemset, which is the result of the last loop and there is no looping again. The k-itemset data is then calculated using the confidence search formula, namely by calculating the total items purchased at once divided by how many transactions bought the first or antecedent item. For example If you buy A then buy B, then how to calculate the value of confidence is if the number of If buy A then buy B there are 3 transactions, while the number of transactions that buy A is as many as 4 transactions, then the formula becomes $3/4 \times 100\% = 75\%$. After the process occurs, the results appear in the form of confidence values and also a structuring solution.



Figure 2.3 Flowchart of the Acquiring Solution Solution System

The program flowchart explains the course of the program. The program is run by the Shop Admin.



B. Data Flow Diagram (DFD)

In Figure 2.6 it is illustrated that DFD Level 0 System has an External entity Shop Admin and System Process Determining Product Arrangement. Between them there are 5 (five) entries from the Shop Admin and 5 (five)

outputs from the System. Enter provided by the Shop Admin to the System is Login Data, Sales Data, Date Range, Min Support and Min Confidence, and History Calculation. While the output obtained by the Shop Admin from the System is Access Rights, View, Display Data, Calculations, and Analysis Results.



Figure 2.6 DFD Level 0 System

Figure 2.7 illustrates the flow of the DFD Level 1 System. Inside there are 1 (one) Calculation, and Calculation Results, and 7 (seven) Data Stores namely Users, Transactions, Process Log, Itemset1, Itemset2, Itemset3, and Confidence.



Figure 2.7 DFD Level 1 System

C. Entity Relationship Diagram (ERD)

Figure 2.8 is a database design or Entity Relationship Diagram (ERD) Database Program. In it there are seven tables consisting of four Fact type tables and three Dimension type tables. Called the Fact table because in it a transaction occurs and there are also attributes or entities that have something measured such as Amount, Amount_A, and Amount_B. While the Dimension table is a table that is the same as the master table, namely a table whose entries have not changed or remain. The following is an explanation per ERD table in table form.



Figure 2.8 ERD Database Program

3. RESULTS AND DISCUSSION

A. SYSTEM IMPLEMENTATION

The results of the implementation of the Best Selling Product Selection system at Warung Sayur Segar Store consist of several pages, namely the Home page, Login, Home (new), Sales Data, Analysis, and Analysis Results. Following is the implementation of pages that can be accessed by users.





() localis	est/status_pron=ss/index.php?m	alah-gup berkantu al 🖉 unu-sil unda n	-			w -
	APLIKASI DATA N	MINING UNTUK PEMILIH STUDI KA	AN PRODU SUS: WAR	K TERLARIS MENGO UNG SAYUR SEGAR	SUNAKAN ALI	GORITMA APRIORI
			HOME	Devisionado.	ANALISA	HASLANAJISA COGDUT
Impo	ort Data (XLS)					
Pah Fie	Tidak ada fic yang da lili					
Conceptor State						
8)mpas	data por ala ber neocharwijia					
Jumish cets	Tanogal	Produk				
1	2018-07-01	BAWANG BOMBAY				
	2014-07-01	BAWAND MERAH				
	2018 01 01	ROWING MERAN				
3	gars 16-04 10 1					

Figure 3.4 Implementation of Sales Data Pages

	APLIKASI DATA	MINING UNTUK PEMILIHA	N PRODUK TERLARIS MENGGUNAKAN ALGORITMA APRIOR	u
		STUDI KAS	JS: WARUNG SAYUR SEGAR	
			HONE DATA PENJUALAN SAGALEA PAGE ANALISA	LOGOU
Ana	isa			
Catatan: Tentukan	Min Support / Minimal transa	kal behang yang sicari kemudian tentuk	ar Min Confidence / Minimal Keyakinannya	
.9			m 04/97/2019 - 25/07/2019	
4			Scarch	
- THE OWNER				
and the second s	han			
No	Tanggal	Produk		
1	2016-07-01	BAWANG BOMBAY		
2	2010-07-04	EXVENIS MERIAL		



Figure 3.6 Implementation of Analysis Results Page

B. Calculation

Data obtained from the company is sales data in the form of an excel file in which there are several columns, namely columns No, Date, Note, Goods, Amount, Price, Discount, and Total Price. Following is the sales data table:

							ΤΟΤΑ
N O	TANGG AL	ΝΟΤΑ	BARANG	JUML AH	HAR GA	DISK ON	L HAR GA
1	01/07/20	PJ180701	BAWAN	3	13,00	0	39,00
	18	010	G		0		0
-			BOMBAY				
2	01/07/20	PJ180701	BAWAN	12	15,00	0	180,0
	18	009	G		0		00
2	01/07/20	D 11 00701		Б	15.00	0	75.00
3	01/07/20 18	PJ160701	G	J	15,00	0	75,00
	10	010	G MERAH		0		0
4	01/07/20	P.1180701	BAWAN	10	15.00	0	150.0
•	18	010	G	10	0	Ū	00
		0.0	MERAH		Ũ		
5	01/07/20	RJ180701	BAWAN	-8,15	22,00	0	-
	18	002	G		0		179,3
			MERAH				00
6	01/07/20	PJ180701	BUAH	15	17,00	0	255,0
	18	006	NAGA A		0		00
7	01/07/20	PJ180701	BUAH	25	17,00	0	425,0
_	18	007	NAGA A		0		00
8	01/07/20	PJ180701	JAHE	3	10,00	0	30,00
•	18	009	EMPRIT	•	0		0
9	01/07/20	PJ180701	JAHE	3	10,00	0	30,00
10	18	010 D 14 00704		10	0 4 050	0	0 40 50
10	01/07/20	PJ180701		10	4,250	0	42,50
	10	010					0
11	01/07/20	P 1180701		0.25	48.00	0	12 00
	18	008		0,20	40,00 0	U	12,00
12	01/07/20	PJ180701	KETAN	10	13.00	0	130.0
. –	18	009	PUTIH B		0	5	00
13	01/07/20	PJ180701	KETAN	15	13,00	0	195,0

Table 3.1 Sales Data

	18	010	PUTIH B		0		00
14	01/07/20	PJ180701	KETUMB	2	5,000	0	10,00
	18	008	AR				0
			BUBUK				
4 -	04/07/00	B 1400704	DESAKU	•	~~~~	0	
15	01/07/20	PJ180701	KRESEK	3	30,30	0	90,90
	18	001	LOS		0		0
4.0	04/07/00	B 1400704	PUTIHA		0 0 0 0	•	0 0 0 0
16	01/07/20	PJ180701	KUNCI	1	6,000	0	6,000
47	18	010				•	=
17	01/07/20	PJ180701	KUNYII	1	5,000	0	5,000
4.0	18	010		0.5	4 0 0 0	0	0 000
18	01/07/20	PJ180701	LAUS	0,5	4,000	0	2,000
40	18	010		-	44.00	0	70.00
19	01/07/20	PJ180701	LENIO	5	14,00	0	70,00
• •	18	009	BESAR	_	0		0
20	01/07/20	PJ180701	LENIO	5	14,00	0	70,00
	18	010	BESAR		0		0

Table 3.2 Sales Data Normalized

N O	TANG GAL	ΝΟΤΑ	BARANG	JUML AH	HAR GA	DISK ON	L HAR
1	01/07/ 2018	PJ18070 1010	BAWANG BOMBAY,BA WANG MERAH,BA WANG MERAH,JAH E EMPRIT,KE CAP TJAPAR 140ML,KETA N PUTIH B,KUNCI,KU NYIT,LAOS, LENTO BESAR,	3	13,00 0	0	39,00 0
2	01/07/ 2018	PJ18070 1009	BAWANG MERAH	12	15,00 0	0	180,0 00

3	01/07/	PJ18070		5	15,00	0	75,00
	2018	1010			0		0
4	01/07/	PJ18070		10	15,00	0	150,0
	2018	1010			0		00
5	01/07/	RJ18070	BAWANG	-8,15	22,00	0	-
	2018	1002	MERAH	,	0		179,3
							00
6	01/07/	PJ18070	BUAH NAGA	15	17.00	0	255.0
	2018	1006	A		0		00
7	01/07/	PJ18070	BUAH NAGA	25	17.00	0	425.0
	2018	1007	A		0	-	00
8	01/07/	PJ18070	JAHE	3	10.00	0	30.00
•	2018	1009	EMPRIT.KET	•	0	Ū	0
					· ·		· ·
			BIENTO				
			BESAR				
9	01/07/	PJ18070		3	10.00	0	30.00
•	2018	1010		•	0	Ū	0
1	01/07/	PJ18070		10	4.250	0	42.50
0	2018	1010			.,_00	Ũ	,00
1	01/07/	P.118070	KEMIRI KET	0 25	48 00	0	12 00
1	2018	1008	UMBAR	0,20	0	Ũ	,
•	2010		BUBUK		Ŭ		U
			DESAKU				
1	01/07/	PJ18070		10	13.00	0	130.0
2	2018	1009			0		00
1	01/07/	PJ18070		15	13.00	0	195.0
3	2018	1010		-	0	-	00
1	01/07/	PJ18070		2	5.000	0	10.00
4	2018	1008		_	-,	-	0
1	01/07/	PJ18070	KRESEK	3	30.30	0	90.90
5	2018	1001	LOS PUTIH	•	0	Ū	0
Ū	2010		A		Ŭ		Ũ
1	01/07/	PJ18070		1	6.000	0	6.000
6	2018	1010		-	-,	-	-,
1	01/07/	PJ18070		1	5.000	0	5.000
7	2018	1010		•	2,200	Ŭ	2,000
1	01/07/	PJ18070		0.5	4.000	0	2.000
8	2018	1010		0,0	.,	Ŭ	_,500
1	01/07/	PJ18070		5	14.00	0	70.00
9	2018	1009		Ű	0	Ũ	0
2	01/07/	PJ18070		5	14 00	0	70 00
0	2018	1010		Ŭ	0	Ũ	0
					v		

Table 3.3 Itemset 1

No	Item	Jumlah	Support	
1	BAWANG BOMBAY	1	5	Tidak Lolos
2	BAWANG MERAH	3	15	Lolos
3	JAHE EMPRIT	2	10	Lolos
4	KECAP TJAPAR 140ML	1	5	Tidak Lolos
5	KETAN PUTIH B	2	10	Lolos
6	KUNCI	1	5	Tidak Lolos
7	KUNYIT	1	5	Tidak Lolos
8	LAOS	1	5	Tidak Lolos
9	LENTO BESAR	2	10	Lolos
10	BUAH NAGA A	2	10	Lolos
11	KEMIRI	1	5	Tidak Lolos
12	KETUMBAR BUBUK DESAKU	1	5	Tidak Lolos
13	KRESEK LOS PUTIH A	1	5	Tidak Lolos

Table 3.4 Items 1 that pass

No	Item	Jumlah	Support				
1	BAWANG MERAH	3	15				
2	JAHE EMPRIT	2	10				
3	KETAN PUTIH B	2	10				
4	LENTO BESAR	2	10				
5	BUAH NAGA A	2	10				

		Table 3.5 Iter	nset 2		
Ν			Jumla	Suppor	
0	Item1	Item2	h	t	
1	BAWANG MERAH	JAHE EMPRIT	1	5	Tidak Lolos
2	BAWANG MERAH	KETAN PUTIH B	1	5	Tidak Lolos
3	BAWANG MERAH	LENTO BESAR	1	5	Tidak Lolos
4	BAWANG MERAH	BUAH NAGA A	0	0	Tidak Lolos
5	JAHE EMPRIT	KETAN PUTIH B	2	10	Lolos
6	JAHE EMPRIT	LENTO BESAR	2	10	Lolos
7	JAHE EMPRIT	BUAH NAGA A	0	0	Tidak Lolos
8	KETAN PUTIH B	LENTO BESAR	2	10	Lolos
9	KETAN PUTIH B	BUAH NAGA A	0	0	Tidak Lolos
10	LENTO BESAR	BUAH NAGA A	0	0	Tidak Lolos

		Table 3.6 Itemset 2 that passes		
No	Item1	ltem2	Jumlah	Support
1	JAHE EMPRIT	KETAN PUTIH B	2	10
2	JAHE EMPRIT	LENTO BESAR	2	10
3	KETAN PUTIH B	LENTO BESAR	2	10

Table 3.7 Itemset 3 Ν Jumla Suppo 0 ltem1 ltem2 Item3 h rt KETAN PUTIH JAHE LENTO Lolo 2 10 1 EMPRIT BESAR В s

		Table 3.8 Iter	n 3 that passe	es	
Ν				Jumla	Suppor
0	ltem1	ltem2	ltem3	h	t
	JAHE	KETAN PUTIH	LENTO		
1	EMPRIT	В	BESAR	2	10

	Table 3.9 Confidence of itemset 3					
No	X => Y	Support X U Y	Support X	Confidence	Keterangan	
1	JAHE EMPRIT, KETAN PUTIH B => LENTO BESAR	10	10	100	Lolos	
2	KETAN PUTIH B, LENTO BESAR => JAHE EMPRIT	10	10	100	Lolos	
3	LENTO BESAR, JAHE EMPRIT => KETAN PUTIH B	10	10	100	Lolos	
4	JAHE	10	10	100	Lolos	

EMPRIT => LENTO BESAR, KETAN					
PUTIH B					
KETAN PUTIH B =>					
JAHE EMPRIT.	10	10	100	Lolos	
LENTO					
LENTO					
BESAR => KETAN	10	10	100		
PUTIH B, JAHE	10	10	100	LOIOS	
EMPRIT					
Table 3.10 Confidence of itemset 2					
X => Y	Support X U Y	Support X	Confidence	Keterangan	
JAHE					
EMPRIT =>	10	10	100	Lolos	
PUTIH B					
JAHE	10	10	100	Lolos	
EMPRIT					
EMPRIT =>	10	10	100	Lolos	
LENTO BESAR	-	-			
LENTO					
JAHE	10	10	100	Lolos	
EMPRIT KETAN					
BUTULB					
PUTH B =>	10	10	100	Lolos	
PUTIH B => LENTO BESAR	10	10	100	Lolos	
ENTO BESAR LENTO BESAR =>	10	10	100	Lolos	
	EMPRIT => LENTO BESAR, KETAN PUTIH B KETAN PUTIH B KETAN PUTIH B => JAHE EMPRIT, LENTO BESAR LENTO BESAR => KETAN PUTIH B, JAHE EMPRIT => KETAN PUTIH B KETAN PUTIH B KETAN PUTIH B => JAHE EMPRIT => LENTO BESAR EMPRIT => LENTO BESAR => JAHE EMPRIT =>	EMPRIT => LENTO BESAR, KETAN PUTIH B KETAN PUTIH B => JAHE EMPRIT, LENTO BESAR LENTO BESAR => KETAN PUTIH B, JAHE EMPRIT Table 3.10 Cor X => Y Support X U Y JAHE EMPRIT => KETAN PUTIH B => 10 KETAN PUTIH B => 10 KETAN PUTIH B => 10 EMPRIT JAHE EMPRIT => 10 EMPRIT JAHE EMPRIT => 10 EMPRIT JAHE EMPRIT => 10 EMPRIT JAHE EMPRIT => 10 EMPRIT JAHE EMPRIT => 10 EMPRIT JAHE EMPRIT => 10 BESAR LENTO BESAR => 10 BESAR => 10	EMPRIT => LENTO BESAR, KETAN PUTIH B KETAN PUTIH B => JAHE 10 10 EMPRIT, 10 10 EMPRIT, 10 10 BESAR LENTO BESAR => KETAN 10 10 PUTIH B, 10 10 PUTIH B, 10 10 EMPRIT => 10 10 PUTIH B => 10 10 FUTIH B => 10 10 EMPRIT JAHE EMPRIT => 10 10 EMPRIT JAHE EMPRIT => 10 10 EMPRIT JAHE EMPRIT => 10 10 BESAR => 10 10 10 BESAR => 10 10	EMPRIT => LENTO BESAR, KETAN PUTIH B KETAN1010PUTIH B => JAHE1010100EMPRIT, LENTO BESAR LENTO BESAR => KETAN1010100BESAR => KETAN LENTO10100100BESAR => KETAN LENTO10100100BESAR => KETAN LEMPRIT1010100JAHE EMPRITTable 3.10 Confidence of itemset 22X => YSupport X U YSupport XConfidenceJAHE EMPRIT => NETAN JAHE1010100PUTIH B KETAN PUTIH B => JAHE1010100EMPRIT JAHE EMPRIT => JAHE1010100EMPRIT JAHE EMPRIT => 1010100100BESAR LENTO BESAR => LENTO10100100BESAR => LENTO BESAR => LENTO10100100BESAR => LENTO10100100BESAR => LENTO10100100BESAR => LENTO10100100BESAR => LENTO10100100BESAR => LENTO10100100	

|--|

No	X => Y	Confidence
1	JAHE EMPRIT, KETAN PUTIH B => LENTO BESAR	100
2	KETAN PUTIH B, LENTO BESAR => JAHE EMPRIT	100
3	LENTO BESAR, JAHE EMPRIT => KETAN PUTIH B	100
4	JAHE EMPRIT => LENTO BESAR, KETAN PUTIH B	100
5	KETAN PUTIH B => JAHE EMPRIT, LENTO BESAR	100
6	LENTO BESAR => KETAN PUTIH B, JAHE EMPRIT	100
7	JAHE EMPRIT => KETAN PUTIH B	100
8	KETAN PUTIH B => JAHE EMPRIT	100
9	JAHE EMPRIT => LENTO BESAR	100
10	LENTO BESAR => JAHE EMPRIT	100
11	KETAN PUTIH B => LENTO BESAR	100
12	LENTO BESAR => KETAN PUTIH B	100

Table 3.12 Arrangement Solutions

If the customer buys JAHE EMPRIT, KETAN PUTIH B, then the

1 customer will also buy LENTO BESAR.

Solution: Place it adjacent

If the customer buysKETAN PUTIH B, LENTO BESAR, then the

2 customer will also buy JAHE EMPRIT.

Solution: Place it adjacent

If the customer buysLENTO BESAR, JAHE EMPRIT, then the

3 customer will also buy KETAN PUTIH B.

Solution: Place it adjacent

If the customer buysJAHE EMPRIT, then the customer will also

- 4 buyLENTO BESAR, KETAN PUTIH B.
 - Solution: Place it adjacent If the customer buysKETAN PUTIH B, then the customer will also buy
- JAHE EMPRIT, LENTO BESAR.
 Solution: Place it adjacent
 If the customer buysLENTO BESAR, then the customer will also buy
- 6 KETAN PUTIH B, JAHE EMPRIT. Solution: Place it adjacent
- 7 If the customer buysJAHE EMPRIT, then the customer will also buy

KETAN PUTIH B. Solution: Place it adjacent If the customer buysKETAN PUTIH B, then the customer will also buy JAHE EMPRIT. 8 Solution: Place it adjacent If the customer buysJAHE EMPRIT, then the customer will also buy LENTO BESAR. 9 Solution: Place it adjacent If the customer buysLENTO BESAR, then the customer will also buy 10 JAHE EMPRIT. Solution: Place it adjacent If the customer buysKETAN PUTIH B, then the customer will also buy LENTO BESAR. 11 Solution: Place it adjacent If the customer buysLENTO BESAR, then the customer will also buy

12 KETAN PUTIH B. Solution: Place it adjacent

4. CONCLUSIONS AND RECOMMENDATIONS

A. CONCLUSIONS

Based on the formulation of the problem in the research conducted by the researcher through the design and discussion in the previous chapters, the conclusions can be taken as follows:

1. The application of Data Mining to determine product arrangement using a priori algorithm is as follows:

- a) Obtain sales data from the research site. The sales data obtained is the original sales data from the research site.
- b) Normalize sales data according to system requirements.
 Namely by ignoring the contents of the column other than the

date column and item. Items are grouped according to the same note.

- c) Applying the a priori algorithm to the application.
- d) Enter normalized sales data into the application then begin the analysis process.
- e) The results of the analysis are obtained by the emergence of calculations and structuring solutions.

2. Display the Product Rank as an application value added in the following ways:

- a. The results obtained are product rankings by sorting the highest number of transactions to the smallest number of transactions.
- b. Ranking The top product is the product that has the most number of transactions, the lower the transaction that occurs in the product is less. By knowing the number of transactions for each product, the user can determine the exact layout and price.

B. RECOMMENDATIONS

The author realizes that in the Application of Data Mining Applications for the Selection of the Best-Selling Products Apriori algorithms still have some disadvantages, for this reason, if the research wants to continue, the authors would like to give some suggestions regarding important parts to be discussed, namely:

- For further research, it is expected that the determination of the value of min confidence can be determined automatically by the system. This is recommended so as not to confuse users and use applications more efficiently.
- 2. Based on the results of trials on 1000 sales data, the time taken is 30 minutes. For further research, it is recommended to use other algorithms that can overcome the weaknesses of a priori algorithm so that research becomes more perfect and the results of execution become faster.

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