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Role of Information
Systems in Industry 4.0



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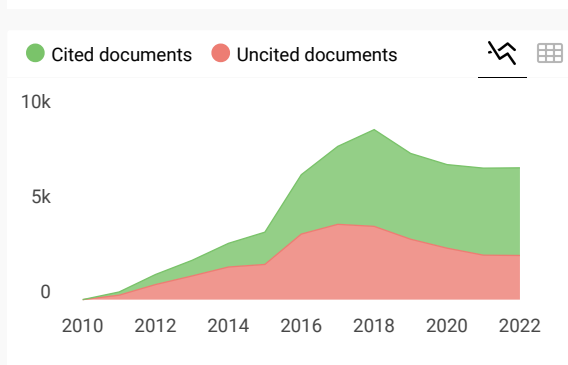
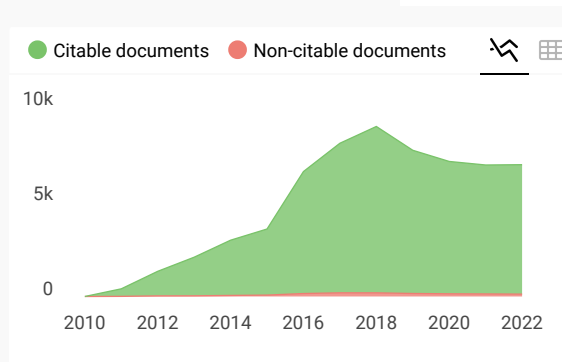
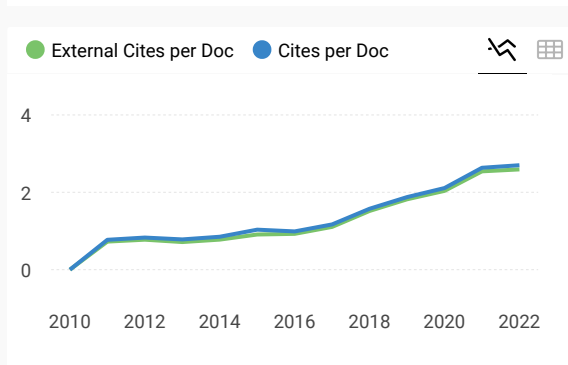
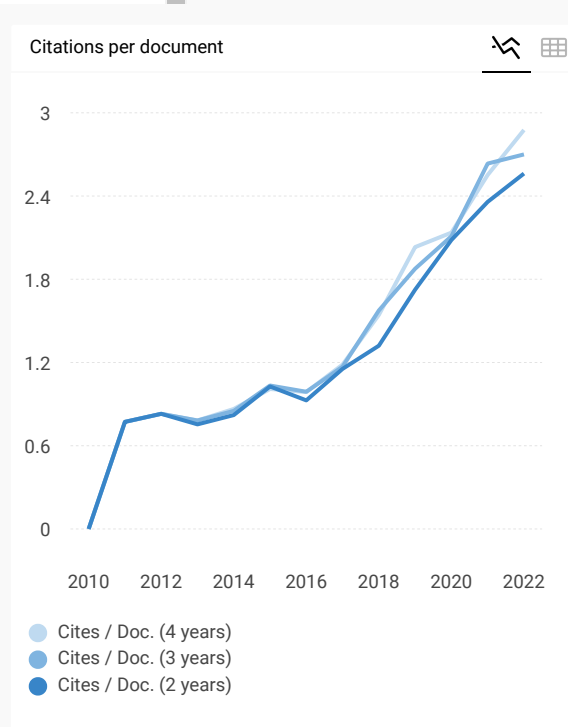
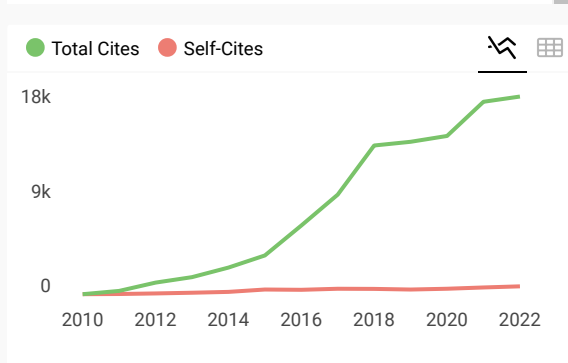
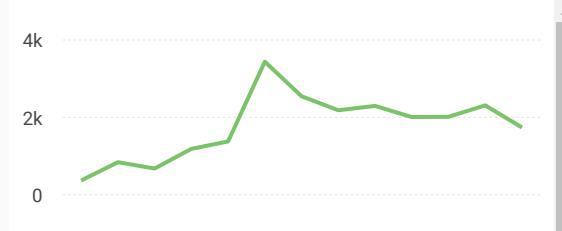
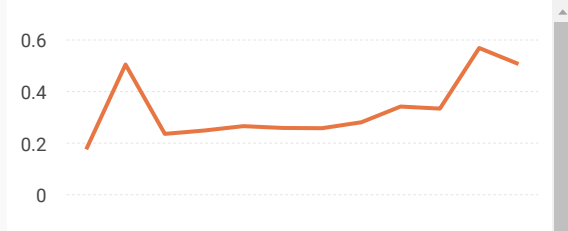
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Business Intelligence for Designing Restaurant Marketing Strategy: A Case Study

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Abstract

In this study we design marketing strategy for a restaurant in Surabaya. The restaurant is a café restaurant, which targets mid-low consumer. The main problem for that restaurant is there has never been a specific marketing strategy. Promotions that have been applied so far are not directed and have no impact on sales because they are not based on any strategy. The application of business intelligence with Power Business Intelligent (Power BI) is expected to be a way for restaurants to predict consumer's consumption patterns. Additionally, we also model the consumption pattern using Market Basket Analysis. Consumer's consumption patterns captured used for designing the right marketing strategy with the 4P (Product, Place, Price, Promotion) concept. It was found that some of the restaurant menu' had a unique relationship and could be used as a promotion to increase sales. Other results, in segment consumers there are children's and office worker. All information is supported using Power BI dashboard that expected to make the restaurant' executives easier to analyze changes in sales based on the occurrence of events and activities.

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Keywords: Business Intelligent; Market Basket Analysis; Power Business Intelligent; 4P concept; Restaurant

1. Introduction

Business owners in any field need to be one step ahead to face new wave consumers [1]. New wave consumers want something that is personalized and customized. Big data analysis makes it easier to find out consumer transaction

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patterns more specifically and personally. Market basket analysis with association rules helps owners know consumer's personal need and if captured can be used to design the right marketing strategy. The object used in this research is a restaurant in Surabaya with mid-low targeting. In the current condition there is never been a specific marketing strategy. The promotions that have been applied are not directed and have no impact on sales because they are not based on any marketing strategy. Moreover, this problem brought the menu variations not conceptualized. The number of menus offered is 106 that may cause confused consumer of ordering food and not optimal profit obtained [2].

The aim of this research is to design marketing design by predicting consumer transaction pattern. Pattern collected processed through one of big data analysis method, market basket analysis. The result will be visualized with dashboard, so the restaurant owner could analyze costumer transaction pattern and faster in act especially at making promotion.

2. Research methods

Business intelligence

Business intelligence is a set of techniques and tools to convert data into meaningful information for analyzing business by utilizing the use of technology and internet [3]. The purpose of implement business intelligence is to make it easier interpret human behavior such as transaction pattern.

Market basket analysis

Market Basket Analysis (MBA) is a method that helpful in the marketing environment for marketing strategies and business decisions. MBA is one of association rules application, knowing which products are purchased simultaneously will be very helpful for shopkeepers or other companies in managing their marketing strategies [4].

Association rules

Association rules is a data mining method to find out association rules between items combination. Determining an association rules, there are several interesting measures obtained from the results of processing data. Generally, there are three measures [5], support, confidence, and lift.

R-Software

Every transaction will be cleaned and processed with R software to look for consumer buying pattern. In this research, R software is a tool for mining data especially association rules [6]. R software makes it easier to make well designed plot, including symbols and mathematical formulas by typing the script and using several packages that help processing data instantly.

Power business intelligence

Power BI is internet-based software, help visualize and analyze raw data. The purpose of data visualization itself is to make it easier for users to get information and make decisions quickly [7]. In this research, Power BI also can be used for restaurant owner tool to make a strategic decision.

Marketing strategy

Marketing strategy is the base of company activities or marketing efforts to achieve the expected goals [8]. Marketing strategies have four marketing mixes which are often referred to as 4P (product, price, promotion, and place).

Research flow

The research flow starts from cleaning dataset, then developed the dashboard using PowerBI, constructing matrix incidence, determine the minimum support and minimum confidence, create association rules, analyse the result and then create the new marketing strategy.

3. Result and discussions

Consumer description

The restaurant consumer mostly are housewives at the age of 27 to 50. They came to shopping cooking ingredients and household needs at a hypermarket and department store. Another consumer are husbands and children look for a place to sit, order food and wait in the restaurant while their housewives are shopping. Other consumer is office worker and entrepreneur who attending economic seminar every Saturday and Sunday.

Data cleaning and aggregation

The raw data obtained in this research are transaction list from a restaurant in Surabaya. Raw data used are in the range from March 23 to June 30, 2018 and then September 23 to October 16, 2018. The restaurant's transaction list is influenced by several events. May 13 and 14 there were tragic bombing in several places in Surabaya. This causes Surabaya's people to feel insecure and prefer to stay at home, including to eat. The same range of data, on May 17 until June 16, 2018 is entering the fasting month. Data within these events will be analysed separately. Fifteen thousand eighty-seven recorded data are used in this research.

Aggregation variables are formed first to facilitate analysis of data. The addition of aggregate variables is the variable day of transactions (Days), whether transactions are carried out on weekends or weekdays (Weekend), whether transactions entered on week with high income or not (Weekly), month of transactions (Month), whether transaction entered on morning or night (Morning), food or beverage (Grouping), and food type (Type).

The results of data cleaning along with the aggregate variables are then processed using Power BI software. Power BI software can display the data summary results by simply typing the script to explore the background of a data with an easier-to-understand visualization. The data to be analysed are data without events of bombings and fasting months. Sales data will be processed on average and other sizes for later analysis. Consumer observations are also processed and analysed. The results of this analysis will be one of the inputs for designing a marketing strategy for the restaurant.

Dashboard business intelligence

It is important for a company to carry out continuous monitoring of its sales movements. The dashboard in Fig. 1(a), Fig. 1(b), and Fig. 1(c) was formed through Power BI software, is a tool for the restaurant executives to carry out monitoring and decision making. Promotion carried out cannot be separated from monitoring.

The first page of the dashboard provides a summary of general information on the restaurant. Important information that can be known is the comparison of profits obtained this month and last month or weekly. Comparison of the two given target limits. The target is determined based on the calculation of the monthly costs incurred by the restaurant. Besides that, it can also be seen that the information on the average ability of consumers in each transaction (willingness to pay) is Rp. 69,370. Seven of the most favourite menus in the restaurant can also be known as a reference for promotion. Distribution of consumer arrivals is also displayed so that the restaurant executives know the right time to do promotion or preparation.

The second page dashboard is a panel for analysing the restaurant menus. This panel will display a glimpse of the most favorite and not favorite menus outside the menu of Teh Manis (sweet tea), Nasi Putih (rice) and Air Putih (water). Using this panel, the executive can find the type of menu that is a favorite. Through this dashboard executive can find out which menu types are most purchased, so that they can be considered in the future to form a menu with that type.

Sales analysis base on occurrence of events can be done with a panel on the third page. Through the event list table along with the profit margins obtained can be known which activity provides high profit. The panel on this page will help executives find out what activities have a positive impact on sales and the possibility that they can be replicated to be carried out continuously or repeated. Building event data is inputted according to the building activity list, while local event data (the restaurant) has not been implemented. National holiday data is also included in the event analysis to find out consumption patterns.



Fig. 1. (a) General Summary Dashboard; (b) Food Reference Dashboard; (c) Event Analysis Dashboard.

3.3.1. Fasting month descriptive data analysis

There may be a difference in the menu ordered during the fasting month, especially at night when Muslims break their fast. Fig. 2(a) is the total food menu sales on May 16 to June 8, 2018 at night (04.00 PM - 10.00 PM). The hours of breaking the fast based on the Ministry of Religion of the Republic of Indonesia for the period 1439 H / 2018 M. Fig. 2(b) represents the total sales of beverage menu fasting night.

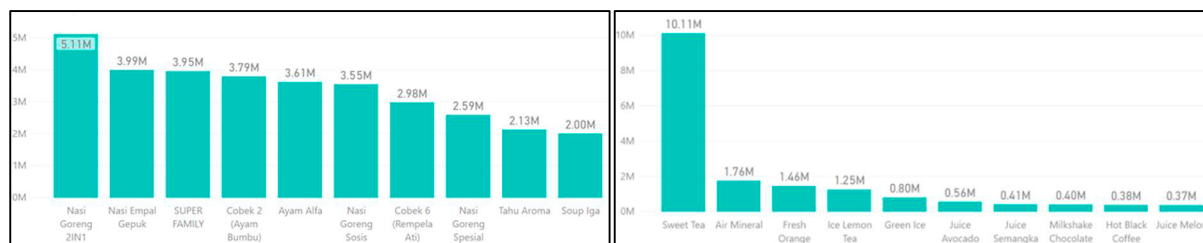


Fig. 2. (a) Total Sales of Food Menu Fasting Night; (b) Total Sales of Beverage Menu Fasting Night.

In Surabaya are 05.22 PM. The most ordered menu still Nasi Goreng 2IN1 followed by Nasi Empal Gepuk (rice and meat) and Super Family. The menu choices that consumers choose when breaking their fast in dishes that have side dishes, especially nutritious meat and eggs, even those that can be eaten with many people. Super Family is a Packet type food containing five pieces of fried chicken and rice, suitable for breaking fast with family or friends.

3.3. Market basket analysis

The output with aggregate variables then analysed by association rules to find out associative relationships between menus. The data used are data without events. This analysis done by using *arules* package in R software. The first step needed is delete duplicate data for each transaction with more than one type of menu name, because analysis using association rules requires only one for each type of menu. The incidence matrix contains relationships between two objects or items so that only one identity is needed for each object or subject.

Associative relationships can be done after forming incidence matrix that will show patterns of consumer habits in choosing menus. Several factors need to be determined, minimum support and minimum confidence. Minimum support and minimum initial confidence are set at 4% and 15% respectively as they tried values as be seen in Table 1. The value of support and confidence needs to be done by several times combination experiment to get the valuable output and used for analysis or improvement.

The results of 10 associative relationships from the value of the predetermined factors, all of which show the lhs menu will increase sales on the Teh Manis (sweet tea) menu. Matrix incidence association rules for Teh Manis (sweet tea) can be seen in Fig. 6. The results of the initial associative relationship still have not shown significant results, the reason being that the restaurant consumers who order any food menu also order menus for Teh Manis (sweet tea), Air Putih (water), and Nasi Putih (rice).

Table 1. Association Rules Support 4% and Confidence 15%.

	lhs	rhs	support	confidence	lift	Count
[1]	{Air Mineral, Ayam Goreng Nasi}	{Teh Manis}	0.0420354	0.8209877	1.248849	133
[2]	{Air Mineral, Nasi Goreng 2IN1}	{Teh Manis}	0.0414033	0.7891566	1.200429	131
[3]	{Nasi Putih, Tahu}	{Teh Manis}	0.0600506	0.781893	1.18938	190
[4]	{Cobek B, Nasi Putih}	{Teh Manis}	0.0420354	0.7777778	1.183212	133
[5]	{Air Mineral, Tahu}	{Teh Manis}	0.0448799	0.7759563	1.180349	142
[6]	{Ayam Goreng Nasi, Nasi Putih}	{Teh Manis}	0.0540455	0.7668161	1.166445	171
[7]	{Nasi Goreng 2IN1, Nasi Putih}	{Teh Manis}	0.0489886	0.7598039	1.155779	155
[8]	{Nasi Goreng, Telur}	{Teh Manis}	0.0840708	0.7578348	1.152783	266
[9]	{Nasi Empal Gepuk}	{Teh Manis}	0.1090392	0.7549234	1.14835	345
[10]	{Ice Lemon Tea, Nasi Putih}	{Teh Manis}	0.0401391	0.7514793	1.143116	127

* Teh Manis (Sweet Tea), Air Mineral (Water), Ayam Goreng (Fried Chicken), Nasi Goreng (Fried Rice), Nasi (Rice), Tahu (Tofu), Empal (Fried Meat)

Adjustment needed for association rules by removing the menu of Teh Manis, Air Putih and Nasi Putih. Minimum support and minimum confidence factor values are determined 1.25% and 15% respectively as the output that is expected to come out after making several attempts to change values (See Fig. 3). The main point of the experiment is to set high confidence values to describing the strength of the two menus. Followed by a high support value, describing the frequency of the combination of the two objects from the overall high transaction data.

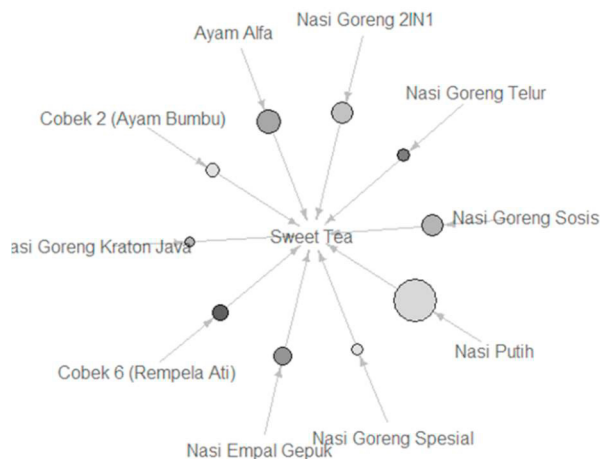


Fig. 3. Matrix Incidence Association Rules Teh Manis.

The bigger value of support factor can describe that the more frequent associative relationships are transactions between Tahu (Tofu) and Ice Lemon Tea and Tahu with Fresh Orange (orange circle in Fig. 4). The other result of associative relationship that was formed after deleting several menus. Several experiments to obtain the values of support and confidence are determined 0.8% and 10% also be done. The results of the associative rules above found several rules that need to be given attention. Rules of Cobek A (chicken and sambal) and Soup Iga menu which have a strong relation (green circle in Fig. 4). A strong link between the two can be caused by consumers who make transactions like Ribs based foods. Other rules is Milkshake Chocolate and Nasi Ayam Goreng (fried chicken and rice, red circle in Fig. 4) has a strong relation because the combination of these two menus is usually purchased for children. These three associative rules that have been mentioned are potential to be some promotions.



Fig. 4. Matrix Incidence Association Rules Adjustment 1 & 2.

Menu list

The results of menu grouping based on the market basket analysis make it easier to analyse which menu types are most desirable for the restaurant's consumers. The food menu with type Nasi and Cobek has the highest average sales from the entire food menu. Menu with Nasi and Cobek type is an Indonesian menu, high average sales proportion for

this type of food describes that the restaurant's consumers like Indonesian menus. Ayam type menu, especially Ayam Goreng Nasi (fried chicken and rice) which is fried chicken is very popular with the restaurant consumers, which shows that the restaurant's consumers like chicken menu. Chicken type menus are generally ordered by consumers who bring children because children like this fast food.

Chinese, Mie and Hotplate menus have a small proportion of sales. Chinese and Mie type menus are typical Chinese menus that has very low demand. It is very possible by the location of the restaurant where is domiciled in a regional mall of indigenous of Indonesian people. The menu Packet type that has been running at the restaurant has also not received a good response from consumers.

All dishes are reviewed by sorting menus on each type. The restaurant focuses on considering total sales and raw materials used. Total sales are an important aspect of a restaurant's life which is can be an indicator what is the most popular menu. The higher total number of sales for each menu in the restaurant is expected to show the consumer menu interest. Menus with low total sales but having the same raw material as the menu with high total sales will be maintained.

Marketing strategy design (4P)

The results of descriptive data analysis that has been done, are used as input to design the right marketing strategy with 4P. The use of the 4P method focuses on satisfying customers and the restaurants. 4P itself analyses 4 interdependent variables which include product, place, price, and promotion (4P). Understanding and using 4P appropriately will be a key factor in achieving successful marketing targets such as profits. The following analysis uses the 4P method:

3.5.1. Product

Observations made by the author while doing research, the restaurant uses too wide product category. This condition known because promotional advertisements carried out by the restaurant (through display menu images, X banners placed in front of the restaurant and front of the cashier table) are menus with too varied food types such as Chinese type and Hotplate type promotion at the same time. This condition underlies the writer to solve the problem with the line extension product marketing strategy by starting to sort the menus to focus on categories from the restaurant. Both food and beverages the restaurant menus are eliminated based on the consideration of total sales and raw materials. Only some menu details that are retained, type, Grouping and considerations for maintaining by the company. The result of the deletion of the menu is 54 menus consisting of 37 food menus and 17 drinks menus. Mostly of the menus that still in the list are Nasi, Cobek and Ayam type of menus that is Indonesian cuisine. The menu with Chinese and Noodle types was mostly removed because it generally had small total sales and was less liked by the restaurant consumers. The remaining 54 menus will be used to focusing the restaurant products to Indonesian cuisine. The restaurant can be said to be an Indonesian restaurant, based on the results of sorting that has been done.

3.5.2. Place

The restaurant located in one of the South Surabaya malls and still at the same place in the mall right now. The location of the restaurant is enough strategic to distribute the products. Located right next to the parking entrance of the 1st floor car and motorbike, consumers who come can immediately see the restaurant at the right side. There is a traditional market located below the mall, a hypermarket that is located right beside the restaurant and a large retail store on the same floor as the restaurant. These places are potential to bringing consumers, especially housewives. The second floor right in front of the escalator has a game centre that also has the potential to bring children's consumers. The location of the restaurant for the past one year has also been strengthened by the existence of a new hotel beside the mall. This hotel has a connected road to the mall right on the parking floor on the 1st floor where the restaurant is located. The existence of this hotel can be an opportunity to channel promotions to hotel residents.

Kartajaya explained that New Wave Culture [4] appeared in Indonesia. Young people, women, and netizens are big contributors to the world of marketing, no longer parents, men and citizens. Young people find it easier to make a choice about a product they have just tried, while women are the decision makers of purchasing in the family. Women are a picture of future consumers who are more responsive to change. The progress of the internet also made it easier for citizens to become netizens who are very easy to recognize and understand the environment quickly.

Catching this new culture, the restaurant is expected to be able to use the internet as a media "location" for product distribution in addition to the current real location.

3.5.3. Price

The price offered by the restaurant is lower compared to similar restaurants with good quality so that it can be said that the strategy used by the restaurant is currently a good value strategy. The restaurant is superior in product price, it is hoped that mid-low consumers are interested to come. Although the restaurant's price is lower than prices at other restaurants, some consumers state that the price offered is still expensive. Based on the same book written by Kartajaya [4], today's citizen has a high lifestyle. Living a high life doesn't mean that the citizen has a high economy but refers to a luxurious lifestyle.

The average price at the restaurant is Rp. 17,701.01 while other similar restaurants offer food with an average of Rp. 25,000 and above. Considering that mid-low-class consumers today are consumers who want to be having a "luxurious" lifestyle, consumers still demand a menu that looks luxurious and quality but at a low price. Knowing this problem, the restaurant needs to do a promotion that gives discounts on certain packages or vouchers for purchases to a certain extent so that consumers feel that they benefit from the price paid.

3.5.4. Promotion

Through observation and random questions to several consumers, the authors found that there were quite several consumers who had never known the restaurant before. Respondents who had known the restaurant, were mostly known because of the mall marketing promotion. This information can be concluded that the promotion run by the restaurant so far less known by customers.

4. Conclusion

Through the results of the analysis found the consumer consumption pattern which mostly leads to the menu with the type of food menu Nasi (rice), Ayam (chicken) and Cobek (chicken and sambal) which is Indonesian food menu. While the menu types of Mie and Chinese are almost not in demand by the restaurant consumers. The restaurant consumers are also consumers who like to transact on weekends. High consumer transactions on weekends are possible because every weekend (Saturday and Sunday) at the mall seminars are held for entrepreneurs.

Tahu (tofu) has a great appeal that shows, it is favorite snack menu at the restaurant. Capturing this condition was formed a promo by combining the Tahu with a drink menu that has a high profit margin. Other analysis results found that other consumer consumption patterns that were seen through a strong attachment between the menu of Ayam Goreng Nasi (fried chicken and rice) and Chocolate Milkshake. This attachment shows that the restaurant has consumers who bring children and has never been realized by the restaurant before.

Consumer consumption patterns will continue to change over time; therefore this pattern needs to be monitored using dashboards. The dashboard can be used as a tool for analyzing the restaurant executives to continuously review menus, activities and promos that have a positive impact or not impact on profits. Through the dashboard, executives can easily make decisions to carry out promotions or activities based on the information displayed.

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