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Prediction of Purchase Behaviors Based on Customer Demand Value Using Factorization Machines

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Abstract—This study focuses on Integrated Marketing Communication which is an important factor of Customer Relationship Management. The purpose of this study is to compare the customer sales data of two different types of stores in the confectionery industry, using the framework which is “Sources of Communication Messages in a Relationship”. In addition, this study proposes the method to categorize the customer’s pursuit values for each store based on sales data. And also this paper proposes the new approach that can promote sales while increasing the value of usage in customers’ living, based on devising a method to improve the efficiency of customer direct mail delivery, using Factorization Machines. The result shows the value that customers desire in each two shops is different from not only on-site investigation but also customer sales data. Furthermore, a customer demand value is defined in this study, the customer direct mail delivery is established to combine customer demand value with the result of purchase behavior prediction using Factorization Machines, makes that the value of usage in customers’ living increase and sales promote.

Index Terms—factorization machines, customer demand value, purchase behavior, direct marketing, customer relationship management, prediction

I. INTRODUCTION

Customer Relationship Management (i.e., CRM) is known to be an important factor in improving the relationship between customers and the company, thus, it is important for companies to provide the demanded value to customers (Grönroos(2007)[1]). Don (1992)[2] has proposed Integrated Marketing Communication (i.e., IMC) which indicates the need to integrate all forms of communications including media marketing, direct marketing and advertising, to the provision and consumption of products and services and the communications between customers as an essential factor to CRM. In addition, Gronroos (2000)[3] proposed “Sources of Communication Messages in a Relationship” (Figure1.) which extended four sources of communication messages proposed by Duncan and Moriarty (1997)[4]. This includes “Planned Messages”, “Product Messages”, “Service Messages”, “Unplanned Messages” and “Absence of Communication”. “Planned Message” is the result of the planned marketing communication campaign and it is often used in Direct Mail (i.e., DM), televisions, and publications. It is generally an one-sided message from companies, but as for DM, companies could send messages according to individual customer’s needs and it is relatively easy to for companies improve it. Therefore, this paper focuses on the DM included in the “Planned Messages”. Although electronic devices are widely used today, DM continues to play an important role in many corporate communication strategies (Simon and Lynda, 2015)[5]. At Canada Post Corporation (2015)[6], DM is said to be more persuasive than digital media, and it is much easier for customers to understand the advertising contents. Also in previous studies on DM distribution, it has been verified that the different types of DM (e.g., the purposes of promotion, the purposes of building customer relationships) lead to different results of purchasing behaviours (Gazquez-Abad, Camanho and Martinez-Lopez, 2011)[7]. Migueis, Camanho and Borges (2017)[8] compared direct marketing reactions at banks using various methods of machine learning. Chen et al. (2016)[9] also described the validity of conducting purchasing behavior predictions using Factorization Machines in e-commerce. However, in CRM in terms of the delivery of DM, none of the previous researches took the value desired by customers into considerations, which is considered important.

![Figure 1. Sources of communications messages in a relationship.](image-url)

Therefore, the purpose of this research is to compare the customer sales data of two different types of stores in the confectionery industry, using the framework proposed by Gronroos (2000)[3], and to categorize the customer’s pursuit values for each store. Therefore, this paper

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proposes the new approach that can promote sales while increasing the value of usage in customers’ living, based on devising a method to improve the efficiency of customer direct mail delivery, using Factorization Machines.

II. ANALYSIS DATA

The data analyzed in this study is data of two different type of confectionary stores in Osaka, Japan. Followed by the analysis, the raw data of each of the two stores are scrutinized to comply with the research purpose. The summary of the data after the examination is shown in Table I.

| TABLE I. SUMMARY OF THE DATA AFTER EXAMINATION |
| --- | --- | --- |
| Variables | Shop Y | Shop M |
| Number of Records | 341,972 | 2,104,036 |
| Term | 1, April, 2014 to 31, May, 2017 | 11, October, 2006 to 7, June, 2017 |
| Number of Customers | 17,410 | 79,236 |
| Number of Products | 1,306 | 7,612 |
| Number of Purchasing Purposes | 6 | 9 |

III. COMPARISON OF TWO SHOPS FROM THE VIEWPOINT OF A SERVICE SCIENCE

A. Comparison and Examination Among Shops Based on Sources of Communication Messages in a Relationship

Table I shows the comparison between two stores based on the source of communication messages in customer relationships proposed by Gronroos (2000)[3].

| TABLE II. COMPARISON OF TWO SHOPS |
| --- | --- |
| Planned Messages | Shop Y | Shop M |
| Product Messages | ☐ | ☐ |
| Service Messages | ☐ | ☐ |
| Unplanned Messages | ☐ | ☐ |
| Absence of Communication | ☐ | ☐ |

The evaluation of two stores are shown in Table II. Confectionary shop Y focuses more on the development of products, by having a famous pastry chef to direct the process of development. It has a strong product message, and it seemed to have provide an unplanned message within the geographical area. On the other hand, confectionary shop M does not conduct any product developments as Shop Y does, however, it actively collects customer requests and feedbacks to improve the operating system of the store. It was also discovered that Shop M continuously distributes DM to customers, which adds another strength to Shop M’s service messages. A more detailed analysis of customer needs will be conducted in the next section, based on the premises shown in Table II.

B. Understanding Customer Needs and Comparing Customer Segmentation Among Shops

Each confectionary store has labeled data on its products, which indicates customer’s purchasing purpose for the product (Table III). This study will first use this data in attempt to understand customer needs.

| TABLE III. PURCHASING PURPOSE AT EACH CONFECTIONARY SHOP |
| --- | --- |
| Shop Y | Shop M |
| Event Gift | Event Gift |
| Event Decoration Cake | Event Decoration Cake |
| Gift | Gift |
| Decoration Cake | Decoration Cake |
| Baked confectionery | Cake |
| Cake | Event |
| Event Cake | Almighty (included Baked Confectionery etc) |
| Cafe Good | Cafe Good |

The distribution of customer’s purchasing purpose at each confectionary shop are shown next.

Fig. 2 show that Customers demand 1st) Cake, 2nd) Baked Confectionery, 3rd) Gift at the confectionary shop Y, and do 1st) Cake, 2nd) Gift, 3rd) Cafe Good at the confectionary shop M. 

Customer’s purchase objective achievement number is shown in Figure 3. Purchase objective achievement number represents the number of purchasing objectives that customers have had when entering the stores.

As shown in Fig. 3, purchasing objective achievement number 3 of shop Y is the peak and the value desired by
customers are diversified. On the other hand, shop M has 1 as its peak, which indicates that each customer has uniquely determined value when entering the store. Therefore, it is possible that the result is directly affected by the use of DM, which ultimately lead to an increase in peak of the number of achievement for purchasing purpose. As result, customer’s utility value in life and store’s sales is further increased.

![Graph](image)

Figure 3. The distribution of customer’s purchase objective achievement number at each confectionary shop.

IV. CUSTOMER DEMAND VALUE AND EXPANSION OF CUSTOMER’S CONSUMPTION BEHAVIOR HYPOTHESIS

Customer demand value is the value that are desired by customers in practice. In other words, each customer aims to achieve the pursuit value by means of goods and services. The difference between value and demand value is that value is a principle or standards that individuals already hold, and the demand value is a value that can be sought to realize the already existing value. One may not be able to realize its value because of economical or environmental constraints. It is also extremely difficult to measure those constraints individually.

In addition, past researches have attempted to analyze the effects of value on customer’s purchasing behavior based on the assumption that there is a strong correlation between customer’s value and their purchasing behavior (Piyathasan et al., 2014[10]; Kim et al., 2012[11]; Miyamoto and Tsubaki, 2016[12]). Questionnaires are often used as a common method when questioning the participants regarding their values, however, it is extremely difficult and expensive to conduct surveys among all customers. There is also a problem of accuracy of the data. Since the surveys are mostly self-assessed, the data could be subjective to participants’ own assessment of selves. Therefore, this paper seek to estimate the customer demand value from purchasing behaviors, based on the hypothesis that customers make purchasing decisions based on their pre-existing values.

Watabe and Tsubaki (2016)[13] proposed a consumption and usage behavioral model by conducting on-site investigation, hearing survey and literature survey based on important factors of consumer purchasing behavior. Figure 4 shows the proposed model confined and extended to the confectionary industry.

![Graph](image)

Figure 4. Consumption and usage behavioral model extended to the confectionary industry.

This study focuses and extends the aspects of Values. Based on the past experiences of pastry chefs and managers of the pastry shops, it is predicted that consumers’ desired values include the following: 1) new preference oriented, 2) conservative oriented, 3) high-quality oriented, 4) “value for price down” oriented (e.g., feeling of getting a good deal), 5) celebration oriented, 6) events oriented, 7) communication oriented.
8) gathering oriented, 9) sweet tooth, 10) sense of security, 11) sharing oriented. The value thought to be derived from purchasing behaviors are shown in Figure5, and the method of calculation for identifying these values from data are shown in Table IV.

![Figure 5. Customer demand value for confectionery.](image)

<table>
<thead>
<tr>
<th>Method of Calculation</th>
</tr>
</thead>
<tbody>
<tr>
<td>Characteristic value is the total number of purchasing new products in 1 month from first sales.</td>
</tr>
<tr>
<td>Characteristic value is the maximum purchasing number for each product.</td>
</tr>
<tr>
<td>Characteristic value is the total number of purchasing ‘Gift’ is more than 2,520 yen and ‘Cake’ is more than 525 yen.</td>
</tr>
<tr>
<td>Characteristic value is the total number of purchasing ‘Gift’ is less than 864 yen and ‘Cake’ is less than 376 yen.</td>
</tr>
</tbody>
</table>

**TABLE IV. THE METHOD OF CALCULATION FOR VALUES**

<table>
<thead>
<tr>
<th>Customer Demand Value</th>
<th>Method of Calculation</th>
</tr>
</thead>
<tbody>
<tr>
<td>New Preference Oriented</td>
<td>Characteristic value is the total number of purchasing new products in 1 month from first sales.</td>
</tr>
<tr>
<td>Conservative Oriented</td>
<td>Characteristic value is the maximum purchasing number for each product.</td>
</tr>
<tr>
<td>High-quality Oriented</td>
<td>Characteristic value is the total number of purchasing ‘Gift’ is more than 2,520 yen and ‘Cake’ is more than 525 yen.</td>
</tr>
<tr>
<td>Value for Price Down</td>
<td>Characteristic value is the total number of purchasing ‘Gift’ is less than 864 yen and ‘Cake’ is less than 376 yen.</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Celebration Oriented</th>
<th>Characteristic value is the total 1), 2), 3) &amp; 4) (unique date).</th>
</tr>
</thead>
<tbody>
<tr>
<td>Events Oriented</td>
<td>Characteristic value is the total 1) &amp; 2) (unique date).</td>
</tr>
</tbody>
</table>

1) As for the preference of new things, characteristic value is the total number of purchasing new products in 1 month from first sales. 2) As for the conservative-oriented, the maximum purchasing number for each product is taken as characteristic value. 3) & 4) As for the high-quality oriented and “value for price down” oriented, characteristic value is taken from number of purchases of ‘Gift’ & ‘Cake’ products that are many in products and relatively higher/lower in price. 3) & 4) As for the high-quality oriented and “value for price down” oriented, characteristic value is taken from number of purchases of ‘Gift’ & ‘Cake’ products that are many in products and relatively higher/lower in price. 3) & 4) As for the high-quality oriented and “value for price down” oriented, characteristic value is taken from number of purchases of ‘Gift’ & ‘Cake’ products that are many in products and relatively higher/lower in price.

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characteristic value is calculated based on the purchase frequency of 'Cake' and/or 'Baked confectionery'. 10) Sense of security: it was considered to be difficult to conduct calculation, since there were no strong indication of the production location of material displayed in stores. 11) Lastly, as for sharing oriented, the calculation was deemed to be impossible since it is difficult to identify consumer’s SNS accounts individually.

V. PROPOSAL OF DM DISTRIBUTION METHOD BASED ON CUSTOMER DEMAND VALUE

A. Factorization Machines

This section explains the framework of the DM distribution method using Factorization Machine (i.e., FM). FM is a model proposed by Rendle (2010)[15] that combines SVM (i.e., Support Vector Machine) and Matrix Factorization. Unlike SVM, FM specializes in dealing with very sparse data. It also deals with many problems that matrix decomposition models often have (e.g., not being able to deal with general prediction problem). Hence, it is a method that specializes in solving general prediction problems corresponding to high dimensional and very sparse data. Equation (1) shows a FM model in the case where the term of interaction is set to two dimensions.

\[
\hat{y}(x) = w_0 + \sum_{i=1}^{n} w_i x_i + \sum_{i=1}^{n} \sum_{j=1}^{n} \langle p_i, p_j \rangle x_i x_j
\]

(1)

\[
\langle p_i, p_j \rangle = w_0 + \sum_{f=1}^{k} p_{i,f}^T v_{j,f}
\]

(2)

where the model parameters that have to be estimated are:

\[
w_0 \in \mathbb{R}, W \in \mathbb{R}^n, V \in \mathbb{R}^{n \times k}
\]

(3)

\[k \in \mathbb{N}_0^+\] is a hyperparameter that defines the dimensionality of the factorization.

B. Extending the DM Distribution Method

1) DM distribution

DM distribution method is suggested in this research has two phases. 1\textsuperscript{st} phase is selecting the customers who will get the DM. 2\textsuperscript{nd} phase is deciding the contents of DM for each customer.

In the 1\textsuperscript{st} phase, not only good customers but also prospective customers for shops have to be selected. In the 2\textsuperscript{nd} phase, it is important for customers to be decided sending the DM which can promote sales while increasing the value of usage in customers’ living. Which means that the DM which strongly recommends products based on customer’s demand, makes customer’s purchasing purpose increase.

In addition, DM has just two recommended products because of space limitations of DM.

2) Extending the method using FM

The step of deciding recommended products in this research is explained below (figure6, 7).

Phase1: The most strongly customer demand is extracted using Chapter IV.

Phase2: Purchasing number of each product that the customer has never purchased before is predicted in each customer using FM.

Phase3-1(for customers who have not achieved all purchasing purposes): (1) The top score of purchasing purposes is elected in purchasing purpose which the customer has never purchased before (yellow). (2) The top score of purchasing purposes is elected in purchasing purpose which the customer has purchased before (blue). (3) The recommended products which are based on customer demands (phase1) are decided in each (1) & (2).

Phase3-2(for customers who have achieved all purchasing purposes): (1) The top 2 score of purchasing purpose is elected in purchasing purposes. (2) The recommended products which are based on customer demands (phase1) are decided in (1).

Figure6. Phase1: Step of deciding strong customer demand value.

Figure7. Phase2,2-1,2-2: Step of deciding recommended purchasing purpose.

The step from model creation to model evaluation using FM is shown below and in Figure8.

Step1 (Preparing training and test data): As training data, customer’s each purchasing products are calculated per one entering the shop in the period A. As test data, customer’s each purchasing products are calculated per one entering the shop in the period B.

Step2 (Model creation): The prediction model (explanatory variables: User/Item, target variables: purchasing number of each products) using FM is constructed. The optimal model is decided with cross validation.

Step3 (Prediction): As explanatory variables are User/Item, the purchasing number of each products are predicted using the step2’s model.

Step4 (Model evaluation): The hyperparameter is decided so that mean squared error between predicted value and actual measured value (i.e., MSE) becomes smallest.
This study aims to compare the customer sales data of two different types of stores in the confectionery industry, using the framework which is “Sources of Communication Messages in a Relationship”, and this study proposed the method to categorize the customer’s pursuit values for each store based on sales data. This paper proposes the new approach that can promote sales while increasing the value of usage in customers’ living, based on devising a method to improve the efficiency of customer direct mail delivery, using Factorization Machines.

In comparison of two shops, the results indicated the strong points in each two shops are different from on-site investigation. Moreover, the value desired by customers to each two shops is also different from customer sales data. Therefore, the value customers desire in practice are defined as “customer demand value”.

In customer direct mail delivery, customer purchase behavior prediction makes 2 parts below possible. 1) recommends purchasing purposes which the customer has never purchased before, 2) recommends purchasing purpose which the customer has purchased before. Further, this paper proposed adding the information of customer demand value to DM which enables recommend more customized products for customers. It can be considered that distribution of DM is able to promote sales while increasing the value of usage in customers’ living.

REFERENCES


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