

## Multivariate analysis of marketing data - applications for bricolage market

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**Abstract:** *By using concepts and analytical tools for computing, marketing is directly related to the quantitative methods of economic research and other areas where the efficiency of systems performances are studied. Any activity of the company must be programmed and carried out taking into account the consumer. Providing a complete success in business requires the entrepreneur to see the company and its products through the consumers eyes, to act as representative of its clients in order to acquire and satisfy their desires. Through its complex specific activities, marketing aims to provide goods and services the consumers require or right merchandise in the right quantity at the right price at the right time and place. An important consideration in capturing the link between marketing and multivariate statistical analysis is that it provides more powerful instruments that allow researchers to discover relationships between multiple configurations of the relationship between variables, configurations that would otherwise remain hidden or barely visible. In addition, most methods can do this with good accuracy, with the possibility of testing the statistical significance by calculating the level of confidence associated with the link validation to the entire population and not just the investigated sample.*

**Key-words:** *marketing, consumer, multivariate statistical analysis, variables*

### 1. Introduction

A good analogy to express the relationship between the information contribution of univariate and bivariate methods on one hand and that of the multivariate, on the other hand, could be for example, just like making a comparison between black&white and color photos. As it can be understood from its name, multivariate means, in the simplest sense, "many variables considered simultaneously" in the analysis, in most cases involving more than two variables. In this respect, multivariate methods can analyze the networking of several variables analyzed simultaneously, based on a specific model for each individual method.

Most of the techniques identify configurations of similarities or links between variables, or make predictions or estimate the relative importance of each variable in predicting or explaining certain variables.

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When the researcher seeks to explain, predict or understand marketing variables (especially those which measure the results or effects) it is recommended the use of multivariate techniques to study the dependencies. Whether it is about sales or market shares, product ratings or services, the satisfaction or purchase intent, the decision maker in marketing will be interested to know which variables can explain or influence these results and to learn more about the nature of the link between them.

An important milestone in capturing the dynamic relationship between marketing and multivariate statistical analysis is the research conducted by Jagdish N. Sheth, entitled "Multivariate revolution in marketing research". The author presented the framework of the most useful multivariate statistical techniques for marketing research, along with a brief description of each. In addition, the article treats the extremely important role of information technology, hardware and software development for the rapid expansion of the widespread use of these methods. Since then, it is considered that the use of multivariate analysis methods began to increase significantly and this trend will continue in the future.

The main objective of this study is seeking to achieve a comprehensive study that capture the place the consumer stands on in the consumer promotion strategy developed by Dedeman, using a number of multivariate analysis methods. Also the study aims at identifying the main promotional techniques the company uses and analysis of the effect those have on consumers.

## **2. Methodology of scientific investigation**

In order to realise this study several methods of multivariate data analysis were used. The multivariate methods can be classified into methods of analyzing dependencies, characterized in that one or more variables are considered dependencies and another group forming the set of independent variables (factors or predictors) and methods for analyzing interdependencies, aiming grouping data based on covariance similarities between them, allowing the identification of structures (configurations) in the data set; they do not analyze the dependence between variables links.

The main categories of methods for analyzing dependence relations that will be treated are:

- factorial analysis – investigates the ties of interdependence between several variables by which to characterize a particular phenomenon by reducing (condensation) the data volume contained in the initial variables and establishing a smaller set of dimensions (factors), aiming at a minimum loss of information. The analysis does not aim to make estimates and predictions, but seeks grouping variables taken into consideration, focusing on the analysis on the interdependencies among them.
- groups analysis – uses the classification of the components of a heterogeneous set (objects, individuals or cases) in relatively homogeneous groups, depending on a

number of variables and interdependencies between them. Cohort Analysis, apart from ensuring homogeneity within groups also creates big differences between groups. Most clustering methods are heuristic-based on algorithms and are relatively simple. They are not based on complex models and exhaustive statistical tools, being from this point of view, substantially different methods of multivariate analysis of variance, regression, multiple linear discriminant or the factor analysis. As far as its purpose and work method, groups analysis is interested in the similarity between objects.

- multidimensional scaling - is a process in which, using various techniques, the analyzed objects (incentives) are represented by points in a multidimensional perceptual space. The axes of this area are interpreted as attributes with which objects are differentiated from each other and it is assumed that they are the expression of psychological springs through which the respondents form their perceptions and preferences related incentives. Being also a method that focuses on studying the similarities between objects, multidimensional scaling is somewhat the reverse of analysis groups: while group analysis classifies objects according to similarities between them, according to some dimensions prespecified multidimensional scaling extracts the fundamental dimensions of similarities and preferences indicated by consumers.

Although these methods are called to study methods dependencies should be noted that they visit and estimate relationships between variable and does not purport to demonstrate causality between variables (except statements that relate directly to this and meet conditions methodological necessary, specific marketing experiments).

### **3. Results and discussions**

Multivariate analysis of marketing data was conducted based on primary data obtained through quantitative research. The quantitative research took place between 15.01.2016-17.06.2016. The research was conducted on a sample of 384 respondents.

#### **3.1. Cluster analysis**

Cluster analysis can be seen as an instrument that aims reduction of sets of objects, or variables to a smaller number of informational entities that are classes or clusters. In table 3.1, for each variable analyzed, we can see the average registered by the members of the two clusters. It can also be observed that the averages for the first variables analyzed in cluster 1 are smaller than those in cluster 2, which means that members of the first group considers variables analyzed (quality, price, attractive appearance, curiosity) underlying the purchase of products while the second group is skeptical and believes that in fact underlies purchasing advertising.

Final Cluster Centers		
	Cluster	
	1	2
Quality	4.10	4.16
Price	3.38	4.24
Attractive appearance	3.25	3.39
Curiosity	2.48	3.23
Necessity	3.90	3.94
Publicity	1.87	4.31

Table 3.1. *The final centers of two clusters*

In Table 3.2 are presented the dimensions of the two clusters of the 370 respondents, 140 are part of cluster 1 (which believes that the analyzed variables underlying the purchase of products of the bricolage company Dedeman) while 121 are part of the cluster 2 and are skeptics.

Number of Cases in each Cluster		
Cluster	1	140.000
	2	121.000
	Valid	261.000
	Missing	119.000

Table 3.2. *The dimensions of two clusters***ANOVA**

	Cluster		Error		F	Sig.
	Mean Square	df	Mean Square	df		
Quality	.081	1	.951	257	.085	.770
Price	8.017	1	.779	257	10.296	.002
Attractive appearance	1.296	1	1.081	257	1.199	.275
Curiosity	36.310	1	1.433	257	25.339	.000
Need	.231	1	1.344	257	.172	.679
Publicity	379.537	1	.639	257	594.258	.000

Table 3.3 *Analysis of variance for each variable*

The table of analysis of variance (ANOVA) contains sums of squares, observing that the division into clusters explains much of the total variance. This situation is common in this type of analysis, as the algorithm produces a maximization of

differences between groups for which there is rarely reason for the differences to be insignificant.

### 3.2. The ANOVA analysis

Table 2.4 notes that, in the analyzed sample, there are differences between the 6 groups formed based on income, regarding the degree of satisfaction created by the products sold in Dedeman.

#### 1. What is your income?

What is your income?	Mean	Std. Error	95% Confidence Interval	
			Lower Bound	Upper Bound
under 700 lei	3.870	.100	3.672	4.067
701-1500 lei	3.787	.107	3.577	3.997
1500,01-2000lei	3.855	.095	3.668	4.042
2000,01-2500lei	3.996	.112	3.776	4.216
2500,01-3000 lei	3.762	.179	3.409	4.115
above 3000 lei	3.670	.252	3.173	4.167

Table 3.4 *Population income*

In table 2.4 is obtained a confidence interval for the total population in those environments. The analysis of these intervals is observed that some overlap, which suggests that between the averages of those categories may not appear differences.

#### 2. What is the highest form of education graduate?

What is the highest form of education graduate ?	Mean	Std. Error	95% Confidence Interval	
			Lower Bound	Upper Bound
secondary education	3.768	.207	3.361	4.176
post secondary school	3.701	.135	3.435	3.966
hight scool	3.806	.149	3.512	4.101
university studies	3.612	.149	3.317	3.906
postgraduate studies	3.521	.211	3.105	3.937

Table 3.5 *The opinion regarding Dedeman products based on revenue*

The means regarding the degree of satisfaction of products in the case of the variable Education are also different.

**Dependent Variable: What about products that are found in shops Dedeman?**

Source	Type III Sum of Squares	df	Mean Square	F	Sig.
Corrected Model	4.307 <sup>a</sup>	10	.431	.751	.676
Intercept	484.556	1	484.556	844.638	.000
Income	2.548	6	.425	.740	.185
Education	1.854	4	.463	.808	.213
Error	145.716	254	.574		
Total	4053.000	265			
Corrected Total	150.023	264			

a. R Squared = .029 (Adjusted R Squared = -.010)

Table 3.6 *Tests of Between-Subjects Effects*

In Table 3.6 the overall average of the image of Dedeman products is computed, which does not take into account the factors group and a confidence interval estimated for the total population. This way we can guarantee a 95% probability that image of Dedeman's products is between 3.432 and 3.93.

If the case of the variable Income  $F_{calc}=0.72 > F_{0.05;6;254}=0.31$ , so we will accept the H1 hypothesis that age has a significant influence on the image of Dedeman products.

In a similar way for the variable Education  $F_{calc}=0.80 > F_{0.05;4;254}=0.31$ , which leads us to the conclusion that education has a significant influence on how the products are perceived.

### 3.3. Main components analysis

A first category of results refers to the average obtained for each of the variables included in the model. In the table below it can be seen that from the 376 respondents were obtained means for the 6 variables.

	Mean	Std. Deviation	Analysis N
From relatives, friends, acquaintances	.5239	.51584	376
Radio	.3245	.48004	376
From TV	.6729	.48652	376
Internet	.3963	.56558	376
Translate			
Newspapers, magazines,	.2686	.44383	376

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Newspapers, magazines,			
Other	.1596	.36670	376

Table 3.7. *The means of the variables introduced into the model*

#### Correlation Matrix

	From relatives, friends, acquaintances	From radio	From Tv	Internet	News- papers, magazines,	Other	
Cor- relation	From relatives, friends, acquaintances	1.000	.205	-.048	-.055	.048	-.048
	From radio	.205	1.000	-.138	-.003	.078	.099
	From Tv	-.048	-.138	1.000	.133	.161	-.125
	Internet	-.055	-.003	.133	1.000	.032	-.074
	Newspapers, magazines,	.048	.078	.161	.032	1.000	-.018
	Other	-.048	.099	-.125	-.074	-.018	1.000

Table 3.8. *The coefficients correlation matrix of the analyzed variables*

There can be observed the relatively high correlations between Internet and TV information sources as well as between the information sources consisting of newspapers and magazines. The small correlations appear between sources of information as radio and TV as well as between sources of information constituted from magazines and other sources.

**Component Matrix<sup>a</sup>**

	Component	
	1	2
From relatives, friends, acquaintances	.406	.580
From radio	.578	.504
From Tv	-.672	.317
Internet	-.426	.225
Newspapers, magazines	-.183	.639
Other	.438	-.229

Table 3.9. *Component matrix*

After extracting those two factors that are the main components, the coefficients of correlation between the variables analyzed and main components are computed. The promotion sources are observed including: radio, relatives, friends, acquaintances and other sources of promotion are strongly correlated with the first component and the information sources include TV, Internet, magazines and newspapers with a second component.

**Total Variance Explained**

Component	Initial Eigenvalues			Extraction Sums of Squared Loadings			Rotation Sums of Squared Loadings		
	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %	Total	% of Variance	Cumulative %
1	1.358	22.627	22.627	1.358	22.627	22.627	1.309	21.823	21.823
2	1.202	20.035	42.661	1.202	20.035	42.661	1.250	20.838	42.661
3	1.001	16.683	59.344						
4	.955	15.922	75.266						
5	.780	13.003	88.269						
6	.704	11.731	100.000						

Table 3.10. *Total variance explained*

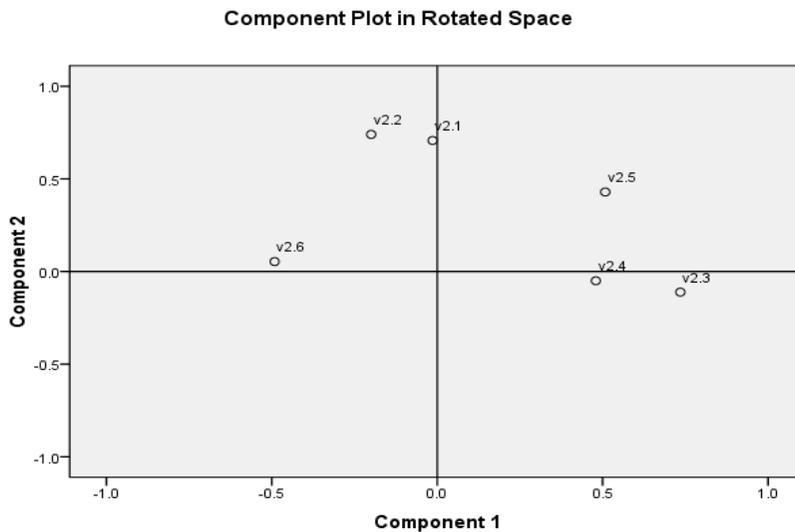


Fig. 3.1 *Graphical representation of associations between variables and factors*

#### 4. Conclusions

This article highlights the utility of applying the multivariate analysis methods in the bricolage market, with the ultimate goal of identifying the factors that influence in significant way the consumer behavior in respect to bricolage products in the current context of globalization and accelerated development. The marketing environment is very complex because of the very nature of the business in the bricolage centers is extremely complex. Bricolage consumer behavior is very dynamic, constantly changing, for this reason it requires constant research.

The multivariate statistical methods of analysis, depending on the relationships which may exist between the variable or dependent variables and independent variables can be grouped into two broad categories: analytical dependencies method (explanatory methods) and methods for analyzing interdependencies (descriptive methods). The analysis method of dependencies explains or provides the evolution of one or more dependent variables based on three or more independent variables.

We are witnessing relatively high correlations between information sources as the Internet and TV and between the sources of information consisting of newspapers magazines and internet. Small correlations are common with sources of information as well as radio and TV between information sources ting up of

magazines and other sources. Education and age have a significant influence on the image of Dedeman products, which leads us to the conclusion that education significantly influences how the products are perceived.

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