Classification Methods in Predicting the Consumers’ the Response to New Product Types

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Abstract. In order to predict the consumers’ response to new product types and development of marketing means for their promotion the task of definition of the closest analogues of the designed item samples. The task was carried out based on the example of whipped confectioneries. The 100-point estimate scale was preliminarily designed for assessment of the marshmallow quality together with the description of the quality levels. An expert sensory analysis was performed for 15 anonymized marshmallow samples obtained from retail organizations and 2 experimental ones manufactured with the use of natural colorants produced of amaranth leaves. The k-average method was used for solving the set task in suggestion, that the consumer would instinctively divide all products within the market into two types: “good” that he will “definitely buy” and “bad” that he “will not buy”. The task solving algorithm was suggested that includes standardization of the indicator values, sample division into two types (“bad” and “good” using the k-average method), repeated clustering, division of the type containing the experimental marshmallow samples in order to further definition of the closes analogues. It was found that the analogues of the experimental marshmallow samples are represented by the products of batch production from popular manufacturers, which allowed the expectation of successful promotion of the new products to the consumer market.

1 Introduction

Food and beverage product expansion through the products with better consumer performance, including the functional products, is one of the priority objectives for population’s health improvement [1,2,3,4,5]. Discussing this matter we should note, that, as a rule, introduction of functional food ingredients in the formulation leads to alteration of the traditional product features, and first among those are taste, flavour, appearance, etc. In such cases the consumer is not always ready for such dramatic changes and often prefers traditional products and technologies. Accordingly, predicting the new product acceptance by the market is one of the fundamental factors defining the viability of the new design and market positioning [6,7,8,9].

2 Research rationale

In relation to the discussed matters, the trend is outlined in confectionery technologies toward the healthy food from the point of view of using the colorants of natural origin [10,11,12]. However, as our researches have proven, natural colorants may often alter sensory properties of the finished goods compared to the best analogues though providing for the better nutrition value [11,12,13,14].

In [15,16] the use of food colorants obtained by selective extraction of pigments from amaranth leaves was suggested as the promising trend of achieving the colour variety of confectionery masses. The formulations for caramels, fondant and whipped confectioneries were designed and tested. However, the viability of the further researches in this direction requires definition of the consumers’ loyalty to new products and prediction of the possibility of its promotion to the consumer market.
3 Problem statement
The task was set for this research to define the rating for confectionery made with the use of colorants made of amaranth leaves (sampled by marshmallow) within one of the types in terms of quality level, which division has been carried out based on the results of sensory analysis of the samples of similar products from the leading manufacturers.

For unbiased assessment of the product quality the 100-point estimate scale for sensory properties was designed together with the description of the quality levels. Simple quality measures have 5-level scale, and the sum of the weighing factors is 20.

15 anonymized marshmallow samples obtained from the retail organizations were presented for sensory analysis. The sensory analysis committee included the experts competent in confectionery products.

4 Results
The obtained values for the simple quality measures were statistically processes and presented as the final total which is the sum of points for all indicators Table 1. The indicators are thus interpreted: x1 - taste; x2 - flavour; x3 - colour; x4 - texture; x5 - shape.

Table 1. Information about the marshmallow quality in retail network (based on the data from unbiased expert evaluation)

| Sample | x1  | x2  | x3  | x4  | x5  | Total grade |
|--------|-----|-----|-----|-----|-----|-------------|
| 3      | 24.0| 13.0| 19.0| 17.5| 22.0| 95.5        |
| 1      | 24.0| 14.0| 20.0| 15.5| 21.5| 95.0        |
| 4      | 23.0| 13.5| 20.0| 17.0| 21.0| 94.5        |
| 5      | 23.0| 13.5| 18.0| 16.0| 21.5| 92.0        |
| 8      | 23.5| 14.0| 19.0| 16.5| 19.0| 92.0        |
| 10     | 22.5| 13.0| 18.0| 17.0| 21.5| 92.0        |
| 11     | 23.0| 13.0| 18.5| 16.5| 21.0| 92.0        |
| 7      | 22.5| 13.0| 18.0| 16.5| 21.0| 91.0        |
| 2      | 23.0| 12.5| 18.0| 16.0| 20.5| 90.0        |
| 14     | 22.5| 12.5| 16.0| 15.0| 19.0| 85.0        |
| 9      | 23.0| 12.0| 16.0| 14.5| 18.5| 84.0        |
| 12     | 22.0| 12.5| 15.5| 14.5| 18.0| 82.5        |
| 13     | 22.0| 12.5| 14.0| 14.5| 18.5| 81.5        |
| 15     | 22.0| 12.0| 15.0| 14.0| 17.5| 80.5        |
| 6      | 22.0| 12.5| 14.0| 13.5| 16.0| 78.0        |

Table 2. Information about the experimental marshmallow sample quality (based on the data from unbiased expert evaluation)

| Sample | x1  | x2  | x3  | x4  | x5  | Total grade |
|--------|-----|-----|-----|-----|-----|-------------|
| 1      | 23.0| 13.0| 14.0| 17.0| 21.5| 88.5        |
| 2      | 22.5| 13.0| 13.5| 19.0| 20.5| 88.5        |

To solve the task of defining the closes analogues of experimental samples for predicting the consumers’ response to the new product types and for designing of the marketing means for their promotion the clustering techniques were analyzed [17]. The k-average method was selected, suggesting that the consumer would instinctively divide all products within the market into two types: “good” that he will “definitely buy” and “bad” that he ‘will not buy’.

The following task solving algorithm was proposed.
1. Indicator value normalization using the formula
\[ y_i = \frac{x_i - x_{\text{min}}}{x_{\text{max}} - x_{\text{min}}} \]  

(1)

2. Retail shop sample division into two types of “bad” and “good” ones applying the k–average technique (using Statistica software). The objective was to unbiasedly divide the samples into two most different groups.

3. Repeat clustering after adding the experimental samples. The objective was to determine to which type the samples would be formally classified.

4. Division of the group including the experimental samples in order to further definition of the closes analogues and finding the indicators where the samples are the closest or the most different. That will further allow the design of marketing means for their promotion.

Here are the results of the task solving for the marshmallow obtained from the retail network and produced experimentally.

1. The normalized values are shown in the Table 3, samples 16 and 17 being the marshmallow with amaranth colorant.

### Table 3. Normalized values of the marshmallow quality

| Sample | \(x_1\) | \(x_2\) | \(x_3\) | \(x_4\) | \(x_5\) |
|--------|--------|--------|--------|--------|--------|
| 3      | 1.00   | 0.50   | 0.85   | 0.73   | 1.00   |
| 1      | 1.00   | 1.00   | 1.00   | 0.36   | 0.92   |
| 4      | 0.50   | 0.75   | 1.00   | 0.64   | 0.83   |
| 5      | 0.50   | 0.75   | 0.69   | 0.45   | 0.92   |
| 8      | 0.75   | 1.00   | 0.85   | 0.55   | 0.50   |
| 10     | 0.25   | 0.50   | 0.69   | 0.64   | 0.92   |
| 11     | 0.50   | 0.50   | 0.77   | 0.55   | 0.83   |
| 7      | 0.25   | 0.50   | 0.69   | 0.55   | 0.83   |
| 2      | 0.50   | 0.25   | 0.69   | 0.45   | 0.75   |
| 14     | 0.25   | 0.25   | 0.38   | 0.27   | 0.50   |
| 9      | 0.50   | 0.00   | 0.38   | 0.18   | 0.42   |
| 12     | 0.00   | 0.25   | 0.31   | 0.18   | 0.33   |
| 13     | 0.00   | 0.25   | 0.08   | 0.18   | 0.42   |
| 15     | 0.00   | 0.00   | 0.23   | 0.09   | 0.25   |
| 6      | 0.00   | 0.25   | 0.08   | 0.00   | 0.00   |
| **16** | **0.50** | **0.50** | **0.08** | **0.64** | **0.92** |
| **17** | **0.25** | **0.50** | **0.00** | **1.00** | **0.75** |

2. As a result of clustering of the 15 samples the two types were obtained interpreted as “good” (nine samples: 1, 2, 3, 4, 5, 7, 8, 10, 11) and “bad” (six samples: 9, 12, 13, 14, 15, 16), the cluster central values are shown in the Table 4.

### Table 4. Marshmallow cluster central values

| Quality measures | “Good” cluster central values (Cluster No. 2) | “Bad” cluster central values (Cluster No. 1) |
|------------------|----------------------------------------------|---------------------------------------------|
| Taste            | 0.58333                                      | 0.12500                                     |
| Flavour          | 0.63889                                      | 0.16667                                     |
| Colour           | 0.80342                                      | 0.24359                                     |
| Texture          | 0.54546                                      | 0.15151                                     |
| Shape            | 0.83333                                      | 0.31944                                     |
3. Repeat clustering of the 17 samples demonstrated, that the marshmallow samples with amaranth colorant were classified into the group identified at the 2nd step of the algorithm as the “good” one.

4. In order to define the closest analogues of the marshmallow with amaranth colorant the values in the “good” group were normalized using formula 1 (Table 5). In the resulting tables the experimental samples are codified as numbers 10 and 11, the rest of the numbers match those in Table 2.

Clustering of the “good” group demonstrated, that the measure “shape” does not affect the difference of the samples within the group. So, in dividing the group into to clusters for definition of the closest analogues this measure was excluded from the assessment.

| Table 5. Normalized values of the marshmallow quality, “good” group |
| --- |
| Sample | $x_1$ | $x_2$ | $x_3$ | $x_4$ | $x_5$ |
| 3 | 1.00 | 0.33 | 0.85 | 0.57 | 1.00 |
| 1 | 1.00 | 1.00 | 1.00 | 0.00 | 0.83 |
| 4 | 0.33 | 0.67 | 1.00 | 0.43 | 0.67 |
| 5 | 0.33 | 0.67 | 0.69 | 0.14 | 0.83 |
| 8 | 0.67 | 1.00 | 0.85 | 0.29 | 0.00 |
| 10 | 0.00 | 0.33 | 0.69 | 0.43 | 0.83 |
| 11 | 0.33 | 0.33 | 0.77 | 0.29 | 0.67 |
| 7 | 0.00 | 0.33 | 0.69 | 0.29 | 0.67 |
| 2 | 0.33 | 0.00 | 0.69 | 0.14 | 0.50 |
| 16 | 0.33 | 0.33 | 0.08 | 0.43 | 0.83 |
| 17 | 0.00 | 0.33 | 0.00 | 1.00 | 0.50 |

As the result of repeated normalization and clustering of the 11 samples (“good” group), the closest analogues of the marshmallow with amaranth (in the resulting tables they are the samples with numbers 10 and 11) are the samples 6 to 9, which are samples 2, 7, 10, and 11 (Fig. 1).

**Fig. 1.** Division of the “good” group and the distance from the cluster centers.

5 Conclusions

So, the analogues of the marshmallow samples are the products of batch production from popular manufacturers, which allowed the expectation of successful promotion of the marshmallow with the colorants made of amaranth leaves to the consumer market.

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