Volume 9, No.4, July – August 2020

International Journal of Advanced Trends in Computer Science and Engineering

Available Online at http://www.warse.org/IJATCSE/static/pdf/file/ijatcse188942020.pdf https://doi.org/10.30534/ijatcse/2020/188942020



Analyzing Soft Cheeses With Vegetable Additives With Neural Network

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ABSTRACT

A wide range of dairy-based products with additives of plant origin is due to the variety of forms of raw materials and methods of processing it. New soft cheeses with vegetable additives have been developed.

Physical and chemical parameters of plant raw materials were studied, and it was found that the wheat germ contains (%) – 28.42 protein, 6.7 – fat, carbohydrates – 19.98. The formula of vegetable additives for the production of soft cheeses has been developed, the ratio of vegetable raw materials and wheat germ is 70:30, while the mass fraction of protein in the additive is 10.6%.

Mathematically, neural networks can be considered as a class of statistical modeling methods, which, in turn, can be divided into three classes: probability density estimation, classification and regression. The problem of estimating the probability density by the method of mixing Gaussian distributions is solved with the help of inverse distribution networks and the generalized d-rule.

Key words: Soft cheese, vegetable Supplement, amino acids, vitamins, neural network.

1. INTRODUCTION

The main consumers of agricultural products are enterprises of the processing industry, which produce important food products for the population. The state policy as the main priorities provides for a significant expansion of domestic production of food products, increasing the range and ensuring their safety. In this regard, the is-sues of further development of the sphere of production and processing of livestock products are strategically important for the agro-industrial complex of the Republic of Kazakhstan [1].

The modern theory of nutrition pays much attention to the design of the composition of food products. A special role is assigned to protein as the main component that determines the nutritional and biological value of the food system [1]. One of the ways to eliminate acute protein deficiency in the diet of people is to create new types of food, including dairy products, using plant proteins [2]. The combination of animal and vegetable proteins in the product allows us to use the principle of complementarity [3].

Cheese is a unique food product, the properties of which are very diverse. There are a huge number of its species in the world. Ordering the assortment of cheeses, improving their specific structure is currently relevant and can ensure the production of cheeses with a high yield, increase the volume of gross output, and obtain high production profitability [4].

To train the neural network, an array of input vectors xi (x1 is the mass fraction of cheese in the recipe mixture; x2 is the ratio B / SZR) was formed, which correspond to the values of the objective function of the score of the consistency of the finished product (Table 1). Creating and learning a neural network is started by a special program created in the MatLab environment [6]. To solve the simulation problem, a three-layer fully interconnected network was formed, including two neurons in the input layer (by the number of components of the input vector), four neurons in the intermediate layer with transfer functions tag-sig and one neuron in the output layer (by the number of components). source vector) with purelin transfer function[6].

The positions of neurons in each layer are determined by the topological function randtop, which forms a random location of neurons in the layer. The linkdist distance function is used to calculate the distances between neurons in the i-th layer. The Levenberg-Marquardt (trainlm) algorithm was chosen as the teaching algorithm. The specified network is formed using the following procedure, written in the program Neiron.m: net = newcf (minmax (P), [6], {'tansig' 'tansig' 'purelin'}, 'trainlm'); where the first argument – the matrix of minimum and maximum values of the components of the input vector – is calculated using the procedure minmax. The result of the newff procedure is a neural network net of a given configuration.

Neural network training is to determine the weights and displacements of all synapses in all layers of the NM. The learning parameters were the number of learning cycles (1000) and the target learning error (0.01). In this case, the rms error was chosen as a function of performance evaluation. The learning process is illustrated by a graph of the dependence of performance assessment on the number of the training cycle [5].

2. MATERIALS AND METHODS

To obtain a fermented vegetable additive from vegetable crops, carrots and pumpkins in the form of pulp were chosen. Carrot pulp was obtained as follows: peeled and washed table carrots are peeled, cut on tops with a diameter of 5 mm, then wiped on a double cleaning machine through sieves with a hole with a diameter of 0.8-0.5 mm and separated carrot juice [7-9].

To process the output data, we use the program code [6], which allows to obtain arrays of values of the output variable for some values of the input variables, which change in the same range as in the experimental study. This three-dimensional model is difficult to understand. In this regard, taking into account the need to obtain a product of functional direction, ie with a high content of biologically active substances, the cross section of the surface (Fig. 6) was a plane parallel to the axis "V / SZR" at the maximum value of cheese in the recipe mixture – 30 %. The obtained graphical dependence is shown in Figure 1.



In the obtained vegetable raw materials, the protein, fat and carbohydrate contents were determined (table 1).

Table 1:- Chemical	composition of	of vegetable raw	materials
	1	0	

The name of components	Carrot cake	Pumpkin pulp
Protein, %	0.84 ± 0.01	0.73±0.02
Fat, %	1.99±0.02	1.23±0.04
Carbohydrates, %	39.75±0.04	40.08±0.03
Moisture, %	36.3±0.03	37.7±0.02
Ash, %	0.62 ± 0.01	0.42 ± 0.01

Since, in the dairy industry, when using food fibers contained in vegetable raw materials, they are pre-subjected to heat treatment to soften the fiber structure and give them an easily digestible form.

We applied biotechnological methods of processing vegetable raw materials to soften the fiber structure and used the enzyme preparation Pectinase to maximize the preservation of all useful nutrients [5].

The studies were performed at pH 6 with a hydrolysis temperature of $48-50^{\circ}$ C within 4 hours. The dry matter content of carrot and pumpkin pulp was initially 63.7% and 62.3%, respectively [11].

When applying an enzyme preparation in the amount of 0.1% of the mass of vegetable raw materials, the fermentation process was slow, the change in the dissolution of dry substances was slow for 6-8 hours.

When increasing the dose of the enzyme preparation to 0.2%, hydrolysis at pH 6 and temperature $48-50^{\circ}$ C was more active for 4 hours and an increase in the destruction of dry substances was observed up to 35-37%.

Then, to prepare a plant Supplement, we studied the chemical composition of the wheat germ. The embryo is almost 80% protein, sugar and fat (table 2).

Table 2 : Chemical composition of wheat germ, %

Product	Moistu	Protei	Fat	Carbo	Fiber	Ash
descriptio	re	n		hydrat		
n				es		
wheat	9.60±0	35.66	11.71	17.40	2.42±	3.8±
Germ	.04	±0.03	±0.04	±0.02	0.03	0.02

Due to the fact that the developed technology for the production of combined soft cheeses involves the use of whole milk, it is necessary to make up for the lack of protein by adding proteins of plant origin in the form of vegetable additives [10].

To obtain a plant Supplement, prepared plant products (fermented carrot and pumpkin pulp, as well as wheat germ) were mixed in a ratio of 70:30, separately carrots+wheat germ and pumpkin+wheat germ.

The result is a plant Supplement containing plant protein, a complex of vitamins, minerals and polyunsaturated fatty acids.

The chemical composition and nutritional value of herbal supplements is presented in table 3.

Figure 1: Neural network model for analyzing cheese

Indicator	carrot-wheat additive	pumpkin-wheat additive
Protein	10.64±0.06	10.61±0.04
Fat	3,16±0.07	2.90±0.05
Carbohydrates	16,7216.72 ±0.05	15.98±0.03
Moisture	68.70±0.03	69.80±0.05
Ash	0.78 ± 0.01	0.7 ± 0.06
Energy value, kcal	138±0.04	132±0.04

 Table 3: Chemical composition of additives, g / 100g

The main criteria for evaluating the quality of food products are indicators of their nutritional and energy value.

The determination of nutritional value (the complex of product properties that provide a person's physiological needs for energy and basic nutrients – proteins, fats, carbohydrates, vitamins, macro – and microelements) and energy value (the amount of energy (kcal, kJ) released in the human body from food substances to ensure its physiological functions) are mandatory in the development of new types of products [12-14].

The nutritional and energy value of the developed new combined soft cheeses was determined (table 4).

Table 4 : Nutritional and energy value of combined soft cheeses with vegetable additives (g / 100 g of the product)

Name of	Soft cheeses with vegetable additives			
indicators	control	with	with	
		carrot-wheat additive	pumpkin-wh eat additive	
Protein	19.50±0.05	24.08±0.07	23.38±0.04	
Fat	23.70±0.06	23.33±0.06	23.20±0.07	
Carbohydrate	2.50±0.05	3.94±0.05	3.87±0.05	
S				
Moisture	50.20±0.08	45.88±0.03	46.69±0.06	
Ash	3.10±0.04	2.87 ± 0.05	2.86 ± 00.4	
Energy	301±0.05	321±0.04	318±0.06	
value, kcal /				
100g				

The combination of whole cow's milk and vegetable additives increases not only the nutritional value, but also the biological value of new products with complex raw materials [15].

The biological value of a product is determined by an indicator of the quality of food protein, reflecting the degree of compliance of the amino acid composition with the needs of the body in amino acids for protein synthesis and is presented in table 5.

Table 5:Biological value of soft cheese proteins (mg/100g protein)

Name of	Soft cheese		
indicators	control	with carrot-wheat additive	with pumpkin-wheat additive
Total amount	19109±0.	23598±0.6	22912±0.5
of amino acids	7		
Irreplaceable,	6944±0.4	10123±0.3	9659±0.8

including			
Valin	993±0.5	1788±0.3	1353±0.5
Isoleucine	809±0.4	1299±0.2	1132±0.3
Leucine	1398±0.7	2250±0.6	1991±0.4
Lizin	1251±0.5	1760±0.7	1661±0.6
Methionine	487±0.2	602±0.3	652±0.5
Threonine	736±0.7	1080±0.4	919±0.5
Tryptophan	304±0.6	317±0.5	845±0.8
Phenylalanine	966±0.3	1027±0.4	1107±0.5
Interchangeab	12165±0.	13475±0.7	13253±0.6
le, including	4		
Alanine	506±0.1	914±0.3	645±0.4
Arginine	727±0.2	823±0.6	918±0.5
Aspartic acid	1646±0.3	1880±0.3	1354±0.2
Histidine	1117±0.8	793±0.2	1139±0.3
Glycine	322±0.7	348±0.1	333±0.5
Glutamic acid	3735±0.4	4486±0.7	4236±0.7
Proline	1739±0.5	2046±0.3	1992±0.6
Serine	1067±0.6	1163±0.5	1316±0.3
Tyrosine	1108±0.7	793±0.3	1158±0.4
Cystine	138±0.1	227±0.4	157±0.5

According to the qualitative composition of amino acids in all soft cheeses, no differences were found. The introduction of a vegetable Supplement has increased the quantitative content of certain amino acids in new soft cheeses [15].

Table 6: Fatty acid composition of soft cheeses (mg/100g of fat)

Name	Soft cheese		
of acids	control	with carrot-wheat	with
		additive	pumpkin-whe
			at additive
Catanata 1	12659.0.2	15616.0.4	12112.0.0
Saturated,	12658 ± 0.3	15616±0.4	13112±0.6
incl			
Oil	-	772±0.2	147±0.4
Nylon	185±0.5	594±0.3	154±0.6
Caprylic	288 ± 0.8	653±0.5	251±0.6
Caprine	843±0.5	1781±0.7	706±0.6
Lauric	875±0.3	1247±0.6	788±0.5
Myristic	1617±0.6	2256±0.7	1968±0.3
Palmitic	5047±0.7	5997±0.5	5389±0.6
Stearic	3811±0.4	2315±0.7	3706±0.5
Monounsatu	8137±0.6	6294±0.8	9090±0.7
rated, incl			
Myristoleic	525±0.3	178±0.5	672±0.6
Palmitoleic	473±0.1	588±0.4	1072±0.8
Oleic	7137±0.2	5527±0.1	7345±0.3
Polyunsatur	762±0.4	1254±0.2	976±0.3
ated, incl			
Linoleic	762±0.3	779±0.5	824±0.4
Acid			
Linolenic	-	475±0.3	152±0.2
Acid			
Total	21557±0.8	23164±0.5	23178±0.7
amount of			
acids			

The vitamin value of milk and dairy products varies significantly depending on the season and conditions of the year, in particular, because of the lack of green feed in winter. Vitamin losses are significant during separation, normalization, pasteurization, sterilization, drying and recovery of milk [13].

For this purpose, the enrichment of soft cheeses with vegetable additives gave positive results. Quantitative content of vitamins and minerals in new dairy products is presented in tables 7 and 8.

Vitamins	Soft cheeses		
	control	with	with
		carrot-wheat	pumpkin-wheat
		additive	additive
A, mcg	205±0.3	250±0.2	238±0.4
β -carotene,	54±0.06	170±0.6	1242±0.5
mcg			
In ₁ , mg	0.030 ± 0.005	0.040 ± 0.005	0.038 ± 0.001
In ₂ , mg	0.340 ± 0.004	0.38 ± 0.004	0.37±0.006
PP, mg	0.40 ± 0.02	0.50 ± 0.01	0.54±0.003
S, mg	0.50 ± 0.04	4.10±0.03	2.72 ± 0.007
E, mg	0.32 ± 0.01	1.20±0.04	0.57 ± 0.002

Table 7: Vitamin Content in soft cheeses, (in 100 g)

Table 8: Mineral content of soft cheeses, mg / 100g

Name	Soft cheeses			
of indicators	control	with carrot-wheat laced	with pumpkin-wheat additive	
Na	476±0.3	800±0.5	559±0.8	
K	73±0.07	155±0.2	111±0.3	
Ca	410±0.3	510±0.3	418±0.7	
Mg	332±0.1	390±0.6	367±0.5	
Fe	0.44 ± 0.01	0.51±0.02	0.55 ± 0.004	
Р	17 ± 0.08	25±0.1	24±0.2	

Thus, from tables 7 and 8, it can be seen that the quantitative content of all vitamins and trace elements increased by 1.5-2 times. New soft cheeses with vegetable additives are enriched with vitamins and trace elements and are biologically complete products, for example, the content of beta-carotene in soft cheeses with carrot and wheat additives has increased by 23 times.

3. CONCLUSION

The creation of a neural network model is based on the use of artificial neural networks, which is one of the directions of development of the theory of artificial intelligence. The neural network model consists of several layers of artificial neurons that mimic the work of their biological counterparts (nerve cells). The functioning of a biological neuron is carried out as follows: input signals are received from synapses through thin densely branching processes of the neuron-dendrites, into the cell body - soma, inside which the nucleus is located. Synapses differ from each other in size and the ability to concentrate special chemical substances near their shell. For this reason, impulses of the same magnitude arriving at the inputs of the nerve cell through different synapses can excite it to different degrees. A measure of cell excitation is considered to be the level of polarization of its membrane, which depends on the total number of chemical substances isolated at all

synapses. If the excitation of the cell is not great, then no changes are registered at its output. If the sum of excitations exceeds the threshold of cell activation, the value of the output signal begins to increase in an avalanche, taking the form of a nerve impulse. This impulse is diverted from the cell by the axon through its numerous nerve endings and gets to the synapse, through which the input signal goes to other neurons connected to this cell.

Data is collected from measuring instruments installed at most stages of the cheese production process. Data is also received from the Central and shop laboratories, where most of the quality parameters of raw materials, semi-finished products and finished products are measured. The collected data is input to an artificial neural network, which, based on the algorithm embedded in it, evaluates the resulting taste index by several criteria on a ten-point scale.

The principle of operation of a neural network is as follows: a set of input parameters is passed through the neural network, for which the output parameters are known. The weight coefficients of the connections of neurons are calculated - the network is trained. Then, if you enter parameters for which the output is not known, the neural network will build a forecast of their values based on weight coefficients. In order for an artificial neural network to work correctly, it must be trained. For training, a training sample consisting of at least one hundred experiments must be submitted to the input of the neural network. The training sample is prepared as follows: all data is collected, which will then be sent to the input of the neural network. And then the expert Commission, consisting of professional tasters, in the conditions necessary for the tasting, evaluates the taste indicators of finished products made with the parameters measured during the experiment, according to a number of criteria. Neural network training is reduced to selecting optimal synaptic weight coefficients. The most popular learning algorithm for MLP-type networks is the back propagation algorithm. Thus, it is possible to evaluate the quality of the finished product, which was previously somewhat difficult based on the subjectivity of the assessment. It is necessary to conduct a number of tests that will allow you to give the most complete assessment of the quality of the product and train the neural network based on the results to evaluate the quality of all subsequent batches of the finished product.

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