Using Artificial Intelligence Algorithms for Speech Therapy Systems

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Abstract: This article analyzes the algorithms used to process speech signals. A structure for processing Uzbek speech signals has been proposed. Based on this structure, the effectiveness of recognizing Uzbek speech sounds, syllables and words was evaluated on the basis of intellectual algorithms. Based on the evaluated indicators, the structure of neural networks that recognize children’s speech pronunciation was developed. Recognition indicators of children’s speech sounds, syllables, and words were identified.

Keywords: speech, recognition, artificial intelligent, Neural network(NN), speech therapy system.

I. INTRODUCTION

It is known that today speech recognition systems are used in many fields. The creation of such systems dates back to the 1950s. Today, scientists use a number of algorithms and mathematical methods in the creation of such systems, such as segmentation of speech, filtering methods, the use of spectral values, Mel Frequency Cepstral Coefficients(MFCC), algorithms of neural networks. Based on the analysis of the algorithms presented in the processing of speech signals in the Uzbek language, speech signals were divided into parameters[1]. Based on these parameters, a decision about the speech signal is required. The decision-making process is carried out by artificial intelligence algorithms. There are several types of artificial intelligence algorithms depending on the functions used. Their currently widely used algorithms for speech signals include dynamic programming algorithms, Hidden Markov models(HMM) and neural network algorithms. These artificial intelligence algorithms perform the task of making decisions based on the parameters of speech signals.

II. MODELING AND EXPERIMENTS

The main purpose of the research work is to propose a generalized model that allows to organize a set of parameters of speech signals and make decisions based on these parameters as a result of effective use in artificial intelligence algorithms and evaluate its effectiveness.

As a result of the experiments, a set of parameters based on spectral analysis of speech signals was generated. Selecting the desired parameters from this set of parameters resulted in a general model based on dynamic programming algorithms, HMM and NN. The structure of this model is shown in Figure 1.

Due to the problem posed in this model, the use of all parameters and decision-making algorithms in the model is not required[2]. That is, the parameters are selected based on the problem and the possibility of decision-making algorithms is used.

![Diagram of speech processing system](image)

Figure 1. General structure of Uzbek language speech processing.

When recognizing speech signals, the user first performs the process of filtering the recorded speech signal in this general structure[3]. This uses the IIR filter algorithm. The filtered signal is then segmented. Segmentation performs the function of splitting the speech signal into frames. The filtered signal is then segmented. Segmentation performs the function of splitting the speech signal into frames. After that, the issue of separating the parameters from...
each frame is considered[4]. Parameter separation is performed using discrete Fourier transform (FT) modification from spectral algorithms. It is sent to the algorithms of the block. Using spectral algorithms, the spectral values, formant frequencies, and spectrograms of a speech signal can be returned as a signal parameter[5]. It is also possible to obtain MFCC parameters using spectral algorithms and to return the sonogram as a signal parameter using the MFCC algorithm. These parameters are formed in a common set of parameters.

The selected parameters are determined by the solution of the problem based on artificial intelligence algorithms. Artificial intelligence algorithms, dynamic programming algorithms, latent marking models, and neural network algorithms were analyzed in the research process. This model has the ability to solve a number of practical problems used in human point processing[5]. As an example, identification based on human speech can be used in systems that are controlled by the content and command of human speech.

The decision-making process based on artificial intelligence algorithms in the recognition of sounds, syllables and words of the Uzbek language were the results of experiments with algorithms DTW (Dynamic time warping), HMM and neural networks.

A reference database for each artificial intelligence algorithm for Uzbek sounds has been created. The results of the analysis were obtained for 29 speech sounds (except ng) in the Uzbek language[7]. The results of artificial intelligence algorithms for vowels and consonants in the Uzbek language are given in Table 1.

### Table 1. The result of artificial intelligence algorithms for sounds

| № | Name | DTW | HMM | NN | № | Name | DTW | HMM | NN |
|---|---|---|---|---|---|---|---|---|---|
| 1 | A | 85 | 83 | 90 | 15 | P | 73 | 83 | 87 |
| 2 | B | 70 | 76 | 86 | 16 | Q | 73 | 88 | 92 |
| 3 | D | 76 | 82 | 93 | 17 | R | 68 | 75 | 76 |
| 4 | E | 89 | 82 | 93 | 18 | S | 60 | 77 | 86 |
| 5 | F | 68 | 76 | 78 | 19 | T | 74 | 66 | 90 |
| 6 | G | 60 | 73 | 90 | 20 | U | 87 | 93 | 90 |
| 7 | H | 74 | 67 | 94 | 21 | V | 73 | 77 | 76 |
| 8 | I | 93 | 97 | 96 | 22 | X | 67 | 79 | 87 |
| 9 | J | 73 | 65 | 65 | 23 | Y | 89 | 83 | 93 |
| 10 | K | 68 | 76 | 97 | 24 | Z | 73 | 76 | 88 |
| 11 | L | 79 | 76 | 86 | 25 | O | 87 | 88 | 91 |
| 12 | M | 81 | 80 | 83 | 26 | G | 68 | 76 | 83 |
| 13 | N | 80 | 86 | 89 | 27 | CH | 65 | 74 | 93 |
| 14 | O | 86 | 87 | 98 | 28 | SH | 66 | 61 | 91 |

The results of the analysis show that experiments on vowels and consonants have made it possible to detect 85-90% of vowels using artificial intelligence algorithms. It gave 60-85% accuracy in consonant sounds. In all of these algorithms, the MFCC coefficients of each frame were obtained as a parameter that converts to artificial intelligence algorithms. The results show that the sound recognition process yielded an average of 73% in the DTW algorithm, 85% in the HMM, and 87% in NN. This in turn proves to be effective in processing speech sounds using neural networks[8].

The results for the open and closed joints of the “R” sound when processing speech joints using these algorithms are given in Table 2.

### Table 2. The result of artificial intelligence algorithms.

| Sound Type | Joints | Degree of detection (%) |
|-----------|--------|-------------------------|
|           | DTW    | HMM | NN |
| R         | Ar     | 78  | 83 | 98 |
|           | Er     | 67  | 75 | 99 |
|           | Ir     | 65  | 76 | 86 |
|           | Or     | 71  | 76 | 87 |
|           | Ur     | 78  | 84 | 97 |
|           | Or'    | 82  | 92 | 97 |
|           | Re     | 72  | 94 | 97 |
|           | Ri     | 74  | 68 | 98 |
|           | Ro     | 79  | 84 | 98 |
|           | Ru     | 76  | 72 | 96 |
|           | Ro'    | 75  | 87 | 99 |
| Average   | 74    | 81 | 95 |

In the recognition of speech joints using artificial intelligence algorithms, open and closed types of joints were obtained. The recognition accuracy of these links yielded an average of 74% in the DTW algorithm, 81% in the Markov chain and 95% in neural networks[9]. In this table, the highest result was given by neural networks. This is because of the process of training joint sounds built on the basis of neuronal tracts. The files selected for training consist of frames with a limited size.

### Table 3. The result of artificial intelligence algorithms for speech words

| № | Uzbek word (English translated) | DTW | HMM | NN |
|---|---------------------------------|-----|-----|----|
| 1 | Bo't(one)                       | 84  | 86 | 95 |
| 2 | Bo(do)                          | 73  | 79 | 87 |
| 3 | Uch(three)                      | 75  | 78 | 88 |
| 4 | To(sth)                         | 83  | 83 | 89 |
| 5 | Bo(ive)                         | 90  | 96 | 92 |
| 6 | Ob'th(six)                      | 79  | 58 | 96 |
| 7 | Ye'i(seven)                     | 87  | 87 | 87 |
| 8 | Sakkato(eight)                  | 82  | 92 | 83 |
| 9 | To(again)                       | 81  | 76 | 92 |
| 10 | Oj'(en)                         | 86  | 81 | 91 |
| 11 | Yng'twenty)                     | 74  | 76 | 97 |
| 12 | Ong(thirty)                     | 83  | 90 | 89 |
| 13 | Ong(forty)                      | 76  | 86 | 92 |
| 14 | El'kill(yfth)                   | 72  | 80 | 84 |
| 15 | Ongish(sixty)                   | 73  | 78 | 88 |
| 16 | Ye'mi(seven)                    | 76  | 85 | 94 |
| 17 | Sakkato(eighty)                 | 87  | 92 | 85 |
| 18 | To(on)(ninety)                  | 89  | 98 | 99 |
| 19 | Yu'hundred)                     | 92  | 96 | 99 |
| 20 | Ming(thousand)                  | 90  | 95 | 99 |
| 21 | Million(million)                | 95  | 97 | 99 |

When processing speech words based on artificial intelligence algorithms, a limited set of vocabulary-based words is required to be selected[10]. Because, based on the selected set of words, the problems of the industry are supposed to be solved as a result of research. Therefore, a set of numerical words was obtained in the experiment. Their results are presented in Table 4.

In the countdown, the results were more effective than the results based on sound and syllable. It yielded 82% in the DTW algorithm, 88% in the Markov chain, and 91% in neural
networks. A general table of these results is given below (Table 4).

| №  | Speech information | D | TW | HMM | NN |
|----|-------------------|---|----|-----|----|
| 1. | Sound             | 73| 77 | 87  |    |
| 2. | The joint         | 74| 81 | 95  |    |
| 3. | The word          | 82| 85 | 91  |    |

From the results of the analysis it can be seen that from the artificial intelligence algorithms used in Uzbek language speech recognition, it was found that neural networks are effective based on experimental results. It is also effective with computational time from other artificial intelligence algorithms due to the simplicity of the computational process in neuronal necks[11]. Experiments have shown that there are clear requirements for the training and testing process when working with neural networks. Therefore, this algorithm is more efficient than other artificial intelligence algorithms. It is recommended to use neural network algorithms in solving field problems in the processing of speech signals.

### III. MATHEMATICAL ALGORITHMS AND APPLICATION

Neural network algorithms are one of the types of artificial intelligence algorithms. These algorithms have also been widely used by scientists in matters of speech recognition. To date, a number of speech recognition issues have been solved by scientists for fields based on neural crests. On the basis of neural networks, a set of speech-based speech, syllables and words based on a limited vocabulary in the recognition of sounds, syllables and words in the Uzbek language was obtained[12].

In this research paper, a speech therapist raises the issue of developing a sequence of algorithms that process children’s speech in ensuring the correct pronunciation of speech in preschool and school-age children with speech impairments. It focuses on speech therapy experiments in determining the solution to the problem. Speech therapy experiments show that the repetition and pronunciation of pronounced words as a result of the occurrence of vowel and consonant sounds at the beginning, middle, and end of a word relative to words leads to the development of speech. In the processing of this type of speech data, a standard database of *.wav, mono and 22050 Hz was created based on the correct pronunciation of sounds, syllables and words by fluent children, recommended by speech therapists. The recorded data was returned as an unwanted speech signal. The main goal is to develop a structure that allows a child’s speech pronunciation to be compared to sounds, syllables, and words spoken by children whose pronunciation in the database is correct. In developing such a structure, mathematical algorithms of neural networks are developed. Based on the results of the analysis and experiment, a structure based on neural networks that provide optimum high accuracy was developed. The general sequence of algorithms for this structure is shown in Fig.2.

![Figure 2. The general structure by which children recognize speech pronunciation.](image)

The possibility of this structure can be used not only in the recognition of children's speech pronunciation, but also in the problems of determining the content of Uzbek speech sounds, syllables and words.

The wide range of issues that neural networks solve today limits the ability to develop a powerful, universal network that is functional with a variety of algorithms. Because each neural network is used in different resources that do not go directly to each other in the problems it solves. Therefore, in the design of neural networks, its structure is developed depending on the type of problem and the field of application. The development of this structure was organized on the basis of the following approach[13].

Speech recognition A multilayer perceptron neural network with a classical multilayer structure with the above-mentioned serial connections and sigmoidal neuron activation functions was used. The input layer of the network consists of 12 elements (MFCC parameters) X1, X2...X12, has a four-layer hidden layer and consists of a single output layer. The first hidden layer consists of F1N N = 8 elements, the second hidden layer F2N N = 6 elements, the third hidden layer F3N N = 4 elements, the fourth hidden layer F4N N = 2 elements. These dimensions are determined experimentally, and the Y dimension of the output layer always consists of 1 element. The fact that the output layer is equal to 1 is based on the child saying a single word during the speech pronunciation.

![Figure 3. The general structure of the neural network used.](image)

In the structure of a neural network, each synapse is characterized by its own synaptic connection or its weight. The current state of the neuron is determined by the randomly found sum of the values at the input.

$$F_{i,N} = \sum_{l=1}^{n} x_{l}w_{l}$$
The output of a neuron has a function for its state:

\[ Y = f(F_{i,N}) \]

One of the most common of these functions is the nonlinear function, i.e., the logistic function in the form \( f(x) \) or the sigmoid function.

\[ f(x) = \frac{1}{1 + e^{-ax}} \]

by decreasing the index a it becomes as if the sigmoid is spreading, at \( a = 0 \) a horizontal line is formed with all points equal to 0.5, and by increasing the value of a pointer, the function allows the unit to be converted to a variable by jumping the unit with the limit \( T \) at the point \( x = 0 \).

This activation function has the following advantages:
- in contrast to the logistics sigma, this function is bipolar and accordingly all signals in the neural network are also bipolar, which reduces the number of iterations of the algorithm for studying this network [14];
- Hyperbolic tangent compared to other bipolar sigmoid due to its simplicity it is computed much faster on a computer, which allows to accelerate the operation of neural networks in both recognition mode and learning mode of recording signals.

The mathematical model of the neural network used in this article is described as follows:

\[ Y = f \left( \sum_{i=1}^{n} F_{i,N} \left( \sum_{j=1}^{m} F_{j,2} \left( \sum_{k=1}^{l} F_{k,1} \left( x_1 \right) \right) \right) \right) \]

Here \( F_{1, N} \) - latent layer neurons, \( X_{112} \) - MFCC coefficients, \( f \) - activation function, \( Y \) - outgoing neuron cells.

In the general structure that recognizes children’s speech pronunciation, we can see that the spectra obtained based on the Fourier modification algorithm are directed directly to the neural network [15]. To do this, the structure of the neural network is constructed based on the following model.

\[ Y = f \left( \sum_{i=1}^{m} F_{i,2} \left( \sum_{j=1}^{l} F_{j,1} \left( x_1 \right) \right) \right) \]

Here \( F_{1, N} \) - latent layer neurons, \( X_{1128} \) - spectral coefficients, \( f \) - activation function, \( Y \) - outgoing neuron cells.

In this formula we can see that the number of spectral coefficients is 128. This is because the speech signal contains 256 values when divided into frames. When spectral values are obtained by changing the Fourier transform from each frame, 128 real frequencies and 128 abstract frequencies are obtained. Only true frequencies were obtained when the spectral coefficients were directed to the neural network.

Based on the sequence of these algorithms, it is possible to determine how accurately the speech signal is recognized in children’s speech development. The sequence of algorithms consists of decision-making through neural networks based on the parametric separation of the speech signal (MFCC and spectral parameters). In this case, there is an 85% high probability that the speech signal is pronounced correctly.

This means that the sequence of algorithms of the Uzbek language processing model characterizes the correct organization. Experimental results show that words with a speech recognition rate of less than 85% are processed using the spectrum of the repetitive speech signal. Based on the experiments, syllables and words with less than 85% familiarity accuracy are listed in Table 5 below. We can see an increase in the level of accuracy of words as a result of the processing of syllables and words based on the signal spectrum.

### Table 5. Description of speech therapy words

| № | Children’s speech sound (in Uzbek language) | Recognition (\%) | Spectrum +NN | Children’s speech sound (in Uzbek language) | Recognition (\%) | Spectrum +NN | Children’s speech sound (in Uzbek language) | Recognition (\%) | Spectrum +NN |
|---|------------------------------------------|-----------------|-------------|------------------------------------------|-----------------|-------------|------------------------------------------|-----------------|-------------|
| 1. | Z | 82 | 93 | ze | 55 | 73 | Dazmol | 60 | 88 |
| 2. | Ch | 82 | 94 | cho | 82 | 85 | Qaychi | 29 | 83 |
| 3. | L | 91 | 97 | lu | 70 | 95 | Olma | 67 | 96 |
| 4. | R | 72 | 97 | ar | 78 | 98 | Rediska | 46 | 76 |
| 5. | P | 92 | 96 | ep | 66 | 97 | Paxta | 77 | 89 |
| 6. | D | 96 | 98 | id | 48 | 79 | Daraxt | 65 | 96 |
| 7. | K | 94 | 96 | uk | 58 | 86 | Kartoshka | 74 | 92 |
| 8. | G’ | 82 | 94 | g’e | 65 | 87 | G’ildirak | 70 | 98 |
| 9. | Q | 76 | 97 | qa | 82 | 90 | Qiwr | 61 | 89 |

The research shows that when a neural network was built for the MFCC and spectral parameters of the speech signal, the neural network was able to solve this problem with high accuracy. Joints and words with a detection rate of less than 85% were re-analyzed in a neural network constructed based on spectral parameters. The results showed that a neural network built on spectral parameters was found to be more efficient than a neural network built on MFCC parameters.

### IV. CONCLUSION

Given that there are several areas and areas of speech signal processing, this article presents mathematical models of algorithms used in children’s speech development and their experimental results. At the same time, a model of speech recognition in the Uzbek language has been developed, the results of experimental research in the recognition of speech signals. The following problems have been solved and the results of experimental research have been obtained in solving the problems of recognizing speech signals.

1. A general model for the recognition of Uzbek speech has been developed and a general model has been developed that provides optimum high accuracy and precision in the processing of speech signals;
2. Solutions of artificial intelligence algorithms based on spectral parameters of speech were analyzed as a result of experiments.
3. The problem of recognizing speech sounds, syllables and words used in children's speech therapy for the Uzbek language was solved using artificial intelligence algorithms; 4. Speech therapy is recommended to use neural network algorithms in solving the problem of recognizing children's speech sounds, syllables and words.

APPENDIX

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№ 24-19-F “Development of computer programs in the rehabilitation of speech defects in preschool and school-age children” research project is to create computer programs for speech development for children with speech development or retardation among preschoolers.

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