



Research article

Strategy of mine ventilation control in optimal mode using fuzzy logic controllers

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Abstract. The issues related to improving the efficiency of automatic ventilation control systems of mines that regulate the air supply to the mine in accordance with the need are considered. During the tests of such a system in the 3RU mine of OAO Belaruskali, the shortcomings of its existing, implementation, associated with the incorrect choice of the most difficult-to-ventilate direction, were revealed. The possibilities of implementing a control strategy, in which the system automatically determines the optimal configuration of the operating modes of fans and regulators, are demonstrated. As an alternative to the implemented algorithms, it is proposed to use a fuzzy control device to account for the nonlinearity of the dependence of the input and output parameters of ventilation equipment and to set the conditions for the optimal operating mode of the system in a declarative form. To assess the effectiveness of the proposed approach, the data of simulation modeling of the current ventilation mode and the transition from one ventilation mode to another are analyzed with comparison with the actual data of the system operation. The simulation results show that the use of an upgraded control scheme for the main ventilation fan based on fuzzy logic in the implementation of automatic ventilation control systems makes it possible to eliminate the possibility of a shortage of fresh air in the regulated directions of its movement, as well as excessive power consumption of the main ventilation fan.

Keywords: automatic ventilation control system; fuzzy control; PID-control; main fan unit; aerodynamic drag; air-depression survey; mine

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Introduction. The approach to ventilation of underground mines (ventilation-on-demand) [1-3] involves the construction of automatic ventilation control systems (AVCS). The systems make it possible to regulate the volume of air supplied to the mine depending on current needs and, as a result, reduce the energy consumption of fans [4, 5]. The following requirements apply to AVCS [6]:

- to provide the required air consumption at the specified points of the distributed ventilation network in accordance with the technological regulations for the ventilation of underground mine workings of the mining enterprise;
- to ensure the minimum possible energy consumption of the main fan unit (MFU) to ensure the specified ventilation conditions;
- do not allow fluctuations in the parameters of the equipment during transients;
- the duration of the change of ventilation modes should not exceed the interval of change in the need for ventilation.

The model of the mine ventilation network is an oriented graph, each branch of which is characterized by aerodynamic drag. If a ventilation regulator is installed in the branch (for example, an automatic



ventilation door – AVD), additional resistance is created. In the branches, there may be sources of air draft – ventilation installations – creating a pressure fall.

The criterion for the optimality of ventilation can be formulated as follows:

$$\begin{cases} Q_i = Q_i^r \quad \forall i = 1 \dots K; \\ N_{\text{MFU}}(\omega, \phi_1, \dots, \phi_K) \rightarrow \min \text{ when } \omega \rightarrow \omega^{\text{opt}}, \phi_i \rightarrow \phi_i^{\text{opt}} \quad \forall i = 1 \dots K, \end{cases} \quad (1)$$

where Q_i , Q_i^r – the actual and required air consumption on the i – ventilation regulator, respectively; N_{MFU} – power consumption MFU; ω – MFU rotation speed; ϕ_i – opening angle of the ventilation regulator; ω^{opt} , ϕ_i^{opt} – optimal values of the MFU rotation speed and the opening angle of the ventilation regulators for the specified conditions; K – number of ventilation regulators.

The first condition in (1) means that the actual air flow in the regulated sections of the ventilation network meets the regulatory requirements that ensure safe and comfortable working conditions. The second condition indicates the dependence of the fan's energy consumption not only on its own operating parameters (rotation speed), but also on the position of the ventilation regulators. The same air consumption in the ventilated areas of the mine can be provided by many combinations of their opening angles, each of which leads to different energy consumption by the fan.

The task of finding the optimal ventilation mode is to find such AVD opening angles and MFU rotation speed at which the required air consumption on the ventilation regulators will be provided, and the fan power consumption will be minimum. Ventilation regulators (AVD) must create the minimum possible additional aerodynamic drag in the workings in which they are installed, while ensuring the required air distribution in the mine.

The approach, which is based on the mine ventilation network model represented by a directed graph, is the most general approach to solving the optimization problem. Data from the current air flow sensors at the control points of the ventilation network are transmitted to a computer model of the mine ventilation network. If a deviation is detected between the required and actual air flow values, the air distribution is calculated using the ventilation network model to determine the new parameters of the control means. The parameters determined as a result of solving the air distribution problem are transmitted to the control system, which sets the calculated values of the MFU rotation speed and AVD positions. The process is repeated each time when the required and actual air consumptions at the control points mismatch.

In the real conditions of mine operation, this approach has a number of disadvantages. Firstly, it requires the installation of a large number of air flow sensors sufficient to calibrate the model, which for an underground network of workings requires significant costs for the arrangement of communication lines and subsequent maintenance. Secondly, due to the constant development of mining operations, the model of the ventilation network requires constant updating in manual mode (the length of existing mine workings is changing, worked out sites are being removed of service, new ones are being introduced) [7]. Such updating requires regular measurements by mine specialists and reflection of all changes in the model, which increases the influence of the human factor on the quality of management.

Existing implementations of ventilation equipment control systems within the framework of ventilation-on-demand projects (Garpenberg, Boliden, Sweden [8]; mines of South Africa [9]; El Teniente, Chile [10]; Totten Mine, Vale Inco Ltd, Canada [11], etc.) are characterized by relative compactness of mines working out polymetallic and iron ore deposits. However, mines on the territory of the EEU (in particular, potash deposits) are characterized by larger



areas of mine fields (more than 100 km²), volumes of air supplied to the mine (more than 50,000 m³/min) and a significant number of simultaneously operating working zones (mining sites) that are very remote from each other. At the same time, the high intensity of mining operations, coupled with high mine productivity (up to 20 million tons of ore per year), leads to a constant change in the configuration of the mine workings and, accordingly, the ventilation network. These features complicate the task of operational redistribution of air while minimizing the energy consumption of the main ventilation fan.

In the approach to the construction of optimal ventilation control systems [12-14], developed at the Mining Institute of the Ural Branch of the Russian Academy of Sciences, the interpretation of operational data on the actual air distribution into the ventilation network model is excluded. At the same time, the control algorithm of the system corrects the tasks of the control means based on data on the current air flow directly to the AVD and MFU without the need to recalculate the total air distribution in the mine on the topology model of the ventilation network [15]. It is established that the necessary condition for the optimal ventilation mode is the presence of at least one fully open ventilation regulator (i.e., a position in which the additional resistance in the corresponding mining is zero). Current implementations of the systems are based on the control of ventilation equipment (AVD and MFU) using PID-control [16]. The use of fuzzy controllers has become widespread in various fields of industry [17].

The traditional use of fuzzy logic controllers in the implementation of automatic control systems is reduced to solving the problem of nonlinear conversion of input signals into output under the following conditions: the dependence of output parameters on input is unknown or difficult to calculate; there are significant dynamic disturbances and measurement noise; constant variability of the system itself, which requires constant correction of control parameters.

Fuzzy sets of the first order [18] allow to take into account the nonlinearity of the relationship of parameters, and fuzzy sets of the second order [19, 20] – to overcome the problems associated with the uncertainty of measurements of input parameters, as well as model parameters. The use of fuzzy controllers is designed to reduce the duration of the system output to the required stable state. Independent of the specific task, algorithms for configuring fuzzy controllers [21-23] allow to adjust the parameters of fuzzy control following changes in the controlled system. The rule base for a fuzzy controller can be either trivial (for example, a small control error corresponds to a small reaction of the system), or contain more complex control logic set by an expert. There are methods of automatic correction of the rule base, allowing achieving the desired characteristics of the equipment [24]. One of the applications of fuzzy control is the determination of coefficients for the PID-controller [25, 26]. Fuzzy controllers are used in a wide range of areas, including when implementing control in hydraulic systems [27, 28] and ventilation control systems (maintaining the required air parameters in stovehouses [29] and intensive care units [30], ventilation of industrial premises [31]).

The purpose of the article is to implement the AVCS management strategy, which assumes ensuring the required air consumption on ventilation regulators while minimizing the energy consumption of the main fan unit based on setting optimality conditions directly in the fuzzy controller rule base.

Methodology. To analyze the work of the existing implementation of the automatic ventilation control system, a number of tests were carried out in the conditions of the 3RU mine of OAO Belaruskali [32, 33]. This makes it possible to assess the speed of the system transition from one ventilation mode to another, its response to disturbances, the stability of the equipment and the efficiency of its functioning.

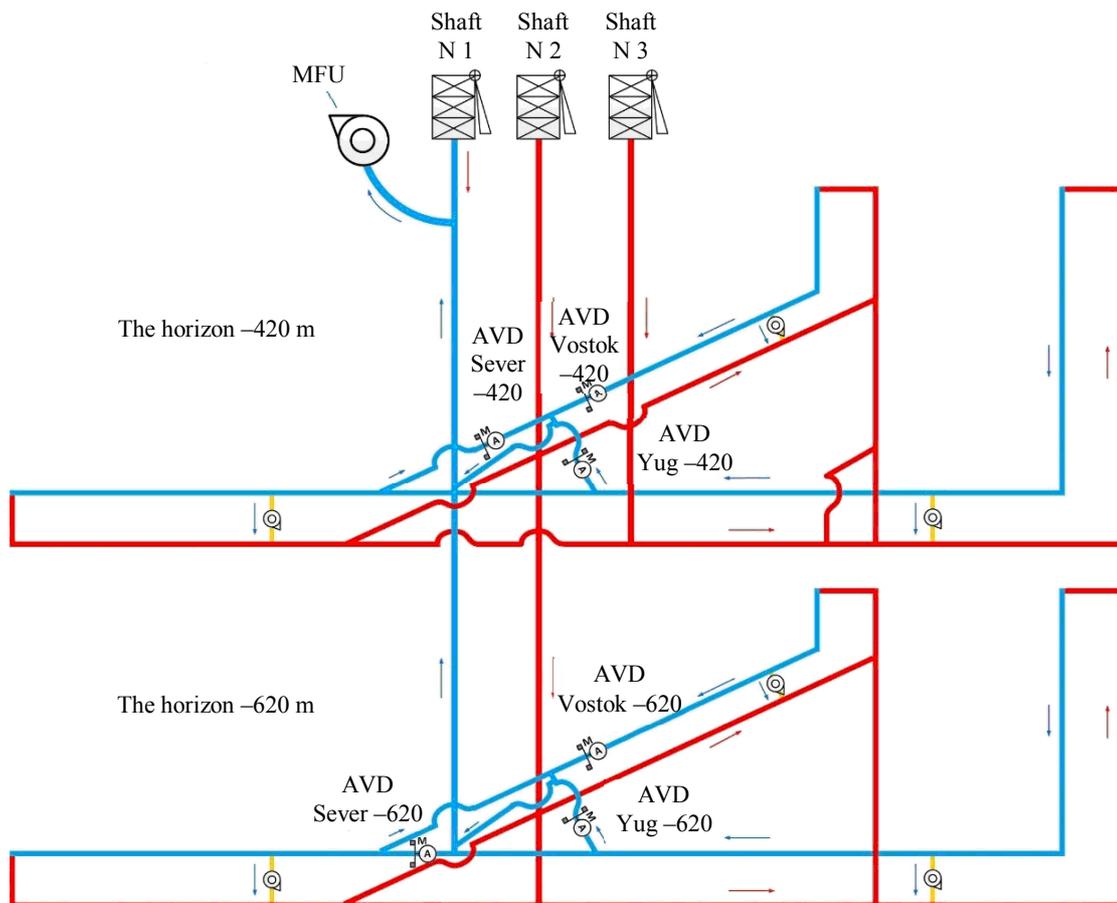


Fig.1. Schematic diagram of the ventilation of the 3RU mine of OAO Belaruskali, indicating the location of the MFU and AVD

	MINING						MAINTENANCE					
	MFU on			MFU off			MFU on			MFU off		
	AVD, m ³ /min	MFU, m ³ /min	Recycle ratio, %	AVD, m ³ /min	MFU, m ³ /min	Remote panel, m ³ /min	AVD, m ³ /min	MFU, m ³ /min	Recycle ratio, %	AVD, m ³ /min	MFU, m ³ /min	Remote panel, m ³ /min
-420 m												
Sever	500	500	0.00	3480	500	0	500	500	0.00	2650	500	1000
Vostok	500	500	0.00	1740	500	0	500	500	0.00	1420	500	270
Yug	4480	1500	0.00	6000	500	0	2630	1500	0.00	4640	500	668
-620 m												
Sever	5040	1600	0.00	4750	500	0	4770	1500	0.00	3550	500	1000
Vostok	1920	1700	0.00	4030	500	0	700	1710	0.43	2560	500	900
Yug	4380	1800	0.00	6000	500	0	2650	1800	0.00	5150	500	900

Fig.2. A fragment of the videogram of the controller interface for managing AVCS operation modes

The mine AVCS consists of one surface MFU, including the main and backup fans, and six automatic louver-type regulators (AVD) installed in each of the main directions (“Sever”, “Vostok”, “Yug”) of the horizons being worked out (-420 and -620 m) [34] (Fig.1).

Mine ventilation control consists in changing the rotation angles of the louver regulators depending on the deviation of the specified air flow from the actual one (if the air flow is greater than the required one, then the regulator is covered, increasing the aerodynamic drag of the mine in which



it is installed, and vice versa), as well as changing the performance of the MFU (by changing the speed of the impeller). For different operating modes of the mine, the required air flow is set for each regulator (Fig.2). At the beginning of each shift, the dispatcher selects the necessary mode for each AVD, depending on whether mining operations (high air consumption) or repair work are being carried out in the direction.

The required supply of the main ventilation fan, adjustable rotation speed ω , is defined as the sum of the specified costs for all regulators Q_{Σ}^r and the set value of external leaks Q^l (the leakage coefficient is calculated by regular measurements by the ventilation service). The calculated fan supply is adjusted for the amount of lack/excess of air ΔQ_{main} in the “main” direction, at which the value of the opening angle is fixed in accordance with the preset value ϕ_{main} (Fig.3).

With optimal ventilation mode, the “main” direction corresponds to the most difficult-to-ventilate direction in which the regulator should be in the fully open position. However, due to the fact that each regulator independently determines the degree of its opening depending on the specified air flow setting, as well as the technical complexity of determining the aerodynamic drag of the corresponding output at the current time, the “main” direction is determined from the maximum value of the required air flow or is set manually by the operator. An incorrectly selected “main” direction leads to one of the situations: the main ventilation fan consumes excessive electricity, since the general configuration of the position of the regulators creates excessive resistance in the ventilation network; in one of the directions (not the “main” one), there is a lack of air when the corresponding regulator is in the fully open position.

In order to determine the possibilities for improving the AVCS control scheme, tests were carried out to assess the possibility of automatically determining the most difficult-to-control direction in the AVCS control cycle. Figure 4 shows the characteristics of the MFU and AVD installed in the mine under study.

As part of the tests, the performance of the main ventilation fan was controlled manually by means of a stepwise change in the speed of the MFU impeller, which leads to the redistribution of air in the shaft and the automatic opening of the regulator in the most difficult-to-ventilate direction. At the beginning of the tests, “-620 Yug” was chosen as the “main” direction (which corresponded to the most difficult-to-ventilate direction for the current ventilation mode according to the simulation results), for which the specified air flow was changed from 4620 to 2770 m³/min.

A sequential decrease in the MFU rotation speed (Fig.5) led to a reconfiguration of the door position, which increased the actual supply of MFU at each segment of a fixed rotation speed (by reducing the aerodynamic drag of the network).

The dynamics of the AVD position first demonstrates their cover (when the actual air consumption on the fan was excessive), then in the process of reducing the fan speed, leading to a lack of air in some directions, a gradual opening to a level corresponding to the optimal position. The fluctuations of the “AVD -620 Yug” controller are caused by the delayed effect of changing the fan operating mode on remote points of the ventilation network (i.e. the controller manages to work out the setpoint change independently before the fan has an effect). As a result, the most difficult-to-ventilate direction was “-420 Yug”, the automatic regulator on which moved to a fully open (technically possible) position.

According to the test results, it can be concluded that the automatic ventilation control system can independently determine the most difficult-to-ventilate direction, i.e. the direction with the greatest aerodynamic drag, at which the regulator will be in the most open state. In case the lack of air in the ventilation network, an increase in the air consumption supplied to the mine can be achieved just by changing the position of the regulators without increasing the fan speed, if none of the regulators is in the fully open position.

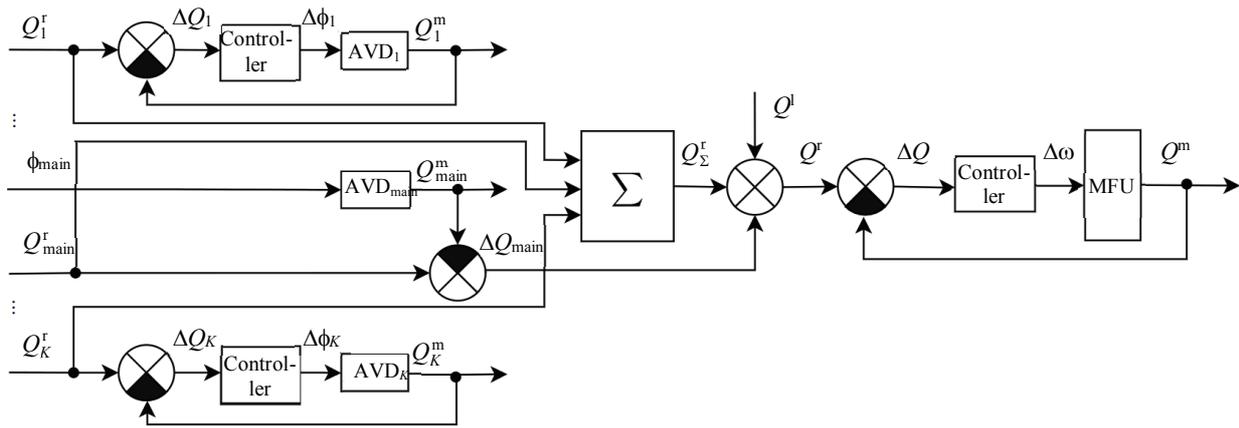


Fig.3. Information structure of the existing AVCS

Q_i^m , Q^m – measured (actual) air consumption for i th AVD and MFU; Q_i^r , Q^r – required (set) air flow on the i th AVD and MFU; ΔQ_i , ΔQ – deviation of air flow on the i th AVD and MFU; Q_{main}^m , Q_{main}^r – measured (actual) air flow and its deviation on the AVD of the “main” direction; $\Delta\phi_i$ – calculated increment of the angle of the i th AVD; $\Delta\omega$ – calculated increment of rotation speed MFU

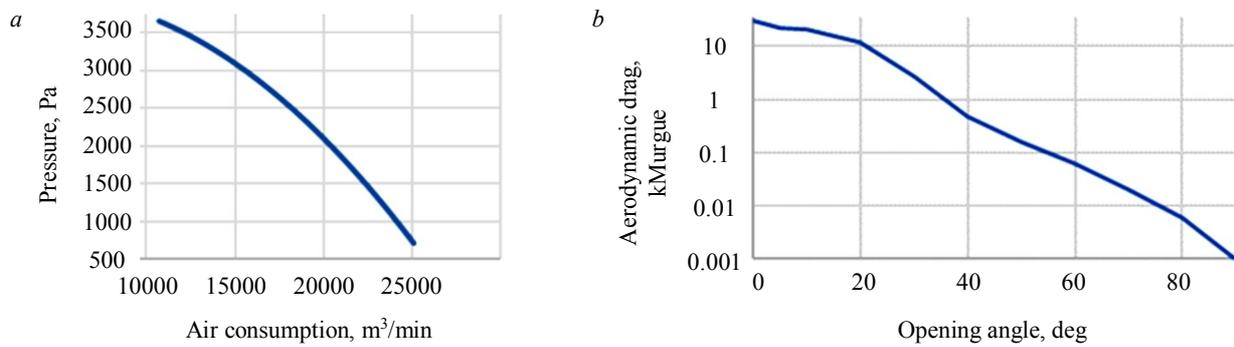


Fig.4. The dependence of the pressure on the air flow MFU (a) and the aerodynamic drag of the regulator on the opening angle (b)

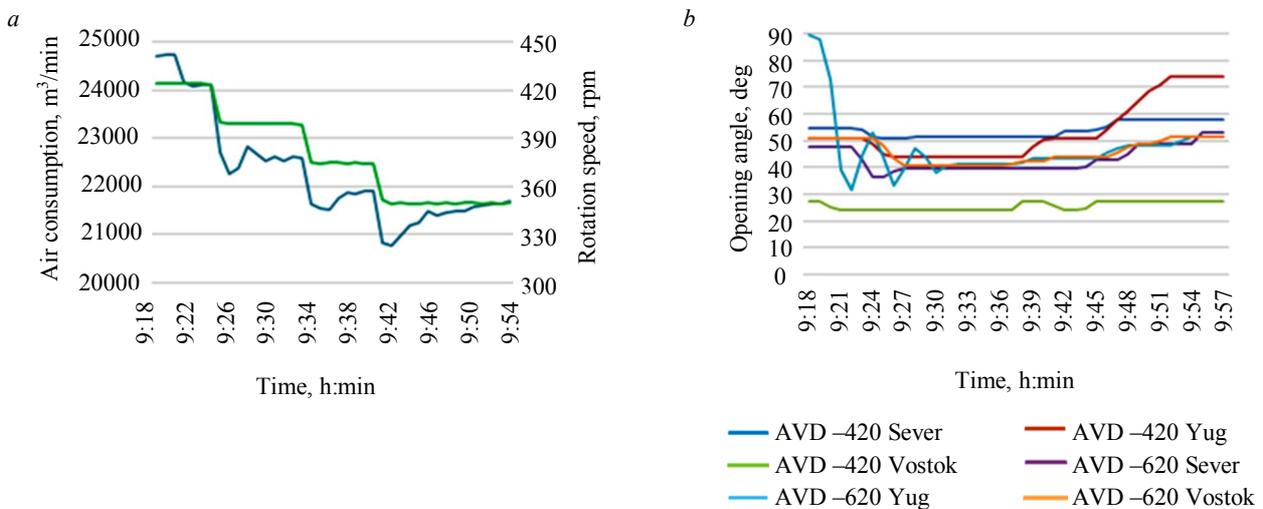


Fig.5. Air consumption per MFU (a) while changing the opening angles of the AVD (b)



To implement a management strategy that makes it possible to compensate for the side effects of excluding the predetermination of the most difficult direction, it is advisable to use a fuzzy control device due to the following circumstances:

- changing the position of the regulator in the most difficult-to-ventilate direction leads to a change in the aerodynamic drag of the ventilation network, which non-linearly affects the air flow provided by the fan;
- the use of linguistic variables and the base of production rules, defined in terms of the subject area, allows to reflect the criteria for choosing the most difficult direction.

Let us consider a mathematical description of a fan control model based on fuzzy logic. It is necessary to set fuzzy variables ΔQ and ϕ_{\max} as input parameters to determine the increment of the fan frequency:

$$\Delta Q = \min_{i=1...K} (Q_i^r - Q_i)^+ - \max_{i=1...K} (Q_i^r - Q_i)^+; \quad (2)$$

$$(x)^+ = \begin{cases} 0, & x < 0 \\ x, & x \geq 0 \end{cases}$$

where ΔQ – lack/overconsumption of air, defined as the difference between the minimum overconsumption and the maximum lack of air for all controllers.

The choice of the maximum shortage and minimum excess consumption for all regulators in expression (2) is due to the need to ensure the actual consumption is not less than the specified one, since compliance with regulatory requirements is a priority task of management;

$$\phi_{\max} = \max_{i=1...K} \phi_i, \quad (3)$$

where ϕ_{\max} – the maximum opening angle among all regulators.

For fuzzy variables, we define the following term sets:

$$\Delta Q - \{NL, NS, Z, PS, PL\};$$

$$\phi_{\max} - \{S, M, L\};$$

$$\Delta \omega - \{NL, NS, Z, PS, PL\},$$

where $\Delta \omega$ – increment of the fan impeller speed; N, P – “negative” and “positive”, Z, S, M, L – “zero”, “small”, “medium”, “large”. For example, the designation NL corresponds to the term “negative large”.

Fuzzy sets that make up term sets for fuzzy variables are defined by membership functions:

$$\mu(x) = e^{-\frac{(x-\bar{x})^2}{\sigma}}, \quad (4)$$

where \bar{x}, σ – parameters that define the center and width of the function graph.

If there is one main ventilation fan, the rules for choosing the optimal ventilation mode for a given air consumption can be formulated as follows: a given air consumption should be provided on each regulator; at least one of the regulators should be fully open (i.e. not to create additional aerodynamic drag).

The rules for fan control can be formulated as follows:



- if there is an excess of air on the controllers in the total, it is necessary to reduce the rotation speed;
- if there is a lack of air, while at least one of the controllers is fully open, it is necessary to increase the rotation speed;
- if there is a lack of air and at the same time no controller is fully open, the rotation speed remains unchanged.

If there is a lack of air in the system, but none of the regulators is fully open, the regulators change their position at a constant fan speed. During the tests, it was demonstrated that in this case the total air consumption supplied to the mine increases. Such a transfer of control from the fan to the controllers can be reflected in the logic of the rule base.

Let us consider the bases of fuzzy inference rules for determining the required increment of the fan impeller speed (Table 1) [35]. The row and column headers correspond to the sets of input variables (ΔQ and ϕ_{\max}), included in the antecedents of the rules, the values at the intersection of rows and columns correspond to the sets of the output variable from the sequence of rules.

Table 1

Rule bases for fan control

	<i>NL</i>	<i>NS</i>	<i>Z</i>	<i>PS</i>	<i>PL</i>
<i>L</i>	<i>NL</i>	<i>NS</i>	<i>Z</i>	<i>PS</i>	<i>PL</i>
<i>M</i>	<i>NS</i>	<i>NS</i>	<i>NS</i>	<i>Z</i>	<i>Z</i>
<i>S</i>	<i>NL</i>	<i>NL</i>	<i>NL</i>	<i>Z</i>	<i>Z</i>

The rule corresponding to the first row and column are as follows: if ΔQ – negative large, ϕ_{\max} – large, then $\Delta\omega$ – negative large.

The cells highlighted in red correspond to a situation when none of the controllers is fully open, while there is a cumulative lack of air in the system. In this case, the control signals are transmitted to the regulators: the fan frequency remains unchanged, and the controllers must be rebuilt in such a way as to reduce the overall resistance of the network to increase the total air consumption supplied to the shaft at the same fan speed.

The degree of truth of the k th rule is calculated by the formula

$$\mu_{B_k}(\Delta\omega) = \mu_{A_k^{\Delta Q}}(\Delta Q) \mu_{A_k^{\phi_{\max}}}(\phi_{\max}), \quad (5)$$

where $\mu_{A_k^{\Delta Q}}$, $\mu_{A_k^{\phi_{\max}}}$ – membership functions of fuzzy sets of input variables; μ_{B_k} – the membership function of the fuzzy set of the output variable.

To calculate the resulting value of the output variable, the formula is used

$$\Delta\omega = \frac{\sum_k \overline{\Delta\omega_k} \mu_{B_k}(\Delta\omega)}{\sum_k \mu_{B_k}(\Delta\omega)}, \quad (6)$$

where $\overline{\Delta\omega_k}$ – the central value of the fuzzy set of the output variable of the k th rule.

Formulas (5) and (6) represent one of the options for implementing fuzzy logical inference, independent of the solution of a specific problem, in which the product operation is used to calculate the membership functions of each rule. The resulting value of the output variable is defined as the weighted average of the degrees of membership of all rules.

Figure 6 shows an upgraded AVCS control scheme: first, the AVD changes the opening angle in accordance with the current air flow mismatch ΔQ_i , then the MFU changes the rotation speed in

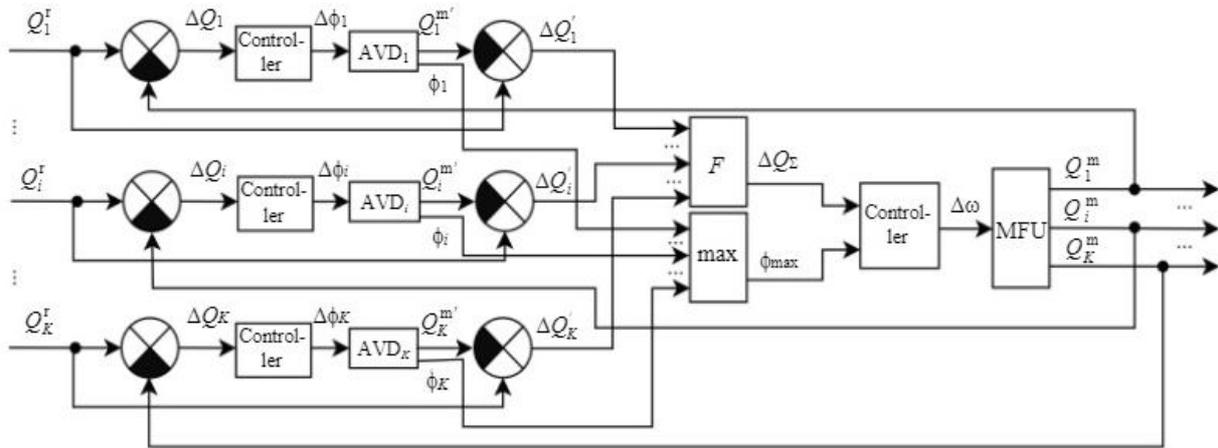


Fig.6. AVCS control scheme with automatic output to the optimal ventilation mode

$\Delta Q_i'$ – deviation of air flow on the i th AVD (before changing the frequency of the MFU); $Q_i^{m'}$, Q_i^m – measured (actual) air flow on the i th AVD (before the MFU frequency change)

accordance with the given fuzzy control model. The air flow values Q_i^m measured as a result of the cumulative change in the opening angle of each AVD and the rotation speed of the MFU are used to calculate the control error on the AVD. The scheme excludes the preliminary selection of the most difficult-to-ventilate direction, in which, according to the logic of the MFU controller (Table 1), the corresponding AVD will automatically switch to a fully open state. The function F is determined according to formula (2) and calculates the total deviation ΔQ_{Σ} for all AVDs.

Discussion. To evaluate the effectiveness of the proposed algorithms, the AVCS operation was modeled in the developed module for the analytical complex “AeroSet” on the ventilation network of the mine, built according to the results of air-depression survey, and the simulation results were compared with the actual data of the equipment in the following situations: the current ventilation mode with the specified air consumptions; transition from one ventilation mode to another, accompanied by a change in the set air consumptions on the ventilation regulators.

Simulation of the current ventilation mode was based on December 18, 2019 AVCS data.

Table 2

Actual and model air consumption for the current ventilation mode

AVD	Set consumption, m ³ /min	Actual consumption, m ³ /min	Deviation from the set consumption, %	Model consumption, m ³ /min	Deviation from the set consumption, %
-420 Vostok	1420	1190	-16.2	1421	0.07
-420 Sever	1830	1790	-2.19	1830	0.0
-420 Yug	4590	5439	18.5	4596	0.13
-620 Vostok	3300	3259	-1.24	3112	-5.7
-620 Sever	3200	3078	-3.81	3202	0.06
-620 Yug	3520	4021	14.23	3401	-3.38

Note. A significant lack of air is highlighted in red, an excess is highlighted in green.

Comparison of actual and specified consumptions (Table 2) shows that in the main directions the deviations of air consumptions are within 10 %, but in some areas, there is a significant overconsumption or lack of air.

In the steady-state configuration of the parameters of the AVCS elements, according to the simulation results, the air consumption on the fan is 19997 m³/min at 352 rpm, while the actual fan performance is 19292 m³/min at 373 rpm.



Comparison of the position of automatic ventilation doors on the design model and in the control system (Table 3) shows that the configuration of the door position in the AVCS is set in a non-optimal way.

Table 3

Actual and model opening angles of the controllers for the current ventilation mode

AVD	Actual opening angle, deg	Model opening angle, deg	AVD	Actual opening angle, deg	Model opening angle, deg
-420 Vostok	24	35	-620 Vostok	47	90*
-420 Sever	45	37	-620 Sever	40	62
-420 Yug	77*	78	-620 Yug	44	90*

* Controllers with the maximum opening angle.

The “main” direction in AVCS is the “-420 Yug” direction (the largest value of the required air consumption), which leads to suboptimal operation of the fan to ensure the specified air consumptions. Based on the simulation results, the most difficult-to-ventilate direction is “-620 Vostok” or “-620 Yug”.

AVCS operation data for February 26, 2019 were used to simulate changes in the ventilation mode. During the day, the required air consumption was changed on several controllers, which led to a significant redistribution of air throughout the mine (Table 4).

Table 4

Actual and model air consumption when changing the ventilation mode

AVD	Initial set consumption, m ³ /min	Final set consumption, m ³ /min	Final actual consumption, m ³ /min	Deviation from the specified consumption, %	Final model consumption, m ³ /min	Deviation from the specified consumption, %
-420 Vostok	1400	1400	1420	1.4	1411	0.8
-420 Sever	2510	2810	2812	0.0	2817	0.2
-420 Yug	4290	2880	3006	4.3	2884	0.1
-620 Vostok	2890	2890	2954	2.2	2955	2.2
-620 Sever	2620	2620	2592	-1.1	2624	0.2
-620 Yug	5310	3800	3929	3.4	3609	-5.0

The use of algorithms based on fuzzy control allows the MFU to reach the specified ventilation mode faster (Fig.7). A significant subsidence of the rotation speed, as well as the opening angle of the controller in the most difficult-to-ventilate direction “-620 Yug”, is due to a significant reduction in the required air consumption on this controller and the working out of this AVD setpoint to reduce the actual air consumption on the MFU, which is justified from the point of view of control stability. The return of the “AVD -620 Yug” controller to the fully open state confirms the ability of the system itself to determine the most difficult direction.

In this situation, the “main” direction in the AVCS is set correctly and corresponds to the condition of the largest air consumption. The set air consumption in the case of actual and model data is within the permissible 5 % range of deviations.

Conclusion. The conducted studies have shown that when implementing AVCS, the preliminary choice of the most difficult-to-ventilate direction based on the largest air consumption (or at the discretion of the operator) can lead to the failure to provide the specified consumptions in the directions and the overconsumption of electricity by the fan.

The proposed approach to the formation of the fuzzy controller rule base makes it possible to realize the possibility of automatically selecting the most difficult-to-ventilate direction, the AVCS output to the optimal ventilation mode is provided, minimizing the energy consumption of the MFU.

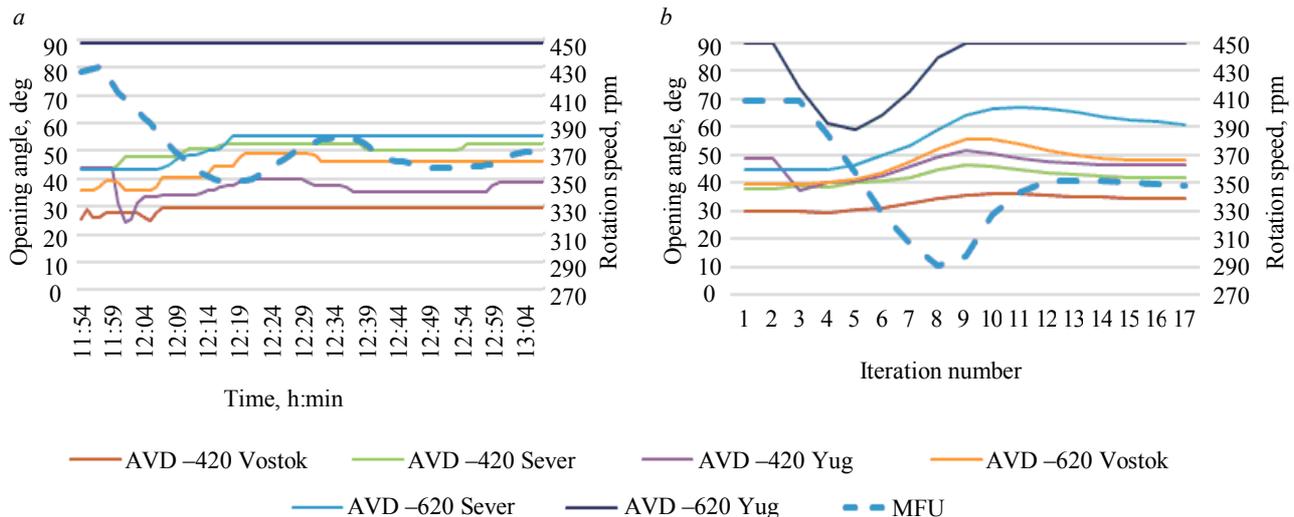


Fig.7. Change in the opening angles of the controllers and the rotation speed of the MFU during the transition from one ventilation mode to another: MFU and AVD AVCS data (a), simulation results (b)

The upgraded control scheme was tested on an up-to-date model of the ventilation network of the 3RU mine of OAO Belaruskali, while the results of model calculations were compared with the actual AVCS data for specific time intervals of its operation.

The proposed ventilation management strategy eliminates the lack of air in regulated directions without the need for external control.

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