

# An improved technique for identification of mathematical model parameters of thermal power equipment and assessment of its performance

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**Abstract.** The problems of state estimation of thermal power system operation and identification of mathematical model parameters have not been acceptably solved due to the complexity of studied objects and their mathematical models, and the lack of effective methods, algorithms and computer programs to solve the required mathematical problems. The results of solving the indicated problems are of importance as such, and play a great part in the qualitative solution to the problems of thermal power equipment control, e.g., the problems of optimal load dispatch among thermal power plant units and optimal control of thermal power equipment operation conditions. The paper describes a technique improved by the author for identification (adjustment, verification) of mathematical model parameters for complex thermal power equipment. The technique allows us to more effectively detect gross errors in measurements of control parameters used for identification of the mathematical model of the studied equipment, to evaluate correctness and rectify errors in the mathematical model construction, and to improve identification accuracy. An improved technique for identification of mathematical model parameters was tested on a detailed mathematical model of the present-day 225 MW generating unit that was constructed by the author. The paper presents results of solving the identification problem of mathematical model parameters of a generating unit and an example of the optimization calculation of the real operation condition in order to reduce specific fuel consumption for electricity generation. In addition, the paper discusses an issue of assessing the identification accuracy of mathematical model parameters of thermal power equipment that depends on the accuracy of measurements of control parameters used to adjust the model, as well as on the correctness of the mathematical model construction and the calculation technique applied.

**Keywords:** identification of parameters, mathematical modeling, accuracy criterion, measured control parameters, model adjustment, real-time control, relative discrepancies, operation conditions, state estimation, coal-fired generating unit.

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## 1 Introduction

Thermal power plants (TPPs) on fossil fuel continue to form the basis of electric power industry in the Russian Federation, especially in the regions of Siberia and the Far East. The thermal power units consume an essential portion of produced fossil fuel and other resources. Hence, the problems of improving the energy and economic operation effectiveness for such units are most topical and noteworthy.

Note that the operation effectiveness of thermal power units (TPUs) directly depends on operating conditions and real-time control of equipment. To improve the control efficiency of power plant equipment, in turn, the operation personnel should have a “feedback”, in other words, monitor changes in equipment parameters and its characteristics difficult or

impossible to meter (burnt fuel consumption, generating unit efficiency, specific fuel consumption, etc.) with change in the control actions.

Besides, the real state of thermal power equipment changes during operation, for example, due to deposition of salts in the turbine flow part, pollution of the heat exchange surfaces of boilers, regenerative heaters and others, which influence operation conditions of equipment and its efficiency. Thus, the problem of assessing the state of main thermal power equipment at thermal power plants (TPPs) is critical for real-time control of operation conditions of TPUs.

The present-day thermal power units such as coal-fired generating units and their boiler units, steam turbines and other auxiliary equipment are technical systems with highly complicated flow diagrams, diverse elementary composition and operation conditions.

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Therefore, the methods of mathematical modeling and optimization of schemes and parameters are the main tools to study thermal power equipment at TPPs.

The keystone for application of the methods of mathematical modeling and optimization of thermal power equipment and thermal power plants was laid in the early works by the scientists at Melentiev Energy Systems Institute. The works by G.B. Levental and L.S. Popyrin focus on optimization of continuous and discrete parameters of TPUs of different types and flow diagrams, present principles of mathematical modeling automation of TPUs and describe approaches to optimization of such equipment under uncertainty of initial information [1, 2]. The methods of mathematical modeling of thermal power units were evolved by the Russian scientists: F.A. Vulman [3], A.A. Palagin [4], V.M. Borovikov [5].

The problems of state estimation and identification of mathematical model parameters for calculation of operation conditions of power systems considering errors in measurements were studied by A.Z. Gamm and his colleagues. Their work [6] describes approaches to detection of gross errors in measurements called “bad” data that are based on the method of test equations.

The problems of state estimation and identification of mathematical model parameters were also treated in the studies of pipeline systems. The work by N.N. Novitsky, which presents a comprehensive analysis of some problems and methods for state estimation that were devised considering specific features of hydraulic circuits, is among such studies [7].

The works by G.V. Nodrenko, Yu.V. Ovchinnikov [8] and G.D. Krokhin, M.Ya. Suprunenko [9] were among the first on this topic as applied to the thermal power industry. They present a technique for coordination of the heat and energy balance equations to solve the state estimation problem. However, the indicated works neither present the statement and solution of the identification problem of control parameters that cannot be metered directly, nor study the relationship between optimal solutions and errors in measurements.

The Department of Thermal Power Systems at Melentiev Energy Systems Institute has gained rich experience in the study of complex thermal power units and thermal power plants. The works by A.M. Kler and N.P. Dekanova are among the early works that focus on approaches to optimization of mathematical models of thermal power units at real-time control of operation conditions of CHPPs [10, 11]. A coordinated technique for the studied equipment diagnostics that is based on the joint solution of extremum optimization problems of state estimation and identification of characteristics of TPUs was proposed by A.V. Mikheev [12]. Moreover, an approach to the improvement of initial information quality by elimination of errors in the bad metered parameters is suggested in the work [13]. The studies by A.M. Kler, A.S. Maximov and E.L. Stepanova [14, 15] are among the latest works devoted to this topic. They focus on development of “high-speed” mathematical models of main equipment at TPPs to perform complex optimization calculations of operation conditions of

CHPPs, and also describe a technique for adjustment of mathematical models to a real state of the studied equipment. The technique allows the adjustment of mathematical model coefficients so that the obtained results correspond most accurately to the real equipment state, which validates optimization solutions.

The insufficiently extensive application of the effective methods of mathematical modeling, in general and their use to control operation conditions of TPPs, in particular are explained by some difficulties, such as significant complexity of mathematical models of current TPUs and need to adjust these models to the real equipment state changing in the course of time.

Thus, in practice the problems of thermal power system state estimation and identification of mathematical model parameters have no feasible solution due to the complexity of the studied objects and their mathematical models, and the lack of effective methods, algorithms and computer programs to solve the required mathematical problems. The results of solving the indicated problems are important as such and play an essential role for the qualitative solution of TPU control problems, e.g., for the optimal load dispatch among TPP units and the optimal control of operation conditions of TPUs and TPPs.

## **2 Description of the improved technique for identification of mathematical model parameters of thermal power equipment**

As was noted above, the works [14, 15] present techniques for adjustment (identification) of the mathematical model parameters of the studied thermal power equipment based on the measurements of control parameters that were taken during the tests on real equipment. These techniques, however, have two shortcomings, which, in specific situations, can prevent from successful solution of the stated problem.

First, the problem is solved successfully in case of lack of gross errors in the measurements among the metered parameters. However, in case of “bad” measurements with gross errors in any considered operation condition, the errors are redistributed among different metered parameters in one operation condition and, which is more important, among different conditions. Such redistribution prevents from the unique determination of an erroneous measurement and leads to incorrect solutions. The numerous calculations using several mathematical models have showed that practically there are always one or several “bad” measurements of control parameters in different operation conditions that cause a raw error in the final identification accuracy. This note is particularly true to old equipment with low accuracy sensors.

Second, the indicated techniques do not take into consideration the mathematical model errors of the studied equipment. The models of the main thermal power equipment at TPPs are based on the standard calculation methods and do not always describe real processes accurately enough. Moreover, at the stage of modeling the author often applies some simplifications,

e.g., neglect of a minor heat carrier flow in the flow diagram of TPU. This fact causes additional errors which should be taken into consideration in identification of the mathematical model parameters.

This paper presents a new improved identification technique. The backbone of the idea is to develop a new comprehensive technique consisting of three stages to solve the above problems and to improve the accuracy of mathematical model identification.

At the first stage of the identification problem solution the incorrect measurements of control parameters are revealed and eliminated from further calculations. The incorrect measurements are the values of the metered parameters that exceed the declared accuracy of sensors used in the tests. Such measurements can be detected by solving the minimization problem of the auxiliary coefficient  $\psi$  for each individual operation condition of equipment. The coefficient  $\psi$  corresponds to the value of the maximum relative deviation among all the metered parameters. The mathematical statement of the first identification stage is the following:

$$\min_{x_m, x_{um}, \theta, \psi} \psi \quad (1)$$

subject to:

$$H(y_m, x_m, x_{um}, \theta) = 0 \quad (2)$$

$$G(y_m, x_m, x_{um}, \theta) \geq 0 \quad (3)$$

$$\overline{x_{mj}} - \psi \cdot \sigma_{xj} \leq x_{mj} \leq \overline{x_{mj}} + \psi \cdot \sigma_{xj} \quad (4)$$

$$\overline{y_{mk}} - \psi \cdot \sigma_{yk} \leq y_{mk} \leq \overline{y_{mk}} + \psi \cdot \sigma_{yk} \quad (5)$$

$$\sigma = \frac{XB \cdot \alpha}{3 \cdot 100} \quad (6)$$

where  $H$  is the function of equality constraints that includes all equations of the mathematical model and its elements (equations of heat transfer, heat balance and others);  $G$  is the function of inequality constraints that includes physical and operating constraints on real equipment operation;  $\psi$  is the coefficient equal to the absolute maximum relative deviation of parameters (the parameters calculated by the mathematical model are with the upper bar, the parameters obtained by measurements on the real equipment – without the bar);  $\sigma_x^2$ ,  $\sigma_y^2$  are the measurement error variances of the vectors  $x_m$  and  $y_m$ , respectively,  $XB$  is the upper limit of the sensor measurement;  $\alpha$  is the class of sensor precision (in %).

The mathematical model parameters of the identification problem can be divided conventionally as follows: the parameters of  $x_m$  that are metered on the studied unit and are the input information for the mathematical model; the metered parameters of  $y_m$  that are the output information for the mathematical and the parameters of  $x_{um}$  that are not metered on the real unit and are the input information for the model. The array of the adjustable coefficients  $\theta$  of the mathematical model is selected for each model individually. They are intended to influence physical processes occurring in the mathematical model elements. Usually such parameters

comprise coefficients of thermal efficiency of boiler heat transfer surfaces, hydraulic resistances of heat exchangers, internal relative coefficients of turbine compartments and others.

The use of the so called “three-sigma” rule in this study is explained by the fact that the probability belief in this case equals 0.997. Therefore, it is possible to reasonably argue that all possible random errors in measurements with the normal distribution do not practically exceed 3 sigma in the absolute value. In equations (4, 5, 10, 11) the minimized auxiliary coefficient  $\psi$  specified initially by the large numbers 60-100 is used instead of the multiplier equal to 3. This substitution is necessary for considering the errors of the applied sensors and the imperfection of the calculation technique, as well as the errors in mathematical models. In the process of the optimization calculation (1, 7) this coefficient tends toward the value 3, but in practice it often takes somewhat lower value. Thus, the suggested technique makes it possible to evaluate an additional error caused by the imperfection of the standard calculation methods and simplifications included in the mathematical model of the studied TPU.

To determine erroneous measurements, it is needed to reveal active constraints on the deviation of the parameter metered on the unit from the parameter calculated on the mathematical model. The measurement value in this constraint can be treated as erroneous and excluded from further consideration. The calculations have showed that such an approach makes it possible to more effectively detect measurement errors and minimizes redistribution of erroneous measurements among the parameters in different operation conditions.

It is noteworthy that the calculations at this stage are not always possible. This concerns thermal power units with an insufficient number of sensors and hence, insufficient amount of initial information for identification for an individual condition. If the number of measurements is approximately the same or exceeds the number of adjustable coefficients, the calculations at this identification stage are possible. Otherwise, the second identification stage that is performed for several conditions concurrently is necessary. In such a case the number of control parameter measurements will be satisfactory to successfully solve the stated problem.

At the second stage of the identification problem solution the mathematical model of the studied equipment is tested for availability of modeling errors and elimination of remaining gross errors in measurements.

The statement of the optimization problem is similar to the problem solved at the first stage, except that it is solved for all considered conditions concurrently (as evidenced by the index  $i$  indicating the sequence number of equipment operation condition).

The problem is formulated as follows:

$$\min_{x_m^i, x_{um}^i, \theta, \psi} \psi, \quad (7)$$

subject to

$$H(y_m^i, x_m^i, x_{um}^i, \theta) = 0; \quad (8)$$

$$G(y_m^i, x_m^i, x_{um}^i, \theta) \geq 0; \quad (9)$$

$$x_{mj}^i - \psi \cdot \sigma_{sj} \leq \overline{x_{mj}^i} \leq x_{mj}^i + \psi \cdot \sigma_{sj}; \quad (10)$$

$$y_{mk}^i - \psi \cdot \sigma_{yk} \leq \overline{y_{mk}^i} \leq y_{mk}^i + \psi \cdot \sigma_{yk} \quad (11)$$

The calculations have showed that the solution to this problem allows the incorrect description of processes occurring in TPUs by the mathematical model to be detected. If the solution yields the parameters with a significant deviation of measurements in different equipment operation conditions, then it points to the lack of the required coefficient in the list of adjustable ones or the inaccurate construction of the mathematical model. Besides, the minor heat carrier flows that are neglected in the mathematical model construction for the studied equipment can cause an additional error in identification. Therefore, at this stage of calculations it is possible to make required changes in the model structure describing specific features of the studied thermal power equipment.

At the third stage of the identification problem solution the following optimization problem is solved.

$$\min_{x_m^i, x_{um}^i, \theta} f(y_m^i, x_m^i, x_{um}^i, \theta) \quad (12)$$

subject to:

$$H(y_m^i, x_m^i, x_{um}^i, \theta) = 0 \quad (13)$$

$$G(y_m^i, x_m^i, x_{um}^i, \theta) \geq 0 \quad (14)$$

$$x_{mj}^i - \psi \cdot \sigma_{sj} \leq \overline{x_{mj}^i} \leq x_{mj}^i + \psi \cdot \sigma_{sj} \quad (15)$$

$$y_{mk}^i - \psi \cdot \sigma_{yk} \leq \overline{y_{mk}^i} \leq y_{mk}^i + \psi \cdot \sigma_{yk} \quad (16)$$

$$f = \sum_{i=1}^R \left[ \sum_{j=1}^N \frac{(\overline{x_{mj}^i} - x_{mj}^i)^2}{\sigma_{sj}^2} + \sum_{k=1}^M \frac{(\overline{y_{mk}^i} - y_{mk}^i)^2}{\sigma_{yk}^2} \right] \quad (17)$$

where  $f$  is the objective function in equation 17 (the parameters calculated on the mathematical model are indicated by the upper bar, the parameters metered on the real equipment are given without the bar);  $R$  is the number of calculated conditions;  $N$  is the dimension of the vectors  $x_m$ ;  $M$  is the dimension of the vectors  $y_m$ .

The third identification stage aims to achieve the maximum possible closeness between the real equipment operation and the calculations on the mathematical model. The objective function  $f$  (17) contrary to the auxiliary coefficient  $\psi$  (1, 7) is the sum of squares of all relative discrepancies of control parameters in all equipment operation conditions. Hence, at the third stage it is possible to reduce both all relative discrepancies of the metered parameters and the maximum discrepancy at the second stage. After identification termination, the values of the adjustable coefficients  $\theta$  of the mathematical model are fixed and not subject to further changes, and the mathematical model is considered adjusted to the real equipment state.

Note that all identification stages are solved strictly in sequence and take into account changes in the

mathematical model that are made at the previous stages.

### 3 Results of identification of mathematical model parameters for the studied generating unit

In this study the present-day generating unit installed at the Kharanor condensing power plant (Yasnogorsk settlement, Trans-Baikal Territory) was taken as a prototype for the mathematical model. It consisted of a 225 MW steam turbine unit K-225-12,8-3P and a high pressure reheat boiler unit EII-630-13,8-565 БТ with a rating of 630 t/h. The flow diagrams and the mathematical models of the turbine and boiler units are described in greater detail in the papers [17, 18].

The mathematical model of a generating unit was constructed by the author using the software "System of computer-based construction of programs" developed at Melentiev Energy Systems Institute of SB RAS [19]. The calculation scheme of the generating unit comprises 100 elements and 169 ties between them. The obtained mathematical model contains 1153 input parameters, 1388 output parameters, 56 parameters of which are calculated iteratively and should be set with an initial approximation.

The calculation results obtained with the help of the improved technique developed by the author for identification of mathematical model parameters of thermal power equipment are given with respect to the described generating unit.

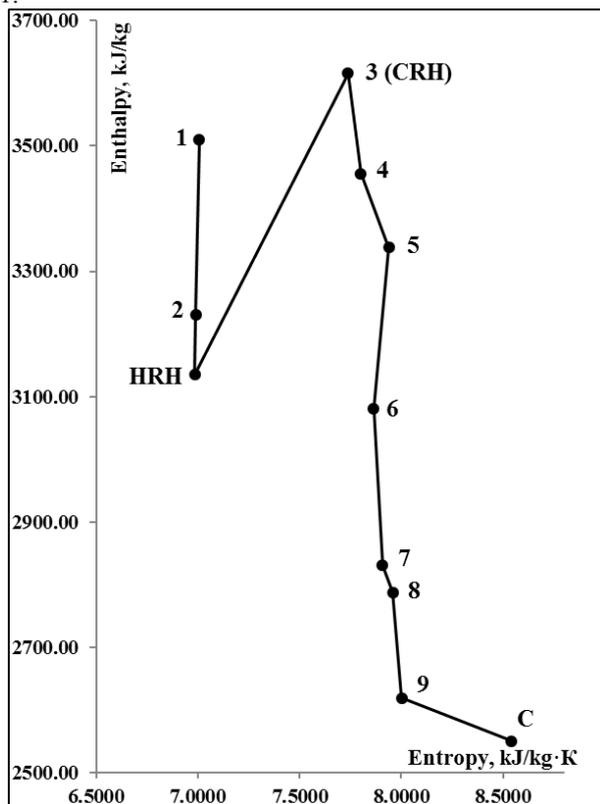
The values of the metered parameters at the scheme control points that are necessary for identification of mathematical model parameters were taken from the sensor readings given by the engineering personnel of the power plant. The accuracy class of the applied sensors are: for the sensors metering electric load - 1%, pressure - 1.5%, water and steam temperature - 2%, flow rate - 3%, gas temperature - 5%. The calculations were made for three selected operation conditions of the generating unit (the minimum load 125 MW, the load 227 MW and the maximum load 235 MW), each containing 60 metered parameter values at different flow diagram points, of which 5 metered parameters are the input (preset) information for the mathematical model and 55 parameters are the output (calculated) information for the mathematical model.

At the first identification stage the formed optimization problem (equations 1-6) comprised 76 optimized parameters and 313 inequality constraints for each considered operation condition. The array of the optimized parameters consisted of 59 adjustable model coefficients ( $\theta$ ); 11 parameters unmetered at the generating unit, such as fuel amount burnt in the boiler, excess air coefficient in the furnace, injections into 4 stages of steam coolers, heads of the main pumps of the main condenser duct, controllers of the superheated steam and minimized auxiliary coefficient ( $\psi$ ); 5 measurements that are the input information for the mathematical model.

The array of the inequality constraints consisted of 120 constraints on the minimum and maximum values of

the calculated control parameter values subject to sensor accuracy (10, 11); 192 physical constraints on the non-negativity of heat carrier flow rate in all model elements, the temperature head, the temperature and mechanical strength of boiler unit elements and some others; 4 operating constraints on the primary and secondary steam temperature, the turbine unit capacity and the flue gas temperature.

The calculations for each operation condition revealed some ambiguous measurements, namely pressure values at the inlet into some turbine compartments and steam flow rates at the condenser inlet. The measurement values presented by the power plant personnel were objectively tested by the steam expansion diagram constructed in the  $h,s$ -diagram in Fig. 1.



**Fig. 1.** Steam expansion process in the turbine unit for one operation condition in the  $h,s$ -diagram

The diagram figures indicate the number of turbine compartment with the steam pressure metered at its inlet (CRH – the steam coming to the cold pipeline of the steam reheat after expansion in the high pressure cylinder, HRH – the steam coming to the intermediate pressure cylinder from the hot pipeline of the steam reheat, C – turbine unit condenser).

There were no pressure measurements of the steam coming to the turbine compartments 7 and 9, as well as temperature measurement at the outlet from compartment 8, because the pressure and temperature sensors were not installed at these points, and therefore their values were set approximately. Besides, the pressure measurements at the inlet to compartments 2 and 5 caused doubt, as far as the internal relative efficiency of the turbine compartment cannot exceed

thermodynamically 100% (slope of the steam expansion diagram toward entropy decrease). Hence, these measurements are inaccurate and should be excluded from further calculations. In addition, it was decided to exclude the water flow rate at the condenser outlet from the metered parameters due to the gross discrepancy of the values of this parameter in two operation conditions.

The similar calculations of two remaining conditions on the mathematical model of the generating unit led to the same results. Thus, at the first identification stage 6 measurements were excluded in each considered operation conditions because of data lack or gross errors in measurements.

At the second identification stage the formed optimization problem (equations 7-11) comprised 109 optimized parameters and 909 inequality constraints. At this stage, contrary to the first, the problem was formed for three considered conditions jointly, which made it possible to refine the values of the array of adjustable coefficients subject to the effect of operation condition change on the thermal and energy efficiency of the generating unit elements.

In the previous work [20] the author suggested that the mathematical model elements be tested for modeling correctness and the corresponding changes be made at this identification stage. For example, the active constraint on the pressure measurement in the deaerator can indicate that this parameter changes nonlinearly with transition to the other condition, which should be taken into consideration in the mathematical model for the generating unit element replacing the adjustable coefficient (the throttling coefficient of steam that is delivered to the deaerator) with the quadratic dependence of type  $k_d = A \cdot x^2 + B \cdot x + C$ , where  $k_d$  is the throttling coefficient,  $x$  is the deaerator feed water flow rate that characterizes the turbine operation condition;  $A, B, C$  are the new optimized coefficients of the identification problem. Similar refinements of the mathematical models can be necessary in case of a sensible change in the internal relative coefficients of the turbine compartments from one condition to the next.

It is noteworthy that at the second identification stage the minimized objective function is the auxiliary coefficient being the maximum relative discrepancy (divergence between the measurement and the calculation of the control parameter subject to the corresponding sensor accuracy) or the relative discrepancies, if they are several. The calculations have showed that the considered minimax criterion vividly indicates the mathematical model “bottlenecks” for the studied TPU and assists in detection of gross errors in the control parameter measurements simply enough.

As applied to the studied generating unit the changes made at the second identification stage allowed the objective function to be minimized to the value 3.83. As was said above, the method is based on the three sigma rule, which means that the error equal to 3.0 or lower is perfectly explained by the declared sensor accuracy. The additional error equal in this case to 0.83 can be explained by the imperfection of the standard methods for calculation of boiler and turbine units, as well as the

required assumptions in the mathematical model of the generating unit.

At the third identification stage the formed optimization problem (equations 12-17) included 108 optimized parameters and 909 inequality constraints. The auxiliary coefficient  $\psi$  minimized at the second identification stage was excluded from the list of optimized parameters and fixed at the value 3.9. The maximum discrepancy as at the second stage and all discrepancies in all operation conditions were optimized at this stage. Thus, optimization of the objective function (17) can lead to the maximum agreement between the calculations on the mathematical model and the measurements on the real equipment without increasing the maximum discrepancy value. Minimization of the objective function (17) with reference to the considered generating unit model (17) produced a meaningful result. The function value reduced from 6940 (the second identification stage) to 1573 (after termination of the third stage).

After termination of the third identification stage, the values of the adjustable coefficients  $\theta$  of the mathematical model were fixed and the mathematical model was considered as adjusted to the real equipment state.

The described technique allows assessing the identification of mathematical model parameters of the studied thermal power equipment. The absolute relative discrepancies of control parameters ( $x$  is the input information for the model,  $y$  is the output information) are taken as the criterion. The discrepancies are calculated by the formulas:

$$k_x = \frac{|x_m - \bar{x}_m|}{\sigma_x}; \quad k_y = \frac{|y_m - \bar{y}_m|}{\sigma_y} \quad (18)$$

The overall accuracy of the identification problem solution can be obtained by calculation of the sum of the absolute relative discrepancies in control parameters in all considered operation conditions by the formula:

$$k = \sum_{i=1}^R \left[ \sum_{j=1}^N \frac{|x_{mj}^i - \bar{x}_{mj}^i|}{\sigma_{xj}} + \sum_{k=1}^M \frac{|y_{mk}^i - \bar{y}_{mk}^i|}{\sigma_{yk}} \right] \quad (19)$$

where  $R$  is the number of calculated operation conditions;  $N$  is the dimension of the vectors  $x_m$ ;  $M$  is the dimension of the vectors  $y_m$ .

In the considered generating unit model the value of the criterion  $k$  after the second identification stage was equal to 511. After minimization of the sum of squares of relative discrepancies at the third stage, the value of criterion (19) considerably decreased and became equal to 332. Thus, the total relative discrepancy of control parameters decreased by 35%, which made it possible to adjust the mathematical model subject to the real generating unit state even more accurately.

Table 1 presents the values of all control parameters (the value calculated on the mathematical model is given above, the value measured on the generating unit is given below). The value of criterion (18) that characterizes the closeness between the calculated and measured values is given to the right of the parameter values. The closer is this value to zero, the lower is the relative discrepancy of the corresponding measurement. The parameters, whose measurements are absent (or were given approximately) or were excluded at one of the identification stages of the generating unit model are showed in the Table by filling. These measurements were excluded from further calculations.

#### 4 Example of the optimization calculation on the adjusted generating unit model

Identification of mathematical model parameters of the studied equipment, inter alia, allows the solution of some critical operational problems, for example, the state estimation of thermal power equipment or optimization of schemes and parameter of the studied equipment at TPP in order to improve efficiency of its operation. The single calculation of operation condition on the adjusted generating unit model takes only several seconds (3-5 seconds), which makes possible its application to real-time control of the generating unit.

The calculation on the generating unit model adjusted to the real equipment state is taken as an example of the optimization calculation. The objective function is the specific consumption of coal equivalent burnt in the boiler unit. The array of inequality-constraints included both physical constraints (on the temperature of pipe metal, mechanical metal stress, non-negativity of steam flow rates and others), and operating constraints (the temperature of primary and secondary steam, pressure in the condenser, turbine capacity). Table 1 presents a composition and values of the optimized parameters (lines 1-9), controlled operating parameters (lines 10-14), and values of the efficiency indicators of generating unit operation (lines 15, 16) in one of the considered conditions and in the optimal condition obtained as a result of optimization calculation. The optimization calculation on the generating unit model takes several tens of minutes (30-60 minutes depending on the number of optimized parameters and inequality-constraints).

As is seen from the Table, the volume of fuel burnt in the boiler unit can considerably be decreased at the same electricity generation influencing the operating parameters of the generating unit, which somewhat increases its operation efficiency. In the above example, the specific consumption of coal equivalent for net electricity generation (227 MW of power) decreased by 3.2%, and the net efficiency of the generating unit increased approximately by the same value.

**Table 1.** Calculation results, measurements of control parameters and values of criterion accuracy for three operation conditions of the studied generating unit model

No	Parameter, measurement unit	1 cond.	$k_i$	2 cond.	$k_i$	3 cond.	$k_i$
1	2	3	4	5	6	5	6
1	Circulating water temperature before condenser	22.23 22.05	0.536	22.34 21.65	2.063	21.07 21.25	0.526
2	Circulating water flow rate before condenser	10253 10211	0.278	10872 10979	0.712	10211 10143	0.451
3	Feed water temperature before condenser	32.2 32.0	0.602	31.0 30.9	0.159	32.3 32.2	0.159
4	Air temperature before primary tubular air heater	48.5 46.1	0.747	47.0 48.7	0.493	46.0 49.5	1.026
5	Steam pressure after second turbine compartment	32.75 32.91	0.794	19.38 19.32	0.293	34.12 34.07	0.292
6	Steam pressure before first turbine compartment	127.26 128.16	0.723	74.52 70.44	3.262	133.56 132.75	0.700
7	Steam pressure before second turbine compartment	47.48 46.39	3.360	27.73 26.99	2.276	49.45 49.50	0.151
8	Steam pressure before third turbine compartment	27.896 28.590	3.469	16.12 16.00	0.596	28.98 29.69	3.787
9	Steam pressure before fourth turbine compartment	15.81 15.82	0.119	9.09 9.17	0.623	16.42 16.41	0.110
10	Steam pressure before fifth turbine compartment	9.46 9.25	4.118	5.46 4.64	16.42	9.83 9.60	4.856
11	Steam pressure before sixth turbine compartment	3.85 3.89	2.018	2.26 2.30	2.154	4.01 4.01	0.139
12	Steam pressure before seventh turbine compartment	1.25 1.20	3.603	0.73 0.70	2.475	1.31 1.50	13.84
13	Steam pressure before eighth turbine compartment	0.92 0.91	1.164	0.54 0.58	3.696	0.96 0.93	3.692
14	Steam pressure before ninth turbine compartment	0.214 0.170	8.726	0.132 0.100	6.340	0.221 0.160	13.05
15	Steam pressure before turbine condenser	0.0686 0.0590	1.912	0.0470 0.0335	2.697	0.0668 0.0575	2.002
16	Steam pressure after first turbine compartment	407.1 411.9	0.904	398.2 400.3	0.395	407.3 412.0	0.884
17	Steam pressure after second turbine compartment	358.7 359.3	0.157	351.0 349.6	0.359	358.4 359.8	0.355
18	Steam pressure after third turbine compartment	493.3 491.1	0.404	488.9 495.4	1.216	491.4 490.3	0.197
19	Steam pressure after fourth turbine compartment	419.7 435.3	3.899	416.6 432.2	3.899	417.8 415.0	0.698
20	Steam pressure after fifth turbine compartment	305.6 298.1	1.866	304.5 307.3	0.696	304.1 301.5	0.656
21	Steam pressure after sixth turbine compartment	185.5 177.2	3.114	185.4 175.3	3.798	184.4 187.6	1.182
22	Steam pressure after seventh turbine compartment	162.5 156.9	2.117	162.8 166.8	1.517	161.5 154.8	2.511
23	Steam pressure after eighth turbine compartment	69.9 70.0	0.055	73.3 70.0	2.493	68.6 75.0	4.799
24	Steam pressure after ninth turbine compartment	37.8 38.5	0.533	31.4 29.3	1.571	37.4 38.2	0.530
25	Circulating water temperature after condenser	29.37 30.20	1.248	26.43 26.35	0.126	28.52 29.5	1.464
26	Temperature of condensate-feed water mixture after condenser	37.2 38.8	2.432	31.1 29.6	2.241	36.8 38.2	2.112
27	Turbine condensate temperature after first low-pressure heater	54.5 55.7	0.895	41.3 46.4	3.861	55.3 55.3	0.027

**Continuation of Table 1**

1	2	3	4	5	6	5	6
28	Turbine condensate temperature after second low-pressure heater	96.3 95.4	0.706	82.5 83.9	1.043	97.48 95.7	1.334
29	Turbine condensate temperature after third low-pressure heater	137.6 139.4	1.368	119.3 122.5	2.391	138.9 139.9	0.719
30	Turbine condensate temperature after deaerator	165.6 164.4	0.901	144.8 148.5	2.763	167.2 164.5	2.052
31	Feed water temperature after fourth high-pressure heater	194.7 198.7	1.991	172.7 175.3	1.301	196.1 200.4	2.142
32	Feed water temperature after fifth high-pressure heater	233.0 236.6	1.800	208.2 209.6	0.720	234.8 238.8	2.013
33	Feed water temperature after sixth high-pressure heater	255.1 261.9	3.411	227.1 230.5	1.719	257.4 264.3	3.471
34	Electric power generated by turbine unit generator	225.09 227.09	2.398	128.16 124.98	3.820	235.12 235.73	0.730
35	Feed water temperature after boiler condensate cooler	273.7 269.6	2.067	255.1 249.0	3.061	270.7 263.9	3.413
36	Steam temperature after high-pressure water economizer	337.3 332.7	1.738	324.3 332.7	3.152	337.3 327.4	3.713
37	Saturated steam pressure after boiler drum	167.75 166.95	0.636	146.20 144.17	1.623	172.7 172.2	0.435
38	Steam temperature before radiant superheater	358.2 358.2	0.012	352.5 360.1	2.286	359.3 358.1	0.351
39	Steam temperature after radiant superheater	417.1 413.3	1.138	438.8 436.3	0.738	410.2 407.6	0.782
40	Steam temperature after injection into primary steam cooler	397.7 399.2	0.376	400.1 401.5	0.358	399.3 405.6	1.568
41	Steam temperature after first chain of middle platen superheater	447.1 444.6	0.623	457.2 454.4	0.698	446.7 446.5	0.060
42	Steam temperature after first chain of outer platen superheater	498.4 494.3	1.019	513.0 506.6	1.602	496.2 492.8	0.859
43	Steam temperature after injection into secondary steam cooler	483.7 483.7	0.012	492.9 493.7	0.189	484.0 488.9	1.224
44	Steam temperature after second chain of middle platen siuperheater	503.2 501.3	0.483	511.9 512.5	0.140	503.0 504.9	0.464
45	Steam temperature after second chain of outer platen superheater	524.5 518.4	1.516	532.9 528.4	1.135	523.7 522.7	0.261
46	Steam temperature after injection into tertiary steam cooler	508.2 509.5	0.334	520.8 519.4	0.341	508.3 512.1	0.943
47	Steam pressure after convection superheater	565.4 566.3	0.222	566.8 566.9	0.036	565.5 566.8	0.322
48	Steam temperature after primary reheat	478.1 481.3	0.801	480.6 485.8	1.297	477.2 476.8	0.094
49	Steam temperature after injection into low-pressure steam cooler	404.4 410.6	1.543	440.7 441.7	0.254	410.9 417.4	1.618
50	Steam temperature after secondary reheat	516.9 514.2	0.669	531.4 528.9	0.630	522.5 515.2	1.822
51	Steam temperature after tertiary reheat	584.1 568.5	3.888	578.0 568.7	2.327	582.2 567.6	3.656
52	Steam temperature after low-pressure water economizer	111.1 109.2	1.420	98.9 101.3	1.785	112.1 107.9	3.134
53	Gas temperature after primary tubular air heater	155.1 149.3	1.767	131.8 139.9	2.417	156.0 146.1	2.983
54	Air temperature after secondary tubular air heater	324.9 328.1	0.588	289.0 304.0	2.804	328.4 321.0	1.400
55	Primary steam flow rate after boiler	175.7 184.2	3.398	102.21 101.96	0.099	184.76 186.61	0.740
56	Flow rate of condensate-feed water mixture after condenser	134.40 154.07	9.838	79.00 77.74	0.629	140.14 161.17	10.52

**End of Table 1**

1	2	3	4	5	6	5	6
57	Gas temperature after high-pressure convection superheater	775.4 696.7	1.218	661.3 650.1	0.269	753.1 703.4	1.244
58	Gas temperature after superheater of secondary reheat	545.6 458.5	2.614	488.1 447.1	1.231	557.5 457.1	3.014
59	Gas temperature after superheater of primary reheat	382.0 339.6	1.590	338.7 313.3	0.954	386.2 332.4	2.018
60	Gas temperature after high-pressure water economizer	219.2 207.0	0.613	188.1 192.5	0.217	221.4 202.4	0.952

**Table 2.** Example of the optimization calculation of generating unit operation condition

No	Parameter, measurement unit	Real condition	Optimal condition
1	2	3	4
1	Consumption of fuel burnt in boiler, kg/s	36.89	36.79
2	Excess air coefficient in boiler furnace	1.21	1.21
3	Steam enthalpy decrease in first steam cooler, kcal/kg	18.11	21.86
4	Steam enthalpy decrease in second steam cooler, kcal/kg	10.39	7.13
5	Steam enthalpy decrease in third steam cooler, kcal/kg	10.90	9.73
6	Steam enthalpy decrease in steam cooler of steam reheat pipeline, kcal/kg	39.86	43.15
7	Head of feeding pump, kgf/cm <sup>2</sup>	175.53	165.45
8	Head of circulating pump, kgf/cm <sup>2</sup>	5.92	1.42
9	Cooling water flow rate before condenser, kg/s	10253	10972
10	Electric capacity at generator terminals, MW	225.09	225.06
11	Primary steam temperature before turbine, C	565.41	575.22
123	Secondary steam temperature before turbine, C	584.05	584.98
13	Flue gas temperature after boiler, C	155.14	155.80
14	Pressure in turbine condenser, kgf/cm <sup>2</sup>	0.0686	0.0653
15	Specific consumption of coal equivalent for electricity generation (net), gce/kW·h	368.58	356.75
16	Net efficiency of generating unit	33.33	34.44

## 5 Conclusions

The paper describes an improved technique developed by the author for identification of mathematical model parameters of complex thermal power equipment. The calculations show that the technique allows the more effective a) detection of gross errors in measurements of control parameters used for identification of mathematical model parameters of the studied equipment, b) assessment of correctness and amendment of errors in mathematical model construction and c)

improvement of the accuracy of identification problem solution.

Besides, the paper describes criteria to assess the identification problem solution accuracy for both individual measurements of control parameters and the total discrepancy of all parameters in the considered operation conditions. The suggested technique also allows assessing an additional error due to imperfection of the standard calculation methods and assumptions in the mathematical model of the studied TPU besides the errors due to the accuracy of sensors used for equipment tests.

The improved technique for identification of mathematical model parameters was tried on the detailed mathematical model of the 225 MW present-day generating unit that was constructed by the author. The paper presents results of the problem solution on identification of mathematical model parameters of the generating unit and an example of the optimization calculation of real operation condition in order to decrease specific consumption of coal equivalent for electricity generation and improve the generating unit efficiency.

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