Control for Reheat Steam Temperature of Power Plant Based on Particle Swarm Optimization

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Abstract-The model of reheat steam temperature of power plant has characteristics such as large delay and large inertia, the traditional PID control is difficult to achieve good results. This paper presents a new PSO with random acceleration factors and verifies its validation, also, this paper presents a new control method, uses the modified PSO to identify the model, tunes parameters SMTH-PID control based on particle swarm optimization and the results of identification. The simulation results show that the method is more effective than the traditional PID control.

KeyWords-Particle swarm optimization, Reheat steam temperature, Identification

I. INTRODUCTION

In order to improve the economy of thermal cycling, large coal-fired units generally use reheat system. Large boiler reheater works at high temperature near the limit of its metal pipes, over high temperature will decrease the safety of the metal pipe, over low temperature will decrease economy, therefore, the reheat steam temperature is one of the most important parameters. [1]

Reheat steam temperature is usually regulated by gas baffle, the model has obvious characteristics of large delay and large inertia, the traditional pid control is difficult to achieve good control results. SMITH control is the solution as the control of large delay [2], which can eliminate some extent delay and greatly improve the quality of control. In process control, combined of PID SMITH control is widely used. This method involves some parameters settings of SMITH control module and the PID control. In this paper, a method of tuning the above parameters was proposed.

II. THE CONTROL OBJECT OF REHEAT STEAM TEMPERATURE

In usually, gas baffle is used to control the reheat steam temperature of thermal power plant, and spray water is used when the temperature is too high. Reheat steam flow through the reheater in the separation gas pass and the reheater in furnace, it was heated by heat transition of gas – metal – steam. Reheat steam temperature refers to the last reheater (High temperature reheater in furnace) outlet steam temperature. Gas baffle is installed in the separation gas pass, changing the gas baffle’s opening, can change the reheater gas flow, thus change the reheat steam temperature of reheater outlet in the separation gas pass, and ultimately change the reheat steam temperature [2].

This shows, the control object of reheat steam temperature has a large delay. The heat transition of gas – metal – steam has large inertia. So, the control object of reheat steam temperature has obvious characteristics of large delay and large inertia. Usually, we describe it as second-order plus delay system:

\[ G(s) = \frac{Ke^{-\tau s}}{(T_1s + 1)(T_2s + 1)} = G_p(s) * e^{-\tau s} \]

The Control Method of Reheat Steam Temperature

Combined of PID and SMITH control is widely used in reheat steam temperature control, the control method is as follows:

\[ G_c(S) \] is PID controller which is generally taken to PI control,

\[ G_p(s) \cdot e^{-\tau s} \] means as formula (1).

By Figure 1 and the principle of SMITH control, when \( G_p(s) \) and \( e^{-\tau s} \) match with the actual model, the effect of delay is eliminated. Combined with PID control, good control results are achieved. In this control method, there are six unknowns, they are gain \( k \) and integral time \( T_1 \) of PID, the gain coefficient \( K \), time constant \( T_1, T_2 \) and delay time \( \tau \) of the control object. The method of parameter setting is given as following, control object parameters are obtained by model identification, PID parameters are obtained by optimizing.

III. PARTICLE SWARM OPTIMIZATION

A. Particle Swarm Optimization [3]

PSO is a population-based evolutionary algorithm, using speed - location model, each particle denotes a candidate solution in D-dimensional space. D is the number of unknowns. Each particle adjusts its flight based on flight experience. The ith particle's velocity
velocity is the PSO with random inertia weight and random acceleration; method 2 is the PSO with random inertia weight, method 3 is more effective than constant acceleration factor. Diversity of groups, random acceleration factor is more effective than constant acceleration factor.

The optimal solution in D-dimensional space is found in the last iteration of the algorithm. Usually, the inertia weight obeys an even distribution in [0,1]. \(c_1, c_2\) are acceleration factors; \(k\) is the iteration number.

The results show that the method 3 is better than method 1 and method 2.

### IV. Model Identification Base on PSO

**A. PSO Model Identification Procedure**

PSO model identification is implemented as follows: get the step response of the model \((y(t)\) or other response) \(\mathrm{y}\), denoted by \(y\), the particle structure is \([K, τ, T_1, T_2]\), each particle denotes a certain model, get the step response of the model, denoted by \(y^*\), select a proper fitness function \(\text{fitness} = f(y, y^*)\), optimize the unknowns by fitness, after many times iterate, the global-optimal solution is the result of the identification.

**Step 1:** Set the population size, upper and lower limits of each parameter, select \(c_1, c_2\), the maximum iteration number.

**Step 2:** In accordance with \([K, τ, T_1, T_2]\) structure, initial population location, velocity vector, the particle’s self-optimum fitness and the global optimum fitness.

**Step 3:** According to the corresponding relations between the particle elements and model parameters, get the model and its step response, calculate the particle’s fitness.

**Step 4:** Update the particle’s self optimum solution and fitness, the global optimal solution and fitness.

**Step 5:** Update position and velocity of each particle. If its location is over limit, then reinitialize it.

**Step 6:** If not reach the maximum number of iterations, then go to step 3.

**B. Identification of Reheat Steam Temperature**

According to experience, we set the range of parameters: \(K \in [0.01-100], T_1, T_2 \in [0.1-100], \tau \in [0-300]\), set speed limit to the difference between maximum and minimum parameters, set population size to 40 set the maximum number of iterations to 300, and generate \(c_1, c_2\) as the part II-B.

The delay time of the model influence SMITH control effect seriously, so in the calculation of fitness, we increase the weight of data in the pure delay area, as:

\[
\text{fitness} = \sum_{i=1}^{n} a * (y_i' - y_i)^2
\]

In the formula 4, \(n\) is current number of iterations, if \(n<\tau *1.1, a=50; \text{if not}, a=1.\)

Take a reheat temperature control object of 300MW unit, step the gas damper by 5%, record the reheated steam temperature response curve and data from the distribute control system (DCS). According to Section 3, we...
identify the model by Matlab, the results of identification are as follows:

\[
G(s) = \frac{0.791 e^{-129.5 s}}{(108s + 1)(145s + 1)}
\]

It can be seen from the Figure 2, the identification curves in good agreement with the actual curve.

\[
(6)
\]

V. PID TUNING BASED ON PSO

The same with the model identification, PID parameter setting can also be carried out by particle swarm optimization.

The expression of PID control is as follow:

\[
G_c(s) = k + \frac{1}{T_i} s
\]

In formula (7): \(k\) is gain, \(T_i\) is integral time. At this point, optimization of parameters turn into \([\text{gain}, \text{integral time}]\). According to experience, we set the range of parameters: \(k \in [0.01, 100]\), \(T_i \in [100, 1000]\), set speed limit to the difference between maximum and minimum parameters, set population size to 40 set the maximum number of iterations to 300, and generate \(c1, c2\) as the part III-B.

Take the fitness function:

\[
\text{fitness} = (\int t * |e(t)| dt)^{-1}
\]

In formula (8): \(t\) is the time coordinate of the step response, \(e(t)\) is the difference between the setpoint and the value of reheat steam temperature.

Several tests show that the fitness hasn’t many local extreme points and it can approach the global extreme points in 50 inertias, so, we make the following improvements:

Take the max inertias to 50, and use the above method to get the best particle \([k_{b1}, T_{b1}]\); Then, take the max inertias to 50, set the range of parameters as \(k \in [0.8, 1.2]\), \(T_i \in [0.8, 1.2]\), use the above method again to get the final global best particle. So, the max inertias reduced to 100 from 300, and the best particle had high accuracy.

The flow diagram of the PID tuning based on PSO shows as Figure 3.

![Figure 2. Identification of reheat steam temperature](image)

**Figure 2. Identification of reheat steam temperature**

**Figure 3. the produce of PID tuning based on PSO**

**Figure 4. the control effect of SMITH-PID and PID**
Use the above method, the final global best particle is $[2.132, 113], k = 2.132, T_i = 113$, the effect of control shows as Figure 4.

Compared with the pid, SMITH-PID has a faster rise time, smaller overshoot and shorter steady time.

VI. CONCLUSION

This paper presents a modified PSO and a control method of reheat steam temperature based on modified PSO and SMITH-PID. Simulation results show that the method of control effect is good.

REFERENCES


