Study on Dynamic Process Characteristics of CHP Unit With Variable Load Based on Working Point Linearization Modeling

Yuehua Huang, Qing Chen*, Jing Ye and Tianlin Lu

College of Electrical Engineering and New Energy, China Three Gorges University, Yichang, China

In view of the difficulty of applying the refine modeling of combined heat and power (CHP) units to the optimization scenario of integrated energy system, a CHP unit model based on working point linearization modeling is proposed, and its variable load characteristics are analyzed. Firstly, the dynamic coupling relationship of CHP unit is analyzed, and the nonlinear dynamic model of the unit is constructed. Then, under the pure condensation and heating conditions, the linearized Laplace transform model of the working point is established, and the variable load capacity under the independent action of control variables is analyzed to test the availability of the Laplace model. On this basis, the dynamic adaptive particle swarm optimization algorithm is used for multivariable cooperative control to test the open-loop characteristics of the variable load capacity of the unit. At the same time, the control strategy of electrothermal cooperation and safety self-test is designed to adjust the control variables, and test the closed-loop characteristics of the unit’s regulation ability. Finally, a 300-MW steam extraction CHP unit is taken as an example to verify the applicability of the unit model and the effectiveness of the control strategy.

Keywords: working point linearization, unit variable load, multivariable cooperative control, dynamic adaptive particle swarm optimization, dynamic regulation characteristics

INTRODUCTION

Clean, low-carbon, and efficient energy use plays an important role in realizing the strategic goal of “double carbon.” Combined heat and power (CHP) plant is becoming the main energy supply source of the electrothermal integrated energy system (IES) with its superior comprehensive energy supply efficiency (Sun et al., 2021). The modeling of the CHP unit variable load capacity is very important for the optimal calculation of power dispatching, frequency regulation, and other scenarios of the IES (Ye et al., 2012; Zhang et al., 2021). However, its refined modeling mechanism is complicated, and the fast dynamic regulation characteristics after optimization control transformation are complex (Shen et al., 2017; Shen et al., 2020a), which makes it difficult to accurately describe the variable load capacity of the unit. Therefore, the research on the description method of the dynamic regulation characteristics of the CHP unit during the variable load process is of great significance to support the application of the CHP unit in the scenario of the IES.

The dynamic characteristics of the CHP unit are not only affected by its internal physical structure attributes but also related to the operation mode, the working condition, the external environment, and other factors (Wang, 2013). At present, it is difficult to obtain an accurate description of the
dynamic characteristics of the unit. A mathematical model that meets certain accuracy requirements and reflects the main dynamic characteristics of the unit is often established by reasonable simplification and approximation, combined with mechanism analysis and experimental modeling (Yang et al., 2018; Yang et al., 2019; Zhu et al., 2020; Li et al., 2021). Algebraic equations are usually used to describe the feasible region, climbing, and standby capacity of the CHP unit model (Liu et al., 2021). This model is mostly used in the calculation scenario of the unit participating in system long-time scale optimal dispatching (Zhang et al., 2018). The algebraic equation linearization model often ignores the continuous time variation characteristics of the unit output, and there is the assumption that the unit output power can change instantaneously. The model undoubtedly expands the rapid load changing capacity of the unit, which is easy to cause the possibility that the dispatching plan cannot be accurately realized, that is, there is the problem of energy nondeliverability (Gao and Yan, 2017). The mathematical description considering the internal characteristics of the unit is in the form of differential algebraic equations, which adopts the optimal control method to improve the short-term rapid regulation ability of the unit and participates in the system frequency modulation calculation scene (Wang et al., 2018; Yang et al., 2021). The constraints of differential algebraic equations will make the optimization problem a highly nonlinear dynamic optimization problem. This kind of an optimization problem is difficult to solve directly. Approximating differential variables through a discrete method will lead to the solution of large-scale optimization problems being easy to fall into a dimensional disaster (Wang, 2012), long solution time, and poor accuracy (Shen et al., 2021a; Shen and Raksincharoensak, 2021a), and cannot be solved online and in real time (Shen et al., 2021b; Shen and Raksincharoensak, 2021b). The modeling of the dynamic characteristics of the unit variable load process faces the challenge of meeting the calculation accuracy and solution speed while taking into account the accurate description of unit dynamic characteristics.

Laplace-transform a real variable function and perform various operations in the complex number field, and then perform the inverse Laplace transform to obtain the corresponding results in the real number field, which is often much easier to calculate than directly obtaining the same results in the real number field (Beerends et al., 2003). Yang et al. (2020) proposed a generalized circuit modeling method based on Laplace transform, which transforms the complex transmission characteristics of multi-energy networks in the time domain into a simple algebraic problem in the Laplace domain. Laplace transform is particularly effective for simplifying differential equations, which can be transformed into easily solved algebraic equations (Hooman and Randolph, 2020), so as to simplify the calculation. The analysis and synthesis of the control system are based on Laplace transform. Gao and Tian (2020) linearized the nonlinear dynamic model of the heating unit with small deviation, obtained the transfer function matrix model, analyzed the coupling relationship of the unit, and studied the thermal power load decoupling control method of the unit. Liu et al. (2005) linearized the nonlinear model at different load/pressure operating points and studied the load pressure nonlinear characteristics of the 660-MW unit. After linearizing the small deviation of the dynamic model of the heating unit (Deng et al., 2017), the transfer function matrix model including the characteristics of the heating side is obtained, the thermolectric coupling characteristics are analyzed, and the decoupler is designed. By introducing Laplace transform into linearization at different working points and using a transfer
function instead of a differential equation to describe the characteristics of the system, the whole characteristics of the control system can be determined intuitively and simply, the motion process of the control system can be analyzed, and the adjustment strategy of the control system can be provided. For the linearization of the CHP unit at the working point, it is rare to use Laplace transform modeling to make the model suitable for the flexible demand of the variable load capacity of the IES.

To solve the above problems, firstly, a simplified nonlinear dynamic model of the CHP unit is constructed. Then, the linearized Laplace transform model of the working point under pure condensation and heating conditions is established to analyze the variable load capacity of the unit when each control variable acts alone and cooperatively. Furthermore, the control strategy of electrothermal cooperation safety self-test is used to test the dynamic regulation ability of the unit in the variable load process. Finally, taking the 300-MW steam extraction CHP unit as an example, the dynamic characteristics of the variable load process of multivariable collaborative control are tested to verify the effectiveness of open-loop and closed-loop control strategies.

**COMBINED HEAT AND POWER UNIT MODEL**

**Dynamic Relationship of Combined Heat and Power Unit**

As shown in Figure 1, the fuel volume $V_B$ of the unit directly controls the boiler combustion to produce high-temperature steam. The high-pressure regulating valve $V_T$ of the steam turbine is connected with the high-pressure (HP) cylinder and the boiler, and the steam extraction regulating butterfly valve $V_H$ is installed in the connecting pipe between the intermediate-pressure (IP) cylinder and the low-pressure (LP) cylinder. The steam exhaust of the IP cylinder of the steam turbine is divided into two parts. One part enters the LP cylinder of the steam turbine through the regulating butterfly valve to continue to work, and the other part enters the heat supply network heater to provide the heat source. After cooling, it is sent to the deaerator through the steam. The high-pressure regulating valve $V_T$ is connected with the high-pressure (HP) cylinder of the steam turbine works in the pure condensation state. By changing $V_T$ and $V_H$, the proportion of heating and generating power of the unit is adjusted to provide heat source and power supply.

During the operation of the extraction type CHP unit, the electric power and thermal power are comprehensively determined by the fuel flow, main steam pressure, extraction steam flow, temperature, and other variables controlled by its valve. The load-pressure simplified nonlinear dynamic model of the pure condensing unit (Liu et al., 2014) is combined with the model in the study by Tian (2005), and the differential equation mathematical model of the dynamic coupling relationship between the electric power, thermal power, and control valve of the CHP unit is obtained. The simplified nonlinear dynamic model of the CHP unit is given by:

$$
\begin{align*}
    r_m &= V_B (t - \tau) \\
    T_f \frac{dr_B}{dt} &= -r_B + r_m \\
    C_o \frac{d\rho_1}{dt} &= -K_3 p_l V_T - K_4 r_B \\
    \rho_1 &= P_A - K_2 (K_1 r_B)^7 \\
    T_f \frac{dN_E}{dt} &= -N_E + K_1 p_l V_T + K_3 p_Z V_H \\
    C_o \frac{dP_Z}{dt} &= -K_6 q_s (96p_2 - t_i + 103) + K_3 p_l V_T (1 - K_4) - K_3 p_Z V_H \\
    q_H &= K_1 K_6 q_s (96p_2 - t_i + 103) \\
    \rho_1 &= 0.01 p_l V_T 
\end{align*}
$$

(1)

where $V_B$ is the coal supply mass flow rate of the unit; $V_T$ is the opening of the steam inlet regulating valve of the HP cylinder of the steam turbine; $V_H$ is the opening of the extraction regulating butterfly valve; $q_s$ is the mass flow of circulating water; $t_i$ is the return water temperature of circulating water; $N_E$ is the generating power of the unit; $p_A$ is the drum pressure; $p_f$ is the front pressure of the steam turbine; $p_z$ is the exhaust pressure of the intermediate pressure cylinder; $p_1$ is the first stage pressure of the steam turbine; $q_H$ is the heating extraction steam flow; $r_m$ is the actual amount of coal entering the pulverizer; $r_B$ refers to the boiler combustion rate; $K_1$, $K_2$, $K_3$, $K_4$, $K_5$, $K_6$, and $K_7$ are the static parameters; $\tau$ is the delay time constant of the milling process; $\varepsilon$ is constant-coefficient; $T_f$ is the milling time constant of inertia; $T_i$ is the inertia time constant of the steam turbine; $C_6$ is the boiler heat storage coefficient; and $C_7$ is the heat storage coefficient of the unit heater.

**Linearization Model of Operating Point of Combined Heat and Power Unit**

The nonlinear model is linearized at a certain working point, and the linearized model can accurately reflect the dynamic and static characteristics of each link of the system near the working point. The model described in Eq. 1 is linearized to measure the action relationship between various inputs and outputs in the model and the influence of system nonlinearity on the controlled object. Firstly, write Eq. 1 in incremental form. The linear model
near the equilibrium point is obtained by small deviation linearization:

\[
\begin{align*}
    r_m &= \Delta V_B(t - \tau) \\
    T_{m} \frac{\Delta m}{dt} &= -\Delta m \\
    C_p \Delta p_t &= -K_p \Delta V_T - K_s \Delta p_t V_T - K_s \Delta V_B \\
    \Delta p_t &= \Delta p_b - K_{p1} \sqrt{K_1 T_{fs}} \\
    T_{m} \frac{\Delta N}{dt} &= -\Delta N_B - K_{p1} \Delta V_T + K_s \Delta p_t V_T + K_s \Delta V_B + K_s \Delta p_t V_H \\
    \Delta N_b &= -K_{qH} (96 \Delta p_2 - \Delta t) + K_s \Delta p_t V_T (1 - K_s) \\
    &+ K_s \Delta p_t V_H - K_{p1} \Delta V_B \\
    \Delta q_B &= K_s \Delta q_H (96 \Delta p_2 - \Delta t) + K_s \Delta q_H (96 \Delta p_2 - \Delta t) \\
    \Delta p_1 &= 0.01 \Delta p_t V_T + 0.01 \Delta p_t V_H
\end{align*}
\]

(2)

Secondly, the Laplace transform of the incremental equation is obtained. Assuming that the initial condition is zero, the Laplace transform is taken for the linear differential Eq. 2 of the system, and the incremental symbol is omitted, the following can be obtained:

\[
\begin{align*}
    r_m &= e^{-\tau} V_B(s) \\
    T_{m} \frac{\Delta m}{dt} &= -r_m(s) + r_m(s) \\
    C_p \frac{\Delta p_t}{dt} &= -K_p V_T(s) - K_s p_t V_T - K_s r_m(s) \\
    p_t(s) &= p_1(s) - K_{p1} \sqrt{K_1 T_{fs}}(s) \\
    T_{m} \frac{\Delta N}{dt} &= -N_B(s) + K_s \Delta p_t V_T(s) + K_s \Delta p_t V_H(s) + K_s \Delta p_t V_H(s) + K_{qH}(s) V_H(s) \\
    C_p \frac{\Delta q_B}{dt} &= -K_{qH} q_H(s) (96 \Delta p_2 - \Delta t) - K_{qH} q_H(s) (96 \Delta p_2 - \Delta t) + K_s p_t V_T(s) (1 - K_s) \\
    &+ K_s p_t V_H(s) (1 - K_s) - K_{p1} \Delta V_B(s) \\
    q_B(s) &= K_s q_H(s) (96 \Delta p_2 - \Delta t) - K_s q_H (96 \Delta p_2 - \Delta t) \\
    p_1(s) &= 0.01 p_t V_T(s) + 0.01 p_t V_H(s)
\end{align*}
\]

(3)

Then, the linear model is appropriately simplified and equivalent to obtain a set of linear equations describing the dynamic characteristics of the unit transfer process, and the system thermoelectric coupling relationship model after linearization of the working point can be obtained, which is expressed in the form of the transfer function matrix as follows:

\[
\begin{bmatrix}
    p_1(s) \\
    p_2(s) \\
    N_B(s)
\end{bmatrix} =
\begin{bmatrix}
G_{11}(s) & G_{12}(s) & G_{13}(s) \\
G_{21}(s) & G_{22}(s) & G_{23}(s) \\
G_{31}(s) & G_{32}(s) & G_{33}(s)
\end{bmatrix}
\begin{bmatrix}
V_T(s) \\
V_B(s) \\
V_H(s)
\end{bmatrix}
\]

(4)

The elements of the thermoelectric coupling relationship matrix are

\[
G_{11}(s) = \frac{M_2}{1 + M_1 s}
\]

(5)

\[
G_{12}(s) = \frac{K_{qH}}{K_{VT} + (1 - M_1 s)} e^{-\tau s}
\]

(6)

\[
G_{13}(s) = 0
\]

(7)

\[
G_{21}(s) = \frac{1 - K_{qH} M_2 C_b s}{(1 + M_1 s)(1 + M_2 C_b s)}
\]

(8)

\[
G_{22}(s) = \frac{K_{VT}(1 - K_{VT}) (1 - M_1 s)}{(1 + M_1 s)(1 + C_b M_2 s)} e^{-\tau s}
\]

(9)

\[
G_{23}(s) = \frac{M_2 (M_1 s^2)}{1 + M_1 s}
\]

(10)

\[
G_{31}(s) = \frac{M_2 C_b s M_3}{(1 + M_3 s)(1 + \frac{C_b M_3 s}{M_1})}
\]

(11)

\[
G_{32}(s) = \frac{K_{VT} M_3 (1 - M_2 s)}{(1 + M_1 s)(1 + M_3 s)(1 + \frac{C_b M_3 s}{M_1})} e^{-\tau s}
\]

(12)

\[
G_{33}(s) = \frac{96 K_{qH} M_5}{(M_4 s)^2} \left(1 + \frac{C_b M_3 s}{M_1} \right)
\]

(13)

where

\[
M_1 = C_b / (K_{VT}), \quad M_2 = K_1 V_B / (K_1 V_T^2), \quad M_3 = 1.5 K_2 C_b \sqrt{K_1 V_B}, \quad M_4 = 96 K_{qH} q_H + K_3 V_H, \quad M_5 = 96 K_4 \Delta q_H + K_3 V_H, \quad M_6 = K_3 [K_6 q_H - 103 K_6 q_H - K_7 V_B (1 - K)],
\]

Under different input operating points, the transfer function of the CHP unit will be different. This nonlinearity only changes the model parameters, and the model structure is unchanged. The Laplace transform model with a linearized working point takes into account the advantages of simplicity, being in real time, and high precision. In this paper, the variable load characteristics of the CHP unit under pure condensation and heating conditions will be studied.

**ANALYSIS METHOD OF VARIABLE LOAD CHARACTERISTICS OF COMBINED HEAT AND POWER UNIT**

**Open-Loop Variable Load Characteristic Test and Multicontrol Variable Synergy Method**

**Variable Load Characteristic Test With Independent Action of Control Variables**

The functional relationship between control variables and output variables is shown in Figure 2. The pure condensation condition
and heating condition are modeled in Simulink to test and analyze the input–output relationship between variable load process variables. See [Supplementary Appendix SB](#supplementary) for the parameters of the working condition model. The dynamic characteristics of the boiler steam turbine heating system and the variable load capacity are obtained by analyzing the output change when the control variables act alone. Under the heating condition, apply step changes of 10 t/h, 10, and 10% on \( V_B \), \( V_T \), and \( V_H \) commands, respectively, and under the pure condensation condition, apply step changes of -10 t/h, -10%, and -10%, respectively. The response curves of electric power, main steam pressure, and extraction pressure under different working conditions are shown in Figures 3, 4. Figure 3 shows the change of the object output under the step disturbance of boiler fuel \( V_B \), turbine high regulating valve \( V_T \) opening, and heating extraction regulating butterfly valve \( V_H \) opening. When \( V_B \) increases, the front pressure of the unit, the electric power of the unit, and the exhaust pressure of the IP cylinder (heating extraction flow) all rise. When \( V_T \) opening increases, the pressure in front of the unit decreases, the boiler releases heat storage, the electric power of the unit first increases and then returns to the original level, and the exhaust pressure (heating extraction flow) of the IP cylinder first increases and then returns to the original level. When \( V_H \) increases, the pressure in front of the turbine will remain unchanged, the electric power of the unit will increase due to the increase in the work share of some steam in the LP cylinder, the IP exhaust pressure will decrease, and the heating extraction flow will decrease due to the decrease in steam extraction from the turbine.

Figure 4 shows the object output changes under the step disturbance of boiler fuel volume \( V_B \), turbine high regulating valve \( V_T \) opening, and heating extraction regulating butterfly valve \( V_H \) opening. When \( V_B \) is lowered, the pressure of the front engine, the power of the generating unit, and the exhaust pressure of the IP cylinder are all decreased. Because \( V_H \) is fully opened, the extraction flow of heating is zero and remains unchanged, which is no different from the traditional pure condensing unit. When \( V_T \) opening decreases, the pressure in front of the unit increases, the boiler releases heat storage, and the electric power of the unit decreases first and then returns to the original level. The variation law of the exhaust pressure of the IP cylinder is similar to that of the electric power. Because \( V_H \) is fully open, the heating extraction flow is zero and remains unchanged. When \( V_H \) opening decreases, because the heating state of the unit is not actually turned on, the electric power of the unit remains unchanged, the exhaust pressure of the IP cylinder increases, and the heating extraction flow is zero.

The dynamic characteristics of the linearized model after univariate action are basically consistent with the original model (Wang, 2013). Therefore, it can be considered that the linearized model has good reproducibility. It is feasible to analyze and study the dynamic regulation characteristics of the variable load process according to the linearized model.

The independent action test of control variables under the two working conditions shows that the control variable regulation has different effects on the load change of the CHP unit. \( V_B \) has obvious effect on the change of electric power, \( V_T \) has a great impact on the change of main steam pressure, and \( V_H \) mainly affects the change of extraction pressure. Although the control variables acting alone have a certain variable load capacity, the regulation time is long and the regulation is differential. The actual system is controlled by pressure, temperature, flow, and other factors, and the final control effect is the superposition of many influencing factors. The control effects of the reverse superposition of influencing factors offset each other, while the control effects of the positive superposition strengthen each other. The multicontrol variable synergy effect of the work point linearization model needs to be further tested and verified.
Multicontrol Variable Synergy Method

In this paper, the optimization algorithm is used to determine the three control variables of $V_T$, $V_B$, and $V_H$ to analyze the change of the system output when the three control variables work together. Compared with other optimization algorithms, dynamic adaptive particle swarm optimization (DAPSO) has relatively excellent convergence ability and solution accuracy (Fu et al., 2017), which is suitable for the variable optimization problem of the control system in this paper. Its mathematical description is as follows:

$$\begin{align*}
\frac{V_{d1}}{V_{d1}} &= \omega_i^{t+1} s_i^{t+1} + \rho_1 r_1 (P_{i}^{t} - x_1^{t}) + \rho_2 r_2 (G_{i}^{t} - x_1^{t}) \\
\frac{x_1}{x_1} &= x_i^{t} + x_1^{t+1}, & s_1 > a \text{ and } s_1 < b \\
\omega_i^{t+1} &= \begin{cases} 
\beta s_1 + b, & \text{if } \rho_1 r_1 < 0, \\
1 - y s_1 + b, & \text{otherwise},
\end{cases} \\
\rho_1 &= \frac{1}{N L} \sum_{i=1}^{N} \left( \sum_{k=1}^{D} (s_k - \bar{s}_k)^2 \right)
\end{align*}$$

(14)

where $v_i^t$ is the velocity of the $i$-th particle in the $t$-th iteration; $x_i^t$ is the position of the $i$-th particle in the $t$-th iteration, and $\rho_1$ and $\rho_2$ are the acceleration coefficients; generally, $\rho_1 = \rho_2 = 2$. $r_1$ and $r_2$ are two random numbers varying in [0,1]. $P_i^t$ is the best location for the $i$-th particle to be searched in $t$ iterations; $G_i^t$ is the best location for the whole population to search in $t$ iterations; $\omega_i^t$ is the inertial weight of the $i$-th particle in the $t$-th iteration; $\gamma$ and $\beta$ are selected in [0,1]; generally, $\gamma = \beta = 0.5$. $a$ and $b$ are the thresholds for controlling the aggregation factor; generally, $a = 0.9$ and $b = 0.5$; $h_i^t$ is the rate factor of evolution; $s_1$ is the fitness aggregation factor; $s_2$ is the spatial aggregation factor. $F(P_i^t)$ is the fitness value of $P_i^t$; $F_i$ is the best fit in the $t$-th iteration; $F_i$ is the average fitness value in the $t$-th iteration; $N$ is the population size; $L$ is the longest radius of the search space; $D$ is the dimension of the solution space; $\bar{s}_k$ is the $d$-dimensional coordinate of the $i$-th particle; $\bar{s}_k$ is the average of the $d$-dimensional coordinates of all particles.

Given the electric power variable load demand, the optimization algorithm flow of the synergy of the three control variables is shown in Figure 5. The connection between DAPSO and the Simulink model is through the particle (i.e., control variables $V_T$, $V_B$, and $V_H$) and the corresponding fitness value of the particle (i.e., the performance index of the matching degree between the output value and the expected value). The optimization process is as follows: DAPSO generates particle swarm optimization (initial particle swarm optimization or updated particle swarm optimization), assigns the particles in the particle swarm to the model parameters $V_T$, $V_B$, and $V_H$ in turn, and then runs the Simulink model under a certain working condition of the unit to obtain the performance index corresponding to the group of parameters, which is transmitted to DAPSO as the fitness value of the particle. Finally, judge whether the algorithm can be exited.

Thermal System Constraints

CHP unit control shall be able to realize accurate variable load tracking, reliable heat supply, and safe pressure fluctuation. Heating and power generation are affected by boiler fuel volume $V_B$, turbine high regulating valve $V_T$ opening, and heating extraction regulating butterfly valve $V_H$ opening. Based on the PID optimization control method, three control variables are matched, and the PID controllers of $V_T(t)$, $V_B(t)$, and $V_H(t)$ are designed, respectively (Chen et al., 2014):

$$\begin{align*}
V_T(t) &= K_{PT} E_T(t) + K_{I} \int_{0}^{t} E_T(t)\,dt + K_{D} \frac{dE_T(t)}{dt} \\
V_B(t) &= K_{PB} E_B(t) + K_{I} \int_{0}^{t} E_B(t)\,dt + K_{D} \frac{dE_B(t)}{dt} \\
V_H(t) &= K_{PH} E_H(t) + K_{I} \int_{0}^{t} E_H(t)\,dt + K_{D} \frac{dE_H(t)}{dt}
\end{align*}$$

(16)

Control scheme 1:

$$\begin{align*}
E_T(t) &= p_1^f(t) - p_1(t) \\
E_B(t) &= p_2^f(t) - p_2(t) \\
E_H(t) &= N_E^f(t) - N_E(t)
\end{align*}$$

(17)

where $K_{PT}$, $K_{I}$, $K_{D}$, $K_{PB}$, $K_{PH}$, $K_{I}$, and $K_{D}$ are parameters of variables associated with the PID. In the control optimization cycle, $p_1^f$, $p_2^f$, $N_E^f$, and $N_E$ are the set values for the main steam pressure, the electric power of the unit, and the IP extraction pressure, respectively. PID parameter commissioning method is determined by engineering experience method, as shown in Supplementary Table S3.

In order to ensure the safe and stable operation of the unit, the influence of output variable fluctuation shall be considered in the regulation process, and the control system path constraint (Tian et al., 2017) shall be met, as shown in Eq. 18a. The steady-state value of each output variable shall not exceed the allowable error range, and the system final value constraint shall also be satisfied (Tian et al., 2015), as shown in Eq. 18b.

$$\begin{align*}
|p_1 - p_1(t)| &\leq M_{p_1} \\
|N_E(t) - N_E^f| &\leq M_{N_E} \\
|p_2 - p_2(t)| &\leq M_{p_2}
\end{align*}$$

(18a)

$$\begin{align*}
|p_1(t) - p_1^f| &\leq m_{p_1} \\
|N_E(t) - N_E^f| &\leq m_{N_E} \\
|p_2(t) - p_2^f| &\leq m_{p_2}
\end{align*}$$

(18b)

where $m_{p_1}$, $m_{N_E}$, and $m_{p_2}$ are the error range of the main steam pressure, electric power, and IP extraction pressure, respectively; $M_{p_1}$, $M_{N_E}$, and $M_{p_2}$ are the fluctuation range of the main steam pressure, electric power, and IP extraction pressure, respectively.

According to the above control process requirements and the optimization control strategy proposed in the study by Wang et al. (2019), the control concepts of disturbance compensation and multivariable coordination are further adopted. Through the design of three key control modules: electrothermal coordination, thermal state reconstruction, and accurate energy balance, and
considering the main influencing factors of main steam fluctuation, a new electrothermal power coordinated distribution and pressure safety self-test control strategy is proposed. The parameters of the control system are manually set by an empirical method to realize the three key functions of the CHP unit control system: accurate electric power tracking, rapid thermal power recovery, and safe and stable operation of the system. The structure of the control system is shown in Figure 6.

Control scheme II:

\[
\begin{align*}
E_T(t) &= p_{tp}^s - \frac{K_1 e^{-t/T_1}}{T_1} p_t(t) \\
E_B(t) &= (N_{E}^{sp} - N_{E}^{p}) - \left\{ N_{E}(t) - \frac{K_2 e^{-t/T_2}}{T_2} p_z(t) \right\} \\
E_H(t) &= N_{E}^{sp} - N_{E}(t)
\end{align*}
\]

Among them, \(K_1, T_1, K_2, \) and \(T_2\) are the parameters of the pressure safety self-test and electrothermal power coordinated distribution control strategy, as shown in Supplementary Table S3.

The variable load regulation process of the CHP unit will lead to large changes in the heating capacity of the unit in a short time, and the heat transfer is delayed. The increase or decrease in long-term accumulated heat will lead to the change of the ambient temperature of the heat load, thus affecting the comfort of users. Therefore, after the step signal of variable load disturbance is given for the dynamic model of the CHP unit, it is necessary not only to test the changes of main steam pressure and electric power but also to further analyze the change process of extraction steam flow in the optimal control cycle. The influence of heat change caused by extraction steam flow fluctuation on heat load demand...
is quantitatively analyzed, and calculate it according to the following method (Wang et al., 2019) (Supplementary Table S2):

\[ q_H(t) = K_7 K_6 q_x(96 p_1(t) - t_1 + 103) \]  

\[ q_{H,\text{equ}} = \int_{t_0}^{t_e} q_H(t) dt \]  

\[ Q_{H,\text{equ}} = q_{H,\text{equ}} \cdot \Delta h \]  

where \( Q_{H,\text{equ}} \) is the thermal power of the equivalent CHP unit, \( \Delta h \) is the enthalpy drop of heating extraction (\( \Delta h = 2.3637 \times 10^3 \)), \( q_{H,\text{equ}} \) is the average equivalent flow of the regulation process, and \( t_0 \) and \( t_e \) are the starting and ending points of optimal control of the CHP unit, respectively.

The CHP unit model has obvious system nonlinearity. The object transfer function parameters contain the input variable information that determines the operating point, so the transfer function has different nonlinear characteristics at different operating points. By using the same PID controller setting parameters to test the control quality of the unit system under different working conditions of heating and pure condensation, how strong the nonlinearity of the CHP unit is measured. Through the analysis of simulation results to determine the method that setting controller parameters can ensure the system control quality under common working conditions, guide the coordinated control system structure or controller parameters to make some adjustment, so as to adapt to the changes of controlled object parameters.

**EXAMPLE ANALYSIS**

In order to verify the effectiveness of the models and methods proposed in this paper, a 300-MW steam extraction heating unit is used to build a model on a MATLAB/Simulink platform for simulation. The variable load open-loop characteristics of the unit are tested by controlling variable disturbance, and the variable load closed-loop dynamic characteristics of the unit are tested by designing control strategy. The response curves of each output are simulated to verify the reproducibility of the linearized model, and the effectiveness of the control method and strategy is analyzed. Relevant parameters of the unit and operating point parameters under pure condensation and heating conditions are shown in Supplementary Appendix SA, SB.

**Open-Loop Characteristic Analysis**

Under the heating condition, when the electric load is given a step signal of 25 MW, the regulation index performance of the system and the optimization process of three control variables are shown in Figure 7. After setting 20 populations of parameters and 20 iterations, the optimization algorithm can stably calculate three control variables after 30 simulation calculations, and the performance fitness value of the regulation index of the system can stably converge to less than 0.2. The \( V_T \) regulation range is [2.5,7]%; the \( V_B \) regulation range is [5.0,13.0]t/h, and they act in the same direction. The \( V_H \) regulation range is widely distributed and inversely correlated with the fitness value.

When the step signal of -25 MW is given for the electric load under the pure condensation condition, the regulation index performance of the system and the optimization process of three control variables are shown in Figure 8. With the same algorithm parameter setting as the heating condition, the adjustment index performance fitness value of the system can stably converge to less than 0.5. The \( V_T \) regulation range is [2.5,7]%; the \( V_B \) regulation range is [5.0,13.0]t/h, and they act in the same direction. The \( V_H \) regulation range is widely distributed and inversely correlated with the fitness value.

Under the heating condition and the pure condensation condition, the output changes of the system under the synergistic action of control variables are shown in Figures 9, 10. Under the heating condition, given the step signal and adjusting the control variables (5.52%, 10.15 t/h, 14.39%), after about 1,500 s, the output power of the system accurately meets the 25 MW required for the variable load, the main steam pressure is adjusted back to \(-0.04\) MPa, and the extraction pressure is...
stabilized to −0.01 MPa. Under the pure condensation condition, given the step signal and the regulated control variable (−5.59%, −10.58 t/h, −8.54%), after about 1,500 s, the system output power accurately reaches −25 MW required by the variable load, the main steam pressure is adjusted back to 0.0007 MPa, and the extraction pressure is stabilized to 0.0015 MPa.

The open-loop regulation eliminates the complex PID setting work and can accurately complete the electric power regulation. However, there is a difference between the safe main steam pressure involved in the variable load process and the extraction pressure of heating, and the calculation time of one-time optimization is about 5 min, which is a long adjustment time.

Closed-Loop Characteristic Analysis

Based on the established s-domain system simulation model of the CHP unit, the step signals of the electric load are given as 5, 15, and 25 MW, respectively. The closed-loop control simulation analysis is carried out by using control scheme I and control scheme II. The changes of electric power, main steam pressure, and extraction pressure of the CHP unit are shown in Figure 11.

Under the two control schemes, the electric power of the unit increases rapidly. In the three variable load scenarios, it takes 10, 35, and 80 s, respectively, to accurately reach the step expected steady-state value. The stronger the variable load, the longer the regulation time. Specifically, the response speed and climbing rate of scheme II are slightly better than scheme I in the first 30 s of regulation.

The dynamic fluctuation of the main steam pressure regulation process is directly related to the safe and stable operation of the unit. It can be seen from the figure that the larger the variable load, the more intense the main steam pressure fluctuation. Scheme II improved by the pressure safety self-test control strategy effectively reduces the pressure fluctuation in the control process. Compared with scheme I, the peak value of pressure fluctuation in scheme II is reduced by about 50%, and all indexes of the control system can return to the steady-state value within 350 s.

In control scheme I, the extraction pressure cannot be adjusted back to the steady-state value (0 MPa, which does not affect heating), and the stronger the variable load capacity, the greater the steady-state deviation of the extraction pressure (the greater the impact on heating). Scheme II improved by the electrothermal cooperative control strategy can realize the rapid recovery function of heating. It can be seen from the
figure that scheme II can recover the extraction steam pressure closer to the steady-state value after 100-s dynamic adjustment. The influence of steam extraction pressure change on heating in the two schemes (compared with rated heating conditions $p_z = 0.35$, $q_h = 400$, $Q_h = 262.63$) can be calculated by Eqs 20–22.

The data in Table 1 shows that, compared with scheme I, the control strategy of scheme II has less impact on the heating power output of the unit, and the heat change in the optimization cycle is reduced by about 60% (reference value: 0 kJ), which can ensure the stability of the thermal output of the unit when the electric power changes to load.

The working point linearization model is established under the pure condensation condition, and the same PID parameters as the control strategy of scheme I of the heating condition are used to analyze the robustness of the control system and the universality of PID parameters. As shown in Figure 12, the control quality of the pure condensation condition is acceptable as a whole, which shows that although the object has system nonlinearity, the robustness of the existing conventional controller is strong enough to overcome the impact caused by this nonlinearity. It also shows that the controller parameters set at any working point of the unit can adapt to the normal load variation range of the unit. However, there are still problems similar to those under rated heating conditions. The control effect of the main steam pressure and the extraction pressure is slightly insufficient. The main steam pressure regulation process fluctuates, and the extraction pressure cannot be adjusted without error. In the simulation test, the control strategy of scheme II cannot be applied to the pure condensation condition, and the safety self-test and electrothermal coordination parameters need to be reset.

CONCLUSION

In this paper, the dynamic characteristic simulation model of the CHP unit based on working point linearization is established, and the open-loop characteristics of the variable load process are analyzed by using the cooperation of improved particle swarm optimization control variables. Through the control strategy based on electrothermal cooperation and safety self-test, the closed-loop characteristics of unit dynamic regulation are studied, and the following conclusions are obtained:

1) When the control variable acts alone, $V_B$ has an obvious effect on the change of electric power, $V_T$ has a great effect on the change of main steam pressure, and $V_H$ mainly affects the change of extraction steam pressure. When the three control variables work together, the electric power can be adjusted accurately, but the safe main steam pressure involved in the variable load process and the extraction steam pressure of heating are adjusted with difference, and the adjustment time is long.

2) The closed-loop characteristic simulation of the control strategy based on electrothermal cooperation and safety self-test shows that the electric power response speed and climbing rate are better, the peak value of pressure fluctuation is reduced by about 50%, and the recovery of extraction pressure closer to the steady-state value has less impact on heating.

3) Experiments on pure condensing and heating conditions with the same PID parameter show that the robustness of the conventional controller is strong enough to overcome the influence of modeling nonlinearity at different working points and adapt to the normal load variation range of the unit.

### TABLE 1 | Influence of steam extraction pressure variation of different heating schemes on heating.

| Variable load (MW) | Scheme | Average value within 600 s | 600 s heat change/kJ |
|--------------------|--------|-----------------------------|----------------------|
|                    |        | $\Delta p_z,\text{equ}$ | $\Delta q_h,\text{equ}$ | $\Delta Q_h,\text{equ}$ |
| 5                  | I      | 0.0121                      | 6.94                 | 4.56                 | 2,734.82               |
|                    | II     | -0.0044                     | -2.54                | -1.67                | -1,001.28              |
| 15                 | I      | 0.0334                      | 19.25                | 12.64                | 7,582.85               |
|                    | II     | -0.0157                     | -9.01                | -5.92                | -3,550.21              |
| 25                 | I      | 0.0541                      | 31.17                | 20.47                | 12,279.48              |
|                    | II     | -0.0205                     | -11.79               | -7.74                | -4,643.11              |
DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Material; further inquiries can be directed to the corresponding author.

AUTHOR CONTRIBUTIONS

YH put forward the main research points. QC completed manuscript writing and mathematical analysis. JY contributed to the framework formation and revision. TL completed the simulation research.

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SUPPLEMENTARY MATERIAL

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