

Research on Temperature Control System of Capacitor Hot Press Based on Improved Particle Swarm Optimization

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1 **Research on Temperature Control System of Capacitor Hot Press**

2 **Based on Improved Particle Swarm Optimization**

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8 **Abstract**

9 In the production of film capacitors, the temperature control of the capacitor hot press is very
10 important. PID control algorithm is widely used in the capacitor hot press. However,
11 traditional PID control algorithm often has difficulty setting parameters. It is not suitable for
12 such a large hysteresis and nonlinear temperature control system. Through the research of
13 traditional PID control algorithm, Particle Swarm Optimization (PSO) algorithm and fuzzy
14 control algorithm, an improved particle swarm algorithm is proposed, and the Improved
15 Particle Swarm Optimization (IPSO) algorithm is applied in the fuzzy PID temperature
16 control system. The temperature control system is simulated by MATLAB. The research
17 results show that the fuzzy PID temperature control system based on IPSO is superior to
18 other temperature control system in terms of adjustment time, overshoot and anti-interference
19 ability.

20 **Introduction**

21 As one of the most basic electronic components in electronic circuits, film capacitors are
22 widely used in consumer electronics, electric vehicles, new energy grid connection, oil and
23 gas exploration, aerospace and other fields. In the production process of film capacitors, the
24 hot pressing process directly determines the quality and service life of film capacitors. If the
25 film capacitor is not well pressed by heat, delamination will occur, so that the film capacitor
26 will have voids, resulting in poor frequency characteristics, increased dielectric loss, and
27 signal distortion. There are three important parameters in the hot pressing process:
28 temperature, time, and pressure. Only under the condition of uniform heating and force, the
29 performance and quality of the film capacitor obtained by hot pressing will be high. During
30 hot pressing, the researchers designed an elastic connection device between the hot pressing
31 plates to achieve uniform pressure on the film capacitor. Because the area of the hot-pressing
32 plate of the capacitor hot-pressing machine is large, the heat balance will be destroyed when
33 entering and leaving the raw materials, and the temperature recovery time of the hot-pressing
34 plate is long. The hot-pressing efficiency and hot-pressing quality further improve the use
35 efficiency of the capacitor hot-pressing machine.

36 The PID controller is often used in the temperature control system of the capacitor heat press.
37 According to the difference between the given value and the measured value of the system,
38 the PID controller realizes the temperature control through proportional integration and
39 differentiation. For the industrial process with relatively simple production process and

40 system structure, the traditional PID control algorithm has better control effect. However,
41 more and more nonlinear and large hysteresis systems appear in modern industry, which
42 requires higher and higher control precision of the system. The traditional PID control
43 algorithm is difficult to meet its control requirements.

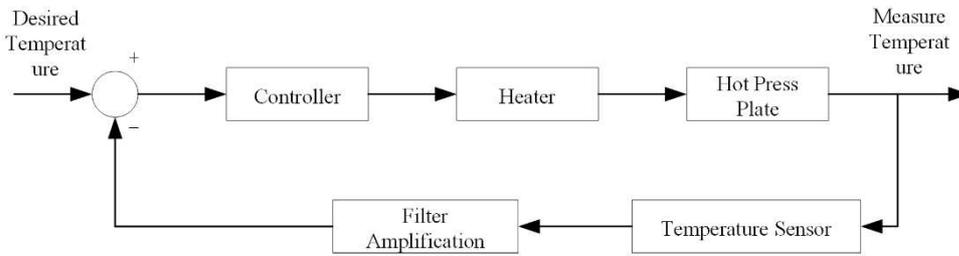
44 With the development of intelligent control technology, many scholars have done a lot of
45 research on temperature intelligent control technology. In order to control the temperature
46 during the transesterification process of edible oil into biodiesel, M F et al. [1] used PID
47 algorithm [2] as the system controller to build a prototype of the biodiesel reactor control
48 system. Lu H et al. [3] developed a multifunctional combined tumour therapy instrument
49 based on improved PID algorithm, using improved PID temperature control algorithm to
50 solve the problem of reduced accuracy due to changes in the process characteristics of
51 different heaters, so as to ensure different The heater is uniformly and smoothly heated by
52 radio frequency when treating tumours in the cavity. Li G [4] et al. used flue gas temperature
53 and coal flue gas flow as the control parameters of the main loop and the secondary loop, and
54 established a linear regression mathematical model of the top temperature of the regenerator
55 and the temperature of the flue side of the machine using linear regression theory. According
56 to the characteristics of the system, a compound fuzzy PID control strategy is proposed. Ye L
57 et al. [5] adopted fuzzy PID control method and designed a self-tuning PID control system to
58 keep the temperature of the honing coolant constant during the production process of the
59 automobile engine. Liu X [6] et al. proposed a loop optimization method (PSO-RCO) based
60 on particle swarm calculation and closed-loop simulation. This method is used to solve the
61 problem of closed-loop optimization of controller parameters of a multi-controller single
62 output thermal system. In order to effectively control the bed temperature of the boiler,
63 Aygun H et al. [7] proposed a particle swarm optimization proportional integral derivative
64 (PSO-PID) controller for the bed temperature of a circulating fluidized bed boiler.

65 In this paper, fuzzy control technology [8-10] and particle swarm optimization technology are
66 combined, applied to the temperature control system of capacitive hot press, and a fuzzy PID
67 temperature control system is constructed. The parameters of the fuzzy PID controller are
68 optimized by particle swarm optimization [11-12], but particle swarm optimization is easy to
69 be premature and fall into local optimization [13-14], which directly affects the performance
70 of the algorithm and leads to inaccurate control accuracy. Aiming at this deficiency, this
71 paper adopts a new method to improve the particle swarm algorithm. The improved particle
72 swarm algorithm is applied to the fuzzy PID control system, and the simulation research is
73 carried out. Compared with the PID temperature control system [15], the fuzzy PID
74 temperature control system [16] and the PSO optimized fuzzy PID temperature control
75 system [17], the IPSO optimized fuzzy PID temperature control system greatly reduces the
76 over-shoot and adjustment time.

77 In Section 2, we introduce the temperature control system structure of the capacitor hot press
78 and establish the mathematical model of the system. In Section 3, we introduce the basic
79 concepts of the PSO algorithm and methods to improve the particle swarm algorithm. In
80 Section 4, we carry out the design of the fuzzy logic of the system. In Section 5, we conduct a
81 simulation study of a capacitor heat press temperature control system by MATLAB. Finally,
82 the full text is summarized in Section 6.

83 **The temperature Model of Capacitor Hot Press**

84 This paper takes the capacitor hot press as the research object. Its temperature control system
 85 mainly includes electric heater, hot pressing plate and temperature sensor. The schematic
 86 diagram of the temperature control of the capacitor hot press is shown in Figure 1. The
 87 controller sends an instruction to the electric heater according to the desired temperature, the
 88 electric heater starts to heat. At the same time, the temperature sensor feeds back the
 89 difference between the real-time temperature value and the set value to the controller.



90

91 Figure 1. The schematic diagram of the temperature control of the capacitor hot press.

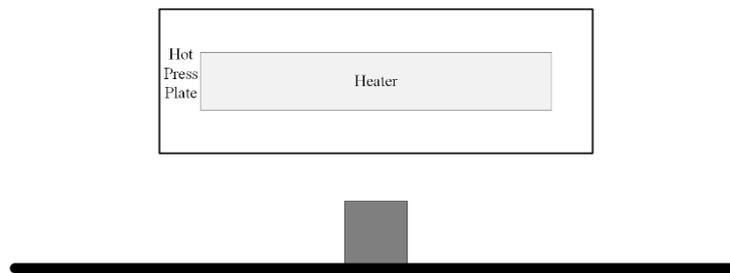
92 The structure of the temperature control system of the capacitor hot press is shown in Figure
 93 2. The transfer function of the heater is:

$$G_R(s) = \frac{K_R}{1 + T_R s} \quad (1)$$

94 The transfer function of the hot press plate is:

$$G_H(s) = \frac{K_H}{1 + T_H s} e^{-\tau s} \quad (2)$$

95 Among them, K_H and K_R are the gain coefficients of the transfer function; T_H and T_R are
 96 the inertia time constants; τ is the pure delay time of the system.



97

98 Figure 2. The structure of the temperature control system of the capacitor hot press.

99 **Improved Particle Swarm Optimization**

100 PSO is a random search algorithm with group collaboration as the core. It was proposed by
 101 Eberhart and Kennedy in 1995 on the basis of the study of bird predation behaviour. In this
 102 algorithm, the particles in the search space represent the solution of the problem, and these
 103 particles form a whole. The direction and distance of the particle's flight are determined by its

104 speed. The particles fly in the solution space along the optimal solution. When a certain
 105 number of iterations is reached, the optimal position vector of the particle is the optimal
 106 approximate solution of the problem. Particle swarm optimization has the advantages of
 107 simple iteration and fast convergence speed, and is suitable for parameter optimization of
 108 PID algorithm.

109 Assuming that K particles are initialized in the D -dimensional solution space, each particle
 110 has its own spatial position $X_m=(X_{m1},X_{m2},\dots,X_{mD})$ and its own flight speed
 111 $V_m=(V_{m1},V_{m2},\dots,V_{mD})$. Among them, $m=1,2,3,\dots,K$. In each iteration of the particle in the
 112 solution space, optimize its own speed through the optimal solution of the individual particle
 113 $P_{best_m}=(P_{m1},P_{m2},\dots,P_{mD})$ and the optimal solution of the particle population
 114 $G_{best_m}=(G_1,G_2,\dots,G_D)$ And location. The update formulas of particle velocity and position at
 115 the N th iteration are can be determined on the basis of the following equations:

$$V_{mD}^N = V_{mD}^{N-1} + c_1 r_1 (P_{mD} - X_{mD}^{N-1}) + c_2 r_2 (G_{mD} - X_{mD}^{N-1}) \quad (3)$$

$$X_{mD}^N = X_{mD}^{N-1} + V_{mD}^{N-1} \quad (4)$$

116 Where, V_{mD}^N denotes the velocity of the m particle, X_{mD}^N denotes the position of the m particle,
 117 c_1 and c_2 denote the learning factor, r_1 and r_2 denote the learning factor random function
 118 within $[0,1]$, and N denotes the number of iterations.

119 In the process of optimizing the parameters, the PSO algorithm is easily trapped in local
 120 optimization, which can directly affect the performance of the algorithm and result in
 121 inaccurate parameters. A new improved particle swarm optimization algorithm is proposed to
 122 avoid this. By introducing two inertia weights w_1 and w_2 in Eq. (3) to enhance the global
 123 search capability of the algorithm. The particle velocity of the N th iteration after the
 124 algorithm is improved can be determined according to the following equation:

$$V_{mD}^N = w_1 V_{mD}^{N-1} + w_2 (c_1 r_1 (P_{mD} - X_{mD}^{N-1}) + c_2 r_2 (G_{mD} - X_{mD}^{N-1})) \quad (5)$$

125 Where,

$$w_1 = w_{max} + (w_{max} - w_{min}) \times n/N_{max} \quad (6)$$

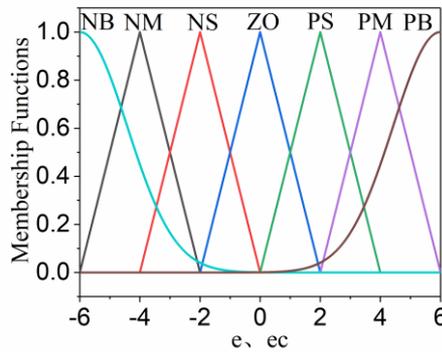
$$w_2 = w_{max} - (w_{max} - w_{min}) \times n/N_{max} \quad (7)$$

126 Where, w_{max} is the maximum inertial weight and w_{min} is the minimum inertial weight, n is the
 127 current number of iterations, and N_{max} is the maximum number of iterations. As the number
 128 of iterations increases, the inertia weight w_1 increases, the flying speed of the particles
 129 becomes faster, and the global search capability of the particles is enhanced, which solves the
 130 problem that the PSO is easy to fall into local optimization. However, when the inertia weight
 131 w_1 is too large, the flying speed of the particles is too fast, and the algorithm will oscillate at
 132 the global optimal solution, leading to premature convergence. In order to constrain the flying
 133 speed of particles, an inertia weight w_2 is introduced. With the increase of the number of
 134 iterations, the inertia weight w_2 decreases, and the flying speed of the particles is balanced by
 135 the inertia weight w_1 and the inertia weight w_2 , thereby obtaining the optimal correction
 136 amount of the PID parameters.

137 **Fuzzy Logic Design**

138 The deviation between the desired temperature and the measured temperature of the fuzzy
 139 temperature control system is set as e , and the deviation change rate is set as ec . Let e and ec
 140 be the input variables of the fuzzy particle swarm PID control system, and the corrections of
 141 the PID parameters K_p , K_i , and K_d , that is, ϕK_p , ϕK_i , and ϕK_d , are set as the output variables.
 142 It is assumed that the basic domains of e and ec are $[-5, +5]$ and $[-10, +10]$ respectively, and
 143 the basic domains of ϕK_p , ϕK_i , ϕK_d are $[-1, +1]$, $[-0.6, +0.6]$, $[-0.8, +0.8]$. Set the fuzzy
 144 universe of e and ec as $[-6, 6]$, and the fuzzy universe of ϕK_p , ϕK_i , ϕK_d as $[-3, 3]$. A seven-
 145 segment fuzzy method is used to divide the input variables and output variables into seven
 146 language values: positive big (PB), positive middle (PM), positive small (PS), zero (ZO),
 147 negative small (NS), negative middle (NM), negative big (NB). The quantization factors for
 148 the input variables e and ec to transform from the basic domain to the corresponding fuzzy
 149 domain are $k_e=6/5=1.2$, $k_{ec}=6/10=0.6$. The scale factor for the output variables ϕK_p , ϕK_i , ϕK_d
 150 to transform from the corresponding fuzzy domain to the basic domain are $k_p=1/3=0.33$,
 151 $k_i=0.6/3=0.2$, and $k_d=0.8/3=0.267$.

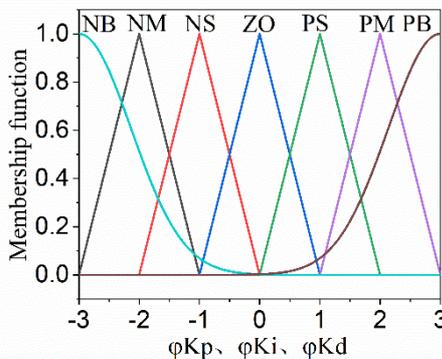
152 The input variables of the fuzzy controller designed in this paper are Gaussian membership
 153 functions, and the output variables are triangular membership functions. The membership
 154 functions of the input variables are shown in Fig. 3, and the membership functions of the
 155 output variables is shown in Fig. 4.



156

(a)

157



158

(b)

159

160 Figure 3. Membership function: (a) Membership of input variables.; (b) Membership of output variables.

161 After analysing the relationship between e , ec and ϕK_p , ϕK_i , ϕK_d , combined with the
 162 knowledge and practical operating experience of experienced old employees, a fuzzy rule
 163 table is obtained. The fuzzy rule tables of ϕK_p , ϕK_i , ϕK_d are shown in Table. 1, Table. 2, and
 164 Table. 3.

165 Table 1. Fuzzy rule table of ϕK_p .

$\begin{matrix} ec \\ e \end{matrix}$	NB	NM	NS	ZO	PS	PM	PB
NB	PB	PB	PM	PM	PS	ZO	ZO
NM	PB	PB	PM	PS	PS	ZO	NS
NS	PM	PM	PM	PS	ZO	NS	NS
ZO	PM	PB	PS	ZO	NS	NM	NM
PS	PS	PS	ZO	NS	NS	NM	NM
PM	PS	ZO	NS	NM	NM	NM	NB
PB	ZO	ZO	NM	NM	NM	NB	NB

166 Table 2. Fuzzy rule table of ϕK_i .

$\begin{matrix} ec \\ e \end{matrix}$	NB	NM	NS	ZO	PS	PM	PB
NB	NB	NB	NM	NM	NS	ZO	ZO
NM	NB	NB	NM	NS	NS	ZO	ZO
NS	NB	NM	NS	NS	NS	PS	PS
ZO	NM	NM	NS	ZO	ZO	PM	PM
PS	NM	NS	ZO	PS	PS	PM	PB
PM	ZO	ZO	PS	PS	PS	PB	PB
PB	ZO	ZO	PS	PM	PM	PB	PB

167 Table 3. Fuzzy rule table of ϕK_d .

$\begin{matrix} ec \\ e \end{matrix}$	NB	NM	NS	ZO	PS	PM	PB
NB	PS	NS	NB	NB	NB	NM	PS
NM	PS	NS	NB	NM	NM	NS	ZO
NS	ZO	NS	NM	NM	NS	NS	ZO
ZO	NS	NS	NS	NS	NS	NS	ZO
PS	ZO						
PM	PB	PS	PS	PS	PS	PS	PB
PB	PB	PM	PM	PM	PS	PS	PB

168

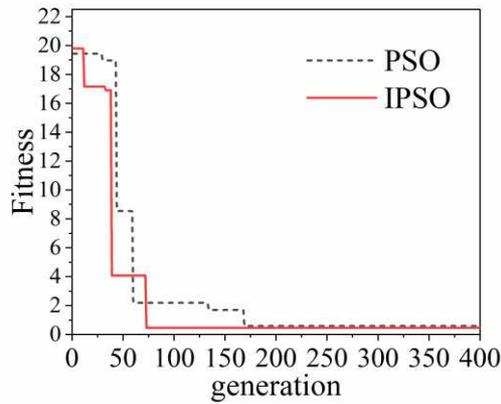
169 Simulation and Results Analysis

170 Algorithm test

171 In order to verify that IPSO algorithm is superior to PSO algorithm, a common standard test
 172 function is selected to compare the performance of the PSO and the IPSO. The expression of
 173 the test function Ackley is can be determined on the basis of the following equation:

$$f(x) = -20 \exp \left[-\frac{1}{2} \sqrt{\frac{1}{n} \sum_{i=1}^n x_i^2} \right] - \exp \left[\frac{1}{n} \sum_{i=1}^n \cos(2\pi x_i) \right] \quad (8)$$

174 The number of selected particles is 20, and the maximum number of iterations is 400. Use
 175 MATLAB software to test the performance of PSO and IPSO, and get the curves of the two
 176 algorithms to optimize the Ackley function. The result is shown in Figure 4. For the test
 177 function Ackley, the convergence curve of the PSO begins to con-erge at the 173rd
 178 iteration, and the convergence curve of the IPSO begins to con-erge at the 73rd iteration.
 179 The IPSO is better than the PSO in terms of convergence speed and optimization accuracy.



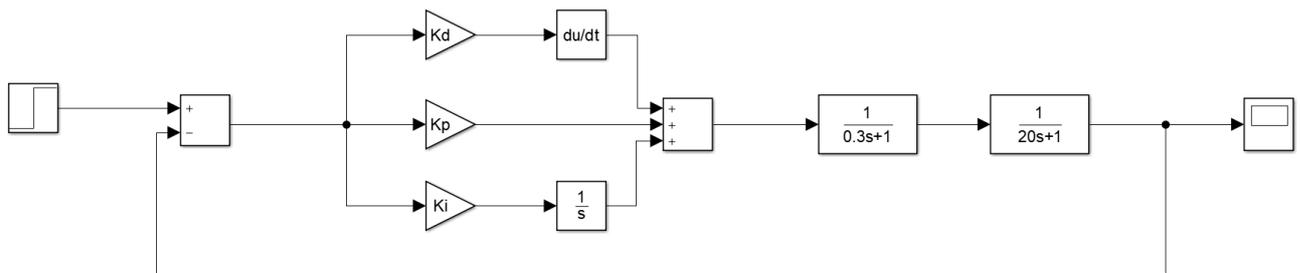
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Figure 4. Ackley function optimization curve.

182 **Control System Simulation**

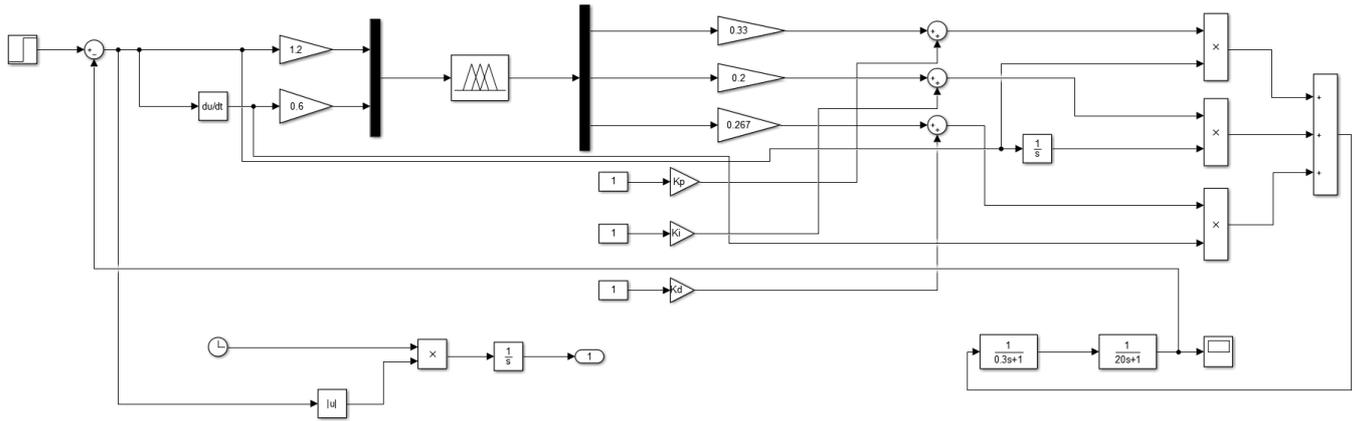
183 In order to verify the advantages of the IPSO algorithm, based on the thermal pressure model
 184 of the capacitor established, the simulation model of the system is established in
 185 MATLAB/Simulink, and the program of the particle swarm algorithm is written in the m file,
 186 through the m file and Simulink. Carry out simulation experiments. The control simulation
 187 model is shown in Figure 5 and Figure 6.



188

189

Figure 5. The simulation model of PID.



190

191

Figure 6. The simulation model of fuzzy PID controller system.

192

Firstly, the particle swarm algorithm is initialized, and the particle swarm size is 5, the number of iterations is 100, the learning factors c_1 and c_2 are both 2, and the inertia weight is 0.6. Then, ITAE is selected as the evaluation index of the controller performance. The expression of ITAE is shown in Equation 9.

193

194

195

$$J_{ITAE} = \int_0^{t_s} t|e(t)| dt \quad (9)$$

196

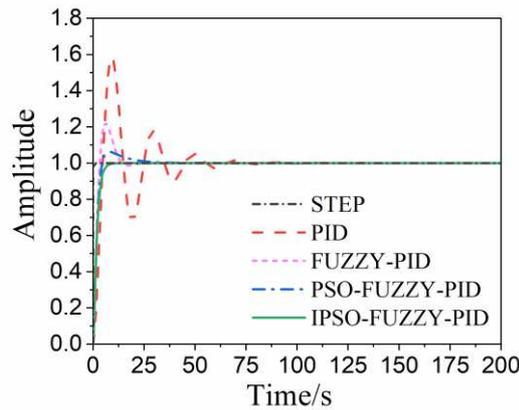
According to the transfer function of the established model, select the gain coefficients K_H and K_R as 1, and the inertia time constants T_H and T_R as 20 and 0.3. Select a conventional step signal as the input signal, and add a step interference of 20% amplitude at 125s. Comparing the performance of IPSO algorithm and other algorithms through the system's ability to adjust the step signal and the adjustment speed.

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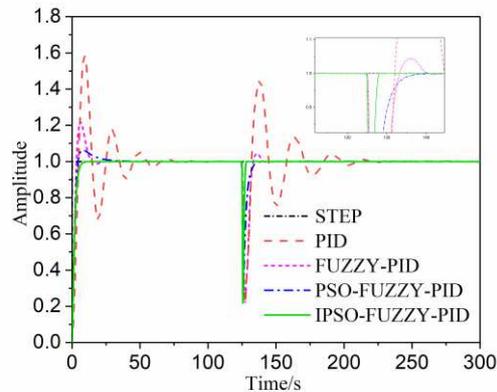
Figure 7. Simulation comparison of four control systems.

203

Table 4. Comparison of performance indicators of four control systems.

Control System	Overshoot	Adjustment time / s
PID	58.173%	66
FUZZY-PID	21.932%	40
PSO-FUZZY-PID	6.295%	34

204 Figure 7 shows the step response signals corresponding to the four control systems. It can be
 205 seen from the simulation curve that the systems controlled by the four algorithms all return to
 206 a stable state within a certain period of time, which indicates that they have anti-interference
 207 capabilities. reach stability. At this time, the interference signal is added into the system, and
 208 the simulation result is shown in Figure 8. As can be seen from the simulation curves in
 209 Figure 8, these four control methods all exhibit different levelling abilities. The traditional
 210 PID control can suppress the system disturbance, but due to the fixed parameters, it is easy to
 211 fall into the empirical optimum. The fuzzy PID control optimized by PSO can make the
 212 system response faster and the anti-interference ability enhanced by optimizing the PID
 213 parameters, but the PSO algorithm is easy to fall into local optimization. The fuzzy PID
 214 control optimized by IPSO solves the problem that the PSO algorithm is prone to premature
 215 maturity, with faster response time and stronger anti-interference ability.



216

217

Figure 8. Simulation comparison after adding interference.

218

Table 5. Comparison of performance indicators after adding interference.

Control System	Overshoot	Adjustment time / s
PID	75.7%	103
FUZZY-PID	75.7%	26
PSO-FUZZY-PID	75.7%	14
IPSO-FUZZY-PID	75.7%	4

219 The performance indicators of the four control systems are shown in Table 4 and Table 5. It
 220 can be seen from Table 4 and Table 5 that under traditional PID control, the system overshoot
 221 is 58.173%, and the system reaches stability within 66s; under fuzzy PID control, the system
 222 overshoot is 21.932%, and the system reaches stability within 40s Stable; under the fuzzy
 223 PID control optimized by PSO, the overshoot of the system is 6.295%, and the system is
 224 stable within 34s; under the fuzzy PID control optimized by IPSO, the overshoot of the
 225 system is 0.001%, and the system is stable within 20s. After adding the disturbance, the
 226 system overshoot is 75.7%. Under traditional PID control, the system reaches stability within
 227 103s; under fuzzy PID control, the system reaches stability within 26s; under PSO optimized
 228 fuzzy PID control, the system reaches stability within 14s; under IPSO optimized fuzzy PID
 229 control, the system reaches stability in Stable within 14s. Stable within 4s. Therefore, the

230 temperature control system using the IPSO algorithm has a faster response speed and has
231 great ad-vantages in anti-interference ability.

232 **Conclusions**

233 In this paper, an improved particle swarm algorithm is proposed through the re-search of
234 particle swarm optimization. The IPSO algorithm is applied to the temperature control system
235 of the capacitor heat press. The temperature control system based on IPSO algorithm can not
236 only make PID parameters self-tuning, but also solve the problem that PSO algorithm is easy
237 to fall into local optimization. The IPSO algorithm realizes the precise control of the
238 temperature control system of the capacitor hot press. It can be seen from the simulation
239 results that compared with the traditional control system, the control system based on IPSO
240 algorithm has great advantages in overshoot and adjustment time. And the system has good
241 adaptive ability and strong robustness and anti-interference ability. The application research
242 of IPSO algorithm in the temperature control system of capacitor heat press provides a
243 solution to the problem of poor temperature control effect of capacitor heat press, which is of
244 great significance to the development of industrial control systems in the direction of
245 intelligence.

246 **Data Availability**

247 Relevant supporting data has been embedded in the paper in the form of charts.

248 **Conflicts of Interest**

249 The authors declare that there are no conflicts of interest regarding the publication of this
250 article.

251 **Funding Statement**

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