Research Article
Precise Active Seeding Downforce Control System Based on Fuzzy PID

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Soil compaction is an important procedure of precision seeding operation. In this paper, a precise downforce control system based on fuzzy PID was proposed in order to improve the quality of the soil compaction and the accuracy of setting working parameters. The conventional mechanism of seeders for soil compaction was optimised. The compressing spring of the compaction mechanism was replaced by a linear motor, which is actively controlled to adjust downforce in real time. A force sensor was connected in series with the linear motor to detect the actual downforce from a press wheel acting on soil. The detected downforce was employed as feedback for the fuzzy PID model. A slave real-time control system was constructed by using an STM32 microcontroller. A user interface was designed for the portable master computer system based on the ForLinx embedded platform to facilitate the setting of target downforce and display the actual downforce in real time. Meanwhile, it was able to adjust the system for different operating requirements, such as soil stiffness, moisture, and crop species. Experiments were conducted on a soil bin, and the results indicated that the active control system has better performance than the conventional passive system in downforce control. The downforce was stable with a variance less than 2.6% under different conditions, and it was 8.11% less than the conventional passive system.

1. Introduction

With the acceleration of agricultural development, researchers had done lots of works on the precise and active seeding [1, 2]. In reference [3], a pneumatic system is designed to realize the active adjustment of sowing depth. In reference [4], the displacement sensor and the ultrasonic sensor are used to measure the trench depth, and the monitoring of the trench depth is realized by controlling the hydraulic cylinder. Thus, the precise and active seeding will be the future development direction.

As an important part of seeding operation, soil compaction (SC) is key to seed germination, emergence, and growth [5–7]. Inconsistent soil compaction will affect the uniformity of crop emergence and then affect crop yield [8]. The existing methods by using springs to apply loading downforce on compaction mechanism to achieve proper SC has been employed for decades. It is difficult to set a target downforce value accurately because it depends on individual experience of an operator to adjust the compress spring force. Meanwhile, the stability of the downforce is affected by uneven soil densities and fluctuating soil surfaces after seeding drills.

In order to improve the performance of seeders, researchers throughout the world have done some research. A profiling elastic press wheel was designed in a recent study [9, 10]. The structure of the profiling elastic press wheel was retrofitted from the conventional press wheel. Traditional rigid spokes replaced with spring-spokes spring-loaded structures were cancelled. During working, the spring-spoke deformation caused wheel deformation when the wheel encountered an uneven soil surface to form a uniform SC. However, in order to achieve required soil compaction strength, the spring should be adjusted according to the experience of operators at the beginning of the work. Another novel innovation was used in a seeder whose unique character was to adjust loading forces on compaction mechanism automatically by using a hydraulic system [11].
A force sensor was added to the press wheel to measure the downforce of the press wheel applied on fields in real time. If the downforce changed because of uneven soil condition, a hydraulic cylinder, part of the hydraulic system, was controlled by operators to ensure a consistent downforce to achieve a uniform SC, yet the results of the SC also depended on the experience of operators.

Therefore, an applicable automated downforce control system is significant to ensure the even downforce loading on the soil ground to achieve the uniform and consistent SC. This study purposed and evaluated a precise active seeding downforce control method based on fuzzy PID. Moreover, a fast, automatic, precise, and self-adjustable control system of the downforce was implemented.

2. Materials and Methods

2.1. System Structure and Operation Principle. The structure of the system is shown in Figure 1. The system consists of a control system and mechanical structures. The control system includes a master computer system and a slave real-time control system, where the slave real-time control system contains a force sensor, a slave computer controller, a motor driver, a sensor signal amplifier, and a power supply. As shown in Figure 2, the structure of the compaction mechanism contains a rack, a wheel frame, a press wheel, a linear motor, and a force sensor.

When the entire system works, the target downforce was set by using a user interface. The slave real-time control system detects loading forces exerted by the linear motor on the compaction mechanism using the force sensor. The slave real-time controller compares the value of the detected downforce with a target value to obtain the difference. Then, the slave real-time controller can adjust the loading force on the press wheel by controlling the linear motor to eliminate the difference, so that the closed-loop control of the downforce is achieved. The master computer system was designed to display and store actual downforce values in real time.

2.2. Mathematical Model of Downforce Detection. As shown in Figure 2, the force sensor was connected in series with the linear motor. One side of the force sensor was articulated with the rack and the other side was articulated with the press wheel frame. As shown in Figure 3, force was analysed with reference to the structure of the compaction mechanism. Segment AB represents the rack. Point A represents the fixed connection point of the press wheel frame with the rack, and Point C represents the hinge point of the force sensor with the rack. Furthermore, Point D represents the hinge point of the linear motor with the press wheel frame support frame, and point E represents the axis of the press wheel.

According to the principle of moment balance, the downforce of the linear motor acting on the press wheel is shown in the following equation:
The ground is the downforce which is expressed as the force exerted by the linear motor on the press wheel frame, as shown in equation (5) to get the mathematical model of downforce detection, as shown in the following equation:

\[ Q = 0.71F_{CD} + 82. \]  

2.3. Design of the Fuzzy PID Control Algorithms. The system was time-varying, hysteretic, and nonlinear. It was difficult to establish a precise mathematical model that can be used. Therefore, according to the characteristics of working, fuzzy PID algorithms with the characteristics of set theory, language variables, and logical reasoning were used to achieve an effective control.

2.3.1. Design of Fuzzy PID Controller. The basic principle of the fuzzy PID controller was briefly described as follows: firstly, the control rules were established according to the relationship between the revised parameters of the PID controller and the displacement error \( E \) and the error rate \( Ec \) of the fuzzy control, as well as the loading characteristics of the linear motor. Then, \( E \) and \( Ec \) were used as input and PID parameters which are automatically adjusted online according to the rules of fuzzy control to meet the requirements of \( E \) and \( Ec \) in different conditions, so that the system can have better response characteristics. According to the analysis above, the diagram of the fuzzy PID controller is designed and shown in Figure 4.

The comprehensive control strategy of the downforce includes a fuzzy controller and a PID controller. The input of the fuzzy controller was the downforce error \( E \) and the error change rate \( Ec \), while the output was three repair parameters of the PID controller, \( \Delta K_p, \Delta K_i, \) and \( \Delta K_d \).

The values of three parameters of the PID controller are calculated by the following equation:

\[
\begin{align*}
K_p & = K_{po} + \Delta K_p, \\
K_i & = K_{io} + \Delta K_i, \\
K_d & = K_{do} + \Delta K_d,
\end{align*}
\]  

where \( K_p, K_i, \) and \( K_d \) represent the initial values of proportion, integral, and differential, respectively.

2.3.2. Variable Fuzzification. The basic domain values of each parameter were set as follows. Downforce error: \( E [-1200, 1200], \) error change rate: \( Ec [-240, 240], \) and three revised parameters: \( \Delta K_p [-120, 120], \Delta K_i [-30, 30], \) and \( \Delta K_d [-30, 30]. \) Set the corresponding fields of input and output to be \([-6, 6]\). The fuzzy subset was \{NB, NM, NS, ZO, PS, PM, PB\}, and they were negative big, negative medium, negative small, zero, positive small, positive medium, and
positive big, respectively. According to the basic universe of each variable and their quantization grade, the quantization factor and the proportion factor could be calculated as follows:

\[
\begin{align*}
    k_E &= \frac{6}{1200} = 0.005, \\
    k_{Ec} &= \frac{6}{240} = 0.025, \\
    k_{\Delta K_p} &= \frac{120}{6} = 20, \\
    k_{\Delta K_i} &= \frac{30}{6} = 5, \\
    k_{\Delta K_d} &= \frac{30}{6} = 5.
\end{align*}
\]

(8)

2.3.3. Determination of the Membership Function. Common membership functions of three fuzzy states are triangle, Gaussian, s, and Z. Currently, there were no mature and effective methods to establish a perfect membership function. Most of the methods of establishing membership functions were still based on experience and experiments. Although the forms are not exactly the same, there are different ways to solve and deal with the problem of actual fuzzy information as long as the same fuzzy concept could be reflected. Since the triangle membership function had several advantages, such as less memory occupied, simple operation, and suitable for online parameter adjustment, it was selected for the fuzzy control, as shown in Figure 5.

2.3.4. Establishment of Fuzzy Control Rules. Fuzzy control rule was key for the fuzzy control. The rule was established on the basis of practical experience, according to the magnitude of input value at different time in the process of noncontrolling and the relationship among three modified parameters. According to the working characteristics of the system and the relationship of the parameters of the fuzzy PID, the fuzzy control rules were designed as given in the following.

The fuzzy rules of $\Delta K_p$. The function of the proportional coefficient, $\Delta K_p$, was to speed up the response of the system and improve the adjustment accuracy of the system. The larger $\Delta K_p$ makes, the faster the response and the higher the adjustment precision of the system were. However, if $\Delta K_p$ was too large, it was easy to produce overshoot and even lead to instability of the system. If the value of $\Delta K_p$ is too small, the adjustment precision would be reduced and the response would be slow. In addition, adjustment time would be prolonged, and static and dynamic characteristics of the system will be changed. The fuzzy rules of $\Delta K_p$ are given in Table 1.

The fuzzy rules of $\Delta K_i$. The function of the integral action coefficient, $\Delta K_i$, was to eliminate the steady-state error of the system. The larger the $\Delta K_i$ makes, the faster the static error of the system would be. However, if $\Delta K_i$ was too large, integral saturation will occur in the initial stage of the response process, which would lead to a larger overshoot during the response processing. If $\Delta K_i$ was too small, it would make the static error of the system difficult to be eliminated so as to affect the adjustment accuracy of the
system. According to the requirements of working, the table of fuzzy control rules for $\Delta K_i$ is as given in Table 2.

The fuzzy rules of $\Delta K_i$. The differential action coefficient, $\Delta K_d$, could improve the dynamic characteristics of the system. Its main function was to restrain the variation of deviations in any directions during response and to forecast the variation of deviations in advance. However, the excessive $\Delta K_d$ would make the response brake ahead of time, so prolonging the adjustment time and reducing the anti-jamming performance of the system. According to the requirements of working, the table of fuzzy control rules of $\Delta K_d$ is as given in Table 3.

### Table 1: Fuzzy rule of $\Delta K_p$.  

| $E$ | NB | NM | NS | ZO | PS | PM | PB |
|-----|----|----|----|----|----|----|----|
| NB  | PB | PB | PB | PB | PB | PM | ZO |
| NM  | PB | PB | PB | PM | PM | ZO | ZO |
| NS  | PB | PM | PM | PS | ZO | NS | NS |
| ZO  | PM | PM | PS | ZO | NS | NM | NM |
| PS  | PS | PS | ZO | PM | NM | NM | NB |
| PM  | ZO | ZO | ZO | NM | NB | NB | NB |
| PB  | ZO | ZO | NB | NB | NB | NB | NB |

### Table 2: Fuzzy rule for $\Delta K_i$.  

| $E$ | NB | NM | NS | ZO | PS | PM | PB |
|-----|----|----|----|----|----|----|----|
| NB  | NB | NB | NB | NM | NM | NS | ZO |
| NM  | NB | NB | NB | NM | NM | NS | ZO |
| NS  | NB | NM | NM | NS | ZO | NM | NM |
| ZO  | NM | NM | NM | ZO | PS | PM | PM |
| PS  | NS | NS | ZO | PS | PS | PM | PB |
| PM  | ZO | ZO | PS | PS | PM | NB | NB |
| PB  | ZO | ZO | PS | PS | PM | PB | PB |

### Table 3: Fuzzy rule of $\Delta K_d$.  

| $E$ | NB | NM | NS | ZO | PS | PM | PB |
|-----|----|----|----|----|----|----|----|
| NB  | PS | NS | NS | NS | PS | PS | PS |
| NM  | PS | NS | NM | NM | NM | NS | ZO |
| NS  | ZO | ZO | NS | NS | ZO | ZO | ZO |
| ZO  | ZO | ZO | ZO | ZO | ZO | ZO | ZO |
| PS  | ZO | ZO | ZO | ZO | ZO | ZO | ZO |
| PM  | PB | NS | PS | PS | PS | PS | PB |
| PB  | PB | PB | PB | PB | PB | PB | PB |

2.5. Software Design of the System

2.5.1. Interface Design of Master Computer. The downforce monitoring interface is shown in Figure 6. It includes communication setting module, system parameter module, parameter setting module, and dynamic curve display module. In the operation, the user sets the downforce target value through the input box, monitors the operation effect in real time through the data display curve, monitors the communication status through the communication data block, and saves the downforce data in the system background for later analysis.

2.5.2. Program Design of the System. The master computer system communicates with the slave real-time control system through the serial port. The default communication baud rate is 115200b/s, 8-bit data and no check bit. The master computer system was used for parameter setting, data display, and save. The slave real-time control system was used to collect the value of the force sensor in real time and to calculate the downforce value according to the force model. According to the target value, the downforce value and the fuzzy PID controller adjusted the linear motor to realize the real-time control of downforce. The operation parameters were returned to the master computer system in real time. The flowchart of the system is shown in Figure 7.

3. Experiments and Result Analysis

3.1. Experimental Equipment and Conditions. The experiments were conducted based on a soil bin and powered by a John Deere 354 tractor. The experimental devices are shown in Figure 8.
Before the experiments, the soil was prepared to simulate the conditions of fields. The soil moisture content, firmness, and the height difference of soil undulation per unit area were measured. Three groups with 60 measurement points in total were evenly selected within the range of the soil through test benches. The measured data are given in Table 4.

### 3.2. Performance Experiments of Control System

The purpose of static experiments was to verify both the accuracy of the downforce detection model and the control performance of the system under static conditions. The experimental setup is shown in Figure 9.

The force sensor \(f\) was situated under the press wheel to detect actual downforces on the ground from the press wheel and recorded the measured value of the force sensor as the actual value. The target downforces were set as 600 N, 1000 N, and 1400 N. The linear motor was adjusted by the control system, and the actual downforce value was measured by the force sensor \(a\), recorded as the detected value. The experimental data curve is shown in Figure 10.

The average value and the correlation between the detected force of the system and the target force calculation equation were shown as equations (10) and (11) [12]:

\[
r = \frac{\sum_{i=1}^{N} (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^{N} (X_i - \bar{X})^2 \cdot \sum_{i=1}^{N} (Y_i - \bar{Y})^2}},
\]

where \(X_i\) is the collected value of the control system, recorded as the detected value; \(\bar{X}\) is the mean of the collected value; \(Y_i\) is the measured value from the force sensor, recorded as the actual value; \(\bar{Y}\) is the mean of the actual force; \(N\) is the total number of data acquisition; \(s\) is the standard deviation of the measurement points; and

\[
\begin{align*}
    \bar{X} &= \frac{\sum X_i}{N}, \\
    s &= \sqrt{\frac{\sum (X_i - \bar{X})^2}{N-1}}, \\
    V &= \frac{S}{\bar{X}} \times 100\%, \\
    U &= 1 - V,
\end{align*}
\]

where \(V\) is the variation coefficient of the downforce and \(U\) is the stability coefficient of the downforce.

According to Figure 10 and equations (10) and (11), the stability of the control system was calculated when the target value was 600 N, 1000 N, and 1400 N, respectively, as given in Table 5.

According to Table 5, through the comparison between the actual downforce values and the system detection values, the correlation coefficient was 97.1% when the set value was 600 N, while it was 98.8% and 98.5% when the set value was 1000 N and 1400 N, respectively. The results indicated that there was a strong correlation between the actual downforce and the system detection value, indicating that the downforce detection model was reliable.
Figure 7: The flowchart of the program. (a) Flowchart of the master computer system program and (b) flowchart of the slave real-time control system program.
According to Figure 10 and equations (10) and (11), the response time $a$ was calculated when the target value was 600 N, 1000 N, and 1400 N, respectively, as given in Table 6.

According to Table 6, in order to measure the system performance, the rise time of the system from 0 N to the target downforce value was the system response time and the system from 0 N to the stability status of the target downforce value was the adjustment time. When the target value was 600 N, 1000 N, and 1400 N, the response time of the system was 0.5 s, 0.8 s, and 1.5 s, respectively, and the adjustment time of the system was 0.8 s, 1.1 s, and 2.4 s, respectively. The stability coefficient of the system was 98.64%

Table 4: The physical parameters of the soil.

| Group | Soil density (kPa) | Soil moisture content (%) | Surface soil fluctuation difference value (cm) |
|-------|--------------------|---------------------------|-----------------------------------------------|
| 1     | 79.08              | 9.79                      | 5.3                                           |
| 2     | 79.56              | 11.23                     | 6.8                                           |
| 3     | 80.37              | 12.92                     | 4.6                                           |

Table 5: Detection accuracy of the control system.

| Target value (N) | Average of detected value | Average of actual value | Correlation coefficient |
|------------------|---------------------------|------------------------|------------------------|
| 600              | 602                       | 608                    | 0.971                  |
| 1000             | 998                       | 997                    | 0.988                  |
| 1400             | 1396                      | 1399                   | 0.985                  |
at the target value of 600 N, while it was 99.06% and 98.96% when the target value was 1000 N and 1400 N, respectively. The results showed that the downforce control system had a good stability under static conditions.

3.3. Stability Experiments under Different Target Downforce Values. The target downforce was set as 600 N, 1000 N, and 1400 N, respectively, and the forward speed of the tractor was 5 km/h during the tests. The downforce value is shown in Figure 11.

According to Figure 11, when the local table was relatively flat, the downforce value was relatively stable, while if the downforce value had a significant fluctuation, the system timely carried out feedback adjustment, so that the downforce was stable within a range of changes although there were fluctuations. Moreover, the downforce stability coefficient was calculated based on equations (10) and (11) as given in Table 7.

When the target value was 600 N, 1000 N, and 1400 N, the average error range was within 200 N and the coefficient of variation was 5.91%, 6.77%, and 4.17%, respectively. In the condition of different target values, the difference of coefficient of variation of the actual pressure values was less than 2.6%. Thus, the impact of pressure values on system performance under different targets was obtained.

3.4. Stability Experiment of Different Downforce Control Regulations. In order to further verify the reliability of the system, the system was compared with the conventional regulation. In this experiment, other parameters were the same except the downforce control mode. The ordinary method usually loads the downforce by springs without downforce detection modules. Therefore, the downforce detection module in this paper was added onto the mechanical compaction mechanism to achieve downforce detection. The mechanical downforce regulating structure cannot be set because of the restriction of spring characteristics. Hence, the downforce was set as 600 N according to experience. Comparative experiments were conducted, and the results are given in Table 8.

According to Table 8, in terms of the conventional method, the stability coefficient of the downforce was 85.98%, while that was 90.84% by using the proposed method. The results showed that the stability coefficient of the downforce based on the control system was higher than that of the conventional method, so the system improves the stability requirements of the downforce and provides a reference for the downforce control method of seeders.

4. Discussion

In the experiment, the response time, adjustment time, and stability of the system were tested, according to the requirements (stable, accurate, and fast) of the control system. The shorter the response time and adjustment time of the system are, the higher the sensitivity of the system is, but lower the stability and anti-interference ability of the system are. The performance of the system was analysed at the speed of 5 km/h, and in actual working, if work speed increased, the interference frequency could be fast, so that the work effect could be reduced. Therefore, how to improve the universality of the system with different operating speeds should be further studied.
5. Conclusions

(1) This paper proposed an active downforce control method to deal with the limitation of the mechanical downforce control structures. A precise active seeding downforce control system based on fuzzy PID was designed, and the control system was developed for the uniform compaction of seeding by controlling the downforce in real time.

(2) A graphic-based user interface was developed on a master computer system. Parameters for seeding operations could be set easily by this interface, which could highly reduce the requirements for operators.

(3) The experimental results showed that the active downforce control mode was stable and reliable. At the operating speed of 5 km/h, the stability coefficient is 90.84%, which is better than 85.96% of the traditional control mode, and downforce is stable with a variance less than 2.6% under different target values. Moreover, downforce could be adjusted for different crops and working conditions without the modification of seeder structures, which improved the generality of the system.

Data Availability

The data used to support the findings of the study are available at https://pan.baidu.com/s/1K2iidaIlMcuVG5LDvP7L9g.

Conflicts of Interest

The authors declare that there are no conflicts of interest regarding the publication of this paper.

Authors’ Contributions

This paper was guided and modified by Yu Tan, Jian Chen, and Xingxing Liu.

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