

# An experimental study on the creep feed grinding of narrow deep grooves of stainless steel

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**Abstract** Experiments were carried out machining SUS321 stainless steel workpieces using a single-layer electroplated cubic boron nitride (CBN) wheel; a WinTec MV-45 machining center and creep feed grinding following orthogonal test method were used. Narrow deep grooves were machined. The surface roughness  $R_a$  values of the groove side walls were collected and analyzed. The influence of grinding parameters (including wheel speed, feed rate, and cutting depth) on the surface roughness of the grooves was studied. The result showed that the feed rate has the most significant influence on the surface roughness  $R_a$  values of the narrow deep groove, the influence of the wheel speed on the surface roughness  $R_a$  values is the second, and the effect of the cutting depth on the surface roughness  $R_a$  values is the least. The optimum parameters were achieved by orthogonal experiment optimum design and taking  $v_w = 1.5$  mm/min,  $a_p = 5$  mm,  $n = 7000$  rpm for grinding SUS321 stainless steel. A model predicting the surface roughness  $R_a$  values of the machined grooves was established by processing the collected data by partial least square regression. The predictions made by the model match well with the experimental data, and thus, the model can be used to predict the surface quality/roughness in future grinding.

**Keywords** Creep feed grinding · Narrow deep groove · Stainless steel · Orthogonal test · Surface roughness · Partial least square regression

## 1 Introduction

Stainless steel is a typical hard processing material, and abrasive dusts adhere to the blade seriously during its grinding process. So it is easy to cause the grinding wheel to block, which has a great effect on the quality of machined surface and decreases cutting efficiency. In recent years, the development of high-speed grinding technology provides a new way for the grinding of stainless steel. Considerable progress has been made in the study on the grinding process of stainless steel. Wang et al. [1] conducted the stainless steel wire tip grinding experiments in a Chevalier Model Smart-B818 three-axis computer numerical control (CNC) surface grinding machine to study the grinding forces, wire tip deflection, and surface roughness. Ohmori et al. [2] investigated in detail a stainless steel mirror-finished surface obtained by a high-precision grinding process using some advanced surface analyzing techniques, and found that ground surfaces exhibited superior surface properties including hardness, tribological and fatigue properties, corrosion, and high-temperature oxidation resistance. Manimarana et al. [3] conducted grinding experiments on stainless steel 316L to study the grinding force and surface roughness. Zhou et al. [4] ground duplex stainless steel (DSS) 2304 to study the effect of abrasive grit size, grinding force, and lubrication on the surface integrity. Baptista et al. [5] studied the fatigue behavior in welded joints of stainless steels (Austenitic 304L and Duplex S31803 type) treated by weld toe grinding. Hadad and Hadi [6] did the surface grinding tests for hardened stainless steel under dry, MQL, and fluid grinding conditions. The results showed that

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**Table 1** Chemical compositions of SUS321 stainless steel

Element	C	Si	Mn	P	S	Cr	Ti	Ni
Chemical composition/wt%	0.08	1	2	≤0.045	≤0.030	17.00–19.00	5C-0.70	9.00–12.00

for MQL grinding of hardened stainless steel, not only a superior surface finish and quality are obtained but also the tangential grinding forces are lower.

In any grinding process, the surface roughness of the machined piece, which is affected by grinding conditions like depth of cut, feed rate, wheel speed, the lubrication effectiveness, and workpiece characteristics such as hardness, toughness and machinability, etc. is one of the most important factors in assessing the quality of the process. Surface roughness influences several functional attributes of parts, such as contact area and characteristics causing surface friction, wearing, light reflection, heat transmission, ability of distributing and holding a lubricant, load bearing capacity, and fatigue resistance [7]. Therefore, the desired surface roughness is usually specified and the appropriate processes are selected to reach the required quality levels.

In the past several years, many researches have been done about investigating the influence of grinding parameters on surface roughness in order to achieve better surface finishes, and many empirical models predicting the surface roughness of ground components under some grinding conditions have been established. Gopal and Rao [8] conducted experiments of grinding of silicon carbide (SiC) to study the effects of wheel parameters (mesh size and grain density) and grinding parameters (depth of cut and feed) on the surface roughness and surface damage. The results indicated that the parameters feed rate, depth of cut, and grit size are the primary influencing factors which affect the surface integrity of SiC during grinding. The surface roughness model was developed using the experimental data considering only the significant parameters. Optimal grinding conditions were also obtained for maximization of material removal using surface roughness and percentage damage as constraints. Rudrapati et al. [7] made experiments by Box-Behnken design matrix and investigated the influence of machining parameters as infeed, longitudinal feed, and work speed on surface roughness in traverse cut cylindrical grinding of stainless steel material (grade SS410). The results showed that surface roughness  $R_a$  values generally

decrease with an increase in longitudinal feed and increase with an increase in infeed and work speed. Infeed, squared combinations of both longitudinal feed and work speed, and interaction effect infeed-longitudinal feed and longitudinal feed-work speed are found to be the most significant for surface roughness. Mathematical modeling had also been done by response surface methodology (RSM) to develop the relationship between process parameters and surface roughness  $R_a$  values. Lin et al. [9] investigated the effects of grinding parameters and grinding wheels on the grinding performance of in situ TiB2/Al composites employing down-grinding style with emulsified lubrication. The results showed that increasing workpiece speed and grinding depth result in the increase of surface roughness, and the most significant grinding parameters on the grinding quality of in situ TiB2/Al composites are wheel speed and grinding depth, while the effect of workpiece speed on the surface roughness is inconspicuous. Yao et al. [10] conducted the experiments of grinding surface integrity using a SA and a CBN wheel, respectively, and compared and analyzed the surface roughness and topography of ground surface with different grinding parameters. The results showed that the SA wheel is suitable for grinding Inconel718, and the grinding depth has a great effect on the surface integrity in grinding Inconel718. It can also be concluded that better surface can be achieved by using a SA wheel, and taking such grinding parameter as  $a_p = 0.005$  mm,  $v_w = 16$  m/min, and  $v_s = 25$  m/s for grinding Inconel718. Fredj et al. [11] conducted the grinding tests in down cut plunge surface grinding mode. For all tests, a 48 runs design of experiment (DoE) rotatable central composite design was selected concerning the grinding parameters, table speed, depth of cut, grinding wheel, grain mesh size, dressing depth, and the number of passes. The results showed that only the table speed, the depth of cut, the dressing depth, the interaction depth of cut, and the number of passes are significant in the case of  $R_a$ . An approach combining the application of the DoE and the neural network methods was developed to establish accurate models for ground surface roughness prediction.

**Table 2** Mechanical properties of SUS321 stainless steel

Property	Tensile strength, $\sigma_b$ (MPa)	Yield strength, $\sigma_{0.2}$ (MPa)	Elongation, $\delta_5$ (%)	Reduction of area, $\psi$ (%)	Hardness, HBR
Value	≥520	≥205	≥40	≥50	≤83

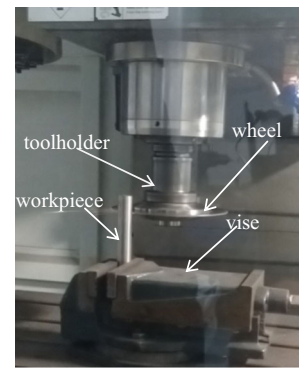
Narrow deep groove is a special type of groove structure with the width of groove to be less than 4 mm and the ratio of the depth over the width (of groove) to be greater than 2. Such grooves are often used on some key parts in the aviation and automotive field. For example, the rotor slot of the rotor pump is a typical part with narrow deep grooves. According to the technical requirements of the manufacturer, the surface roughness  $R_a$  value of the slot wall is  $0.4 \mu\text{m}$ . Machining such narrow deep grooves of high precision has been proven to be challenging. In recent years, the creep feed grinding technology has shown a unique advantage in the process of machining narrow deep grooves [12–16]. Referring back to the rotor pump, when transporting corrosive material, rotor material must be a corrosion-resistant material, such as SUS321 austenitic stainless steel, which has good low-temperature strength as well as good elevated temperature strength and is widely used in various fields of chemical industry and light industry.

Although there exist many literatures relating the grinding parameters with the part roughness and some models for the surface roughness generated by the grinding process, they are not the results of the creep feed grinding of SUS321 stainless steel. In order to find the influence of the grinding parameters such as cutting depth, feed rate, and wheel speed on the surface roughness and achieve precise prediction of surface roughness for effective control of grinding quality during the creep feed grinding of SUS321 stainless steel, experiments were conducted by an orthogonal test method in this work. A cubic boron nitride (CBN) grinding wheel with a single-layer electroplating was used to perform the creep feed grinding of a SUS321 stainless steel workpiece. Straight grooves with a width of 2 mm and a depth of 8, 12, and 5 mm were machined. Each groove was machined according to different grinding parameters. The relationship between the surface roughness  $R_a$  values of the walls of the grooves and the grinding parameters is investigated. The effect of grinding parameters on the surface roughness/quality is studied to make a prediction model of the surface roughness of the machined narrow deep grooves.

## 2 Experimental

### 2.1 Experimental conditions

A cylindrical sample of SUS321 austenitic stainless steel with a size of 100 mm (length) by 20 mm (diameter) was selected for this investigation. The chemical composition and mechanical properties of SUS321 austenitic stainless steel are given in Table 1 and Table 2, respectively. The narrow deep groove grinding experiments were carried out in a WinTec MV-40 machining center with a maximum operational spindle speed of 10,000 rpm using the setup shown in Fig. 1. The single-

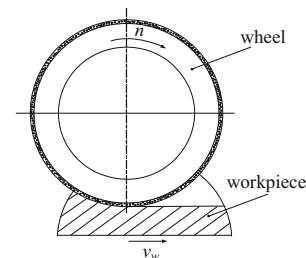


**Fig. 1** Experiment setup for grinding the narrow deep grooves

layer electroplated CBN wheel with 45 steel as the hub body was mounted on the spindle of the machining center by a special toolholder. The wheel size is about 180 mm in diameter and 1.8 mm in depth. The width of the coating is 5 mm. The average grit size is 120 mesh. The volume fraction of CBN in the coating is 22.3 %. The workpiece is secured by a bench vise mounted on the machine table. Processing sketch map (down) is shown in Fig. 2. The grinding experiments were performed in dry condition. After the grinding process of the narrow deep grooves on the SUS321 stainless steel bar, each groove bottom was cut by a wire electrical discharge machining (WEDM) to observe surface morphology of the groove side walls by scanning electron microscope (SEM). The surface roughness  $R_a$  values of the side walls of the narrow deep grooves were measured using a TR220 contact type surface profilometer.

### 2.2 Experimental program

The surface roughness in a creep feed grinding process is affected by many factors such as the rigidity of the machine tool, the material of the workpiece, the grit material, the grit size, the cooling condition, and the grinding parameters. In order to investigate the influence of the grinding parameters (including peripheral wheel speed, feed rate, and cutting depth) on the ground surfaces of the narrow deep grooves, the grinding experiments were conducted in this paper by an orthogonal test method. The orthogonal test is an effective measurement to assay the comprehensive effect of multiple



**Fig. 2** Processing sketch map (down)

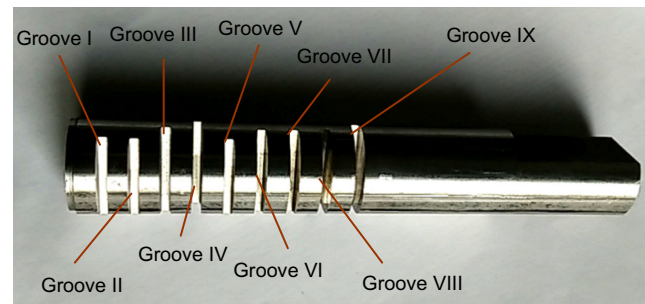
**Table 3** The factors and levels of the orthogonal experiment

Level	Factor		
	Feed rate $v_w$ (A) (mm/min)	Cutting depth $a_p$ (B) (mm)	Wheel speed $n$ (C) (rpm)
1	1	5	3000
2	1.2	8	4500
3	1.5	12	7000

factors [17]. Table 3 lists the factors and levels of the orthogonal experiment in which three grinding parameters including (A) feed rate, (B) cutting depth, and (C) wheel speed were selected. Each parameter has three levels to be optimized. For the interaction between the factors is ignored, and three factors are selected,  $L_9(3^4)$  is the suitable orthogonal array for the experiment [18–20], as is shown in Table 4. Nine trials were carried out according to the  $L_9(3^4)$  array to machine nine narrow deep grooves to complete the optimization process. Each row of orthogonal array represents a run, which is a specific set of factor levels to be tested. The run order of the trials was randomized to avoid any personal or subjective bias. The machined workpiece was shown in Fig. 3. Each groove bottom was cut along its middle line by a wire electrical discharge machining (WEDM) to observe surface morphologies of the groove side walls by SEM. The surface roughness  $R_a$  values of the side walls of each groove were measured across the tool feed direction using a TR220 contact type surface profilometer. The sampling length was given by  $L = 0.25$  mm, and the assessment length is given as  $5L$ . The measurements were taken at different locations on each sample. The surface roughness  $R_a$  values written in the corresponding column of the orthogonal array were the average of the five measurements.

**Table 4**  $L_9(3^4)$  orthogonal test array for the grinding experiments

Trial No.	Factor			Surface roughness $R_a$ value ( $\mu\text{m}$ )	
	Feed rate $v_w$ (A) (mm/min)	Cutting depth $a_p$ (B) (mm)	Wheel speed $n$ (C) (rpm)	left groove walls	right groove walls
1	1	5	3000	0.636	0.479
2	1.2	5	4500	0.891	0.604
3	1.5	8	3000	0.658	0.514
4	1	12	7000	0.842	0.583
5	1.5	5	7000	0.714	0.221
6	1.2	8	7000	1.109	0.509
7	1	8	4500	0.549	0.51
8	1.2	12	3000	1.038	0.502
9	1.5	12	4500	0.469	0.48

**Fig. 3** Machined workpiece

### 3 Results and discussion

From the experiment result in Table 4, it can be seen that the surface roughness  $R_a$  value of No. 9 trial is minimum for the left groove wall and the relatively optimal combination is  $A_3B_3C_2$ . It is also evident that the surface roughness  $R_a$  value of No. 5 trial is minimum for the right groove wall and the relatively optimal combination was  $A_3B_1C_3$ . But these tests may not be the optimum combination scheme. The deep analysis should be done to find the better scheme.

#### 3.1 Range analysis

Range is the difference between the maximum and the minimum of the average of the target, and it is represented as  $R$  which can be obtained by:

$$R_i = \max(p_{i1}, \dots, p_{ij}, \dots, p_{in}) - \min(p_{i1}, \dots, p_{ij}, \dots, p_{in}) \quad (1)$$

where,  $p_{ij} (j = 1, 2, \dots, n; i = 1, 2, \dots, n)$  is the arithmetic mean of the test results obtained when the same levels  $j$  of the factor column  $i$  are selected in the orthogonal table. The value of  $R_i$  is used for evaluating the importance of the factors, i.e., a larger

$R_i$  means a greater importance of the factor. By comparing the different  $p_{ij}$  under the same factor, we can know that the trend of the influence of this factor on test result and get the optimum level of this factor column. The optimal combination of the whole experiment can be obtained by combining the optimal level of each factor column. The range analysis of the surface roughness  $R_a$  values of the left groove walls was shown in Table 5. The trend curve for each factor was shown in Fig. 4.

From the range analysis in Table 5 and the trend curve in Fig. 4, it can be seen that the feed rate has the greatest influence on the surface roughness  $R_a$  values of the left groove walls; the wheel speed has less influence on that, and the cutting depth has the smallest influence. The optimum combination is selected as  $A_3B_1C_2$ . The optimum parameters by orthogonal experiment optimum design are  $v_w = 1.5$  m m/min,  $a_p = 5$  mm, and  $n = 4500$  rpm.

According to the surface roughness  $R_a$  values of the right groove walls in Table 4, the range analysis of the surface roughness  $R_a$  values of the right groove walls was obtained according to the formula (1) and shown in Table 6. The trend curve for each factor was shown in Fig. 5.

From Table 6 and Fig. 5, it can be found that the influence of grinding parameters on the surface roughness  $R_a$  values of the right groove walls is about the same as that on the surface roughness  $R_a$  values of the left groove walls: the effect of feed rate is the most significant, the second important factor is the wheel speed, and the effect of the cutting depth is the least. The optimum combination is selected as  $A_3B_1C_3$ , which is consistent with the result obtained through the intuitive

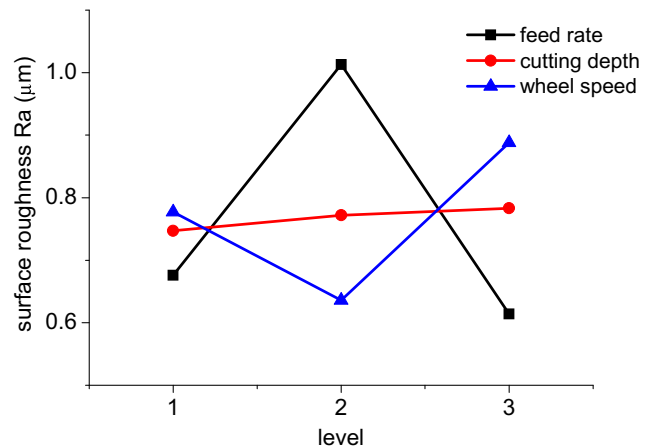


Fig. 4 Trend curve of the level on each factor of the left groove walls

analysis of the surface roughness  $R_a$  values of the right groove walls. So the verification test need not be carried out. The optimum parameters by orthogonal experiment optimum design are  $v_w = 1.5$  mm/min,  $a_p = 5$  mm, and  $n = 7000$  rpm.

### 3.2 Micrograph analysis of surface topography

The surface morphology of the grinding is mainly formed by the superposition of the groove marks generated by the interference of the abrasive cutting edge and the workpiece [21]. The typical SEM picture of the ground groove walls were shown in Fig. 6.

Table 5 Range analysis of the surface roughness  $R_a$  values of the left groove walls

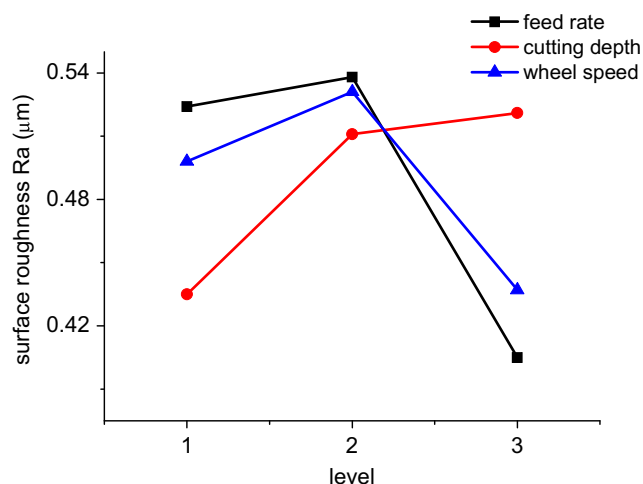
Trial No.	Factor			Surface roughness $R_a$ value of the left groove wall ( $\mu\text{m}$ )
	Feed rate $v_w$ (A) (mm/min)	Cutting depth $a_p$ (B) (mm)	Wheel speed $n$ (C) (r/min)	
1	1	1	1	0.636
2	2	1	2	0.891
3	3	2	1	0.658
4	1	3	3	0.842
5	3	1	3	0.714
6	2	2	3	1.109
7	1	2	2	0.549
8	2	3	1	1.038
9	3	3	2	0.469
$p_{11}, p_{21}, p_{31}$	0.676	0.747	0.777	
$p_{12}, p_{22}, p_{32}$	1.013	0.772	0.636	
$p_{13}, p_{23}, p_{33}$	0.614	0.783	0.888	
Range $R_1, R_2, R_3$	0.399	0.036	0.252	
Order of the factor	A > C > B			
Optimum combination	$A_3B_1C_2$			

**Table 6** Range analysis of the surface roughness  $R_a$  values of the right groove walls

Trial No.	Factor			Surface roughness $R_a$ value of the right groove wall ( $\mu\text{m}$ )
	Feed rate $v_w$ (A) (mm/min)	Cutting depth $a_p$ (B) (mm)	Wheel speed $n$ (C) (rpm)	
1	1	1	1	0.479
2	2	1	2	0.604
3	3	2	1	0.514
4	1	3	3	0.583
5	3	1	3	0.221
6	2	2	3	0.509
7	1	2	2	0.510
8	2	3	1	0.502
9	3	3	2	0.480
$p_{11}, p_{21}, p_{31}$	0.524	0.435	0.498	
$p_{12}, p_{22}, p_{32}$	0.538	0.511	0.531	
$p_{13}, p_{23}, p_{33}$	0.405	0.521	0.437	
Range $R_1, R_2, R_3$	0.133	0.086	0.094	
Order of the factor	A > C > B			
Optimum combination	A <sub>3</sub> B <sub>1</sub> C <sub>3</sub>			

It can be seen from Fig. 6 (a)-(h) that the surface topographies of the narrow deep grooves machined with different grinding parameters are different. Obvious wide scratch, ditch, and lateral uplifts caused by grinding plow can be seen in the SEM pictures. Some of the apparent burns can be also seen on some of the pictures, and abrasive dusts adhere to the ground surface seriously, as shown in Fig. 6 (e), which indicates that with the increase of grinding wheel speed, the temperature of the grinding zone increases, and the ground surface gets burned. Grinding process is inevitably accompanied with the wear of grinding wheel. The CBN grinding debris shedding can be seen clearly in Fig. 6 (g).

It is also learned from Fig. 6 that the surface quality of the left and right groove walls were slightly different, which was

**Fig. 5** Trend curve of the level on each factor of the right groove walls

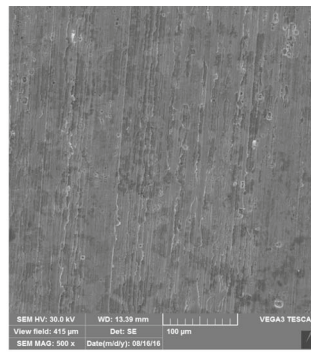
consistent with the change rule of the surface roughness  $R_a$  values in Table 4. Taken together, the surface quality of Groove V is the best. The optimum combination is that  $v_w$  is given by 1.5 mm/min,  $a_p$  is given by 0.5 mm, and  $n$  is given by 7000 rpm. In this grinding condition, the peripheral wheel speed reached at 65.94 m/s, which belongs to high-speed grinding range.

#### 4 Establishment of empirical prediction model of surface roughness

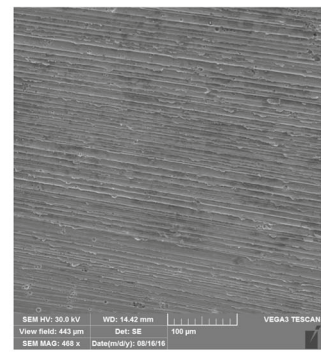
Similar to other machining processes, the surface roughness of grinding can be predicted theoretically by establishing the model of the interference between the grinding wheel and the workpiece.

Fig. 7 shows the ideal longitudinal profile of the surface formed by the abrasive cutting edge with different convex height, where the relevant parameters are labeled, and  $R_t$  represents the peak-valley roughness (also known as the overall roughness) [21]. If cutting depth  $a_p > R_t$ , then the cutting depth had no effect on the roughness. The arithmetic mean roughness  $R_a$  value is the mean deviation from the average roughness on the sampling length and is much smaller than  $R_t$ . Creep feed grinding in particular involves large depths of cut, ranging from 1 to 25 mm [21, 22]. In the process of narrow deep grooves of SUS321 stainless steel, this grinding technology was adopted. So the cutting depth  $a_p$  can be neglected when establishing the regression model of the surface roughness  $R_a$  value.

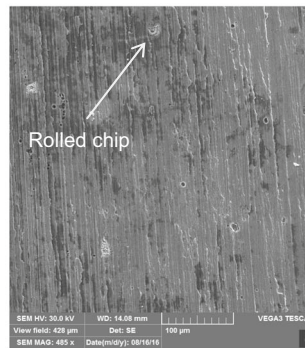
**Fig. 6** Surface topography of the narrow-deep-groove walls of SUS321 stainless steel



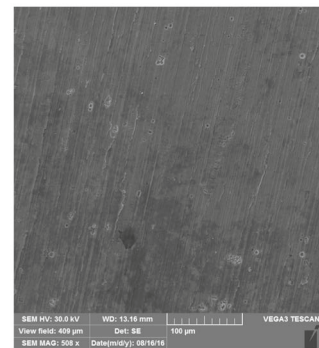
(a) the left groove wall of Groove I



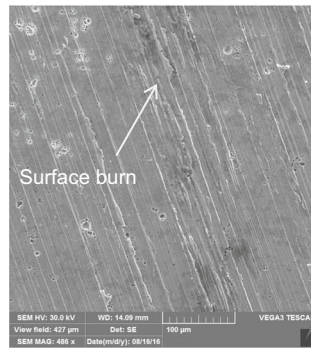
(b) the right groove wall of Groove I



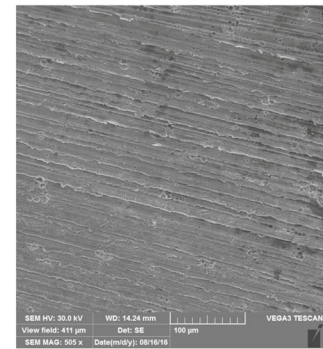
(c) the left groove wall of Groove V



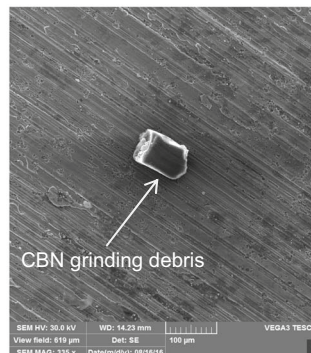
(d) the right groove wall of Groove V



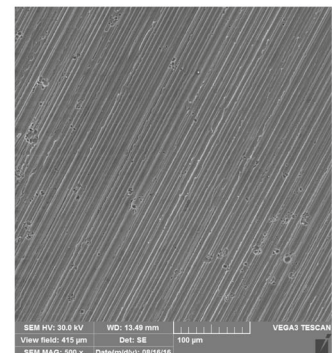
(e) the left groove wall of Groove VI



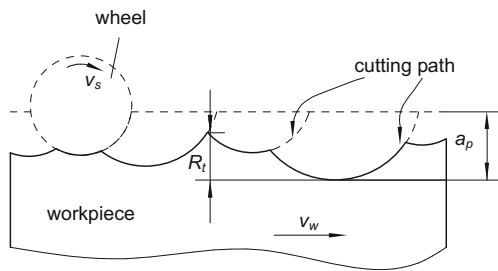
(f) the right groove wall of Groove VI



(g) the left groove wall of Groove IX



(h) the right groove wall of Groove IX



**Fig. 7** Longitudinal theoretical profile formed by non-uniform grinding wheel topography

**4.1 Establishment of the regression model**

Partial least squares is a method for constructing predictive models, and it has been accepted and successfully applied in the control of industrial processes [23]. In this study, this technique was proposed to establish the regression model of surface roughness  $R_a$  value. There are two abnormal large values in the surface roughness  $R_a$  values of the left groove walls due to workpiece burn, so the regression analysis was carried out with the surface roughness  $R_a$  values of the right groove walls as the object of investigation.

A typical regression equation is as shown:

$$\hat{y} = \beta_0 + \beta_1x_1 + \beta_2x_2 + \dots + \beta_px_p \tag{2}$$

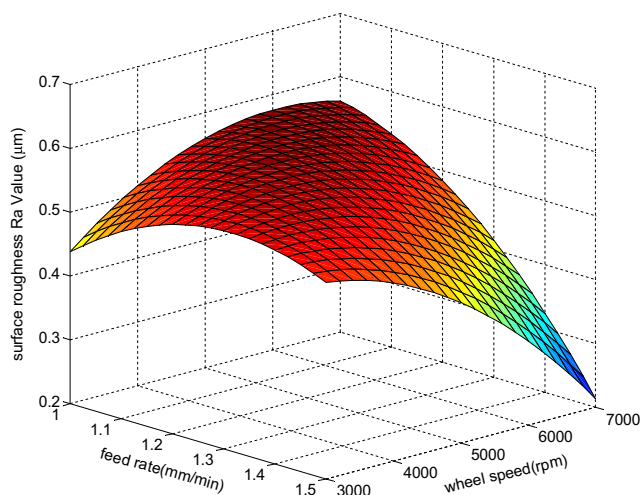
where  $\hat{y}$  is the fitted value and  $\beta_0, \beta_1, \dots,$  and  $\beta_p$  are the estimations of the regression parameters.

The real value for  $y$  is

$$y = \beta_0 + \beta_1x_1 + \beta_2x_2 + \dots + \beta_px_p + \varepsilon \tag{3}$$

where  $y$  is the response parameter,  $x_1, x_2, \dots, x_p$  are the experimental factors, and  $\varepsilon$  is the random error.

According to the regression model in the above, relationship between the surface roughness  $R_a$  value and the grinding



**Fig. 8** Trend of the surface roughness with the feed rate and the wheel speed

**Table 7** Results of the regression statistics

Item	Result
Multiple R	0.97357
R square	0.94656
Adjusted R square	0.86089
Standard error	0.04079
Number of points	9

parameters can be expressed by a quadratic polynomial model as is shown:

$$R_a = -2.2874 + 3.3509v_w + 0.00039v_s - 1.0322v_w^2 - 1.4867E-08v_s^2 - 0.00021v_wv_s \tag{4}$$

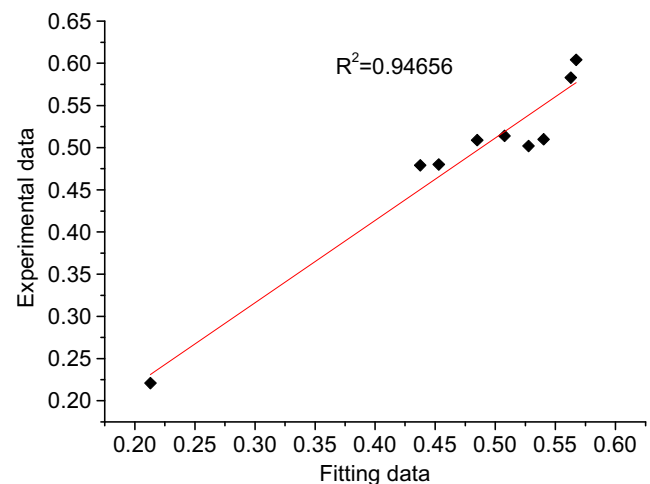
where  $v_w, v_s$  is the feed rate (mm/min) and  $v_s$  is wheel speed (rpm), equivalent to the wheel speed  $n$  in Section 2 and 3.

From the above equation (4), it is easy to see that the feed rate ( $v_w$ ) strongly affects surface roughness  $R_a$  value, followed by the wheel speed ( $v_s$ ), whereas the interaction of feed rate and wheel speed has a little influence on surface roughness  $R_a$  value. This result is consistent with that obtained by the range analysis above.

Fig. 8 shows the trend of the surface roughness with the feed rate and the wheel speed. This trend is consistent with the experimental data. It is also known from Fig. 8 that the interaction between the feed rate and the wheel speed has a little effect on the surface roughness  $R_a$  value.

**4.2 Test of regression model**

After the quadratic polynomial model of surface roughness  $R_a$  value is established, it is necessary to test the regression effect by mathematical statistics methods such as R-test and F-test.



**Fig. 9** Correlation between experimental data and the fitting data



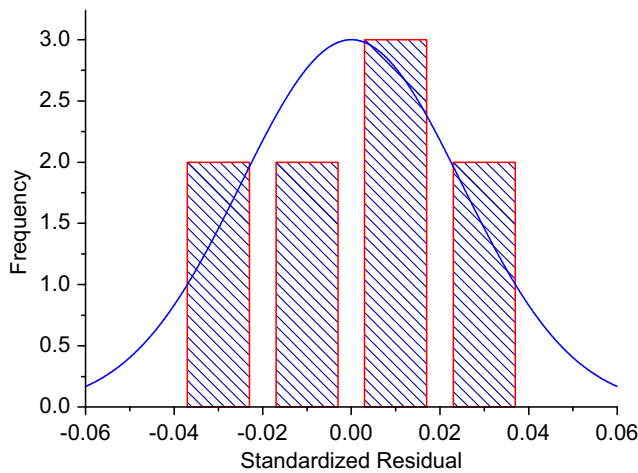


Fig. 10 Histogram of the residuals for the model

R-test is the goodness of fit test. The results of the regression statistics is shown in Table 7.

The coefficient of determination ( $R^2$ ) is used to test the goodness of fit of the regression equation and the sample value. It is defined as follows:

$$R^2 = 1 - \frac{\sum (y_i - \hat{y}_i)^2}{\sum (y_i - \bar{y})^2} \tag{5}$$

where  $y_i$  is the predicted quality variable,  $y_i$  is the measured quality variable, and  $\bar{y}$  is the mean.

The closer the coefficient of determination ( $R^2$ ) is to 1, the better the regression equation and the sample fit. Table 7 shows that the regression function fits well with the data for  $R^2 = 0.94656$ . Figure 9 shows the correlation between experimental data and the fitting data.

The goodness of fit test can only show the approximate degree of the model to the sample data. The linear relationship established between the explanatory variables and the explained variable in the model must be made a significant judgment in general by F-test, which is the variance analysis. The residual graphics is taken into account when analyzing the variance of the model, and it can be seen in Fig. 10. It is known from Fig. 10 that the distribution of the residues of the model follows a normal distribution approximately.

Table 8 shows results of variance analysis for the surface roughness  $R_a$  values of the narrow deep groove walls.

The quadratic polynomial regression model of surface roughness  $R_a$  value in the above was obtained at 95 %

confidence level and the significant level  $\alpha = 5 \%$ . It is known from the Significance  $F$  column of Table 8 that the  $P$  value = 0.038574 <  $\alpha$ . Therefore, the regression model is significant on the whole; thus, it can predict the roughness  $R_a$  values.

### 5 Conclusion

The creep feed grinding experiments designed by the orthogonal test method were carried out with a single-layer electroplated CBN wheel and SUS321 stainless steel workpiece. The narrow deep grooves were machined. The surface roughness/quality of each groove wall was analyzed. The influence of grinding parameters on the surface quality of the narrow deep grooves was explored, from which the prediction model of the surface roughness  $R_a$  value of the narrow deep groove walls of SUS321 stainless steel was established by partial least square regression.

The conclusions are summarized as follows:

- (1) The feed rate has the greatest influence on the surface roughness  $R_a$  values of groove walls, the wheel speed has less influence on that, and the cutting depth has the smallest influence.
- (2) The optimum combination is selected as  $A_3B_1C_3$ . The optimum parameters by orthogonal experiment optimum design are  $v_w = 1.5$  mm/min,  $a_p = 5$  mm, and  $n = 7000$  rpm.
- (3) It is desirable that the quadratic polynomial regression model of surface roughness  $R_a$  value was established by partial least square regression. The regression function fits well with the data for  $R^2 = 0.94656$ , and the model is significant on the whole with the data for  $P$  value = 0.038574 at 95 % confidence level.
- (4) In order to prevent the grinding burn caused by the increase of grinding wheel speed, the problem of heat transfer in creep feed grinding should be considered in future grinding experiments.

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Table 8 Variance analysis for the surface roughness  $R_a$  values

	<i>df</i>	SS	MS	<i>F</i>	Significance <i>F</i>
Regression analysis	5	0.090708	0.018142	10.90142	0.038574
Residual	3	0.004992	0.001664		
Total	8	0.095701			

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