

A New Detection Method for Oxygen, Nitrogen and Hydrogen on Superalloy Milling Surfaces

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Abstract: Oxygen, nitrogen and hydrogen from air greatly affect the surface quality of milled superalloys—a difficult-to-cut material. Here, we described a novel infrared-thermal conductivity method to measure oxygen, nitrogen and hydrogen on superalloy surfaces during milling. A milling experimental program was projected via the uniform design method—this contained a qualitative factor. The quadratic regression model of the superalloy was established using the best regression subset method. Here, the cutting speed, feed per tooth, axial cutting depth, radial cutting depth, and tool nose radius were the independent variables. The results were feasible with an acceptable regression effect.

Keywords: Superalloy, Oxygen, Nitrogen, Hydrogen, Detection Method

1. Introduction

The aero-engine is a core part of an aircraft. It is often called the pearl on the crown of aviation. Superalloys are generally used in the manufacture of aero-engine components, such as turbine blades, combustors, and high-pressure turbines. They have excellent high-temperature strength, thermal stability, and resistance to corrosion under severe service environments [1].

Superalloys have very strict quality requirements during production. The blank superalloy is obtained by a near-net forming method, such as precision forging [2], integral precision casting [3], single crystal casting [4], isothermal forging [5], hot isostatic pressing [6], or powder metallurgy [7]. These are then finished via precision machining, such as by computerized numerical control (CNC) milling. During the superalloy precision milling process, large quantities of heat are generated due to the drastic plastic deformation and severe friction. The low thermal conductivity characteristics of the superalloys lead to a high cutting temperature (1000°C) and a narrow cutting range [8]. Thus, oxygen, nitrogen and hydrogen in the air intrude on the cutting interface at high

temperatures. They will react with the titanium, chromium, and other elements in the material to form compounds with low melting points or eutectic points (hereafter referred to as inclusions).

These inclusions degrade the mechanical properties at high temperature [9-10] in two ways. First, small, hard, and brittle inclusions with sharp edges will produce stress concentrations after long-term use—this initiates cracks [11]. Second, oxygen, nitrogen and hydrogen may gather on the surface to react with titanium and chromium, leading to the segregation of alloying elements on the freshly machined surface. This weakens the surface strength and initiates cracks on the surface [12]. Therefore, oxygen, nitrogen and hydrogen are considered harmful, and their elemental contents are strictly controlled. For instance, the nitrogen content is held below 0.2% [13-14] to enhance fatigue and creep failure resistance as well as durability during the smelting and forging processes. Nevertheless, only a few studies have focused on methods to decrease oxygen, nitrogen and hydrogen contents to control the surface quality during cutting. No papers have measured oxygen, nitrogen and hydrogen effectively.

In this study, CNC milling experiments were designed via a

uniform design method. The surface oxygen, nitrogen and hydrogen contents were measured based on the infrared-thermal conductivity method. Statistical models were used due to difficulties in building a physical model of surface oxygen, nitrogen and hydrogen contents [15]. A quadratic model was proposed to fit the experimental data. The veracity of the measurement and the feasibility of the detection method were verified via regression analysis.

2. Materials and Method

2.1. Experimental Design

There are six cutting parameters in the design of experiments: cutting speed, v_c (30–60 m/min, arithmetically divided into six levels); feed per tooth, f_z (0.1–0.2 mm/z, arithmetically divided into six levels); radial cutting depth, a_e

(2–12 mm, arithmetically divided into six levels); axial cutting depth, a_p (0.5–1.0 mm, arithmetically divided into six levels); tool nose radius, r_e (0.8 mm, 1.2 mm, and 1.6 mm); and the material of the cutting blade A (pseudo-variable design factor). The WIDIA and Zhuzhou cutting tools are represented as A_1 ($A_1 = 1, A_2 = 0$) and A_2 ($A_1 = 0, A_2 = 1$), respectively. Thus, all parameters must be optimized, but the total number of experiments becomes very large when all six parameters are considered. The uniform design method is a good method to reduce the number of experiments—it is considered a qualitative factor to design milling tests [16]. Following repeated optimization computations using the Uniform Design Software 5.0, an experimental program with minimum central deviation containing a qualitative factor is projected (Table 1). Terms ω_O , ω_N and ω_H are the surface oxygen, nitrogen and hydrogen contents, respectively.

Table 1. Uniform design test and results of the superalloy in milling.

NO.	v_c [m/min]	f_z [mm/z]	a_e [mm]	a_p [mm]	r_e [mm]	A_1	A_2	ω_O [%]	ω_N [%]	ω_H [%]
1	48	0.10	8	0.8	1.2	1	0	0.0027	0.0674	0.0027
2	30	0.14	10	1.0	1.6	1	0	0.0072	0.0802	0.0042
3	36	0.18	10	0.5	0.8	1	0	0.0210	0.0665	0.0108
4	60	0.20	12	0.8	1.6	0	1	0.0044	0.0207	0.0015
5	36	0.20	6	0.7	0.8	0	1	0.0116	0.0525	0.0061
6	42	0.16	2	0.9	1.2	0	1	0.0161	0.1020	0.0025
7	42	0.16	4	0.7	1.6	1	0	0.0179	0.0995	0.0064
8	30	0.10	6	0.6	1.2	0	1	0.0231	0.1350	0.0057
9	60	0.12	2	0.6	0.8	1	0	0.0107	0.2430	0.0051
10	54	0.14	8	0.5	1.6	0	1	0.0128	0.0239	0.0062
11	48	0.12	12	0.9	0.8	0	1	0.0140	0.0646	0.0035
12	54	0.18	4	1.0	1.2	1	0	0.0090	0.0441	0.0034

2.2. Experimental Procedure

This study used a NiCr16TiAl (Wuxi Chuantie Metal Material Co. Ltd., China) rod (Ni=71.35%, Cr=16.78%, Ti=1.96%, Al=1.54%, and Mn=1.51%; remaining balance contains Fe and other elements; Figure 1). The milling tests were performed along the top surface of the specimen via the dry down milling process—each specimen was used for two experiments. The samples' top surface was premachined with a larger axial depth of cut. This prevented the machined surface from scratching. The plunging cutter XDPT110408 (12/16) PDSRMM (German WIDIA Company) and APKT11T308 (12/16)-PM (Zhuzhou Cemented Carbide Group Co. Ltd., China) with uncoated carbide inserts were used in these experiments (Figure 2). The diameter of the plunging center is 16 mm with two teeth. Only one insert was used to avoid interference of the inserts. A new insert was used in each cut to avoid tool wear affecting surface integrity. The VDL-850D Vertical Machining Center (Dalian Machine Tool Group, China) was used for milling tests.

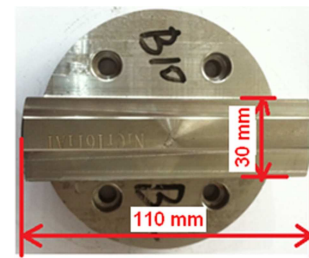


Figure 1. Sample for the milling test (color online only).



Figure 2. Toolbar and movable milling blades used for milling test: (a) toolbar and movable milling blades made in Zhuzhou and (b) toolbar and movable milling blades made by German WIDIA (color online only).

2.3. Data Measurement Procedure

The infrared-thermal conductivity method detected oxygen, nitrogen and hydrogen. It has good sensitivity and stability. It is traceable and easily automated, which makes it practical for high throughput inspection. Here, a sample is placed in a graphite crucible and then melted in an electrode oven under helium. Oxygen from the sample is reduced to carbon dioxide with carbon and is detected by infrared absorption method. Nitrogen from the sample is decomposed into the gaseous state and is detected via a thermal conductivity method. Hydrogen is also resolved into the gaseous state. It is then reduced to water vapor with hot copper oxide and detected via an infrared absorption method.

However, the infrared-thermal conductivity method has strict demands, and the samples are generally processed to a regular 1-gram cylinder. After milling, the superalloy is finished into a regular detection sample. This requires a working procedure of line cutting, polishing, and cleaning. Thus, the preparatory detection is challenging. Otherwise, the oxygen, nitrogen and hydrogen coming from the air are concentrated on the newly finished superalloy surface. The distribution of these elements in the material is not uniform, and this leads to severe distortion in the results. The new surface is simultaneously formed with a new chip surface under milling. The thickness of the chips is relatively small because the f_z and a_p are small; thus, the oxygen, nitrogen and hydrogen contents of the chip surfaces are theoretically the

same as the newly finished surface. Thus, the oxygen, nitrogen and hydrogen contents of the superalloy's newly machined surface are indirectly detected via the oxygen, nitrogen and hydrogen chip contents.

After each test, the chips were immediately collected and sealed in three bags. We confirmed that the weight of the chips in each bag was more than one gram. The surface oxygen, nitrogen and hydrogen content measurements were performed in triplicate using the ONH836 oxygen/nitrogen/hydrogen Elemental Analyzer (America Leco Company). The average value of the three measurements was used. All data are listed in Table 1. To ensure accuracy, the ONH836 elemental analyzer was calibrated with a NiCr16TiAl sample containing oxygen, nitrogen and hydrogen; the specimen was supplied by AVIC Beijing Institute of Aeronautical Materials.

3. Results and Discussions

3.1. Quadratic Fit Model

Based on the experimental data, the unknown regression coefficients were determined using Uniform Design Software 5.0, and the empirical equations were used to estimate the ω_O , ω_N , and ω_H considering the following input process parameters: v_c , f_z , a_e , a_p , r_e , and A . The quadratic polynomial equations for the ω_O , ω_N , and ω_H response variables are expressed as follows:

$$\begin{cases} \omega_O = 0.0072 - 0.0194(a_p - 0.75) - 0.0019A_1 + 0.0001(v_c - 45)^2 + 0.0001A_1(v_c - 45) \\ + 0.0543A_1(f_z - 0.15) + 0.0002(a_e - 7)^2 - 0.0056(a_e - 7)(r_e - 1.19) - 0.0013A_2(a_e - 7) \\ + 0.0061A_1(r_e - 1.19) \\ \omega_N = 0.0729 - 0.0134A_1 - 0.0184(v_c - 45)(r_e - 1.19) + 14.8150(f_z - 0.15)^2 \\ + 5.0021(f_z - 0.15)(a_p - 0.75) + 0.2738A_2(f_z - 0.15) - 0.0161(a_e - 7)(a_p - 0.75) \\ - 0.3067A_1(a_p - 0.75) \\ \omega_H = 0.0039 + 0.0014A_1 - 0.0047(a_p - 0.75) - 0.1099(f_z - 0.15)(r_e - 1.19) \\ + 0.0453A_1(f_z - 0.15) - 0.0166A_2(f_z - 0.15) + 0.0007(a_e - 7)(a_p - 0.75) \\ + 0.0001A_1(a_e - 7) - 0.0056A_1(a_p - 0.75) + 0.0038(r_e - 1.19)^2 \end{cases} \quad (1)$$

The P values and fit error of the quadratic fit models are listed in Table 2. The P value is the probability of obtaining a test statistic that is at least as extreme as the actual calculated value. Smaller P values indicate a better correlation. Items with P values lower than 0.05 are closely related. The fit model is updated according to the P value, which represents the statistical significance. If the P value of the item in the quadratic model is higher than 0.05, then this item is removed from the quadratic model, and the fit process was performed again. Finally, quadratic fit models were obtained as Eq. 1. The fit precision of the quadratic fit model was higher indicating that the quadratic regression model was credible (Table 2).

3.2. Fit Precision and Significance

The F-test determines the critical value if a significance level α is given. The significance of the fit models or the

significance of the regression coefficients can be judged by comparing the F value with the critical value. Here, a significance level of $\alpha=0.03$ was considered to be the critical value of $F_{0.03}(10,1)=672.55$. This F-test was followed by analysis of variance (ANOVA) to validate the model precision of ω_O , ω_N , and ω_H (Table 3). Table 3 exhibited a larger F value for all the fit models compared to the critical value. This indicates that the regression effect of the models was significant. The R^2 determines the fit precision and higher R^2 values imply higher fit precision. All models in Table 3 had an R^2 value varying from 0.999 to 1.000. This implied that the input variables were closely related to the response variables.

The F values of the mono factor and their interactions are listed in Table 4; the multiplication sign represents the interactions of the cutting parameters. The significance of the regression coefficients was distinct according to the P value in

Table 2 and the F value in Table 4. If the P value is between 0.01 and 0.05, then the significance is considered to be strong. Values below 0.01 are very significant. Table 5 shows the

factors that most strongly influence the surface oxygen, nitrogen and hydrogen contents as judged by the P and F values.

Table 2. P value of Eq. 1 in the quadratic fit model.

Item	P value									
	Constant	a_p	A_1	v_e^2	f_z^2	a_e^2	r_e^2	$v_e \times r_e$	$v_e \times A_1$	$f_z \times a_p$
ω_O	0.000	0.000	0.000	0.000	/	0.000	/	/	0.010	/
ω_N	0.003	/	0.009	/	0.007	/	/	0.003	/	0.000
ω_H	0.000	0.004	0.000	/	/	/	0.010	/	/	/

Item	$f_z \times r_e$	$f_z \times A_1$	$f_z \times A_2$	$a_e \times a_p$	$a_e \times r_e$	$a_e \times A_1$	$a_e \times A_2$	$a_p \times A_1$	$r_e \times A_1$	Fit model
ω_O	/	0.000	/	/	0.000	/	0.000	/	0.004	0.000
ω_N	/	/	0.006	0.019	/	/	/	0.000	/	0.000
ω_H	0.000	0.003	0.005	0.010	/	0.011	/	0.003	/	0.000

Table 3. ANOVA results.

Item	Source	Regression	Residual	Total
ω_O	Sum of squares	0.0004	0	0.0004
	Mean square	0	0	/
	F value	100,000,000.00	/	/
	Multiple regression correlation coefficients R^2	0.9999		
ω_N	Sum of squares	0.0396	0	0.0396
	Mean square	0.0040	0	/
	F value	162,440.50	/	/
	Multiple regression correlation coefficients R^2	0.9998		
ω_H	Sum of squares	0.0001	0	0.0001
	Mean square	0	0	/
	F value	100,000,000.00	/	/
	Multiple regression correlation coefficients R^2	0.9997		

Table 4. Factors affecting the surface oxygen, nitrogen and hydrogen contents of the superalloy.

Influencing factor	F value			Critical value $F_{0.05}(10,1)$	Significance		
	ω_O	ω_N	ω_H		ω_O	ω_N	ω_H
Constant	805,816.50	44,028.61	228,312.40	672.55	Highly significant	Highly significant	Highly significant
a_p	424,203.26	/	23,861.53	672.55	Highly significant	/	Highly significant
A_1	105,040.34	5,336.01	117,759.24	672.55	Highly significant	Highly Significant	Highly significant
v_e^2	2,231,006.63	/	/	672.55	Highly significant	/	/
f_z^2	/	8,065.04	/	672.55	/	Highly Significant	/
a_e^2	177,145.85	/	/	672.55	Highly significant	/	/
r_e^2	/	/	3,744.53	672.55	/	/	Significant
$v_e \times r_e$	/	53,962.01	/	672.55	/	Highly significant	/
$v_e \times A_1$	4,360.92	/	/	672.55	Significant	/	/
$f_z \times a_p$	/	163,474.43	/	672.55	/	Highly significant	/
$f_z \times r_e$	/	/	175,645.23	672.55	/	/	Highly significant
$f_z \times A_1$	146,607.86	/	61,400.23	672.55	Highly significant	/	Highly significant
$f_z \times A_2$	/	10,667.66	18,308.49	672.55	/	Highly significant	Highly significant
$a_e \times a_p$	/	1,121.46	4,413.25	672.55	/	Significant	Significant
$a_e \times r_e$	4,542,450.99	/	/	672.55	Highly significant	/	/
$a_e \times A_1$	/	/	4,413.25	672.55	/	/	Significant
$a_e \times A_2$	2,847,675.24	/	/	672.55	Highly significant	/	/
$a_p \times A_1$	/	464,330.98	44,041.76	672.55	/	Highly significant	Highly significant
$r_e \times A_1$	30,842.30	/	/	672.55	Highly significant	/	/

Table 5. The impact of different factors on the surface oxygen, nitrogen and hydrogen contents in the superalloy.

Item	Ranking
ω_O	$a_e \times r_e > a_e \times A_2 > v_e^2 > a_p > a_e^2 > f_z \times A_1 > A_1 > r_e \times A_1 > v_e \times A_1$
ω_N	$a_p \times A_1 > f_z \times a_p > v_e \times r_e > f_z \times A_2 > f_z^2 > A_1 > a_e \times a_p$
ω_H	$f_z \times r_e > A_1 > f_z \times A_1 > a_p \times A_1 > a_p > f_z \times A_2 > a_e \times a_p > r_e^2 > a_e \times A_1$

3.3. Analysis of Model Error

The oxygen, nitrogen and hydrogen contents in the chip and machined surface were most strongly related to cutting temperature. The highest temperature was located at the root of the cutting edge contacting with the chip, and the other area of the chip had a higher temperature due to its slow cooling rate, which made the temperature of the chip much higher than the machined surface. Thus, the detection method for the oxygen, nitrogen and hydrogen had some errors. In addition, a cutting tool with a large tool nose radius would increase the temperature, leading to a increase of the solubility of the oxygen, nitrogen and hydrogen theoretically, but the large tool nose was poorly coordinated with other cutting parameters, and thus lowered the impact on the results to some extent. Although there were some errors in experimental results, the fit precision and significance of the models were high. Therefore, the measurement method was feasible and could be used to guide the practical production of superalloys.

4. Conclusions

This subject of this study consisted in the new detection method for oxygen, nitrogen and hydrogen on superalloys milling surface, and in a careful analysis of results by statistical methods. The following items may be emphasized.

- (1) The oxygen, nitrogen and hydrogen contents of the machined surface were indirectly detected as oxygen, nitrogen and hydrogen in the chip. Although this detection method had theoretical error, it was suitable for rapid and practical testing.
- (2) In this study, a uniform test containing a qualitative factor was designed and conducted. A mathematical model for surface oxygen, nitrogen and hydrogen content was developed via an optimal subset regression method. This made the regression effect significant and implied that the experimental results were precise and credible.
- (3) ANOVA data showed that the interaction between radial cutting depth and tool nose radius had the most significant impact on the surface oxygen content. Similarly, the axial cutting depth using the WIDIA milling tooth had the most significant effect on surface nitrogen, and the interaction between the feed per tooth and the tool nose radius had the strongest influence on the surface hydrogen content.

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