

# Optimization of Surface Roughness and Hole Diameter Accuracy for Drilling of EN- 31 Based on GRA

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Abstract: In the present work drilling parameters has been optimized for EN-31 alloy steel using GRA (Grey Relational Analysis). The parameters optimized are spindle speed (SS), feed rate (FR) and point angle (PA) on bases of surface roughness and hole diameter accuracy responses. Drilling is performed with the help of cemented carbide twist drills. On the bases of GRA along with identification, significant contribution of parameters has been carried out by using ANOVA. The optimal values obtained are PA ( $118^0$ ), SS (800 rpm), and FR (0.18 mm/rev). Out of three variables considered point angle has significant effect on responses as compare to other drilling parameters.

Keywords: Drilling. EN-31 alloy steel; Orthogonal array; Grey relational analysis; ANOVA; Optimization

# I. INTRODUCTION

With the concern of quality of drilled products, numbers of the problems arise, such as hole surface roughness (SR), hole diametrical accuracy (HDA), burr height (BH) and tool wears (TW). SR and HDA have most effect on performance of a drilled products. These important characteristics depends on the drilling parameters for a specific combination of material and drill tool. Some of the parameters have been optimized by different researchers for different materials and drilling tool combinations. An experiment and numerical study for cutting forces (CF), TW and SR has been done by Davim et al. [1] for drilling of composite A356/20/SiCp-T6. Kurt et al. [2] and Kilickap [3] made use of Taguchi for optimizing SR and HDA in the dry drilling of Al 2024 and Al 7075 respectively. Genetic algorithm (GA) has also been used for optimization of multi-objective drilling by Gaitonde et al. [4], also Karnik et al. [5] developed an artificial neural network (ANN) model for high speed drilling by considering SS, FR and PA as parameters. Tosun [6] and Rajmohan et al. [7] optimized drill parameters (FR, CS, PA) by using Taguchi based GRA for SR and burr height for composites. Mathematical model has been developed by Pirtini and Lazoglu [8] and Furness et al. [9] for drilling process to estimate of CF and hole quality. Effect of coating on drilled cast Al 356 alloy has been investigated by Kalidas et al. [10] for both dry and wet conditions. Further, Nouari et al. [11] investigated the effect of drill parameters on SR and holes dimensions. The study of many researchers in this direction has led to the present work. In this presented work Taguchi based GRA has been applied for the optimization of drilling parameters (PA, SS and F). The responses consider for optimizing are SR and HAD. The most effective parameter has been identified along with the range and confirmative test has been conducted.

#### II. EXPERIMENTAL PROCEDURE

Experimentation has been conducted on the bases of Taguchi design for GRA. The correlation has been done between variables and responses.

## A. Description of Experimental set up And Measurements

The material selected for study is EN-31 alloy and chemical composition is given in table 1. The drilling tests are carried out on KMC-11VC CNC VERTICAL MACHINE CENTRE, with FANUC OiMD-PB Controller, made by KENT Industry. The material has been prepared by cutting the plates of size 200×75×16 mm and facing has been performed on CNC machine with face milling cutter to obtain flat surface and reduce the thickness up to 14 mm. This plate is then mounted rigidly on the table and holes have been drilled in the plate. The surface roughness (SR) and hole diameter accuracy (HAD) are responses considered for study. Measurement of SR has been taken in surface roughness tester (Mitutoyo Surf test 4) and have been repeated three times. The diameters of holes created are measured on co-ordinate measuring machine (CMM) having accuracy of 0.1 µm



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Table 1: Chemical	composition	of EN 31	alloy steel
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SYMBOL	CHEMICAL COMPOSITION (Wt %)							
EN 31	С	Cr	Si	Mn	Р	S		
	0.95-1.05	1.30-1.65	0.15-0.35	0.25-0.45	< 0.027	< 0.025		

## B. Cutting Tool

In this study, drilling operations were performed using three cemented Carbide twist drill of 10.08 mm diameter with two flutes and three different point angles (118<sup>0</sup>, 127<sup>0</sup>, 135<sup>0</sup>) shown in Figure 1.



Figure 1: Drill bits having different point angles

## C. Plan of Investigation

Experiments have been performed as per L27 OA by considering the levels of parameters given in Table 2. Table 3 shows the results obtained for the responses.

Table 2: Factors and levels of independent variables					
Factors	Unit	Levels			
		1	2	3	
Point angle (PA)	Degree	118	127	135	
Spindle speed (SS)	Rpm	800	1200	1600	
Feed (F)	mm/rev.	0.10	0.14	0.18	



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Experiment no.	Point angle	Spindle Speed	Feed	Surface roughness value/ μm	Diametrical error/ μm
1	135	1600	0.1	1.12	20.3
2	135	1600	0.14	2.67	9.5
3	135	1600	0.18	2.74	8.5
4	135	1200	0.1	2.06	10.6
5	135	1200	0.14	2.39	13.4
6	135	1200	0.18	2.42	8.9
7	135	800	0.1	3.37	21.2
8	135	800	0.14	3.7	7.5
9	135	800	0.18	2.2	4
10	118	1600	0.1	3.09	15 3
11	118	1600	0.14	2.67	8.2
12	118	1600	0.18	3.33	10
13	118	1200	0.1	3.11	14.2
14	118	1200	0.14	2.91	11.3
15	118	1200	0.18	3.03	24.1
16	118	800	0.1	3.94	19.1
17	118	800	0.14	4.55	25.5
18	118	800	0.18	5.54	20.5
19	127	1600	0.1	1.17	-3.8
20	127	1600	0.14	1.91	5.5
21	127	1600	0.18	2.69	8
22	127	1200	0.1	2.36	-6.2
23	127	1200	0.14	2.43	4.5
24	127	1200	0.18	2.87	10.6
25	127	800	0.1	4.23	7.5
26	127	800	0.14	2.67	9.2
27	127	800	0.18	4.45	15.5

#### Table 3: Experimental results for Surface roughness and diametrical error

# III. OPTIMIZATION STEPS USING GREY RELATIONAL ANALYSIS

The smaller-the-better methodology for GRA has been implemented for the considered responses of drilling process. Accordingly, S/N ratios and respective normalized values has been calculated and further deviation sequences have been determined to calculate the grey relational co-efficient. Table 4 represents the Grey relational grade with rank and Grey relational co-efficient for each experiment.



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	Grey relationa	l coefficient		
Experiment			Grey Relational	
no.	Surface Roughness	Diametrical error	Grade	Rank
1	0.333333333	0.806713883	0.570023608	10
2	0.522362017	0.490834495	0.506598256	16
3	0.53133796	0.464209436	0.497773698	18
4	0.446727379	0.520226194	0.483476786	21
5	0.487134312	0.596665897	0.541900104	14
6	0.49086168	0.474859017	0.482860349	22
7	0.615982158	0.837505784	0.726743971	5
8	0.663709713	0.437503453	0.550606583	12
9	0.463749494	0.339430483	0.401589989	26
10	0.577427575	0.650756761	0.614092168	7
11	0.522362017	0.456214749	0.489288383	19
12	0.61037212	0.504169995	0.557271057	11
13	0.580128261	0.619172615	0.599650438	8
14	0.553458735	0.53906718	0.546262958	13
15	0.569371571	0.943998498	0.756685034	3
16	0.700208551	0.767099124	0.733653838	4
17	0.801111585	1	0.900555793	2
18	0.997751445	0.813473199	0.905612322	1
19	0.339509904	0.333333333	0.336421619	27
20	0.428636809	0.382915175	0.405775992	25
21	0.524919094	0.450878543	0.487898819	20
22	0.483417104	0.402304153	0.442860629	23
23	0.492106452	0.354312317	0.423209385	24
24	0.548210846	0.520226194	0.53421852	15
25	0.746605531	0.437503453	0.592054492	9
26	0.522362017	0.482844872	0.502603445	17
27	0.783676744	0.656586683	0.720131714	6

Table 4: Grey relational co-efficient and grade values





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From Table 4 and Figure 2, it is observed that experiment 18 has best results for multi performance characteristics. Further, mean of grey relational grade for all levels has been summarized as shown in Table 5. It also shows the response table for average grey relational grade by factor level. Bold numerals represent optimized value of factors with respect to their corresponding levels. So, the optimized level of parameters for good SR and HDA is  $(PA1=118^{\circ}, SS1=800 \text{ rpm}, \text{ and } F3=0.18 \text{ mm/rev})$  as shown in Table 5.

Table 5: Response table for grey relational grade; Main effects on Grey grade						
		Level				
	1	2	3	maxmin ( $\Delta$ )	Rank	
Point angle	0.678119	0.493908	0.529064	0.18421082	1	
Spindle speed	0.670395	0.534569	0.496127	0.174267616	2	
Feed	0.566553	0.540756	0.593782	0.053026734	3	

Further, Table 5, shows difference of maximum and minimum grey relational grade i.e. 0.18 (point angle), 0.17 (spindle speed), and 0.05 (feed). It is observed that point angle is most effective and spindle speed and feed follows it.



(c). Point angle Vs Grey grade Figure 3 (a, b, c): Effect of drilling parameters on the multi-performance characteristics

ANOVA has been formulated as per the grey grade values obtained in previous steps for finding the substantial factors. From Table 6 it is clear that PA (32.69%) effects more on drilling of EN 31 alloy steel then by SS (28.65%) and FR (2.40%). The interaction between PA x SS and PA×FR had lesser effect on multiple performance characteristics (minimum surface roughness and minimum hole diametrical error) i.e. 4.2% and 5.4%. Further, interaction of spindle speed and feed rate (S×F) has no effect on multiple performance characteristics.

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					Percentage	
Factors	DOF	Sum of square	Mean Square	F Ratio	contribution	
Point angle(PA)	2	0.172161129	0.086080565	5.004487596	0.326927493	
Spindle speed(S)	2	0.150886593	0.075443297	4.386066058	0.286527952	
Feed(F)	2	0.012656331	0.006328165	0.367902155	0.024033895	
PA×S	4	0.022019352	0.005504838	0.320036162	0.041813919	
S×F	4	0.002661959	0.00066549	0.038689746	0.00505496	
PA×F	4	0.028612635	0.007153159	0.415864996	0.054334314	
Error	8	0.1376054	0.017200675		0.261307466	
Total	26	0.526603399			1	

The predicted value of GRG at the optimal level as per standard calculation [12] is obtained as **0.8082**. The 95% confidence interval for obtained for present study is between 0.654 and 0.962. In last confirmation test has been conducted by setting drilling parameters and two trials have been conducted. The values corresponding to predicted and confirmation test of surface roughness, hole diametrical error and grey relational grade have been given in Table 7.

Table 7: Optimal values of machining and response parameters					
	Optimal dril	ling parameters			Confidence Interval
			Final	%	
	Prediction	<b>Confirmation Test</b>	gain	improvement	Range
Setting level	PA1, SS1, F3	PA1, SS1, F3			
Grey relational					
grade	0.8082	0.8321	0.0239	2.9 %	$0.654 \le \mu \le 0.962$

The value of grey relational grade has increased from 0.8080 to 0.8321, i.e. 3% of improved grade has been archived. Thus, confirmation tests expose that GRA for optimization is efficient for drilling parameters of EN31.

## IV. CONCLUSIONS

Drilling experiments have been performed on a CNC vertical machining centre using carbide twist drill on EN 31 alloy steel as work material.  $L_{27}$  orthogonal array was used for different combinations of drilling experiments. The surface roughness and hole diametrical error have been selected as responses for different combinations of drilling parameters. Taguchi based Grey relational analysis optimization technique has been used for multi response optimization. The recommended level of parameters for better SR and HDA simultaneously are PA1 (118<sup>0</sup>), SS1 (800 rpm), and F3 (0.18 mm/rev). Out of three parameters considered, PA has most effect on responses as compare to other considered drilling parameters. Order of importance of factor is A (point angle), B (spindle speed), C (feed rate). Main contribution percentages for multiple performance characteristics in drilling EN31 steel alloy are 32.7% (PA), 28.6% (SS) and 2.4% (F). Predicted weighted GRG has increased from 0.8080 to 0.8321, which confirms the enhancement in the performance of drilling EN31 with optimal values of process parameters.

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