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Assessment of cutting force and surface roughness in LM6/SiC_p using response surface methodology

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Abstract

Metal matrix composites are replacing the traditional material world because of their superior mechanical properties. Consequently, the necessity for accurate machining of the composite has increased drastically. The major problems while machining metal matrix composites are surface roughness and cutting force. The present work focuses on the study of the cutting conditions which influence the cutting force and surface roughness in terms of spindle speed, feed rate, axial, radial depth of cut and weight percentage of silicon carbide particle (SiC_p). Central composite rotatable second order response surface methodology has been employed to create the mathematical model. The adequacy of the model has been verified using analysis of variance. The direct and interaction effects of the process parameters have been studied to keep the cutting force and surface roughness minimal.

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Keywords: Cutting force; Surface roughness; Metal matrix composites; End milling; Response surface methodology

1. Introduction

Milling is a metal removal process of feeding the work against a rotating multipoint cutter. The ratio of the metal removal rate is rapid as the cutter rotates at a high speed and has many cutting edges. It is employed to produce slots, pockets, precision molds and dies, automotive and aerospace components. Cutting force cause deflections of the part or machine structure and supply energy to the machining system which results in excessive cutting temperature or unstable vibrations. The reason for analyzing cutting force of the machine tool is to estimate tool forces that must be resisted by machine tool components, bearing loads, jigs and fixtures and power requirement. To satisfy the demands on product capabilities and functions, natural materials

are insufficient and hence metal matrix composites (MMC) are considered. MMC consists of at least two components, of which one is a metal matrix and the other is a reinforcement. The metal matrix is generally an alloy. MMC has been studied mainly for the aerospace industry and space component due to a higher specific modulus, higher specific strength, lower coefficient of thermal expansion, better wear resistance, better properties at elevated temperature. Recently, electronic and automotive industries have been concentrating the composites (Rosso, 2006).

The cutting force model in turning of LM6/SiC_p metal matrix composite has been proposed by Joardar, Das, Sutradhar, and Singh (2014).

The cutting speed, depth of cut and weight percentage of silicon carbide were specified as machining parameters. Sequential approach in face central composite design saves the number of experimentations needed. The author pointed that tangential cutting force and radial cutting force are more sensitive to cutting speed. Jeyakumar, Marimuthu, and Ramachandran (2013), carried out the mathematical model of cutting force, tool wear

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and surface roughness during end milling of Al6061/SiC_p composite under dry condition. The cutting force component in the z-direction was at significantly higher magnitudes than that of the x-direction. The results indicated that the depth of cut was the dominant factor affecting tool wear, cutting force, and surface roughness. Palanikumar and Muniaraj (2014) employed the response surface method to analyze the thrust force in the drilling of hybrid metal matrix composites by coated carbide drill. The burr formed because of thrust reduced the hole quality. They demonstrated that the thrust force and burr formation were most influenced by the feed rate. Kalla, Sheikh-Ahmad, and Twomey (2010) put forward a mechanistic modeling technique to predict the cutting force by transforming specific cutting energies from orthogonal to oblique cutting in helical end milling. They concluded that proposed model was reasonable for unidirectional but less desirable for multi-directional composites.

Pramanik, Zhang, and Arsecularatne (2006) built up a systematic model for predicting the forces using Merchant's analysis, slip line field theory of plasticity and the Griffith theory of fracture. They stated that the force during the chip formation is higher than the ploughing and particle fracture. Sivasakthivel, Vel Murugan, and Sudhakaran (2010) brought forward a central composite rotatable second order response surface methodology to develop a mathematical model to predict cutting forces in terms of helix angle, axial depth of cut, radial depth of cut, feed rate and spindle speed of Al6063 of high speed steel end mill cutter. The experimental results showed that increase in feed rate and axial depth of cut reduces the infeed and crossfeed force. Babu, Selladurai, and Shanmugam (2008) established the effects of cutting parameters on the variations of cutting forces of Al SiC_p metal matrix composite material. The author concluded that cutting forces were sensitive in the high speed and full immersion condition. They stated that with the increase in the depth of cut, cutting force in tangential direction increases.

Seeman, Ganesan, Karthikeyan, and Velayudham (2010) studied the tool wear and surface roughness during the machining of Al metal matrix composite. The result showed that the tool wear was affected by BUE formation at low speed and better surface finish was achieved at low feed rate with high speed. Anandakrishnan and Mahamani (2011) experimentally investigated on in situ MMC using uncoated tungsten carbide turning insert. They concluded that with the increase in the depth of cut, the rate of flank wear, cutting force and surface roughness increase. Palanisamy, Rajendran, and Shanmugasundaram (2007) optimized the machining parameters such as speed, feed, depth of cut using genetic algorithm, yield to minimize machining time while considering the technological constraints as cutting force, tool life in end milling process. Davim and Baptist (2000) discussed the relationship between cutting force and tool wear when machining the composite. In drilling, correlation was obtained between evolution of flank wear of the drills and the feed force. Similar correlations have been obtained by turning between the evolution of flank wear of the insert and the feed and depth forces.

Valarmathi, Palanikumar, and Latha (2013) analyzed the thrust force for composite panel based on Taguchi's design of experiments and response surface methodology. The authors

have taken input parameters such as spindle speed, feed rate and point angle. Their results indicated that high spindle speed with low feed reduced the thrust force. Arokiadass, Palaniradja, and Alagumoorthi (2012) studied the effects of speed, feed depth of cut and % weight of silicon carbide on tool wear in machining LM25 Al alloy reinforced with SiC_p in end milling operation. They analyzed the process constraints on performance characteristics using ANOVA. The decision attained by them is that the spindle speed and the content of SiC_p are the influencing factor on tool wear.

Mahesh, Muthu, and Devadasan (2015) propounded a methodology to study the optimum cutting parameter aiming to minimize the surface roughness through response surface methodology and genetic algorithm (GA). They analyzed the direct and interaction effect of cutting parameter using design expert software. The conformity test have shown that there is a good agreement with predicted and observed values. Makadia and Nanavati (2013) investigated the influence of machining parameters as feed rate, tool nose radius, cutting speed and depth of cut of turning on the surface roughness of AISI 410 steel. Three level full factorial design of experiment has been used to conduct the experiment. Response surface optimization was employed to get optimum machining condition. Shihab, Khan, Mohammed, and Siddiquee (2014), studied the surface integrity in terms of surface roughness and microhardness of the dry hard turning process. Three levels, three parameter central composite design was employed to collect the experimental data. Good surface integrity has been achieved by keeping feed rate and depth of cut as lower level.

Raju, Janardhana, Kumar, and Rao (2011), applied the multiple regression analysis and Genetic algorithm to predict the surface roughness under dry and wet condition using high speed steel and carbide tools. Feed rate was the significant factor which effects the surface roughness. Through experiments, it has also been found that surface roughness decreases with the application of coolant by using a carbide tool. Klilekap, Cakir, Aksoy, and Inan (2005) experimentally investigated the tool wear and surface roughness of 5%SiC_p Al MMC using uncoated and TiN coated cutting tools in turning process. Cutting parameters were considered as spindle speed, feed rate and depth of cut to predict the surface roughness and tool wear. They found that increase in cutting speed, increases tool wear and surface roughness. They also noticed that built up edge was absent while machining directly cast materials.

Chen, Wang, and Lee (2013) presented an optimal cutting parameter design of high speed cutting of the DIN 1.2344 tool steel. The orthogonal array with gray relation analysis was applied to optimize the end milling process with multiple performance characteristics. The selected cutting parameters were surface roughness on relief face, cutting speed, feed per tooth, axial cutting depth and radial cutting depth, while the considered performance characteristics are tool life and metal removal rate. The results of confirmation experiments revealed that gray relation analysis can effectively acquire an optimal combination of the cutting parameters.

In the present work, an attempt is made to investigate the effect of cutting variables on cutting force in end milling of

LM6 aluminum alloy composites. The main aim is to develop the mathematical model for cutting force in terms of spindle speed, feed rate, axial depth of cut, radial depth of cut and various weight percentage of SiC_p (wt% of SiC_p) by the response surface method. The infeed, crossfeed and thrust force are measured in tangential, radial and axial direction during end milling. The mathematical model aided us to study the direct and interaction effects of each parameter.

Very few authors (Arokiadass et al., 2012) have considered weight percentage of silicon carbide as one of the parameters to analyze the wear characteristics. Most of the authors studied the effect of cutting parameters such as speed, feed and depth of cut, but did not embark on the investigation by varying the weight percentage of silicon carbide on cutting force and surface roughness. Tool wear strongly correlates the surface roughness and cutting force while machining of composite. In the current work, experiments have been conducted with the purpose of specifically identifying the relationship between surface roughness and cutting force. Henceforth, the aims of this present work are

- To implement the five factor five level DoE (Design of Experiments) technique in the measurement of cutting force (axial, radial and thrust force) and surface roughness in machining of various weight percentage of SiC_p.
- To generate the mathematical model for cutting force considering the process parameter as Spindle speed, Feed rate, Axial depth of cut, Radial depth of cut and varying weight percentage of SiC_p.
- To study the direct and interaction effect of process parameters. Specifically, to describe the relationship between surface roughness and cutting force.

2. Experimental procedure

2.1. Machining parameters and response variables

Machining parameters that are taken into account during machining in end milling process are spindle speed, feed rate, axial depth of cut, radial depth of cut and weight percentage of SiC_p. The responses are cutting force and surface roughness (Ra). The three components of force in milling are infeed force (F_x), crossfeed force (F_y), thrust force (F_z) acts tangential, normal and parallel to the machine tool respectively.

2.2. Limits of machining parameters

The upper and lower limit of each process variables have to be identified from the relation.

$$X_i = \frac{2(2X - (X_{max} + X_{min}))}{X_{max} - X_{min}} \tag{1}$$

where X_i is the required coded value of the variable X . X is any value of a variable from X_{min} to X_{max} . The upper limit of the process parameter is coded as 2, lower limit as -2. The intermediate values were calculated from Eq. (1) and machining parameters and their levels were presented in Table 1.

Table 1
Machining parameters and their level.

Parameters	Unit	Notation	Factor Levels				
			-2	-1	0	1	2
Spindle Speed	rpm	N	1500	2000	2500	3000	3500
Feed rate	mm/rev	F	0.02	0.03	0.04	0.05	0.06
Axial depth of cut	mm	X	1	1.5	2	2.5	3
Radial depth of cut	mm	Y	1	1.5	2	2.5	3
Silicon carbide	wt%	W	5	10	15	20	25

Table 2
Chemical composition of LM6 aluminum alloy.

Elements	Si	Ni	Pb	Fe	Mn	Zn	Cu	Mg	Sb	Ti	Al
Percentage	10–13	0.1	0.1	0.6	0.5	0.1	0.1	0.1	0.05	0.2	Remaining

2.3. Preparation of MMCs

Aluminum metal matrix composite finds vast applications in the fields of construction, automotive and marine fields. Especially, LM6 is used for marine components, water-cooled manifolds and jackets, motor car and road transport fittings and pump parts. LM6 aluminum Alloy chemical composition is given in Table 2. Silicon carbide particulates of 30 micron size were used as reinforcement material. The particle size is also an important factor, wherein if the particle size increases, then the property will decrease because of inter particle spacing. However, the addition of SiC particles will increase the wear resistance, hardness and brittleness of the composite material (Dwivedi, Kumar, & Kumar, 2010; Ozben, Kilickap, & Cakir, 2008). The Stir casting technique was employed in order to prepare composite material. LM6 Alloy is preheated to 750 °C and kept in a graphite crucible in an electric induction furnace. SiC particles were preheated to 350 °C in furnace in order to release moisture and other toxic gases. SiC particles of amount weighing 5%, 10%, 15%, 20%, 25% were slowly added to the molten LM6 aluminum alloy inside crucible. Mechanical stirrer at 1250 rpm is used to prepare the composite for a period of 15 min. The molten material was poured inside the die at a temperature of 850 °C. Cast composites were obtained after the die is allowed to cool in still air at room temperature. The SEM micrograph of 5%, 10%, 15%, 20% and 25% silicon carbide particles in the parent metal are shown in Figure 1. It indicates that the silicon carbide particles are evenly distributed in the matrix material.

2.4. Experimental setup

Experiments were performed on the HAAS CNC vertical machining center with two carbide insert end mill cutter with the diameter of 12 mm and has the following specification: table length 1070 mm, width 230 mm, maximum spindle speed 4000 rpm, feed rate 5.1 m/min and the power of spindle motor 5.6 kW. The dimension of the work specimen was 100 mm × 100 mm × 25 mm.

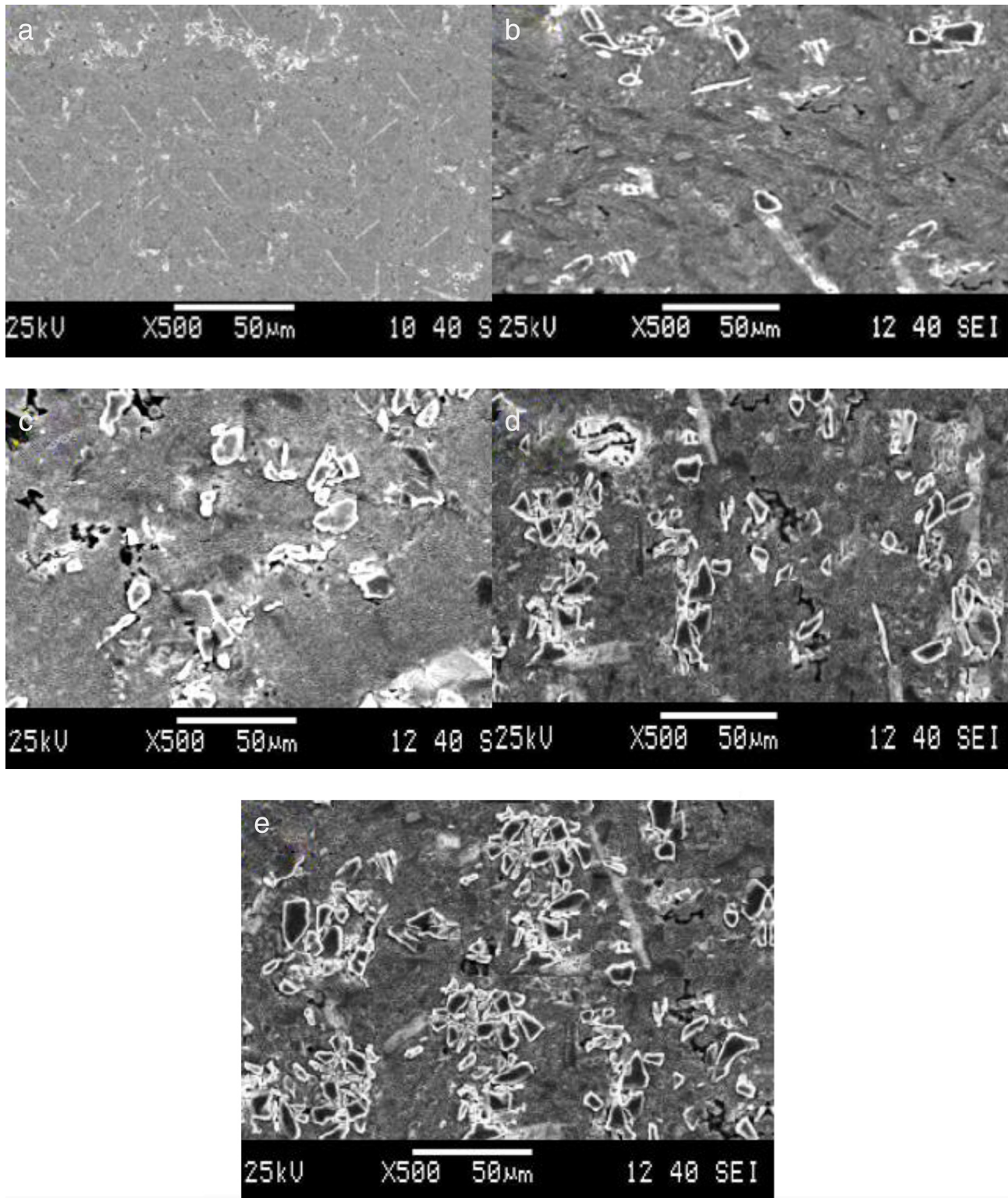


Fig. 1. SEM micrograph of (a) cast LM6/5 wt% SiC_p, (b) cast LM6/10 wt% SiC_p, (c) cast LM6/15 wt% SiC_p, (d) cast LM6/20 wt% SiC_p and (e) cast LM6/25 wt% SiC_p.

The work piece was rigidly mounted on a specially designed vice which contains sensors to measure the cutting forces. The cutting force of the tool point in *X*, *Y* and *Z* directions are measured as infeed, crossfeed and thrust force respectively, using three axis milling tool dynamo-meter made by Syscon

instruments which is pre-calibrated to 1 kgf. Sensing element bonded with strain gauge is used to sense these forces. These forces from strain gauges are processed with instrumentation amplifiers then converted into digital data by 8 bit ADC. This digital output has been send out serially through RS232C to store



Fig. 2. Experimental setup in Haas vertical machining center.

the data in the data acquisition software (MSoffice one note). The acquired data are tabulated to predict the mathematical model.

Arithmetic average roughness Ra, root mean square deviation Rq, maximum height of the profile Ry, ten point height of irregularities Rz, maximum profile peak height Rp are some of the roughness parameters to characterize the surface roughness. Among these, Ra is the commonly used roughness parameter and best suited for monitoring the consistency of the machining process. Ra is the average absolute deviation of the workpiece from the center line and the same is used in this study. After machining of each workpiece, surface roughness value is measured using MITUTOYO SJ-201 surface roughness tester as a cutoff length of 2.5 mm. For each specimen three sets of trails are measured and graphs are saved. Average surface roughness of each specimen have been noted in terms of Ra. Figure 2 shows the experimental setup used for conducting experiments.

2.5. Generation of mathematical model

The regression procedure was used for the development of the mathematical model to predict the cutting force. The second order polynomial (Bhattacharyya, 1998) representing the response surface for “k” factors is given by Eq. (2)

$$y = a_0 + \sum_{i=1}^k a_i x_i + \sum_{i=1}^k a_{ii} x_i^2 + \sum_{i < j}^k a_{ij} x_i x_j + \varepsilon_i \quad (2)$$

where a_0 = free term, a_i = linear term coefficient, a_{ii} = quadratic term coefficient, a_{ij} = interaction term coefficient and ε = error term. The response Y is expressed as a function of process parameters spindle speed (N), feed rate (Z), axial (X) and radial depth of cut (Y), wt% of SiC_p (W) shown in Eq. (3)

$$Y = \Phi(N_{iu}, Z_{iu}, X_{iu}, Y_{iu}, W_{iu}) + e_u \quad (3)$$

where Φ = response surface, e_u = residual, u = no. of observations in the factorial experiment and iu represents the level of the i th factor in the u th observation. When the mathematical form

of Φ is unknown, this function is approximated satisfactorily within the experimental region by polynomials in terms of process parameter variables. Box and Hunter proposed the central composite rotatable design for fitting a second order response surface based on the criterion of rotatability. The central composite rotatable design for five parameters with five levels consists of 32 experiments with the combination of sixteen factorial design points (lie at the vertices of the regular polyhedral), ten star points (to form sphere with α radius, consisting of equally spaced points from the center) and six replicated center points (also known as axial points which provide roughly equal precision of standard error). The design matrix with measured and predicted value of cutting force and surface roughness has been shown in Table 3. The Mini Tab statistical software (Version 15) package has been used to develop the response equations and evaluate the coefficient values. This software is also used to perform the data analyses (Montgomery, 2009). The statistical empirical equation has been developed using only the significant coefficients and is shown in Eqs. (4)–(7). R² value for in feed, cross feed and thrust force in the x , y and z directions are approaches to unity as indicated in Table 4. Because of high R² values, good correlations exist between the experimental and the predicted values.

$$\begin{aligned} F_x = & 114.673 - 27.203N + 16.24F + 23.528X + 12.21Y \\ & + 29.986W + 3.124N^2 - 0.839F^2 + 7.309X^2 \\ & + 8.452Y^2 + 5.945NF - 1.219NX + 8.057NY \\ & - 4.141NW + 6.479FX + 10.079FY + 5.846FW \\ & + 4.244XY + 11.232XW + 5.562YW \end{aligned} \quad (4)$$

$$\begin{aligned} F_y = & 95.13 - 14.422N + 19.637F + 18.874X + 5.398Y \\ & + 30.792W + 1.46N^2 + 6.67X^2 + 1.369Y^2 + 4.357NF \\ & + 7.193NY - 9.554NW + 3.248FX + 5.351FY \\ & + 3.21FW + 4.861XY + 8.329XW + 6.242YW \end{aligned} \quad (5)$$

$$\begin{aligned} F_z = & 90.466 - 18.722N + 12.568F + 18.602X + 5.894Y \\ & + 29.97W + 8.517X^2 + 4.489Y^2 + 2.395W^2 \\ & + 7.626NF + 11.235NY - 6.229NW + 3.809FX \\ & + 5.278FY + 4.487FW + 6.437XY + 12.794XW \\ & + 8.543YW \end{aligned} \quad (6)$$

$$\begin{aligned} Ra = & 5.628 - 0.62N + 0.29F + 0.45X - 0.07Y + 0.52W \\ & + 0.019N^2 + 0.64F^2 - 0.85X^2 + 0.37Y^2 - 0.46W^2 \\ & + 0.57NF + 0.06NX + 0.19NY + 0.7NW - 0.48FX \\ & + 0.09FY + 0.07FW - 0.1XY - 0.68XW + 1.1YW \end{aligned} \quad (7)$$

Table 3
Design matrix with measured and predicted values of cutting force and surface roughness.

Run	Coded Factor					Cutting force (N)									Surface Roughness Ra(μm)		
	X1	X2	X3	X4	X5	Infeed force F_x (N)			Crossfeed force F_y (N)			Thrust force F_z (N)			Observed	Predicted	% error
						Observed	Predicted	% of error	Observed	Predicted	% of error	Observed	Predicted	% of error			
1	-1	-1	-1	-1	1	153.712	153.016	0.45	121.644	122.717	-0.88	131.454	131.389	0.08	4.21	4.281	-1.68
2	1	-1	-1	-1	-1	59.306	58.352	1.61	44.145	44.751	-1.37	48.069	47.931	0.07	1.21	1.203	0.57
3	-1	1	-1	-1	-1	105.802	105.824	-0.02	83.385	84.529	-1.37	93.164	93.388	-0.21	5.62	5.537	1.47
4	1	1	-1	-1	1	87.644	87.708	-0.07	87	88.875	-2.16	73.575	74.966	-1.02	5.36	5.303	1.06
5	-1	-1	1	-1	-1	136.306	135.626	0.50	88.29	82.459	6.60	101.043	101.776	-0.74	8.92	9.075	-1.73
6	1	-1	1	-1	1	111.04	110.398	0.58	88.29	89.853	-1.77	84.366	86.078	-1.44	3.59	3.619	-0.80
7	-1	1	1	-1	1	228.273	228.61	-0.15	194.238	193.679	0.29	187.371	189.609	-4.20	3.26	3.258	0.12
8	1	1	1	-1	-1	84.462	84.538	-0.09	85.347	86.985	-1.92	65.555	67.531	-1.30	5.28	5.229	0.96
9	-1	-1	-1	1	-1	99.302	98.578	0.73	39.101	41.089	-5.08	60.543	59.454	0.66	2.83	2.807	0.81
10	1	-1	-1	1	1	88.46	87.774	0.78	66.2	68.907	-4.09	71.613	71.692	-0.06	5.18	5.209	-0.55
11	-1	1	-1	1	1	182.991	183.282	-0.16	149.112	152.373	-2.19	135.742	136.159	-0.57	6.67	6.791	-1.81
12	1	1	-1	1	-1	98.962	98.994	-0.03	80.018	82.799	-3.48	75.679	76.025	-0.26	2.04	1.981	3.17
13	-1	-1	1	1	1	211.7	211.288	0.19	168.55	168.819	-0.16	181.485	182.411	-1.68	6.07	6.073	-0.04
14	1	-1	1	1	-1	77.47	76.8	0.86	58.86	61.341	-4.22	57.879	58.545	-0.39	2.46	2.423	1.50
15	-1	1	1	1	-1	162.845	163.152	-0.19	106.4	106.743	-0.32	94.189	95.410	-1.15	4.23	4.197	0.78
16	1	1	1	1	1	239.22	239.564	-0.14	194.39	198.145	-1.93	205.082	207.276	-4.50	6.70	6.683	0.25
17	-2	0	0	0	0	182.265	181.575	0.38	126.234	129.814	-2.84	124.084	124.354	-0.34	6.92	6.925	0.02
18	2	0	0	0	0	73.013	72.763	0.34	74.405	72.126	3.06	51.012	49.466	0.79	4.48	4.517	-0.82
19	0	-2	0	0	0	77.574	78.837	-1.63	49.05	55.856	-13.88	62.258	65.330	-1.91	7.62	7.767	-1.92
20	0	2	0	0	0	146	143.797	1.51	130.29	134.404	-3.16	116.121	115.602	0.60	8.81	8.703	1.21
21	0	0	-2	0	0	96.991	96.853	0.14	85.347	84.062	1.51	85.347	87.330	-1.69	1.31	1.303	0.53
22	0	0	2	0	0	191.763	190.965	0.42	156.96	159.558	-1.66	165	161.738	5.38	3.10	3.003	3.12
23	0	0	0	-2	0	124.295	124.061	0.19	86.138	89.81	-4.26	98.1	96.634	1.44	4.24	4.139	2.54
24	0	0	0	2	0	173.607	172.901	0.41	113.772	111.402	2.08	120.024	120.210	-0.22	3.98	3.907	1.83
25	0	0	0	0	-2	56.805	54.701	3.70	29.43	33.546	-13.99	39.507	40.106	-0.24	2.77	2.751	0.68
26	0	0	0	0	2	177.346	174.645	1.52	154.998	156.714	-1.11	161.865	159.986	3.04	4.79	4.739	1.06
27	0	0	0	0	0	113.00	114.673	-1.48	97.309	95.13	2.24	88.29	90.466	-1.92	5.60	5.625	-0.44
28	0	0	0	0	0	114.85	114.673	0.15	91.423	95.13	-4.05	90.252	90.466	-0.19	5.60	5.625	-0.44
29	0	0	0	0	0	114.803	114.673	0.11	95.347	95.13	0.23	90.252	90.466	-0.19	5.61	5.625	-0.26
30	0	0	0	0	0	114.92	114.673	0.21	94.366	95.13	-0.81	91.233	90.466	0.70	5.68	5.625	0.96
31	0	0	0	0	0	114.803	114.673	0.11	96.328	95.13	1.24	92.214	90.466	1.61	5.63	5.625	0.08
32	0	0	0	0	0	114.721	114.673	0.04	97.309	95.13	2.24	89.271	90.466	-1.07	5.66	5.625	0.61

Table 4
Summary of regression analysis.

Responses	S value	R2%	Adjusted R2%
Infeed force (F_x) (N)	1.20327	99.98	99.94
Crossfeed force (F_y) (N)	3.44895	99.80	99.44
Thrust force (F_z) (N)	2.23813	99.90	99.72
Surface roughness (Ra) (μm)	0.03312	99.99	99.97

2.6. Checking the adequacy of the model

Adequacy of the experimental design has been verified by Analysis of variance (ANOVA). In ANOVA analysis lack of fit, F and R ratio have been used to check the adequacy of a model. Tables 5 and 6 indicates the ANOVA of regression parameters of

Table 5
Analysis of variance for cutting force.

Source	Infeed force(F_x) (N)					Crossfeed force (F_y) (N)				
	DOF	SS	Adjusted mean square	F value	P value	DOF	Sum of square	Adjusted mean square	F value	P value
Regression	20	73 732.6	3686.6	2546.28	0.000	20	53 241.7	2662.1	277.75	0.000
Linear	5	62 533.4	12 506.7	8638.11	0.000	5	46 251.8	99 250.4	965.15	0.000
Square	5	3651.2	730.2	504.36	0.000	5	1488.3	297.7	31.06	0.000
Interaction	10	7548	754.8	521.33	0.000	10	5501.6	550.2	57.4	0.000
Residual error	11	15.9	1.4			11	105.4	9.6		
Lack-of-fit	6	13.1	2.2	3.94	0.077	6	80.4	13.4	2.68	0.15
Pure error	5	2.8	0.6			5	25	5		
Total	31	73 748.5				31	53 347.2			

Table 6
Analysis of variance for cutting force and surface roughness.

Source	Thrust force (F_z) (N)					Surface roughness (Ra) (μm)				
	DOF	SS	Adjusted mean square	F value	P value	DOF	Sum of square	Adjusted mean square	F value	P value
Regression	20	54 707.1	2735.4	546.07	0.000	20	114.255	5.7127	5207.74	0.000
Linear	5	42 898	8579.6	1712.7	0.000	5	22.944	4.5887	4183.09	0.000
Square	5	2788.2	557.6	166.46	0.000	5	46.102	9.2204	8405.35	0.000
Interaction	10	9020.8	902.1	180.08	0.000	10	45.209	4.5209	4121.25	0.000
Residual Error	11	55.1	5	11	0.	012	0.0125			
Lack-of-Fit	6	45.5	7.6	3.94	0.077	6	0.006	0.0076	0.96	0.528
Pure Error	5	9.6	1.9			5	0.006	0.0011		
Total	31	54 762.2				31	114.267			

the predicted surface response model for cutting forces (infeed, cross feed and thrust force) and surface roughness. The calculated values of F -ratio for lack of fit are compared with the standard values of F -ratio corresponding to their degrees of freedom. The standard percentage point of F distribution for 95% confidence level is 4.95. From Tables 5 and 6 the obtained F values were 3.94, 2.68, 3.94, 0.96 for infeed, crossfeed, trust force and surface roughness respectively, smaller than the standard value, indicating that the model is adequate. It is also found that from the P values, cutting force and surface roughness models linear, square and interaction effects are significant.

2.7. Validation of the model

The validity of the model is checked for the levels of parameters which has not been included in the experimental design. The results of conformity are shown in Table 7. The test result shows that developed model computes the cutting force and surface roughness with reasonable accuracy. The difference in percentage between experimental and predicted value relative to the predicted value is estimated as the percentage error in the regression model. The error between the experimental and predicted value is less than 5%, which confirms the validity of the model.

3. Results and discussion

A mathematical model, as indicated in Eqs. (4)–(7), has been developed to predict the infeed force, cross feed force, thrust force and surface roughness in relation with spindle speed, feed rate, axial depth of cut, radial depth of cut and wt% SiC_p. The influence of the process parameters on the cutting force have been studied using the developed model. Direct and interaction effects of the process parameters on different cutting forces were plotted in Figures 3–7.

3.1. Direct effects of machining parameters on cutting forces and surface roughness

3.1.1. Effect of spindle speed

The effect of spindle speed on infeed, cross feed force, thrust force and surface roughness is shown in Figure 3. It is clear that an increase in spindle speed resulted in decreasing trend in cutting forces in all the three directions. When the spindle

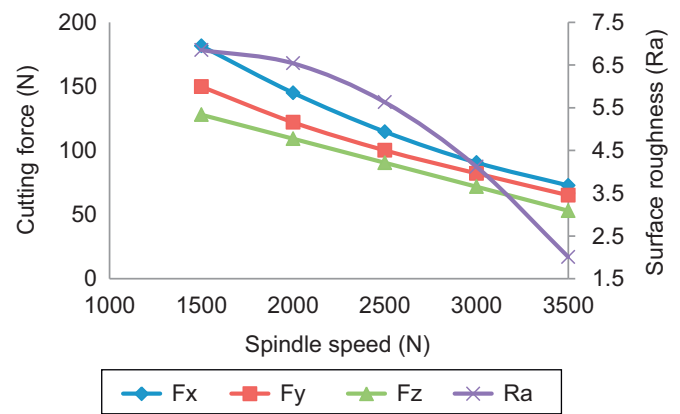


Fig. 3. Direct effect of spindle speed.

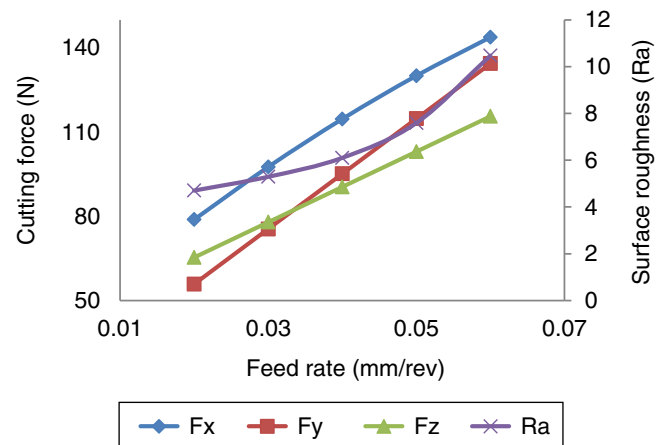


Fig. 4. Direct effect of feed rate.

speed is increased, the chip-tool contact length got decreased, the material got heated up, and softened the matrix. Hence the cutting force in end milling process is reduced. Further, it is inferred that high cutting force is produced at extremely slow speeds (Stephenson & Agapiou, 2006). With the increase in spindle speed, the surface roughness is on the decline. At lower spindle speed, the unstable larger built-up-edge is formed and the chip fracture is produced which leads to the rough surface. Whereas at higher cutting speed, the built-up-edge formation is

Table 7
Result of conformity.

S. No.	Control factors					Infeed force (N)			Crossfeed force (N)			Thrust force (N)			Surface roughness (μm)		
	N	F	X	Y	Z	Observed value	Predicted value	% Error	Observed value	Predicted value	% Error	Observed value	Predicted value	% Error	Observed value	Predicted value	% Error
1	-2	-2	1	1	2	266	265.21	0.3	227.2	229.66	-1.1	255.13	255.71	-1.48	7.143	7.217	-1.03
2	0	1	2	1	0	256.68	258.55	0.43	208.9	207.53	0.66	211.99	210.46	3.24	2.472	2.482	-0.40
3	1	1	2	2	1	380.81	375.51	1.41	290.2	291.48	-0.4	332.43	332.65	-0.73	3.089	3.092	-0.09
4	2	2	-2	-2	0	39.46	40.26	-3	70.53	68.71	2.57	56.78	56.154	0.36	4.216	4.244	-0.66

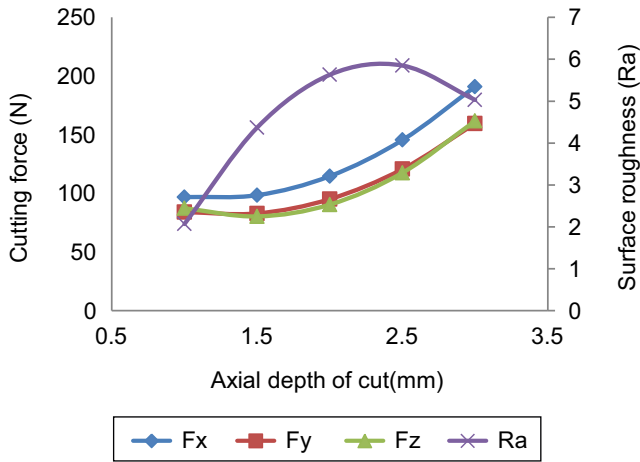


Fig. 5. Effect of axial depth of cut.

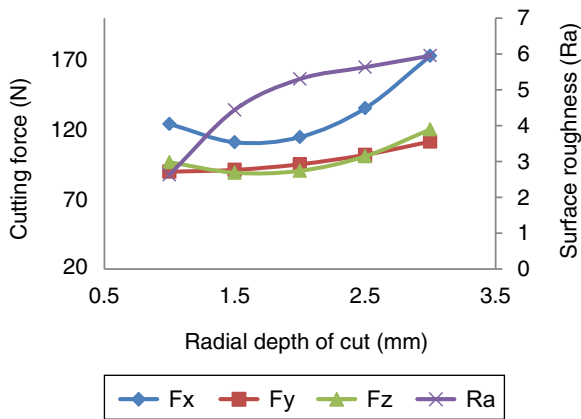


Fig. 6. Effect of radial depth of cut.

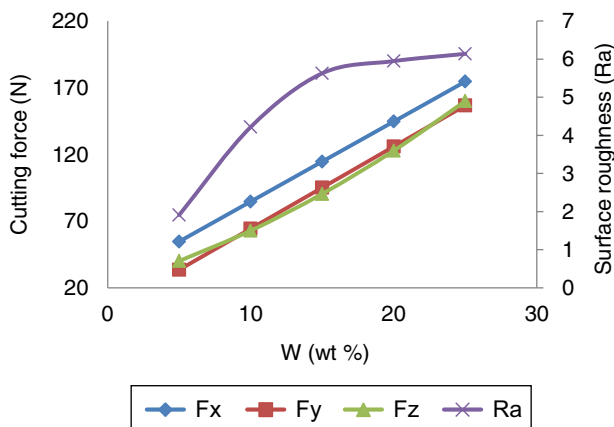


Fig. 7. Effect of wt% of SiCp.

disappearing and the chip fracture is reduced, and hence surface roughness is decreased. The same inference was derived by Seeman et al. (2010). The minimum cutting force and surface roughness is obtained at 3500 rpm of spindle speed.

3.1.2. Effect of feed rate

From Figure 4, it is evident that the feed rate has an important effect on the cutting force and surface roughness. The feed rate is directly proportional to the cutting force in tangential, radial and axial direction. The reason is that the area of contact and load carried by the cutter increases because of the increase in the feed rate. Increasing the feed rate increases the roughness of the machined surface. At a higher feed rate, the material removal rate is also high which results in the creation of friction between work and tool, leading to a pitting mark. A similar observation was made by Dwivedi et al. (2010).

3.1.3. Effect of axial depth of cut

The increase in the axial depth of cut resulted in an increased infeed, crossfeed, thrust force and surface roughness, as shown in Figure 5. When the axial depth of cut increases, more length of the flute gets engaged which results in more cutting force in the rotating direction. It is inferred from the figure that, when the axial depth of cut ranges from 1 mm to 2 mm the surface roughness increases because of the higher chip thickness of the work material.

3.1.4. Effect of radial depth of cut

From Figure 6, it is evident that the radial depth of cut has the significant effect on all the three forces and surface roughness. An increase in the radial depth of cut during the milling operation resulted in increased cutting force. The reason being that, higher radial depth of cut increases the volume of metal removal and increase the contact area of work piece, which resulted in increased cutting force in rotational direction. The surface roughness increases with the increase in the radial depth of cut due to and increase in the vibration and wear.

3.1.5. Effect of wt% of silicon carbide

The effect of wt% of SiC_p on cutting force is shown in Figure 7. While the wt% of silicon carbide particle increases, the cutting forces and surface roughness also increases. This is due to the addition of harder and stiffer reinforcing material into the soft metallic material which becomes more difficult to machine. The cutting force and surface roughness is minimum at the lowest wt% of SiC_p combination. Arokiadass et al. (2012) derived the same inference in his research work.

3.2. Interactive effects of machining parameters on cutting force

The change in the effect of one variable, when the second variable is changed from one level to another, is known as Interaction effect. The interaction effects of the process variables are useful in understanding the process behavior. The two-way interactive effects of the process variables which have strong interaction with cutting force are shown in Figures 8–14

3.2.1. Interactive effects of spindle speed and feed rate for thrust force

Figure 8 shows the interactive effect of spindle speed and feed rate on thrust force. It has been inferred from the direct

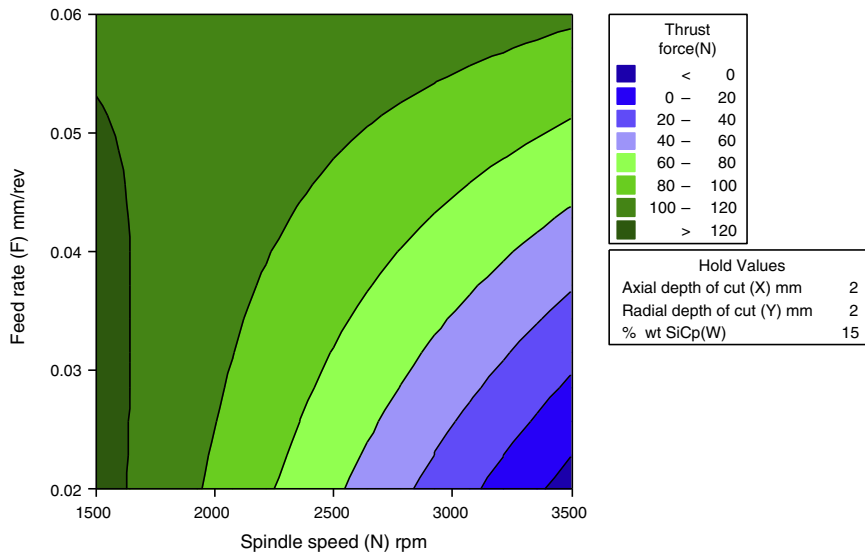


Fig. 8. Interactive effect of spindle speed and feed rate on thrust force.

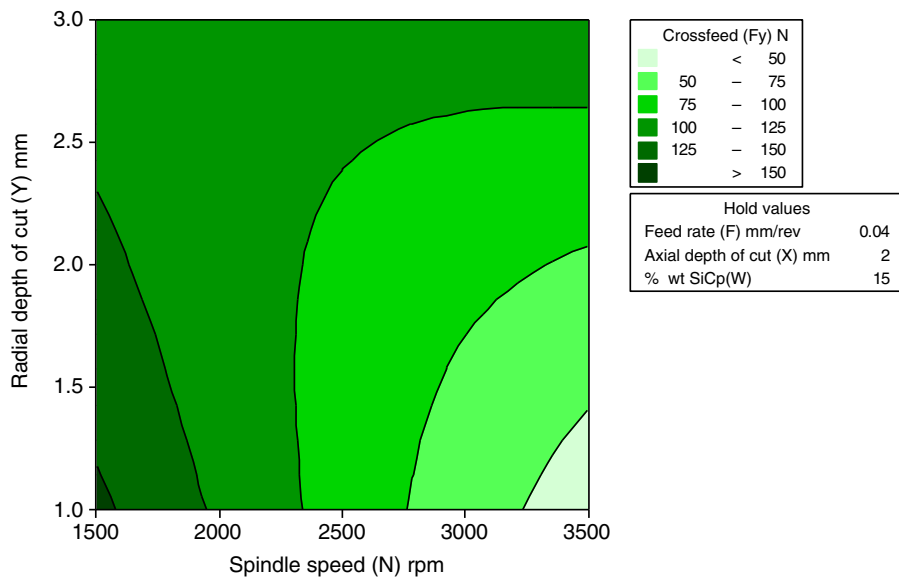


Fig. 9. Interactive effects of spindle speed and radial depth of cut for crossfeed force.

effect analysis that the spindle speed has a negative effect and feed rate has a positive effect on infeed force, crossfeed force and thrust force. It is observed that the thrust force decreases with increase in cutting speed from 1500 rpm to 3500 rpm. The trend gets similar for all the levels of feed rate from 0.02 mm/rev to 0.06 mm/rev.

3.2.2. Interactive effects of spindle speed and radial depth of cut for crossfeed force

Figure 9 indicates the interactive effect of spindle speed and radial depth of cut for crossfeed force. It has been inferred from the direct effect analysis that the spindle speed has the negative effect and radial depth of cut has a positive effect. It reveals

that crossfeed force increase with increase in spindle speed for all the level. The trend gets same for the radial depth of cut from 1 mm to 2 mm. The crossfeed force is minimal for the 3500 rpm of the spindle speed and 1 mm of the radial depth of cut.

3.2.3. Interactive effects of axial depth of cut and wt% of SiC_p for Infeed force

Figure 10 shows the relation between axial depth of cut and wt% of SiC_p on Infeed force. The direct effect analysis indicates that both axial depth of cut and wt% of SiC_p have a positive effect on infeed force, crossfeed force and thrust force. It is observed that axial depth of cut increases resulted in increase of infeed

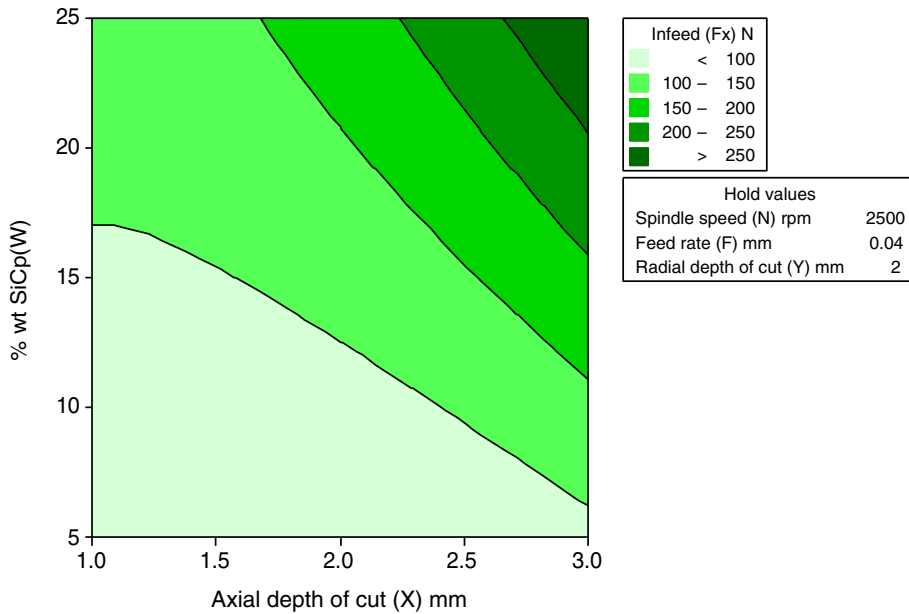


Fig. 10. Interactive effects of axial depth of cut and wt% of SiC_p for infeed force.

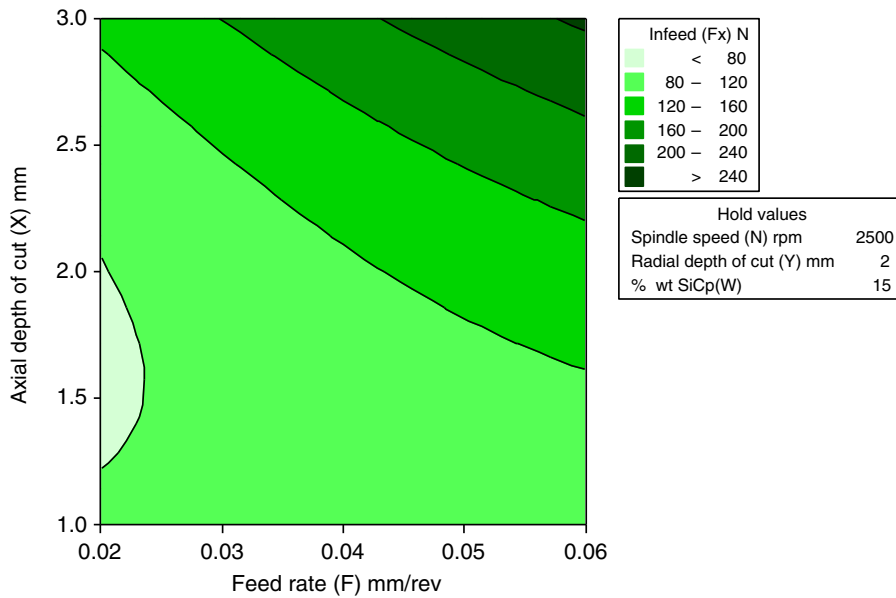


Fig. 11. Interactive effects of feed rate and axial depth of cut for infeed force.

force. A similar trend is followed for all the levels of weight percentage of silicon carbide.

3.2.4. Interactive effects of feed rate and axial depth of cut for infeed force

Figure 11 displays the interactive effect of the feed rate and the axial depth of cut on the infeed force. The direct effect analysis shows that both feed rate and axial depth of cut have a positive effect on the cutting force, which indicates that an increase in the feed rate increases the infeed force from 0.02 mm/rev to

0.06 mm/rev for all the level of axial depth of cut. The cutting force is minimum when 0.02 mm/rev to 0.03 mm/rev of the rate and 1–2.3 mm of the axial depth of cut.

3.2.5. Interactive effects of spindle speed and feed rate for surface roughness

Figure 12 shows the effect of the spindle speed and the feed rate on surface roughness. The inference from the direct effect analysis is that the spindle speed has a negative effect and the feed rate has a positive effect. It is also observed that an increase

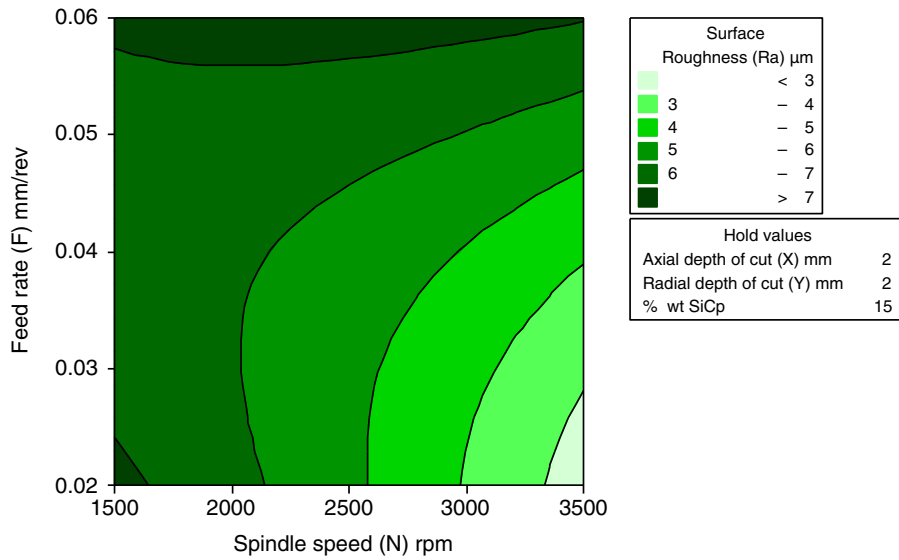


Fig. 12. Interactive effects of spindle speed and feed rate for surface roughness.

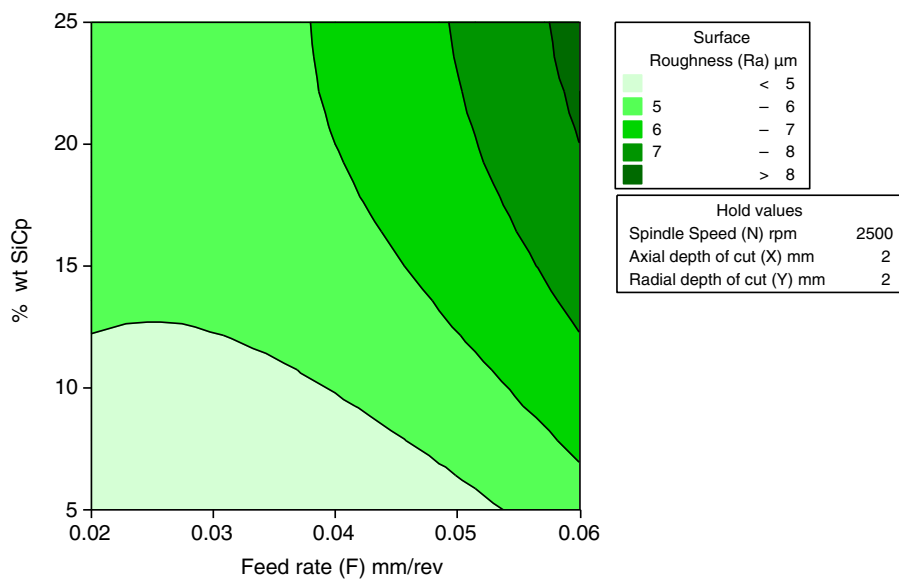


Fig. 13. Interactive effects of feed rate and wt% of SiC_p for surface roughness.

in the spindle speed resulted in a better surface finish. The same trend follows for the change of level of the feed rate from 0.02 mm to 0.04 mm. It is observed that the surface roughness is minimum at 3500 rpm of spindle speed and 0.02 mm/rev of feed rate

3.2.6. Interactive effects of feed rate and wt% of SiC_p for surface roughness

Figure 13 describes the interactive effect of feed rate and wt% of SiC_p surface roughness. The interpretation from the direct effect analysis is that both feed rate and wt% of SiC_p have a positive effect on surface roughness. The graph indicates that, decrease in feed rate, results the better surface finish for the

change of wt% SiC_p from 5% to 25%. The better surface finish can be maintained at 5 wt% SiC_p.

3.2.7. Interactive effects spindle speed and wt% of SiC_p for surface roughness

Figure 14 indicates the interactive effect of spindle speed wt% of SiC_p on surface roughness. From the direct effect analysis, the spindle speed has a negative effect and wt% of SiC_p has a positive effect. It infers that, the increase in spindle speed decreases the surface roughness at all levels of weight percentage of silicon carbide. At 3500 rpm of spindle speed and 5 wt% SiC_p ensures better surface finish.

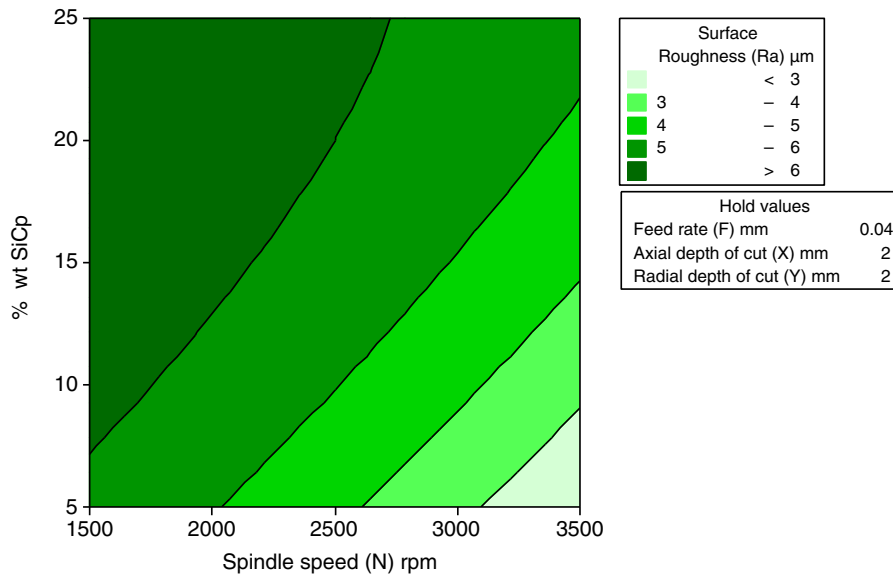


Fig. 14. Interactive effects of spindle speed and wt% of SiC_p for surface roughness.

4. Conclusion

The experimental investigation presents the effect of the machining parameters (spindle speed, feed rate, axial depth of cut, radial depth of cut and weight percentage of SiC_p) on the cutting forces (infeed force, crossfeed force and thrust force) and surface roughness in end milling of LM6 Al/SiC_p metal matrix composites.

The following conclusions have been obtained from the results of the present research work:

- Infeed force is the most dominant of the three components and shows extensively higher magnitudes than that of the cross feed and thrust force.
- Increase in spindle speed decreases the cutting forces (infeed force, crossfeed and thrust force). Additionally, the axial depth of cut on cutting forces is more sensitive compared to wt% of SiC_p, radial depth of cut and feed rate due to excessive contact area.
- The main effect plot indicates that, the feed rate is the most significant parameter on surface roughness. Because of less metal removal rate, friction and pitting mark, the surface roughness in end milling of MMC is minimum at the lower feed rate. At a higher spindle speed, built-up-edge formation is vanished, therefore the surface roughness decreases.
- The results of ANOVA and confirmation experiments have proved that the mathematical models of the cutting force are well-fitted and the estimated values of the responses are closer to the investigation's results with 95% confidence level.

Conflicts of interest

The authors have no conflicts of interest to declare.

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