

Parameter Optimization in Incremental Forming of Titanium Alloy Material

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Abstract Incremental forming is a flexible and adaptable process with a high scope in future for prototyping sector and batch shop production. It finds application in almost every engineering field. One such prominent field is the research and development. The research and development are often associated with prototyping of a variety of products for evaluating and testing the design and clarifying production costs and issues. In this paper, experimentation of Ti-6Al-4V sheets using SPIF was studied, and the influence of tool feed (f), incremental step depth (d) and spindle speed (s) to the surface roughness (R_a), wall angle (θ), and average thickness (t) were evaluated. The method was carried out using CNC Milling Machine with the help of a fixture and hemispherical end tool. Response surface methodology was used to design the experiments, and ANOVA was performed to find the factor which affected the selected method significantly. Finally, the input parameters were optimized to achieve maximum wall thickness, minimum surface roughness, and maximum wall angle.

Keywords Sheet metal · CNC machine tool · CAD/CAM · Response surface methodology · ANOVA

1 Introduction

Incremental sheet forming (ISF) is a dieless sheet forming process which can reduce the high tooling cost associated with the traditional process and increase the customizability making it suitable for prototyping and in low volume production industries like aerospace, automotive, biomedical, etc. [1]. The incremental sheet forming process has better formability than other conventional sheet metal forming techniques due to localized deformations in ISF. Kopac and Kampus [2] used the ball-type forming tool with 10 mm diameter and used grease as lubricant which improved tribological characteristics. Park et al. [3] have compared the traditional sheet forming with the incremental sheet forming and have found that forming limit curve appears in a different pattern in ISF. The low step depth has increased the formability limit. Cerro et al. [4] used finite element analyses to predict accurately the response parameters such as geometrical accuracy, sheet thickness, and roughness of formed component. The FEA was carried out using ABAQUS explicit software, and results have been compared to the actual experimentation. The low incremental depth and application of lubricant between the contact surfaces have significantly improved the surface finish. Araghi et al. [5] combined the stretch forming process and incremental sheet metal forming process. The combined process has been observed to be similar to the two-point incremental forming. Sheet thinning in SPIF and combined process have been compared and studied in detail. Finite element simulation procedure has been set up for pure incremental sheet forming and combined process. Uniform thickness distribution and reduction in forming time have been found in the combined process. Results of Minutolo et al. [6] indicate that higher wall angles can be formed in cone shapes when compared

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Fig. 1 Experimental setup in LV 45—CNC Vertical Machining Centre



Fig. 2 Incrementally formed sheet metal



to pyramid shapes. Numerical simulations performed using LS-DYNA reveals that it is possible to form free surfaces relevant to different strain conditions including the most stressed zone that acts as the fracture points. Yao et al. [7] optimized the input parameters such as tool diameter, step depth, and sheet thickness over response parameters such as deformation energy, geometric error, and surface roughness. Experiments have been carried out in Al1060 sheets using a hemispherical end tool made of X210CrW12. The process variables relation has been studied using response surface methodology and a regression model has been developed. It is found that increasing the tool diameter has increased the deformation energy but

decreased the accuracy and surface roughness. Jadhav [8] used the helical tool path for incrementally forming the sheet metal which results in twist and dents in the final formed components. To overcome this defects, the sheet is formed using either the tool path with distributed increment or bi-directional tool path and also has improved the geometric accuracy of the part. This selected tool path has distributed the forces uniformly along the edge of the geometry which has been the factor for increasing geometric accuracy. Reddy et al. [9] analyzed the formability and surface finish of Al 5052 alloy by incrementally forming the truncated cones and truncated pyramids using Box–Behnken method. It has been reported that for all

Table 1 Input parameters and output response values

Run	Spindle speed (rpm)	Tool feed (mm/min)	Incremental step depth (mm)	Formed wall thickness (mm)	Wall angle (degree)	Surface roughness (microns)
1	100	2000	0.3	0.38	27.16	1.04
2	200	2000	0.3	0.36	27.52	1.08
3	150	1500	0.2	0.42	28.43	0.94
4	150	2000	0.2	0.38	28.38	0.85
5	150	1500	0.2	0.38	28.65	0.92
6	200	2000	0.1	0.42	29.63	0.72
7	200	1000	0.3	0.38	27.92	1.19
8	150	1500	0.3	0.36	27.67	1.1
9	100	1000	0.3	0.4	27.45	1.14
10	200	1000	0.1	0.44	29.74	0.83
11	100	1500	0.2	0.38	28.20	0.88
12	150	1500	0.2	0.38	28.45	0.92
13	100	1000	0.1	0.46	29.39	0.8
14	150	1500	0.2	0.4	28.32	0.96
15	150	1000	0.2	0.4	28.85	1.02
16	200	1500	0.2	0.38	28.88	0.98
17	150	1500	0.2	0.38	28.78	0.92
18	150	1500	0.1	0.44	29.61	0.77
19	150	1500	0.2	0.4	28.70	0.9
20	100	2000	0.1	0.42	29.10	0.74

Table 2 ANOVA table for surface roughness

Source	Sum of squares	Df	Mean square	F value	P value	Prob > F
Model	0.32	9	0.036	52.03	< 0.0001	Significant
L-spindle speed	4.000E-003	1	4.000E-003	5.81	0.0366	Significant
M-incremental step depth	0.29	1	0.29	415.08	< 0.0001	Significant
N-tool feed	0.030	1	0.030	43.96	< 0.0001	Significant
LM	8.000E-004	1	8.000E-004	1.16	0.3062	
LN	4.500E-004	1	4.500E-004	0.65	0.4375	
MN	2.000E-004	1	2.000E-004	0.29	0.6016	
L^2	9.091E-006	1	9.091E-006	0.013	0.9108	
M^2	1.278E-004	1	1.278E-004	0.19	0.6756	
N^2	1.278E-004	1	1.278E-004	0.19	0.6756	
Residual	6.881E-003	10	6.881E-004			
Lack of fit	4.748E-003	5	9.495E-004	2.23	0.2003	Not significant
Pure error	2.133E-003	5	4.267E-004			
Cor total	0.33	19				
SD = 0.026		Mean = 0.93				PRESS = 0.043
R-Squared = 0.9791		Adj R-Squared = 0.9603				Pred R-Squared = 0.8703

incremental depths, surface roughness decreases with increases in tool diameter. Surface roughness up to certain angle increases with increase in incremental depth and then decreases. The surface roughness value decreases as the wall angle increases. Hussaina et al. [10] have found that

Al-1060-H24 (hardening exponent = 0.042) provides 7.5% higher formability than Stainless Steel 304A (hardening exponent = 0.53). Hence, it has been concluded that less hardening exponent produces high formability of the sheet metal.

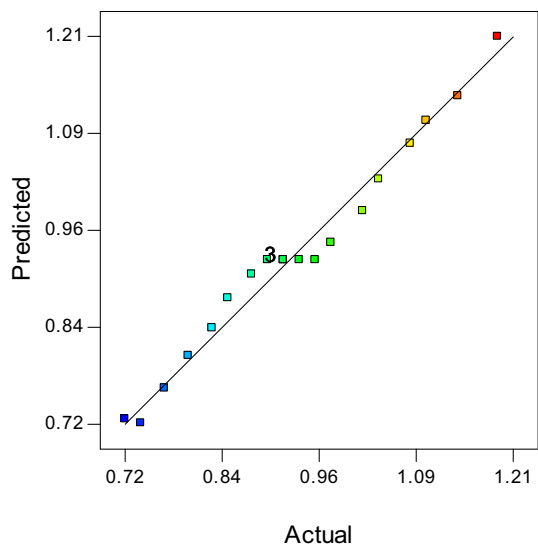


Fig. 3 Predicted response versus actual response for surface roughness

2 Materials and Methods

The Ti-6Al-4V alloy material of thickness 0.6 mm and 16 mm hemispherical tool were selected for the optimization. The experiments were conducted using a 3-Axis CNC milling machine (LMW LV45) which is shown in Fig. 1. A specialized fixture setup was designed and manufactured for the sole purpose of incrementally forming the 150 × 150 mm sheets. The formed cone specimen is shown in Fig. 2. The design of experiment's layout containing the input process parameters and the output responses are given in Table 1.

3 Results and Discussion

The experiment based on central composite design (CCD) is conducted, and the obtained responses are analyzed for optimal conditions for the surface roughness, wall angle, and thinning. ANOVA is performed for the design of experiment to obtain the factors that significantly affect the single point incremental sheet forming (SPIF) of Ti-6Al-4V alloy material.

3.1 Surface Roughness

Table 2 shows the ANOVA for surface roughness at 95% confidence interval ($\alpha = 95\%$), respectively. The P value in the range of 0–0.05 indicates that the factors are statistically significant in affecting the process and P value in the range of 0.05 to 0.1 indicates that factors are marginally significant, whereas factors whose P value is above 0.1 indicate their insignificance in affecting the process. Significant model terms are L , M , N . The obtained value of R^2 is 0.9791 for surface roughness which indicates that the model is 97.91% capable to predict the response value. The R^2 value is in good agreement with the adjusted R^2 (0.9603), and the “Pred R -Squared” of 0.8703 is in reasonable agreement with the “Adj R -Squared” of 0.9603. Therefore, this model can be used to predict the surface roughness within the selected parameter ranges. PRESS (stands for “Prediction Residual Sum of Squares”) is a measure of how well a selected model fits every point in a design. The model is desirable if PRESS value is less, and hence, the obtained value of 0.043 makes the model desirable.

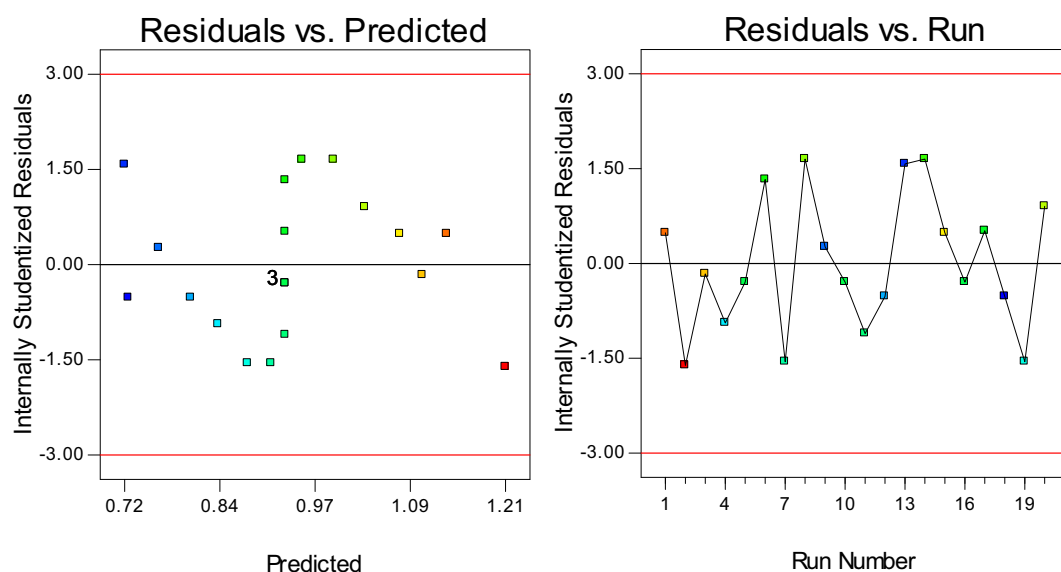


Fig. 4 Residuals versus predicted and run number for surface roughness

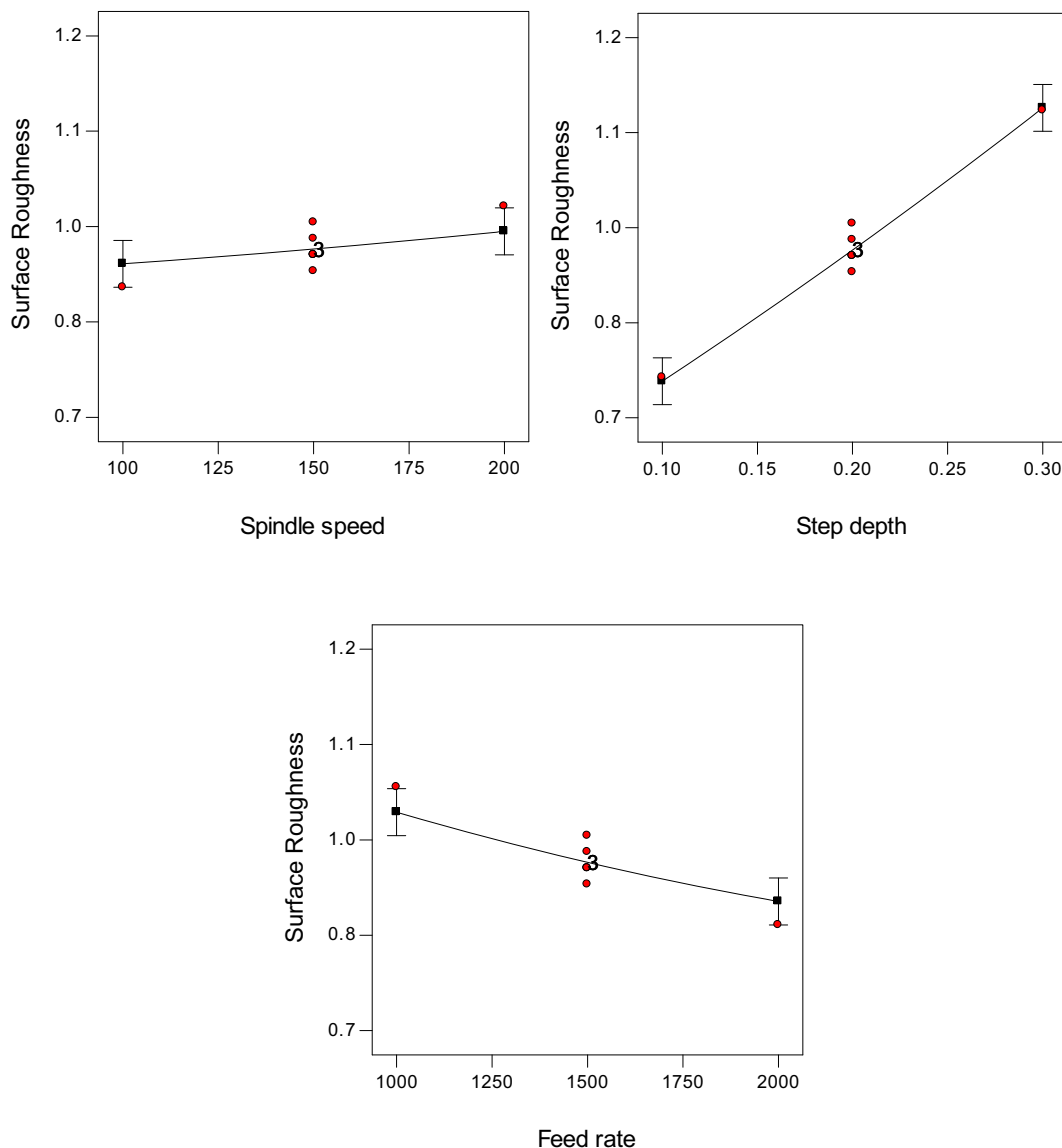


Fig. 5 One-factor graph for surface roughness

The quadratic equation which fits the experimental model is given in the equation in terms of coded factors. The quadratic equation for R_a is given in Eq. 1.

$$\begin{aligned}
 \text{Surface Roughness } R_a &= 0.93 + 0.020 * L + 0.17 * M - 0.055 * N \\
 &+ 0.010 * L * M - 0.0075 * L * N - 0.005 * M * N \\
 &+ 0.00182 * L^2 + 0.00682 * M^2 + 0.00682 * N^2
 \end{aligned}
 \tag{1}$$

. Figure 3 shows the graph for surface roughness plotted against experimental and predicted values. Figure 4 shows the graph plotted for residuals versus predicted and run number for surface roughness.

One-factor analysis explains how individual factors affect the surface roughness upon changing their levels. From Fig. 5, the following information can be found: (1) With the increase in spindle speed from 100 to 200 rpm, there is only a small change in surface roughness from 0.9 μ m to 1 μ m. This indicates that friction does not play a major role in increasing the formability of this material. (2) With the increases in step depth from 0.1 mm to 0.3 mm, the surface roughness increases from 0.75 μ m to 1.15 μ m. The better surface finish is achieved when decreasing the step depth from 0.3 mm to 0.1 mm. (3) With the increases in feed rate from 1000 mm/min to 2000 mm/min, the surface roughness decreases from 1.05 μ m to 0.9 μ m.

Table 3 ANOVA table for wall angle

Source	Sum of squares	Df	Mean square	F value	P value	Prob > F
Model	10.37	9	1.15	49.87	< 0.0001	Significant
<i>L</i> -spindle speed	0.57	1	0.57	24.72	0.0006	Significant
<i>M</i> -incremental step depth	9.51	1	9.51	411.46	< 0.0001	Significant
<i>N</i> -tool feed	0.24	1	0.24	10.53	0.0088	Significant
LM	3.125E−004	1	3.125E−004	0.014	0.9097	
LN	6.125E−004	1	6.125E−004	0.027	0.8739	
MN	0.011	1	0.011	0.46	0.5153	
<i>L</i> ²	0.021	1	0.021	0.89	0.3683	
<i>M</i> ²	5.114E−004	1	5.114E−004	0.022	0.8847	
<i>N</i> ²	3.551E−004	1	3.551E−004	0.015	0.9038	
Residual	0.23	10	0.023			
Lack of fit	0.068	5	0.014	0.42	0.8177	Not significant
Pure error	0.16	5	0.033			
Cor total	10.60	19				
SD = 0.15		Mean = 28.54			PRESS = 0.84	
<i>R</i> -Squared = 0.9782		Adj <i>R</i> -Squared = 0.9586			Pred <i>R</i> -Squared = 0.9203	

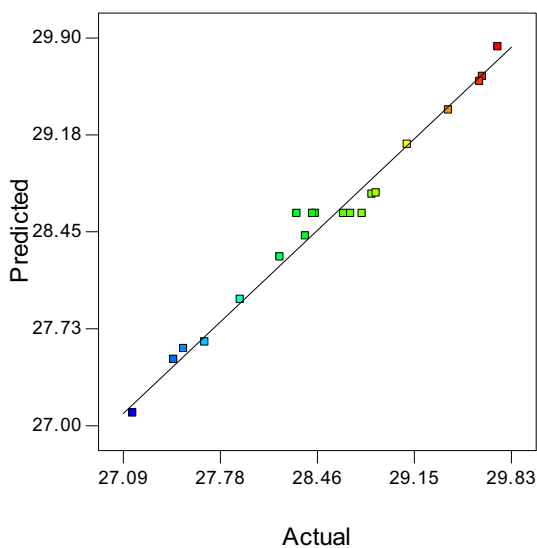


Fig. 6 Predicted response versus actual response for wall angle

3.2 Wall Angle

Table 3 shows the ANOVA for wall angle at 95% confidence interval ($\alpha = 95\%$), respectively. Significant model terms are *L*, *M*, *N*. The obtained value of R^2 is 0.9782 for wall angle which indicates that the model is 97.82% capable to predict the response value. The R^2 value is in good agreement with the adjusted R^2 (0.9586), and the “Pred *R*-Squared” of 0.9203 is in reasonable concurrence with the “Adj *R*-Squared” of 0.9586. Therefore, this model can be

used to predict the wall angle within the selected parameter ranges. The obtained PRESS value is lesser which makes the model desirable. Figure 6 shows the graph for wall angle plotted against experimental and predicted values. Figure 7 shows the graph plotted for residuals versus predicted and run number for wall angle.

The quadratic equation which fits the experimental model is given in the equation in terms of coded factors.

$$\begin{aligned}
 \text{Wall angle} = & 28.53 + 0.24 * L - 0.97 * M - 0.16 * N \\
 & - 0.00625 * L * M + 0.00875 * L * N \\
 & - 0.036 * M * N - 0.086 * L^2 + 0.014 * M^2 - 0.011 * N^2
 \end{aligned}
 \tag{2}$$

One-factor analysis explains how individual factors affect the wall angle upon changing their levels. From Fig. 8, the following information can be found: (1) The spindle speed at 100 rpm has produced 28° wall angle and at 200 rpm has produced 28.7° wall angle which indicates that increase in spindle speed does not produce high variation in wall angle. (2) With the increase in step depth, the wall angle decreases greatly from 29.5° to 27.6° when using 0.1 mm and 0.3 mm step depth, respectively. (3) The feed rate also has not produced significant variations in wall angle as feed rate is associated with crossfeed and infeed.

3.3 Thinning (Measured Thickness)

Table 4 shows the ANOVA for thinning at 95% confidence interval ($\alpha = 95\%$), respectively. Here, *M*, *N* are significant

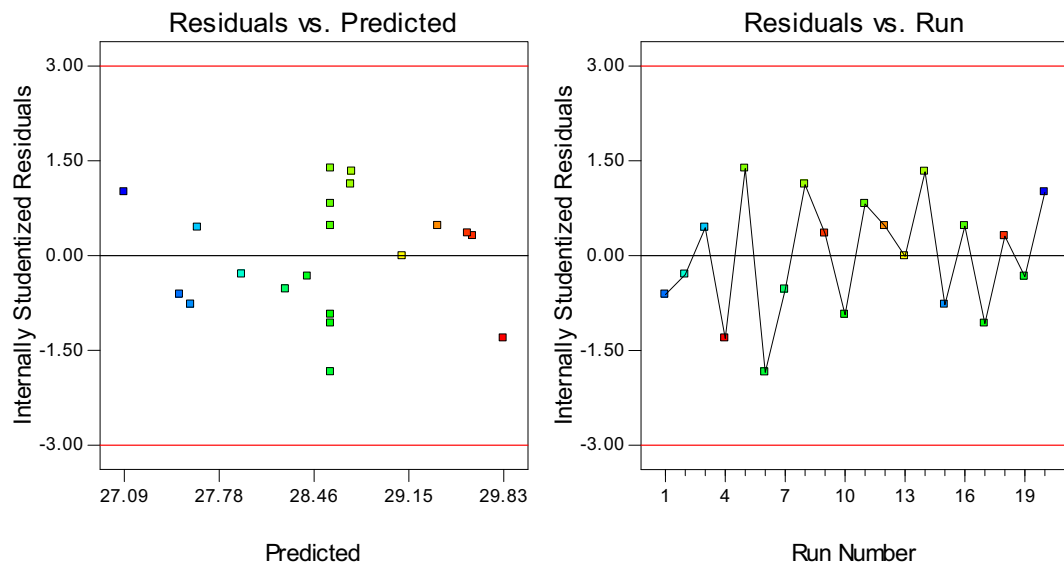


Fig. 7 Residuals versus predicted and run number for Wall angle

Fig. 8 One-factor graph for wall angle

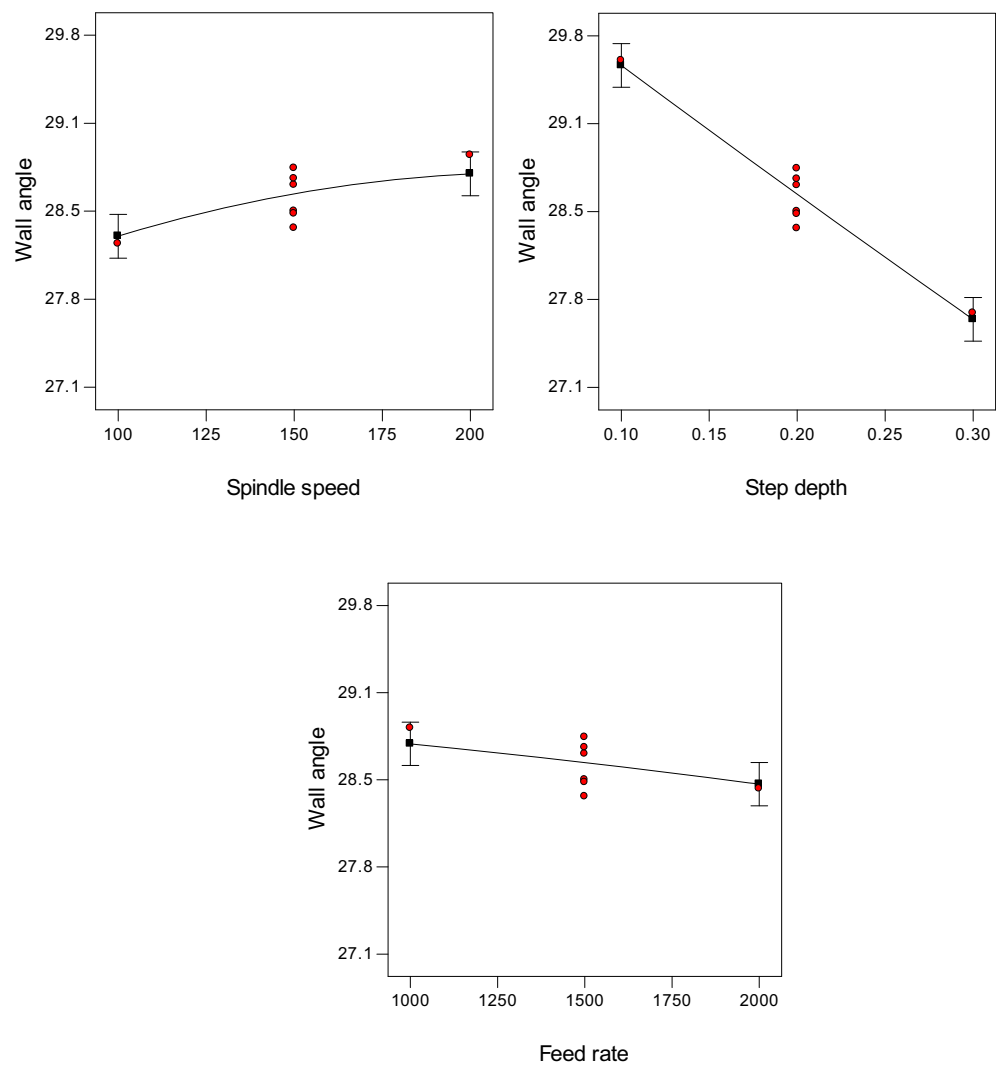
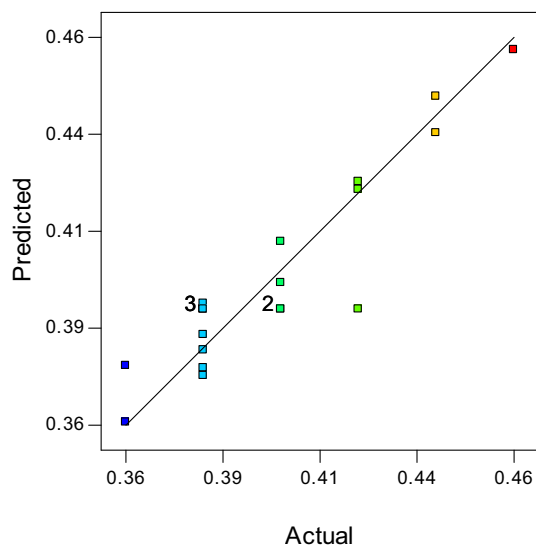


Table 4 ANOVA table for thinning

Source	Sum of squares	df	Mean square	F value	P value	Prob > F
Model	0.012	9	1.370E-003	6.89	0.0029	Significant
<i>L</i> -spindle speed	3.600E-004	1	3.600E-004	1.81	0.2081	Not significant
<i>M</i> -incremental step depth	9.000E-003	1	9.000E-003	45.27	< 0.0001	Significant
<i>N</i> -tool feed	1.440E-003	1	1.440E-003	7.24	0.0227	Significant
LM	5.000E-005	1	5.000E-005	0.25	0.6269	
LN	5.000E-005	1	5.000E-005	0.25	0.6269	
MN	5.000E-005	1	5.000E-005	0.25	0.6269	
<i>L</i> ²	5.682E-005	1	5.682E-005	0.29	0.6046	
<i>M</i> ²	6.568E-004	1	6.568E-004	3.30	0.0992	Marginally significant
<i>N</i> ²	8.182E-005	1	8.182E-005	0.41	0.5356	
Residual	1.988E-003	10	1.988E-004			
Lack of fit	6.548E-004	5	1.310E-004	0.49	0.7731	Not significant
Pure error	1.333E-003	5	2.667E-004			
Cor total	0.014	19				
SD = 0.014			Mean = 0.40			PRESS = 0.006664
<i>R</i> -Squared = 0.8612			Adj <i>R</i> -Squared = 0.7362			Pred <i>R</i> -Squared = 0.5346

**Fig. 9** Predicted response versus actual response for thinning

model terms and M^2 is marginally significant model term. The obtained value of R^2 is 0.8612 for thinning which indicates that the model is 86.12% capable to predict the response value. The R^2 value is in reasonable concurrence with the adjusted R^2 (0.7362). Therefore, this model can be used to predict the thinning within the selected parameter ranges. The lesser value of PRESS is desirable, and hence, the obtained PRESS value is 0.006664 which makes the model desirable.

The quadratic equation which fits the experimental model is given in the equation in terms of coded factors.

$$\begin{aligned} \text{Thinning} = & 0.39 - 0.006 * L - 0.030 * M \\ & - 0.012 * N - 0.0025 * L * M \\ & + 0.0025 * L * N + 0.0025 * M * N \\ & - 0.004545 * L^2 + 0.015 * M^2 + 0.005455 * N^2 \end{aligned} \quad (3)$$

. Figure 9 shows the graph for thinning plotted against experimental and predicted values. Figure 10 shows the graph plotted for residuals versus predicted and run number for thinning.

One-factor analysis explains how individual factors affect the average thickness of the formed sheet metal upon changing their levels. From Fig. 11, the following information can be found: (1) As the spindle speed increases from 100 to 200 rpm, the thickness of the sheet decreases slightly from 0.395 mm to 0.385 mm which shows that increase in spindle speed does not produce high variation in thinning of the sheet metal. (2) With the increase in step depth, the thickness of the formed sheet metal decreases greatly from 0.43 mm to 0.38 mm when using 0.1 mm and 0.3 mm step depth, respectively. (3) The feed rate also produces only significant amount of variations in thinning of sheet metal.

3.4 Desirability-Based Optimization

In the RSM, desirability-based optimization has been performed for the multiresponse optimization. The desirability $d = 0$ denotes that the response is completely intolerable, whereas $d = 1$ denotes that the response is closely of the target value. Table 5 shows the upper limits, the lower

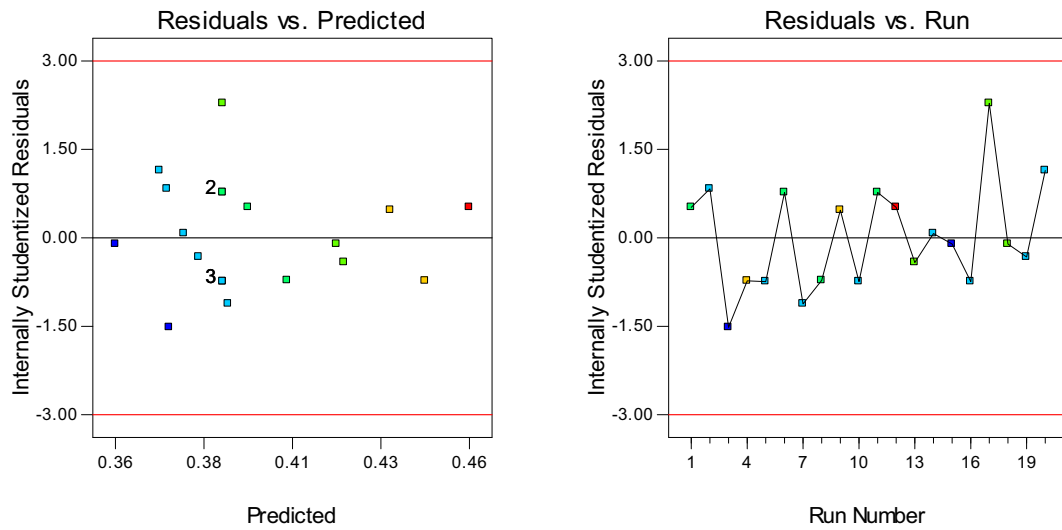


Fig. 10 Residuals versus predicted and run number for thinning

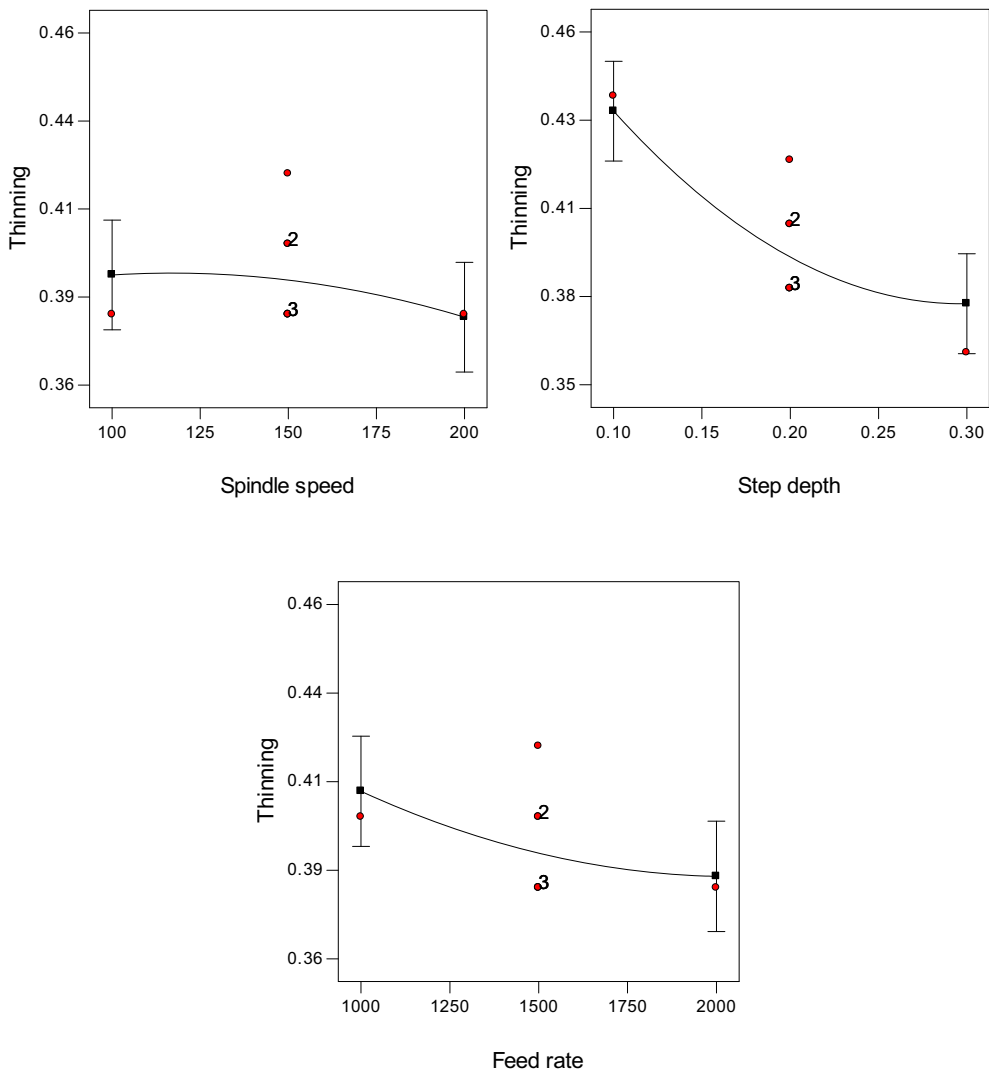


Fig. 11 One-factor graph for thinning

Table 5 Goals set and limits used for optimization

Constraints name	Goal	Lower limit	Upper limit
Spindle speed	Is in range	100	200
Incremental step depth	Is in range	0.1	0.3
Tool feed	Is in range	1000	2000
Surface roughness	Minimize	0.72	1.19
Wall angle	Maximize	27.16	29.74
Thinning	Maximize	0.36	0.46

limits, goals set used for optimization. The highest desirability solution is selected as optimal value, and the same is shown in Table 6. The histogram of the desirability of the best solution is shown in Fig. 12.

Numerical optimization is an algorithm which is similar to hill climbing technique. The desirability is a measure of how possibly the optimized response can be obtained. The desirability should always be closer to 1. Considering output responses such as surface roughness, thinning, and wall angle, the best optimized combination of values is 0.8 μm , 0.46 mm, and 29.7° which can be obtained when formed with 150 rpm spindle speed, 0.1 mm step depth, and 1000 mm/min tool feed which is shown in Fig. 13. The desirability to obtain the same responses when run using the optimized process parameters is 90%.

3.5 Confirmatory Experiment

The confirmatory experiment was performed with the optimized process parameters obtained from the numerical optimization. The confirmatory experiment has produced component with surface roughness of 0.78 μm , thinning of 0.44 mm, and wall angle of 29.76° which has deviate by only 2.5%, 4.5%, and 0.2% respectively.

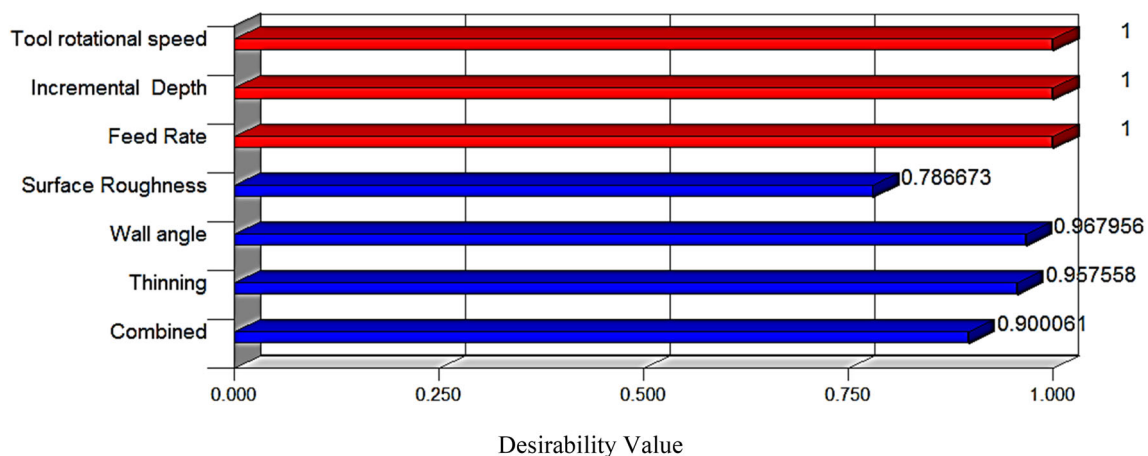
4 Conclusion

- The obtained value of R^2 is 0.9791, 0.9782, 0.8612 for surface roughness, wall angle, thinning, respectively, which indicates that the model is 97.91%, 97.82%, 86.12% capable to predict the response value.
- Incremental depth and tool feed are the significant parameters that affects the selected response parameters.
- The best optimal global solutions are as follows

Spindle speed	Incremental step depth	Tool feed
146.09 rpm	0.10 mm	1000.00 mm/mm
Surface roughness	Wall angle	Thinning
0.82 μm	29.7°	0.46 mm

Table 6 Best global solutions for optimization

Number	Spindle speed	Incremental step depth	Tool feed	Surface roughness	Wall angle	Thinning	Desirability
1	146.09	0.10	1000.00	0.82	29.7	0.46	0.900

**Fig. 12** Histogram of the best solution

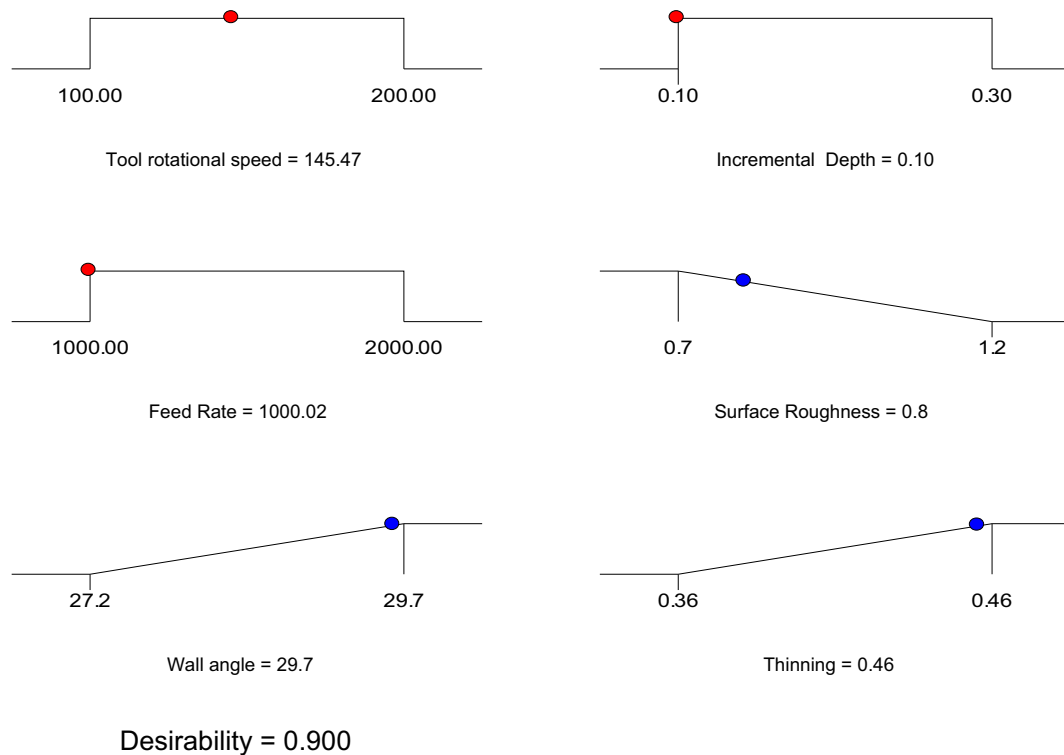


Fig. 13 Ramp plot for optimal responses

- d. Numerical optimization has been performed, and running confirmatory experiment with optimal parameters results in 2.5%, 4.5%, and 0.2% deviation from the actual desired values.

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