

Analysis on Multi Responses in Face Milling of Ammc Using Fuzzy-Taguchi Method

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Abstract

In this paper, Fuzzy-Taguchi Method has been used to identify the optimal combination of influential factors by analyzing the multi responses in the face milling. Milling experiment has been performed on AMMC (Aluminium Metal Matrix Composite), according to Taguchi orthogonal array (L27) for various combinations of influential parameters: speed, feed, depth of cut and coolant. Fuzzy logic is applied for the analysis of experimental response data of vibrations, temperature, surface roughness and resultant forces. The Fuzzy grade is calculated from this data and Fuzzy grade is optimized using Taguchi method in order to get the optimal parameter values, and also influence of parameters on individual responses is studied using Taguchi S/N ratio analysis. This work is useful for analysis of machining parameters in face milling.

Keywords

Face Milling, AMMC, Fuzzy Logic, Taguchi S/N Ratio Analysis

1. Introduction

Conventional materials have the limitations in achieving good combination of strength, stiffness, toughness, density, etc. To overcome these limitations and to meet the ever increasing demand of modern day technology, composites are most promising materials in recent days. Metal matrix composites (MMCs) possess high strength, hardness, toughness, and good thermal resistance properties as compared to unreinforced alloys.

Milling is the process of machining flat, curved or irregular surfaces by feeding the work piece against a rotating cutter containing a number of cutting edges. The literature review related to machining of AMMC is presented in the following.

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Shivanand *et al.* (2004) [1] compared Powder Metallurgy method and stir casting method for producing the AMMC through testing of mechanical properties and concluded that stir casting method was best suitable for preparation of AMMC. Dalalah and Bataineh (2009) [2] presented a fuzzy logic approach to the selection of the best silicon crystal slicing technology. Fuzzy reasoning is used to model the expert's comprehension and uncertainty in the factors used in the decision criteria. Kuttolamadom *et al.* (2010) [3] studied the effects of machining feed on surface roughness in milling Al-6061. Increase the feed up until a cut-off surface roughness limit is reached and then increase the speed within the roughness range, to maximize productivity. Yazdi and Khorram (2010) [4] investigated the selection of optimal machining parameters for face milling operations in order to minimize the surface roughness and to maximize the material removal rate using RSM and ANN methods. Abuthakeer *et al.* (2011) [5] carried out a study to obtain the surface roughness and vibration responses were investigated at various parametric levels and combinations using LabVIEW software. On the completion of the experimental test, ANN is used to validate the results. Gunay *et al.* (2011) [6] focused on study of machining parameters on the cutting forces and surface roughness during face milling of Ti-6Al-4V alloy with carbide tools under dry conditions. Resultant cutting forces and surface roughness increased with an increase in feed rate, whereas decreasing with increase in cutting speed. Çalışkan *et al.* (2012) [7] showed the influence process parameters on the cutting forces (F_x , F_y , and F_z) and R_a in hard milling. According to the results of variance analysis, the cutting forces are the most sensitive to feed rate f_z and then depth of cut. The cutting speed is only influential on F_x . Globocki Lakica *et al.* (2013) [8] carried out Experimental Research Using of MQL in Metal Cutting. The analysis shows that turning with MQL is a good alternative for conventional lubrication. Al-Zubaidi *et al.* (2013) [9] proposed a new multi objective optimization approach in the face milling. It is showed that the method provides a robust way of looking at the optimum parameter selection problems. Jatin (2013) [10] studied the effect of different machining on Surface Roughness in milling by Taguchi analysis. Low cutting speed should be used for long cutter life. High cutting speed and low feeds give best surface finishes. Venkata Ramaiah *et al.* (2013) [11] made an attempt to obtain optimum machining parameters for minimum cutting forces and cutting temperature by using Fuzzy Logic. It is showed that the method provides good results in machining of Al 6061. Das *et al.* (2014) [12] investigated the application of traditional Taguchi method with fuzzy logic for multi objective optimization of the machining process of Al-5Cu. Experimental results are demonstrated to present the effectiveness of this approach.

To address the lack of research in this issue, the present work has been done on face milling of AMMC with the following objectives:

- 1) To study the influence of machining parameters on multi responses;
- 2) To identify the optimal setting of milling process parameters (coolant, cutting speed, feed rate and depth of cut) for optimal responses: vibrations, temperature, surface roughness and resultant forces.

2. Taguchi Orthogonal Array for Conducting Experiments

In this experiment four process parameters at three levels have been considered are shown in **Table 1**.

L_{27} orthogonal array as shown in **Table 2** has been chosen for conducting experiments. Experiments are performed according to this design and the values of surface roughness, resultant force, vibrations and temperature are recorded and their Normalized responses and response values are shown in **Table 3**.

3. Milling of AMMC Material

3.1. Experimental Procedure

Step by Step procedure used in the experimental work.

- 1) Keep the milling machine ready for performing the machining operation;
- 2) Connect the DAQ system to milling machine;
- 3) Connect the milling tool dynamometer to the milling machine;
- 4) Prepare the AMMC work piece sample and fix in machine vice;
- 5) Fix the milling cutter to an arbor and make machine ready for experiment;
- 6) Perform milling experiments as per Taguchi design on work piece for various combinations of process control parameters like coolant, spindle speed, feed and depth of cut;
- 7) Measure surface roughness with the help of a portable stylus-type Talysurf (Taylor Hobson, mitutyo);

- 8) Measure forces such as thrust force, feed force, cross feed force by using milling tool dynamometer;
 9) Measure vibrations by using accelerometer sensor (PCB Accelerometer having Sensitivity 100.5 mV/g) and temperature by using temperature sensor (NI-9211 Temperature Module) of LabVIEW based DAQ system.

3.2. Measurement of Responses

Experimental responses: surface roughness, vibrations, temperature and resultant forces are measured for different combinations of influential parameters. The measuring instruments and procedure is presented in the following.

Table 1. Process parameters and their levels.

Sl. No.	Parameters	Level 1	Level 2	Level 3
1	Coolant	Dry	Kerosene	Soluble oil
2	Speed, rpm	900	1120	1400
3	Feed, mm/rev	315	500	800
4	Depth of cut, mm	0.8	1.0	1.2

Table 2. L₂₇ orthogonal array.

Exp. No.	Process parameters			
	Coolant	Speed (rpm)	Feed (mm/min)	Depth of cut (mm)
1	Dry	900	315	0.8
2	Dry	900	500	1
3	Dry	900	800	1.2
4	Dry	1120	315	0.8
5	Dry	1120	500	1
6	Dry	1120	800	1.2
7	Dry	1400	315	1
8	Dry	1400	500	1.2
9	Dry	1400	800	0.8
10	Kerosene	900	315	1.2
11	Kerosene	900	500	0.8
12	Kerosene	900	800	1
13	Kerosene	1120	315	1
14	Kerosene	1120	500	1.2
15	Kerosene	1120	800	0.8
16	Kerosene	1400	315	1.2
17	Kerosene	1400	500	0.8
18	Kerosene	1400	800	1
19	Soluble oil	900	315	1
20	Soluble oil	900	500	1.2
21	Soluble oil	900	800	0.8
22	Soluble oil	1120	315	1.2
23	Soluble oil	1120	500	0.8
24	Soluble oil	1120	800	1
25	Soluble oil	1400	315	0.8
26	Soluble oil	1400	500	1
27	Soluble oil	1400	800	1.2

Table 3. Normalized responses.

Exp. No.	Responses				Normalized responses			
	Resultant force (Kgf)	Vibrations (m/sec ²)	Temperature (°C)	Surface roughness (µm)	Resultant force (Kgf)	Vibrations (m/sec ²)	Temperature (°C)	Surface roughness (µm)
1	7.28	8.29	35.1	0.26	0.84860	0.77143	0.8537	0.93631
2	9.27	8.77	35.3	0.27	0.77821	0.63429	0.8469	0.92994
3	10.82	8.94	37.1	0.53	0.72338	0.58571	0.7857	0.76433
4	3.00	9.24	41.7	0.69	1.00000	0.50000	0.6293	0.66242
5	7.87	9.48	43.9	1.15	0.82773	0.43143	0.5544	0.36943
6	8.06	9.68	50.7	0.84	0.82101	0.374 29	0.3231	0.56688
7	12.88	10.35	56.4	0.59	0.65051	0.18286	0.1293	0.72611
8	16.40	10.57	58.4	0.60	0.52600	0.120 00	0.0612	0.71975
9	8.66	10.69	60.2	0.58	0.79979	0.08571	0.0000	0.73248
10	7.87	8.12	30.8	0.66	0.82773	0.82000	1.0000	0.68153
11	8.77	8.43	31.5	0.52	0.79590	0.73143	0.9762	0.77070
12	11.70	8.62	31.8	0.43	0.69225	0.67714	0.9660	0.82803
13	23.02	8.97	39.2	0.16	0.29183	0.57714	0.7143	1.00000
14	26.70	10.93	41.3	0.36	0.16166	0.01714	0.6429	0.87261
15	24.35	9.67	41.7	0.65	0.24478	0.37714	0.6293	0.68790
16	29.09	10.99	43	0.55	0.07711	0.00000	0.5850	0.75159
17	27.29	9.69	43.7	0.51	0.14079	0.37143	0.5612	0.77707
18	31.27	10.89	44.9	0.24	0.00000	0.02857	0.5204	0.94904
19	4.90	7.87	33.6	1.20	0.93279	0.89143	0.9048	0.33758
20	4.90	8.57	33.9	1.41	0.93279	0.69143	0.8946	0.20382
21	4.90	7.49	35	0.81	0.93279	1.00000	0.8571	0.58599
22	8.49	8.89	34.7	1.34	0.80580	0.60000	0.8673	0.24841
23	7.55	7.65	37	1.09	0.83905	0.95429	0.7891	0.40764
24	8.12	8.27	38.4	1.32	0.81889	0.77714	0.7415	0.26115
25	16.55	9.87	37.5	1.73	0.52069	0.32000	0.7721	0.00000
26	16.03	10.43	39.9	0.36	0.53909	0.16000	0.6905	0.87261
27	16.58	10.99	41.6	0.50	0.51963	0.00000	0.6327	0.78344

3.2.1. Measurement of Surface Roughness

The surface roughness values of the machined surface are measured in order to analyze the surface finish quality. Surface Roughness is measured with the help of Talysurf (**Figure 1**).

3.2.2. Measurement of Vibrations Using PCB Accelerometer

Spindle vibrations are measured using LabVIEW based DAQ. To measure the vibrations of the spindle, PCB Accelerometer (sensitivity 100.5 mv/g) is placed on spindle as shown in **Figure 2**.

3.2.3. Measurement of Temperature Using NI-9211 Thermocouple

The temperature at the contact of tool and work piece is measured using LabVIEW software based NI-9211 temperature thermocouple. According to the design of experiments at various conditions Dry, Kerosene and soluble oil (**Figure 3**).

3.2.4. Measurement of Cutting Force Using Milling Tool Dynamometer

In order to measure the forces of thrust, feed and cross feed force, milling tool dynamometer (**Figure 4**) the resultant force [7] from these forces calculated.



Figure 1. Talysurf (surface roughness measuring machine).

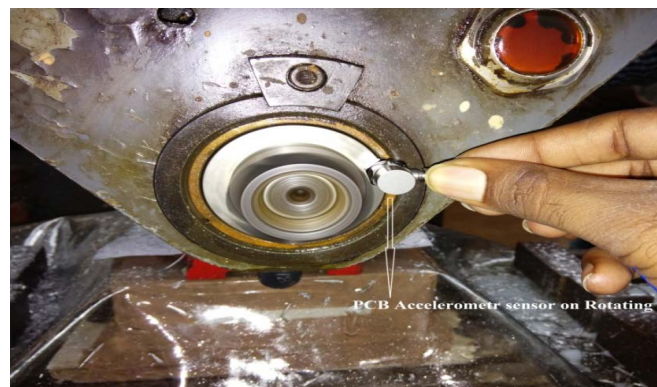


Figure 2. PCB accelerometer mounted on moving spindle.

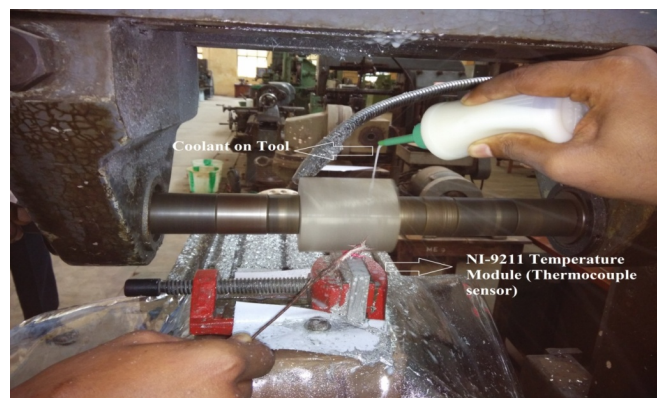


Figure 3. Thermocouple placed at contact of work piece and tool.



Figure 4. Milling tool dynamometer for measuring cutting forces.

3.3. Data Normalization

Data Normalization is done on data which has different range and unit in one data sequence may differ from the

others. Data preprocessing is also necessary when the directions of the target in the sequences are different.

If the target data value characteristic is “smaller the better”. The original sequence can be normalized using the Equation (1) as follows:

$$x_i^*(k) = \frac{\max x_i^o(k) - x_i^o(k)}{\max x_i^o(k) - \min x_i^o(k)} \tag{1}$$

where $i = 1, \dots, m$; $k = 1, \dots, n$. m is the number of experimental data items and n is the number of parameters. $x_i^o(k)$ Denotes the original sequence, $x_i^*(k)$ the sequence after the data pre-processing, $\max x_i^o(k)$ the largest value of $x_i^o(k)$, $\min x_i^o(k)$ the smallest value of $x_i^o(k)$, and x^o is the desired value.

4. Analysis of Multi Responses

Deals with analysis of multi responses data shown in **Table 3** and optimization of process parameters in milling of AMMC using Fuzzy logic and Taguchi analysis. And also influence of process parameters on individual responses is studied using Taguchi S/N ratio analysis.

4.1. Determinations of Optimum Process Parameters Using Fuzzy Logic

The experimental data is analyzed using Fuzzy logic to determine optimum process parameters as in the following.

4.1.1. Creation of Membership Functions

Figures 5-8 shows the membership function for vibrations, Temperatures, Surface roughness input values in the process parameter.

Figure 9 shows the membership function selected to defuzzify the output (performance), calculated using the simplifying rules. The rules for process parameter for some rules are shown in **Table 4**.

Using more than three fuzzy sets would cause an explosion in the number of possible expressions. For the current case study 3 fuzzy sets and 4 inputs are considered. This results in a possible $3^4 = 81$ expressions. The five fuzzy sets used in the performance membership function are “very low”, “low”, “medium”, “high”, and “very high”. Again, the trimf shape is employed to map the fuzzy sets. The use of the centroid defuzzification method

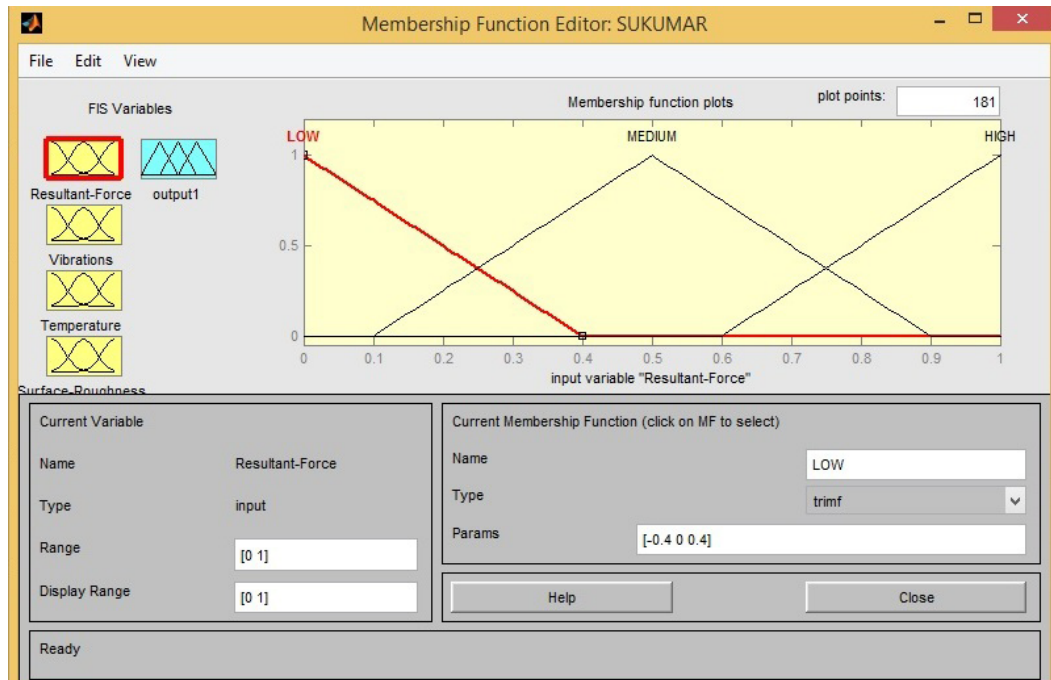


Figure 5. Membership function for resultant force.

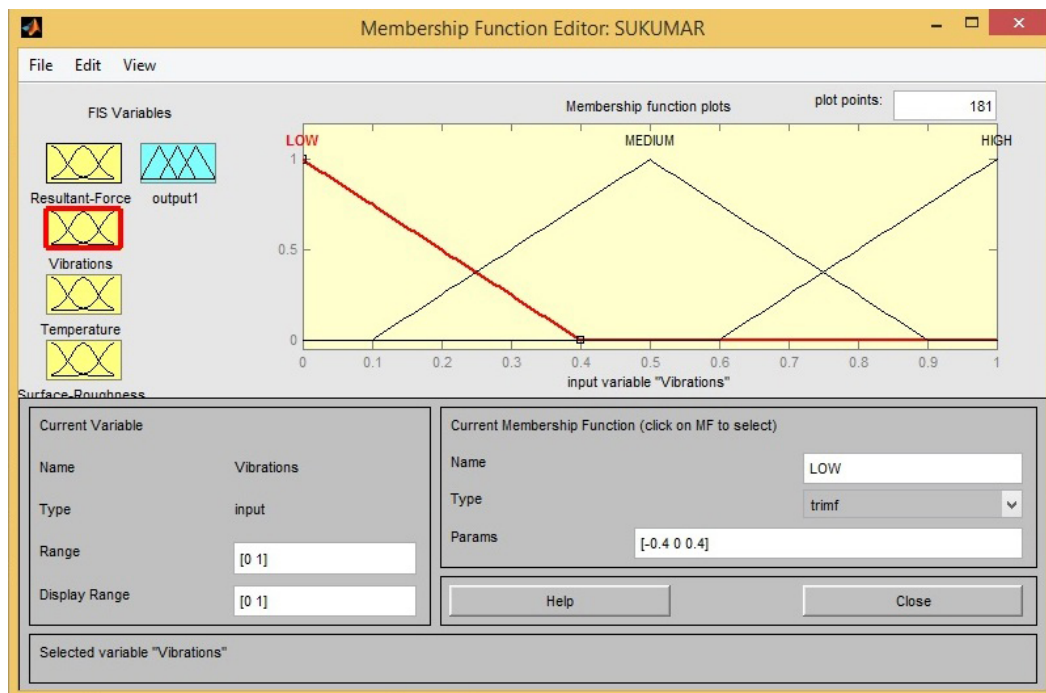


Figure 6. Membership functions for vibrations.

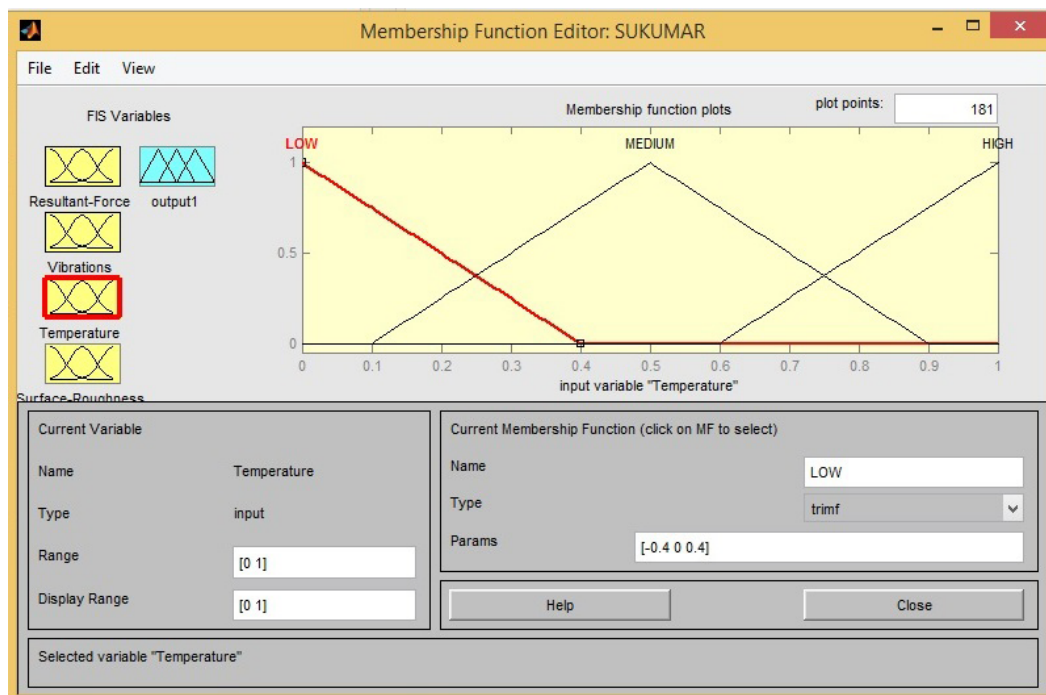


Figure 7. Membership functions for temperature.

is recommended as it results in a more smoothly shaped rule surface. In other words, the output performance index is less sensitive to slight variations in input values which occur near the fuzzy set overlaps. After the input and output membership functions are all defined and their fuzzy sets properly configured, the next step is to write the simplifying rules used to transform the input into output. As shown in the next section, this is the most crucial step in creating a fuzzy logic process parameter system.

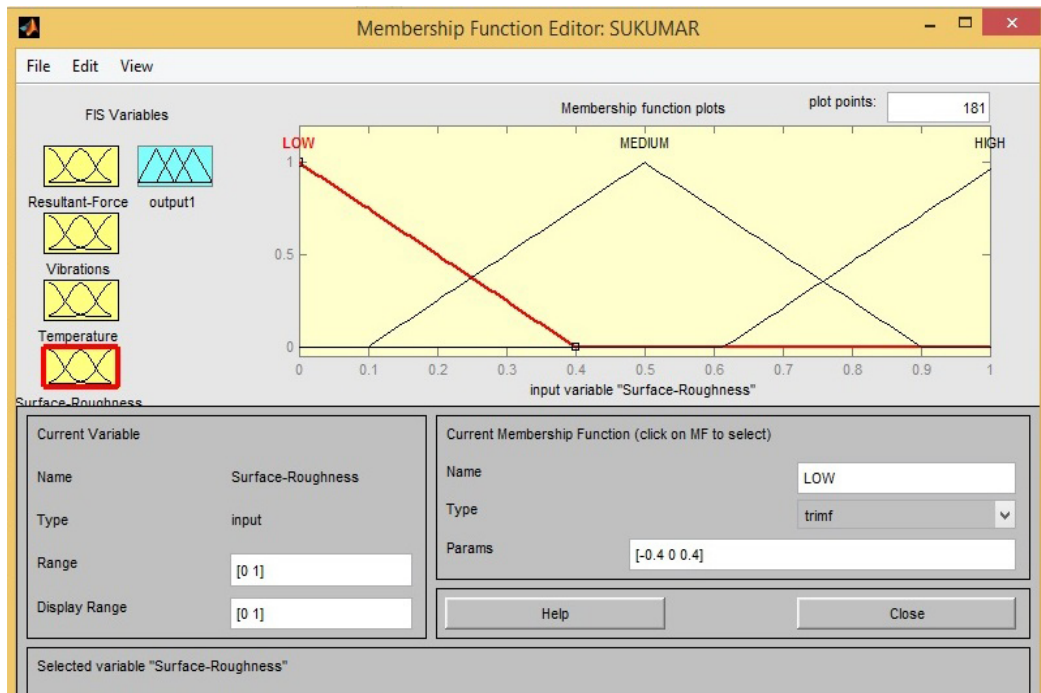


Figure 8. Membership functions for surface roughness.

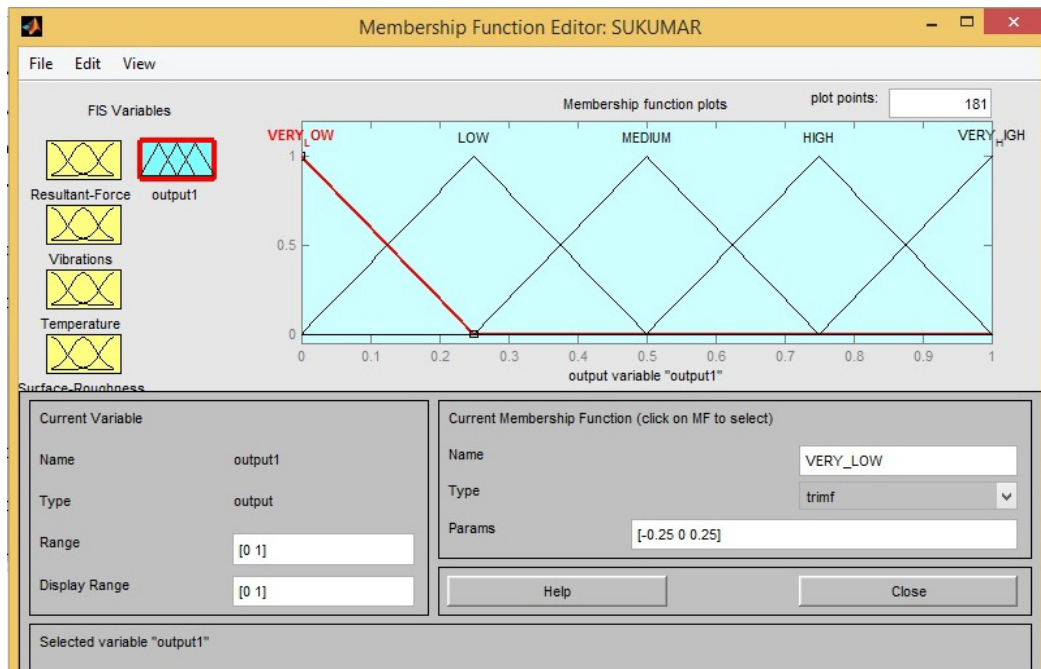


Figure 9. Membership functions for performance.

4.1.2. Evaluation of Fuzzy Grade

Fuzzy grade values are determined from Fuzzy logic using Fuzzy rules (Table 4) and normalizing data (Table 3). By using evaluation function of the MATLAB editor.

The evaluation function is: $b = [\text{experimental data}]$; $a = \text{readfis}(\text{"File name"})$, $t = \text{evalfis}(b, a)$.

After executing above code, the output of FIS editor is obtained as shown in Table 5. These Fuzzy grade values are used for determining optimum parameter values by applying Taguchi techniques as in the following section.

Table 4. Rules for process parameter.

S. No.		Resultant force		Vibrations		Temperature		Surface roughness		Performance
1	If	Low	And	Low	And	Low	And	Low	Then	Very high
2	If	Low	And	Low	And	Low	And	Medium	Then	High
3	If	Low	And	Low	And	Low	And	High	Then	Medium
4	If	Low	And	Low	And	Medium	And	Low	Then	Very high
5	If	Low	And	Low	And	Medium	And	Medium	Then	High
6	If	Low	And	Low	And	Medium	And	High	Then	Medium
7	If	Low	And	Low	And	High	And	Low	Then	High
8	If	Low	And	Low	And	High	And	Medium	Then	Medium
9	If	Low	And	Low	And	High	And	High	Then	Low
10	If	Low	And	Medium	And	Low	And	Low	Then	Very high
-	-	-	-	-	-	-	-	-	-	-
-	-	-	-	-	-	-	-	-	-	-
-	-	-	-	-	-	-	-	-	-	-
78	If	High	And	High	And	Medium	And	High	Then	Very low
79	If	High	And	High	And	High	And	Low	Then	Medium
80	If	High	And	High	And	High	And	Medium	Then	Low
81	If	High	And	High	And	High	And	High	Then	Very low

Table 5. Fuzzy grade for normalized process parameters.

Exp. No.	Input parameters				Fuzzy grade
	Resultant force (Kgf)	Vibrations (m/sec ²)	Temperature (°C)	Surface roughness (µm)	
1	0.84860	0.77143	0.8537	0.93631	0.312
2	0.77821	0.63429	0.8469	0.92994	0.3091
3	0.72338	0.58571	0.7857	0.76433	0.3424
4	1.00000	0.50000	0.6293	0.66242	0.3988
5	0.82773	0.43143	0.5544	0.36943	0.541
6	0.82101	0.37429	0.3231	0.56688	0.4524
7	0.65051	0.18286	0.1293	0.72611	0.416
8	0.52600	0.12000	0.0612	0.71975	0.4777
9	0.79979	0.08571	0.0000	0.73248	0.5
10	0.82773	0.82000	1.0000	0.68153	0.3263
11	0.79590	0.73143	0.9762	0.77070	0.3286
12	0.69225	0.67714	0.9660	0.82803	0.3252
13	0.29183	0.57714	0.7143	1.00000	0.3893
14	0.16166	0.01714	0.6429	0.87261	0.4943
15	0.24478	0.37714	0.6293	0.68790	0.473
16	0.07711	0.00000	0.5850	0.75159	0.5686
17	0.14079	0.37143	0.5612	0.77707	0.447
18	0.00000	0.02857	0.5204	0.94904	0.5144
19	0.93279	0.89143	0.9048	0.33758	0.4091
20	0.93279	0.69143	0.8946	0.20382	0.4605
21	0.93279	1.00000	0.8571	0.58599	0.2908
22	0.80580	0.60000	0.8673	0.24841	0.4837
23	0.83905	0.95429	0.7891	0.40764	0.4157
24	0.81889	0.77714	0.7415	0.26115	0.4906
25	0.52069	0.32000	0.7721	0.00000	0.734
26	0.53909	0.16000	0.6905	0.87261	0.3976
27	0.51963	0.00000	0.6327	0.78344	0.4159

4.2. Taguchi Analysis

Taguchi S/N ratio analysis is performed on Fuzzy grade data shown in **Table 6** using Minitab software and optimum parameter values are found (**Table 7**) and the main effects plot is shown in **Figure 10**.

From the results (**Table 7** and **Figure 11**), optimum process parameters combination for Fuzzy grade is Speed 3-Coolant 3-Depth of cut 3-Feed 1

Which means

Speed at level 3 (1400 rpm)

Table 6. Factors and fuzzy grade for process parameters.

S. No.	Coolant	Speed (rpm)	Feed (mm/min)	Depth of cut (mm)	Fuzzy grade
1	Dry	900	315	0.8	0.312
2	Dry	900	500	1	0.3091
3	Dry	900	800	1.2	0.3424
4	Dry	1120	315	0.8	0.3988
5	Dry	1120	500	1	0.541
6	Dry	1120	800	1.2	0.4524
7	Dry	1400	315	1	0.416
8	Dry	1400	500	1.2	0.4777
9	Dry	1400	800	0.8	0.5
10	Kerosene	900	315	1.2	0.3263
11	Kerosene	900	500	0.8	0.3286
12	Kerosene	900	800	1	0.3252
13	Kerosene	1120	315	1	0.3893
14	Kerosene	1120	500	1.2	0.4943
15	Kerosene	1120	800	0.8	0.473
16	Kerosene	1400	315	1.2	0.5686
17	Kerosene	1400	500	0.8	0.447
18	Kerosene	1400	800	1	0.5144
19	Soluble oil	900	315	1	0.4091
20	Soluble oil	900	500	1.2	0.4605
21	Soluble oil	900	800	0.8	0.2908
22	Soluble oil	1120	315	1.2	0.4837
23	Soluble oil	1120	500	0.8	0.4157
24	Soluble oil	1120	800	1	0.4906
25	Soluble oil	1400	315	0.8	0.734
26	Soluble oil	1400	500	1	0.3976
27	Soluble oil	1400	800	1.2	0.4159

Table 7. Rank of process parameters for fuzzy grade.

Level	Coolant	Speed	Feed	Depth of cut
1	7.766	9.334	7.256	7.592
2	7.516	6.794	7.456	7.649
3	7.074	6.227	7.643	7.115
Delta	0.692	3.106	0.387	0.534
Rank	2	1	4	3

Coolant at level 3 (Soluble oil)
 Depth of cut at level 3 (1.2 mm)
 Feed at level 1 (315 mm/rev)

4.3. Individual Response Analysis

Taguchi S/N ratio analysis is applied for data shown in **Table 5** to study the influence of process parameters on individual responses. The results are shown in **Figures 11-14** and **Tables 8-11**.

4.3.1. S/N for Vibration versus Coolant, Speed, Feed and Depth of Cut

From **Figure 11** and **Table 8**, the optimum process parameters combination for individual response (Vibration) is

Speed 3-Depth of cut 3-Coolant1, 2 -Feed 3

Which means

Speed at level 3 (1400 rpm)

Depth of cut at level 3 (1.2 mm)

Coolant at level 1, 2 (Dry, Kerosene)

Feed at level 3 (800 mm/rev)

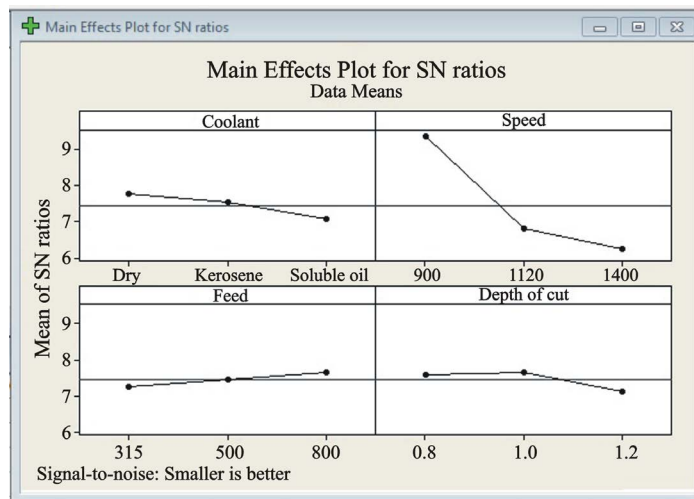


Figure 10. Main effects plot for vibration.

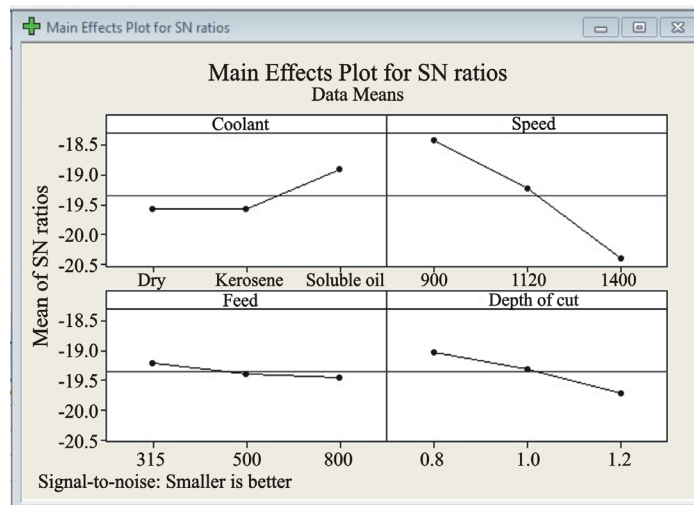


Figure 11. Main effects plot for fuzzy grade.

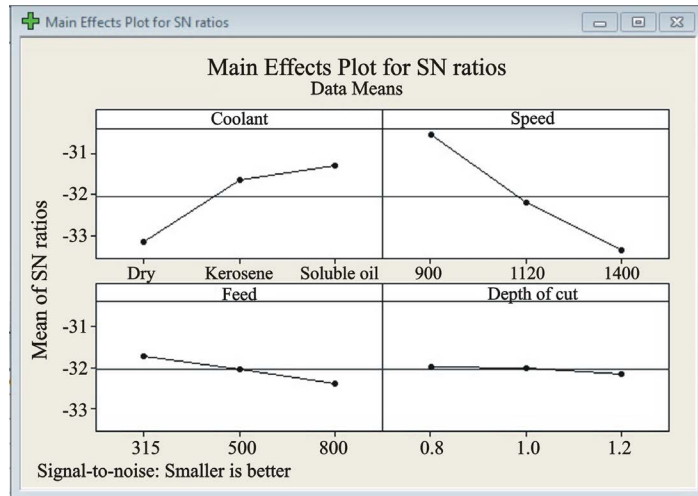


Figure 12. Main effects plot for temperature.

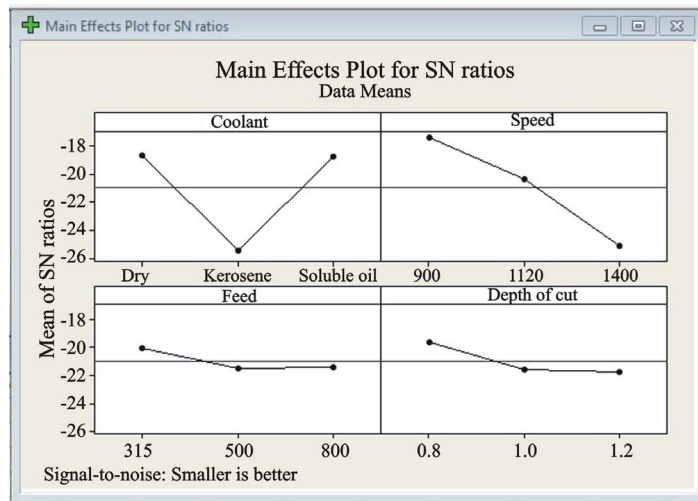


Figure 13. Main effects plot for resultant force.

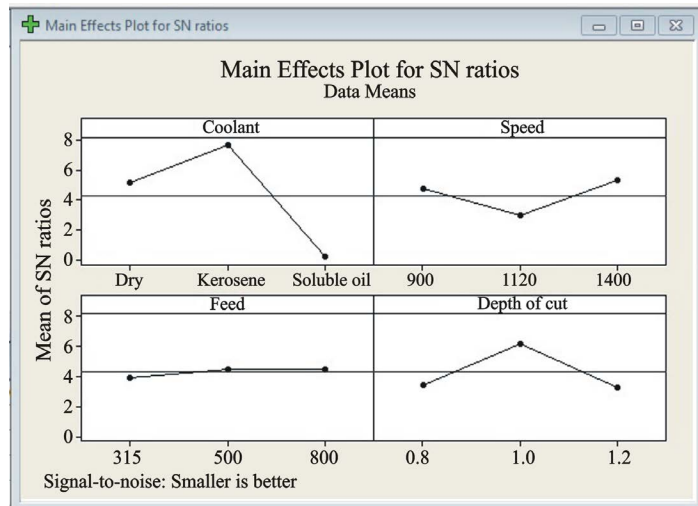


Figure 14. Main effects plot for surface roughness.

Table 8. Rank of process parameters for vibration.

Level	Coolant	Speed	Feed	Depth of cut
1	-19.58	-18.42	-19.20	-19.03
2	-19.58	-19.23	-19.40	-19.32
3	-18.90	-20.41	-19.46	-19.72
Delta	0.68	2.00	0.26	0.69
Rank	3	1	4	2

Table 9. Rank of process parameters for temperature.

Level	Coolant	Speed	Feed	Depth of cut
1	-33.17	-30.56	-31.72	-31.98
2	-31.66	-32.20	-32.02	-32.00
3	-31.31	-33.37	-32.39	-32.15
Delta	1.87	2.81	0.67	0.17
Rank	2	1	3	4

Table 10. Rank of process parameters for resultant force.

Level	Coolant	Speed	Feed	Depth of cut
1	-18.68	-17.42	-20.02	-19.62
2	-25.50	-20.38	-21.46	-21.56
3	-18.73	-25.11	-21.44	-21.73
Delta	6.83	7.69	1.44	2.11
Rank	2	1	4	3

Table 11. Rank of process parameters for surface roughness.

Level	Coolant	Speed	Feed	Depth of cut
1	5.0958	7.7189	3.9266	3.4718
2	7.6211	2.9283	4.4599	6.1405
3	0.1882	5.2580	4.5187	3.2929
Delta	7.4329	2.3297	0.5921	2.8476
Rank	1	3	4	2

4.3.2. S/N for Temperature versus Coolant, Speed, Feed and Depth of Cut

From **Figure 12** and **Table 9**, the optimum process parameters combination for individual response (Temperature)

Speed3-Coolant1-Feed3 -Depth of cut3

Which means

Speed at level 3 (1400 rpm)

Coolant at level 1 (Dry)

Feed at level 3 (800 mm/rev)

Depth of cut at level 3 (1.2 mm)

4.3.3. S/N for Resultant Force versus Coolant, Speed, Feed and Depth of Cut

From **Figure 13** and **Table 10**, the optimum process parameters combination for individual response (Resultant force) is

Speed3- Coolant2-- Feed2-Depth of cut3

Which means

- Speed at level 3 (1400 rpm)
- Coolant at level 2 (Kerosene)
- Feed at level 2 (500 mm/rev)
- Depth of cut at level 3 (1.2 mm)

4.3.4. S/N for Surface Roughness versus Coolant, Speed, Feed and Depth of Cut

From **Figure 14** and **Table 11**, the optimum process parameters combination for individual response (Surface roughness) is

- Coolant3- Depth of cut3 -Speed2- Feed1
- Which means
- Coolant at level 3 (Soluble oil)
- Depth of cut at level 3 (1.2 mm)
- Speed at level 2 (1120 rpm)
- Feed at level 1 (315 mm/rev)

4.4. Conformation Test Results

Conformation experiment is conducted for optimum parameter combination and the values of Vibrations (shown in **Figure 15** and **Figure 16**), Temperature (shown in **Figure 17** and **Figure 18**), surface roughness, and resultant forces are recorded (**Table 12**).

- According to Fuzzy based Taguchi S/N ratio analysis, the optimal combination of input parameters is Coolant = Soluble oil
- Speed = 1400 rpm
- Depth of cut = 1.2 mm
- Feed = 315 mm/rev

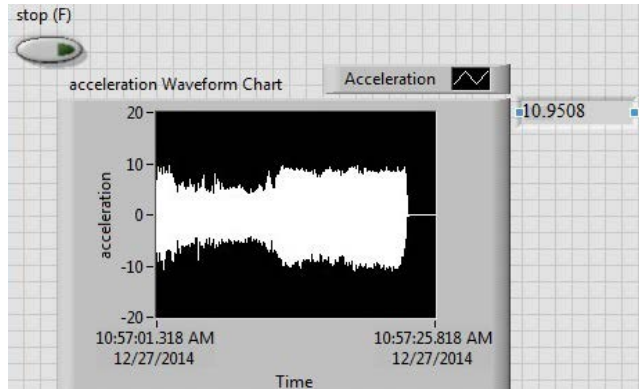


Figure 15. Lab VIEW front panel of acceleration(output).

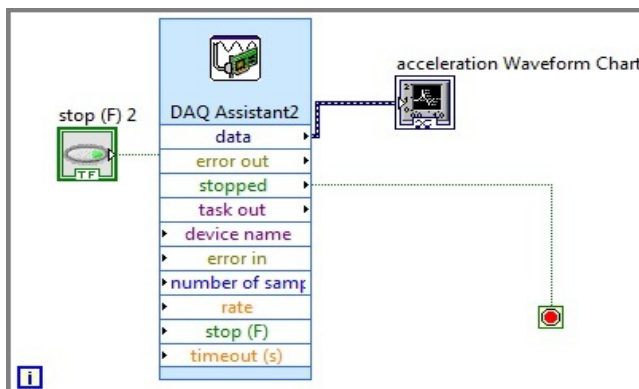


Figure 16. Lab VIEW block diagram of acceleration.

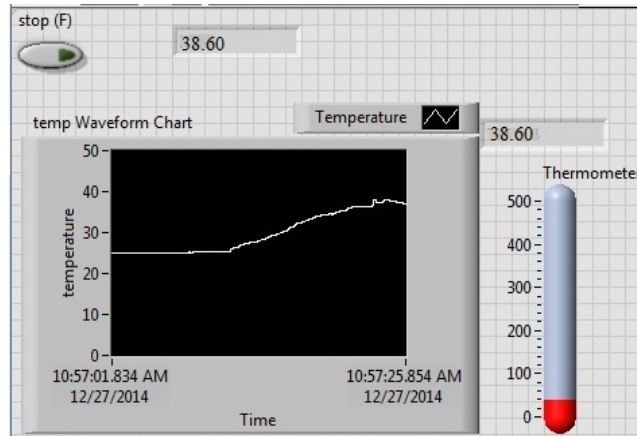


Figure 17. Lab VIEW front panel of temperature(output).

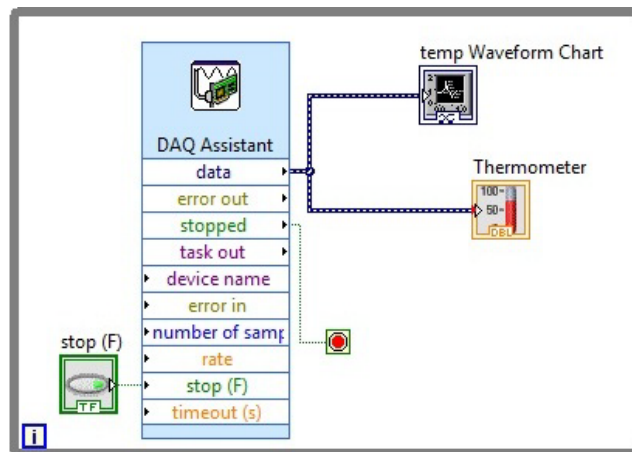


Figure 18. Lab VIEW block diagram of temperature.

Table 12. Experimental results.

Coolant	Speed (rpm)	Feed (mm/min)	Depth of cut (mm)	Resultant force (Kgf)	Vibrations (m/sec ²)	Temperature (°C)	Surface roughness (µm)
Soluble oil	1400	315	1.2	19.28	10.95	38.6	0.52

5. Conclusions

The influence of machining parameters on the multi responses is studied and the following conclusions are drawn from the results.

- 1) The order of influenced parameters found from Fuzzy-Taguchi analysis is as follows:
 - Speed (most influential);
 - Coolant (moderately influential);
 - Depth of cut (least influential);
 - Feed (very least influential).
 - 2) Taguchi analysis shows that speed has more influence on vibrations, forces and temperature and that coolant has more influence on surface roughness.
 - 3) Confirmation test has been conducted and results are satisfactory.
- However, this work can be extended further by considering the followings:
- Accuracy of predictions will be enhanced by generating more experimental data for training;
 - Tools with coated materials like Titanium, diamond, etc., are to be used in order to get the best results.
 - Use of CNC machines is for automatic adjustments of parameter values.

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