

# Optimization of Material Removal Rate during Dry Turning of EN 354 Alloy Steel Material using Taguchi Methods

Girish Kumar<sup>1</sup> and Hari Singh<sup>2</sup>

<sup>1</sup>M.tech Scholar Department of Mechanical Engineering National Institute of Technology Kurukshetra-136119, India

<sup>2</sup>Department of Mechanical Engineering National Institute of Technology Kurukshetra-136119, India

E-mail: <sup>1</sup>hsingh\_nitk@rediffmail.com, <sup>2</sup>girish.gh2009@gmail.com

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**Abstract**—This paper focuses on optimizing turning process parameters for maximizing material removal rate (MRR) using Taguchi's parameter design approach. Experiments were conducted based on Taguchi's L9 orthogonal array. Dry turning experiments are carried out on EN 354 alloy steel bars with tungsten carbide cutting tool. Three cutting parameters— Cutting Speed, Feed rate and Depth of Cut— have been selected for investigation. The signal to noise ratio (SNR) and the analysis of variance (ANOVA) are applied to optimize process parameters for maximizing material removal rate. The confirmation experiment is also conducted at the predicted optimal setting of the process parameters and the average value is found to lie within 95% confidence interval.

**Keywords:** Material Removal Rate (MRR), ANOVA, Taguchi methods, Optimization, Dry Turning, Orthogonal array

## 1. INTRODUCTION

Taguchi method is one of the most effective systems of off-line quality control where the quality is in-built at the product design stage instead of controlling it at the manufacturing stage or through the inspection of final products (Ross, 1996). Taguchi introduces his approach for designing products/processes so as to be robust to environmental conditions, designing and developing products/processes so as to be robust to component variation, and minimizing variation around a target value [1].

Akhyar et al. [2] used Taguchi method for optimization of cutting parameters in turning Ti-6%Al-4%V extra low interstitial with coated and uncoated cemented carbide tools under dry cutting condition and high cutting speed. Kaladhar et al. [3] investigated the effect of turning process parameters on surface roughness of AISI 202 austenitic steel. It was revealed that the feed is the most significant parameter influencing surface roughness.

Negrete et al. [4] optimized the cutting parameters for minimizing cutting power whereas Cayda [5] varied the cutting tool to evaluate the machinability of AISI 4340 steel. Mustafa Gunay and Emre Yucel [6] used Taguchi technique for

determining optimum surface roughness in turning of high-alloy white cast iron on CNC lathe using ceramic and cubic boron nitride (CBN) cutting tools. Dave et al. [7] studied the effect of machining conditions on MRR and surface roughness during CNC turning of different grade of EN materials using TiN-coated cutting tools. Aggarwal and Singh [8] optimized the radial and feed forces in CNC machining of P-20 tool steel material using TiN coated tungsten carbide inserts. Liu et al. [9] applied the Taguchi's parameter design approach to determine an ideal feed rate and desired force combination; the experimental results indicated that surface roughness decreases with a slower feed rate and larger grinding force.

The metal cutting studies focus on the features of tools, work material composition and mechanical properties and all the machine parameter settings that influence the process efficiency and output quality characteristics/responses. A significant improvement in process efficiency can be obtained by process parameter optimization that identifies and determines the regions of critical process control factors leading to desired outputs or responses with acceptable variations ensuring a lower cost of manufacturing. The performance of any machining process is evaluated in terms of machining rate.

## 2. SELECTION OF PROCESS PARAMETERS

The objective of this work is to obtain optimal settings of turning process parameters to yield optimal material removal rate (MRR). The selection of machining parameters was done based upon review of literature. The process parameters selected are cutting speed, feed and depth of cut.

## 3. TAGUCHI METHOD

Taguchi developed a method for designing experiments to investigate how different parameters affect the mean and variance of a process performance characteristic that defines

how well the process is functioning. The method is based on orthogonal arrays which provide a reduced variance and set of well-balanced experiments. Taguchi recommends the use of the loss function to measure the performance characteristic deviating from the desired value. The value of the loss function is further transformed into a signal-to-noise (S/N) ratio. Usually, there are three categories of the performance characteristic in the analysis of the S/N ratio, that is, the lower-the-better, the higher-the-better, and the nominal-the-better. Signal to noise ratios, the log functions of desired output quality with emphasis on variation, provide a set of well-balanced experiments to accommodate many design factors simultaneously [10].

Furthermore, a statistical analysis of variance (ANOVA) is performed to see which process parameters are statistically significant. With the S/N and ANOVA analyses, the optimal combination of the process parameters can be predicted. Finally, a confirmation experiment is conducted to verify the optimal process parameters obtained from the parameter design. In this paper, equation (1) is used to calculate S/N ratio for material removal rate. Here,  $y$  denotes the measured value of MRR in each trial and  $n$  denotes the number of observations in that trial.

$$SNR(\text{larger the better}) = -10 \log \frac{1}{n} \sum \frac{1}{y^2} \quad (1)$$

Traditional experimental methods like full factorial experiments are very complicated and difficult to implement; as they require a large number of experiments [11]. To minimize the number of tests, Taguchi developed a particular design of orthogonal arrays to study the entire parameter space with small number of experiments.

#### 4. EXPERIMENT SETUP

Turning has been done on centre lathe using tungsten carbide inserts available at Central Workshop of NIT Kurukshetra. After each experiment run a new carbide insert is used.

##### 4.1 Workpiece Material

The work material selected for the study was EN354 steel. It is used in the manufacturing of forged gears, camshaft, and other heavy-duty machinery components. Chemical composition of the work material EN354 (SAE/AISI 4320) alloy steel is given in table 1:

**Table 1: Chemical Composition of EN 354 alloy steel**

| Elements     | Carbon | Silicon | Manganese | Sulphur | Phosphorus | Chromium  | Molybdenum | Nickel    |
|--------------|--------|---------|-----------|---------|------------|-----------|------------|-----------|
| Range (in %) | 0.20   | 0.35    | 0.50-1.00 | 0.04    | 0.04       | 0.75-1.25 | 0.10-0.20  | 1.50-2.00 |

#### 4.2 Process parameters and their levels

Experiment consists of dry turning of EN 354 steel alloy on a centre lathe machine. The process parameters along with their 3 levels are given in table 2.

**Table 2**

| Process Parameters | Unit   | Level1 | Level2 | Level3 |
|--------------------|--------|--------|--------|--------|
| Speed (s)          | Rpm    | 550    | 715    | 930    |
| Feed (f)           | mm/rev | 0.05   | 0.07   | 0.10   |
| Depth of cut (d)   | Mm     | 0.4    | 0.6    | 0.8    |

#### 5. RESULTS AND ANALYSIS

Taguchi technique [12] is a powerful tool for identification of effect of various process parameters based on orthogonal array (OA) experiments which provides much reduced variance for the experiments with an optimum setting of process control parameters. In this work L9 array was used to carry out the experiment.

The response, material removal rate, was measured by varying the machining parameters and the corresponding values are shown in table 3. MINITAB version 17 software was used for analysis of variance (ANOVA).

**Table 3: Experiment Results and SNR**

| Exp. No. | CONTROL PARAMETER LEVELS |                   |                       | MRR (mm <sup>3</sup> /min) | SNR     |
|----------|--------------------------|-------------------|-----------------------|----------------------------|---------|
|          | Speed (s) (rpm)          | Feed (f) (mm/rev) | Depth of cut (d) (mm) |                            |         |
| 1        | 550                      | 0.05              | 0.4                   | 1253.09                    | 61.9596 |
| 2        | 550                      | 0.07              | 0.6                   | 2476.27                    | 67.8760 |
| 3        | 550                      | 0.10              | 0.8                   | 4612.53                    | 73.2788 |
| 4        | 715                      | 0.05              | 0.6                   | 1786.55                    | 65.0403 |
| 5        | 715                      | 0.07              | 0.8                   | 3225.01                    | 70.1706 |
| 6        | 715                      | 0.10              | 0.4                   | 2473.37                    | 67.8658 |
| 7        | 930                      | 0.05              | 0.8                   | 4354.24                    | 72.7782 |
| 8        | 930                      | 0.07              | 0.4                   | 1332.49                    | 62.4933 |
| 9        | 930                      | 0.10              | 0.6                   | 6349.71                    | 76.0551 |

**Table 4: Response for Signal to Noise Ratio**

| Level | Speed (rpm) | Feed (mm/rev) | Depth of cut (mm) |
|-------|-------------|---------------|-------------------|
| 1     | 67.70       | 66.59         | 64.11             |
| 2     | 67.69       | 66.85         | 69.66             |
| 3     | 70.44       | 72.40         | 72.08             |
| DELTA | 2.75        | 5.81          | 7.97              |
| RANK  | 3           | 2             | 1                 |

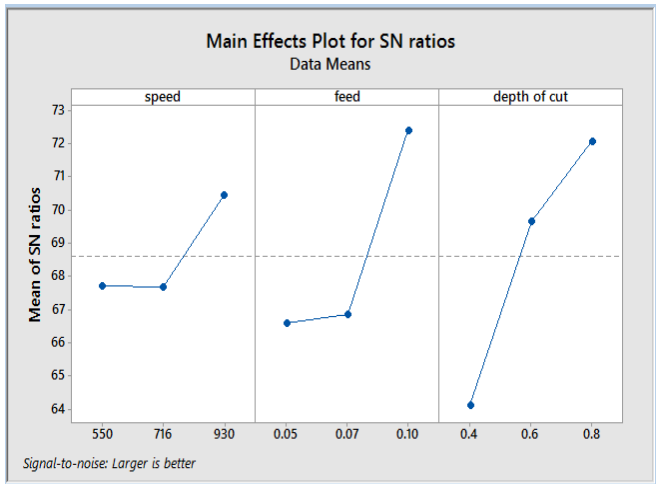


Fig. 1: Effects of process parameters on MRR (SNRdata)

Table 5: Response Table for Means

| Level | Speed(rpm) | Feed(mm/rev) | Depth of cut(mm) |
|-------|------------|--------------|------------------|
| 1     | 2781       | 2465         | 1686             |
| 2     | 2495       | 2345         | 3538             |
| 3     | 4012       | 4479         | 4064             |
| Delta | 1517       | 2134         | 2378             |
| Rank  | 3          | 2            | 1                |

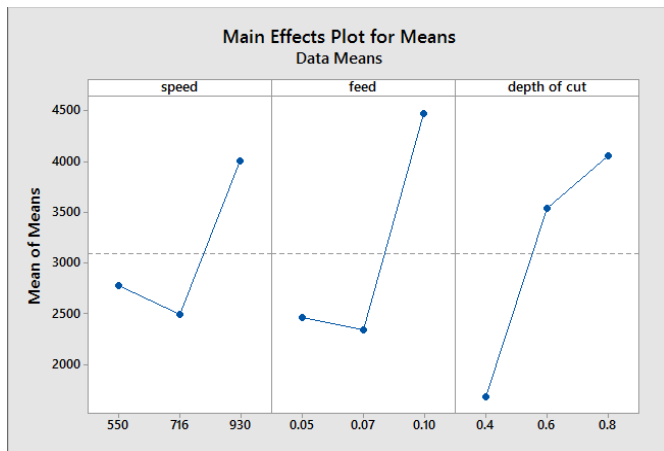


Fig. 2: Effects of process parameters on MRR (Mean data)

Table 6: ANOVA for MRR

| Source       | DF | Adj SS   | Adj MS  | F-Value | P-value |
|--------------|----|----------|---------|---------|---------|
| Speed        | 2  | 3900036  | 1950018 | 2.44    | 0.291   |
| Feed         | 2  | 8623971  | 4311985 | 5.40    | 0.156   |
| Depth of cut | 2  | 9357061  | 4678530 | 5.86    | 0.146   |
| Error        | 2  | 1597914  | 798957  |         |         |
| Total        | 8  | 23478981 |         |         |         |

S= 893.844, R-sq=93.19%, R-sq (adj) = 72.78%

6. ESTIMATION OF OPTIMUM VALUE OF MRR

The optimum value of material removal Rate (MRR) is predicted at the optimal levels of significant variables which have been selected as speed (s1), feed (f3)and depth of cut (d3).

The estimated mean of the response characteristic can be computed for material removal rate:

$$\mu_{MRR} = \bar{s3} + \bar{f3} + \bar{d3} - 2\mu \tag{3}$$

$$= 4012 + 4479 + 4064 - 2 \times 2953.69$$

$$= 6641.62 \text{ mm}^3/\text{min}$$

Where  $\mu$  = overall mean of MRR

$\bar{s3}$ ,  $\bar{f3}$  and  $\bar{d3}$  are taken from table 5.

For MRR, 95 % confidence interval of confirmation experiment[1] is calculated as:

$$CI = \sqrt{f_{0.5}(1, f_e) V_e \left[ \frac{1}{N_{eff}} \right]} \tag{4}$$

$$= \pm 3391.78$$

Where,  $f_e=2$ , is taken from ANOVA table 6,

$f_{0.5}(1, 2)= 18.512$  (taken from F- distribution table)

$$N_{eff} = N / (1 + \text{TotalDF associated in mean estimation})$$

$$= 9 / (1 + 6)$$

$$= 1.2857$$

$$V_e = \text{Error of Adj MS} = 798957$$

So the Confidence Interval is:

$$3249.84 < \mu_{MRR} (\text{mm}^3/\text{min}) < 10033.4$$

7. CONFIRMATION EXPERIMENT

In order to validate the results obtained, two confirmation experiments were conducted at optimal level of the process variables. The average value of the characteristics was obtained and compared with the predicted value. The average value of MRR was obtained  $7013.56 \text{ mm}^3/\text{min}$  which is within the 95% of confidence interval of response characteristic. It is to be pointed out that this optimal value is within the limit of process variables.

## 8. Conclusion

The following conclusions are drawn from the study:

(a) The optimal setting of process parameters in turning for maximum MRR within the selected range is as follows:

i) Depth of cut should be 0.8mm.

ii) Feed rate should be 0.1mm/rev.

iii) Speed should be 930 rpm.

(b) The predicted optimal range (95% CI) of the material removal rate is:

$$CI: 3249.84 < \mu_{MRR} (mm^3/min) < 10033.4$$

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