

Optimization of Surface Roughness and Cutting Force in MQL Hard-Milling of AISI H13 Steel

The-Vinh Do^{1(⊠)} and Nguyen-Anh-Vu Le²

¹ Thai Nguyen University of Technology, Thai Nguyen, Viet Nam thevinh8880@tnut.edu.vn
² Nha Trang University, Nha Trang, Viet Nam

Abstract. Minimum quantity lubrication (MQL) has been applied successfully to hard milling as an alternative to flood coolant processing and dry cutting. The objective of this research is to optimize process parameters to find the minimal values of surface roughness and cutting force during MQL hard milling of AISI H13 steel with coated carbide (TiAlN) cutting tool. The characteristics of the cutting force and the surface roughness obtained under MQL condition were experimentally investigated. The experimental design technique. The response surface methodology (RSM) and analysis of variance (ANOVA) were employed to analyze the influence of cutting parameters (i.e., cutting speed, feed rate, depth-of-cut and hardness of workpiece) on the cutting force and the surface roughness. The statistical models to predict cutting force and surface roughness under MQL condition were established.

Keywords: MQL \cdot Hard milling \cdot Cutting force \cdot Roughness RSM \cdot ANOVA

1 Introduction

Many studies have addressed the question "*Why perform MQL cutting*?". According to Diniz et al. [1], MQL is an acronym used to describe a procedure in which a very small amount of lubricant (<50 ml/h) is pulverized in a flow of air directed at the cutting zone during milling. MQL has been widely applied in the machining processes (i.e., milling, turning and drilling) due to efficiency and environmental issues. The effectiveness of MQL has already been demonstrated with the improvement of surface roughness [2–5], reduction of tool wear, enhancement of tool life, a decrease in cutting temperature, and a reduction in lubricant-related costs [2, 3, 5–10]. Many studies proved for the benefits of using MQL in machining in comparison with dry cutting and wet cutting methods. In the research of Dhar et al. [3], the effect of MQL on tool wear and surface roughness in turning AISI-4340 steel was significant. There was a noticeable reduction in tool wear and surface roughness by MQL due to a reduction of temperature in the cutting zone and a favorable change in the chip–tool and work–tool interaction. In comparison with wet and dry cutting [5], MQL effects using vegetable oil-based cutting fluid were

presented. The significant contributions of MQL to turning AISI 9310 alloy steel were a reduction of cutting temperature, a decrease in tool wear, and an improvement of surface roughness. Similarly, in research of Dhar et al. [6] reduction of cutting temperature was presented in turning AISI-1040 steel by employing MQL. In the milling process, the effectiveness of MQL was also demonstrated in many other studies. The tool life enhanced by an application of MQL was expressed in high-speed end milling of AISI D2 cold-worked die steel with 62 HRC in a study by Kang et al. [7], and in the research by Iqbal et al. [8]. In a study by Inconel 718 steel milling [11], Thamizhmanii et al. concluded that surface roughness obtained by using MQL is lower than that obtained by dry cutting. The tool life was improved by 43.75% by MQL rather than by dry cutting. Rahman et al. [12] concluded that the surface roughness obtained by MQL is equivalent to what was obtained through wet cooling means. The difference in cutting force between that of flood cooling and MQL was considered to be insignificant.

The application of MQL in hard-milling of AISI H13 steel has not been adequately studied to date. Consequently, the author continued to respond to the question of "*Why use MQL cutting*?". In this research, the values of surface roughness and cutting force components were collected in a series of meticulous experiments. The experimental design was performed by using the L27 orthogonal array of Taguchi's experimental design technique. The second-order models for prediction of surface roughness and cutting force under MQL were established by means of RSM. The results expressed optimal values of cutting parameters to achieve minimal values of surface roughness and cutting force.

2 Experimental Procedure

The L27 orthogonal array of Taguchi's experimental design technique was used to design the experiment. The cutting parameters are cutting-speed (v), feed-rate (f), depth-of-cut (d) and hardness-of-workpiece (h). Each parameter includes three levels (1, 2, and 3). The cutting parameters with three levels are shown in Table 1.

| Levels | Cutting parameters | | | | | |
|--------|--------------------|--------------|---------------|---------|--|--|
| | v (m/min) | F (mm/tooth) | <i>d</i> (mm) | h (HRC) | | |
| 1 | 40 | 0.01 | 0.2 | 40 | | |
| 2 | 55 | 0.02 | 0.4 | 45 | | |
| 3 | 70 | 0.03 | 0.6 | 50 | | |

Table 1. Cutting parameters with levels

A series of meticulous experiments related to the hard-milling of AISI H13 steel was conducted under MQL condition. The cutting tool used is $\Phi 10$ TiAlN coated end mill. MQL parameters applied were water soluble oil used for the lubricant, 50 ml/h for the flow rate, and 3 kg/cm² for the pressure [13]. In order to reduce the possibility for experimental errors to occur, each experiment was repeated five times.

3 Results and Discussion

Table 2 shows the results of the experiment for surface roughness and cutting force components.

| No. | v | f | d | h | Ra (µm) | Fx (N) | Fy (N) | Fz (N) |
|-----|----|------|-----|----|---------|--------|--------|--------|
| 1 | 40 | 0.01 | 0.2 | 40 | 0.151 | 46.2 | 40.4 | 6 |
| 2 | 40 | 0.01 | 0.4 | 45 | 0.202 | 136.5 | 86.7 | 15.6 |
| 3 | 40 | 0.01 | 0.6 | 50 | 0.285 | 160.5 | 187.9 | 24.5 |
| 4 | 40 | 0.02 | 0.2 | 45 | 0.181 | 84.1 | 70.7 | 10.8 |
| 5 | 40 | 0.02 | 0.4 | 50 | 0.221 | 170.3 | 182.4 | 24.8 |
| 6 | 40 | 0.02 | 0.6 | 40 | 0.308 | 190.1 | 199.6 | 27.3 |
| 7 | 40 | 0.03 | 0.2 | 50 | 0.251 | 122.9 | 105.8 | 16 |
| 8 | 40 | 0.03 | 0.4 | 40 | 0.273 | 150.5 | 136.3 | 20.1 |
| 9 | 40 | 0.03 | 0.6 | 45 | 0.405 | 241.6 | 244.9 | 34 |
| 10 | 55 | 0.01 | 0.2 | 45 | 0.142 | 51 | 46.1 | 6.8 |
| 11 | 55 | 0.01 | 0.4 | 50 | 0.209 | 132.2 | 99 | 16.2 |
| 12 | 55 | 0.01 | 0.6 | 40 | 0.163 | 120.6 | 132.9 | 17.8 |
| 13 | 55 | 0.02 | 0.2 | 50 | 0.239 | 82.9 | 62.5 | 10.2 |
| 14 | 55 | 0.02 | 0.4 | 40 | 0.207 | 116.1 | 114 | 15.9 |
| 15 | 55 | 0.02 | 0.6 | 45 | 0.254 | 149.3 | 180.1 | 23.1 |
| 16 | 55 | 0.03 | 0.2 | 40 | 0.229 | 78.7 | 80.4 | 11.1 |
| 17 | 55 | 0.03 | 0.4 | 45 | 0.334 | 139.2 | 135.8 | 19.3 |
| 18 | 55 | 0.03 | 0.6 | 50 | 0.416 | 259.7 | 201.2 | 32.4 |
| 19 | 70 | 0.01 | 0.2 | 50 | 0.107 | 44 | 58.7 | 7 |
| 20 | 70 | 0.01 | 0.4 | 40 | 0.108 | 71.2 | 60.5 | 9.2 |
| 21 | 70 | 0.01 | 0.6 | 45 | 0.126 | 104.6 | 114.2 | 15.4 |
| 22 | 70 | 0.02 | 0.2 | 40 | 0.164 | 51.7 | 38.4 | 6.3 |
| 23 | 70 | 0.02 | 0.4 | 45 | 0.214 | 94.4 | 121.9 | 15.2 |
| 24 | 70 | 0.02 | 0.6 | 50 | 0.326 | 203.8 | 216.2 | 29.5 |
| 25 | 70 | 0.03 | 0.2 | 45 | 0.25 | 60.3 | 73.6 | 9.1 |
| 26 | 70 | 0.03 | 0.4 | 50 | 0.39 | 176.2 | 162.2 | 23.5 |
| 27 | 70 | 0.03 | 0.6 | 40 | 0.305 | 131.3 | 115.5 | 17.2 |

Table 2. Experimental results for surface roughness and cutting force components.

3.1 The Analysis of Variance and the Mathematical Model

An ANOVA for surface roughness is shown in Table 3. Based on ANOVA, feed rate and depth-of-cut are the most influential variables regarding surface roughness. They contribute 52.76% and 21.79% to the total effect, respectively. On the other hand, the influences of input factors as cutting speed (v), feed rate (f), depth of cut (d) and hardness of workpiece (h) on surface roughness have statistical significance.

| Source | DF | Seq SS | Adj SS | Adj MS | F | Р | % | | |
|---------------------------------------|--------------------------------|----------|----------|----------|--------|----------------------|-------|--|--|
| ANOVA for Ra, R -Sq = 95.72% | | | | | | | | | |
| Model | 14 | 0.186438 | 0.186438 | 0.013317 | 19.18 | 0.000^{a} | 95.72 | | |
| v | 1 | 0.004576 | 0.004576 | 0.004576 | 6.59 | 0.025^{a} | 2.35 | | |
| f | 1 | 0.102756 | 0.102756 | 0.102756 | 147.97 | 0.000^{a} | 52.76 | | |
| d | 1 | 0.042438 | 0.042438 | 0.042438 | 61.11 | 0.000^{a} | 21.79 | | |
| h | 1 | 0.015961 | 0.015961 | 0.015961 | 22.98 | 0.000^{a} | 8.19 | | |
| ANOVA for Fx , R - $Sq = 96.54\%$ | | | | | | | | | |
| Model | 14 | 84499.1 | 84499.1 | 6035.7 | 23.89 | 0.000^{a} | 96.54 | | |
| v | 1 | 7409.5 | 7409.5 | 7409.5 | 29.33 | 0.000^{a} | 8.47 | | |
| f | 1 | 13535.6 | 13535.6 | 13535.6 | 53.58 | 0.000^{a} | 15.46 | | |
| d | 1 | 49057.6 | 49057.6 | 49057.6 | 194.21 | 0.000^{a} | 56.05 | | |
| h | 1 | 8716.4 | 8716.4 | 8716.4 | 34.51 | 0.000^{a} | 9.96 | | |
| ANOVA | ANOVA for Fy, $R-Sq = 95.93\%$ | | | | | | | | |
| Model | 14 | 85755.5 | 85755.5 | 6125.4 | 20.22 | 0.000^{a} | 95.93 | | |
| v | 1 | 4785.7 | 4785.7 | 4785.7 | 15.80 | 0.002^{a} | 5.35 | | |
| f | 1 | 10238.8 | 10238.8 | 10238.8 | 33.80 | 0.000^{a} | 11.45 | | |
| d | 1 | 57336.3 | 57336.3 | 57336.3 | 189.27 | 0.000^{a} | 64.14 | | |
| h | 1 | 7116.2 | 7116.2 | 7116.2 | 23.49 | 0.000^{a} | 7.96 | | |
| ANOVA for Fz , $R-Sq = 98.11\%$ | | | | | | | | | |
| Model | 14 | 1660.53 | 1660.53 | 118.61 | 44.56 | 0.000^{a} | 98.11 | | |
| v | 1 | 121.16 | 121.16 | 121.16 | 45.52 | 0.000^{a} | 7.16 | | |
| f | 1 | 228.98 | 228.98 | 228.98 | 86.03 | 0.000^{a} | 13.53 | | |
| d | 1 | 1056.47 | 1056.47 | 1056.47 | 396.90 | 0.000^{a} | 62.42 | | |
| h | 1 | 157.24 | 157.24 | 157.24 | 59.07 | 0.000^{a} | 9.29 | | |
| ^a Significant | | | | | | | | | |

Table 3. Analysis of variance for experimental values under MQL condition

On the one hand, Table 3 also shows analysis of variance for Fx, Fy, Fz. The cutting forces get affected mostly by depth of cut followed by feed rate. Depth of cut contributes 56.05%, 64.14% and 62.42% to the total effect of the factors to Fx, Fy, Fz, respectively. Feed rate's contributions are 15.46%, 11.45% and 13.53% to the total effect of the factors to Fx, Fy, Fz, respectively. On the other hand, the influences of input factors have statistical significance on the cutting force.

A mathematical model of Ra established using RSM is shown in the following Eq. (1):

 $\begin{array}{l} {\rm Ra} = 1.28831 \ - \ 0.00744864 \ v - \ 15.8793 \ f - \ 0.0482593 \ d - \ 0.0404059 \ h - \ 2.93827 {\rm e} - 005 \\ v^2 + \ 65.5556 \ f^2 \ - \ 0.0194444 \ d^2 + \ 0.000302222 \ h^2 + \ 0.156148 \ v \ \ast \ f - \ 0.00582963 \ v \ \ast \ d \\ + \ 0.000196148 v \ \ast \ h + 7.41111 \ f \ \ast \ d + 0.205778 \ f \ \ast \ h + 0.0106444 \ d \ \ast \ h \end{array}$

(1)

The coefficient of determination expresses that the mathematical model is a good model for predicting of the surface roughness under MQL condition.

Using RSM, the cutting force models are shown in the following equations:

 $\begin{array}{l} {\rm Fx} &= 1141.03\,-\,3.08148\,v-\,5162.44\,f+54.413\,d-\,45.7427\,h-\,0.00474074v^2-\,32333.3f^2-263.75\,d^2+\,0.415333\,h^2-\,30.2222\,v*f-\,2.3763v*d+\,0.084563\,v*h+\,4465.56\,f*d+201.644\,f*h+\,10.2\,d*h \end{array}$

 $\begin{array}{l} {\rm Fy} \ = \ 527.606 \ - \ 7.87711 \ v + \ 6821.74 \ f + \ 157.309 \ d - \ 18.8253 \ h + \ 0.0276296 \ v^2 \ - \ 160833 \ f^2 \ - \ 39.5833 \ d^2 \ + \ 0.102 \ h^2 \ - \ 45.3185 \ v \ * \ f \ - \ 4.3437 \ v \ * \ d \ + \ 0.142104 \ v \ * \ h \ - \ 832.222 \ f \ * \ d \ + \ 107.156 \ f \ * \ h \ + \ 9.15778 \ d \ * \ h \end{array}$

(2)

 $Fz = 116.209 - 0.733062 v + 147.333 f + 12.3204 d - 4.52904 h + 0.00145679 v^{2} - 13888.9 f^{2} - 20.9722 d^{2} + 0.0364444 h^{2} - 5.55556 v * f - 0.459259 v * d + 0.015437 v * h + 251.111 f * d + 21.5556 f * h + 1.4 d * h$ (4)

The coefficient of determinations expresses that the mathematical models are good models for predicting of the cutting force component under MQL condition.

3.2 Optimization of Surface Roughness and Cutting Force

Figure 1(a) shows the result of optimization for surface roughness. The optimal cutting parameters are 70 m/min for the cutting-speed, 0.01 mm/tooth for the feed-rate, 0.2 mm in the depth-of-cut, and 40 HRC for workpiece-hardness. According to these optimal cutting parameters, the minimum surface roughness of the MQL condition is 0.0868 μ m, with a desirability value of 1.000. The result indicates that under MQL-cutting conditions, a higher cutting-speed, lower feed-rate, lower depth-of-cut, and lower workpiece-hardness will lead to optimal surface roughness.



Fig. 1. The optimization plot of surface roughness (a) and cutting-force components (b)

Figure 1(b) shows the result of optimization for cutting-force components. The optimal cutting parameters are 63.636 m/min for the cutting- speed, 0.01 mm/tooth for the feed-rate, 0.2 mm in the depth-of-cut and 40 HRC for workpiece-hardness under MQL conditions. According to these optimal cutting parameters, a higher cutting-speed, lower feed-rate, lower depth-of-cut and lower workpiece-hardness will lead to minimal cutting force.

4 Conclusions

Under MQL conditions, feed rate and depth-of-cut are the most influential variables related to surface roughness. The feed rate contributes 52.76% and depth-of-cut contributes 21.79% to the total effect.

The cutting force components are principally affected by depth-of-cut, followed by the feed rate. Depth-of-cut contributes 56.05%, 64.14% and 62.42% to the total effect of the factors to Fx, Fy, and Fz, respectively. The feed rate's contributions are 15.46%, 11.45% and 13.53% to the total effect of the factors to Fx, Fy, and Fz, respectively.

A higher cutting speed, lower feed rate, lower depth-of-cut and lower workpiece hardness realized better surface roughness and minimum cutting force.

References

- 1. Diniz, A.E., Micaroni, R.: Cutting conditions for finish turning process aiming: the use of dry cutting. Int. J. Mach. Tools Manuf. **42**(8), 899–904 (2002)
- Weinert, K., et al.: Dry machining and minimum quantity lubrication. CIRP Ann. Manuf. Technol. 53(2), 511–537 (2004)
- Dhar, N., Kamruzzaman, M., Ahmed, M.: Effect of minimum quantity lubrication (MQL) on tool wear and surface roughness in turning AISI-4340 steel. J. Mater. Process. Technol. 172 (2), 299–304 (2006)
- Hwang, Y.K., Lee, C.M.: Surface roughness and cutting force prediction in MQL and wet turning process of AISI 1045 using design of experiments. J. Mech. Sci. Technol. 24(8), 1669–1677 (2010)
- Khan, M., Mithu, M., Dhar, N.: Effects of minimum quantity lubrication on turning AISI 9310 alloy steel using vegetable oil-based cutting fluid. J. Mater. Process. Technol. 209(15), 5573–5583 (2009)
- Dhar, N., et al.: The influence of minimum quantity of lubrication (MQL) on cutting temperature, chip and dimensional accuracy in turning AISI-1040 steel. J. Mater. Process. Technol. **171**(1), 93–99 (2006)
- Kang, M., et al.: Effect of the minimum quantity lubrication in high-speed end-milling of AISI D2 cold-worked die steel (62 HRC) by coated carbide tools. Surf. Coat. Technol. 202 (22), 5621–5624 (2008)
- Iqbal, A., He, N., Li, L.: Empirical modeling the effects of cutting parameters in high-speed end milling of hardened AISI D2 under MQL environment. In: Proceedings of the World Congress on Engineering, London, UK (2011)
- Duchosal, A., et al.: An experimental investigation on oil mist characterization used in MQL milling process. Int. J. Adv. Manuf. Technol. 66(5–8), 1003–1014 (2013)

- 10. Rahim, E.A., Dorairaju, H.: Evaluation of mist flow characteristic and performance in Minimum Quantity Lubrication (MQL) machining. Measurement **123**, 213–225 (2018)
- 11. Thamizhmanii, S., Rosli, S.H.: A study of minimum quantity lubrication on Inconel 718 steel. Arch. Mater. Sci. Eng. **39**(1), 38–44 (2009)
- 12. Rahman, M., Kumar, A.S.: Evaluation of minimal of lubricant in end milling. Int. J. Adv. Manuf. Technol. **18**(4), 235–241 (2001)
- 13. Do, T.-V., Hsu, Q.-C.: Optimization of minimum quantity lubricant conditions and cutting parameters in hard milling of AISI H13 steel. Appl. Sci. 6(3), 83 (2016)