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Semi empirical modeling of cutting temperature and surface roughness in turning of engineering materials with TiAlN coated carbide tool

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Introduction. In manufacturing, obtaining a given surface roughness of the machined parts is of great importance to fulfill functional requirements. However, the surface roughness significantly affected by the heat generated during the machining process, which can lead to a decrease in dimensional accuracy. The surface roughness significantly affects the fatigue characteristics of the part, and the service life of the cutting tool is determined by the cutting temperature generation. The purpose of the work. The purpose of this study is to create semi-empirical models for predicting surface roughness and temperature of various work materials. Enhanced cutting performance is achieved by accurately determining the cutting temperature in the machined zone. However, calculating the cutting temperature for each specific case is fraught with difficulties in terms of labor resources and financial investments. This paper presents a comprehensive empirical formula designed to predict both theoretical temperature and surface roughness. Methodology, The performance of the surface roughness and temperature generation was evaluated for the EN 8, Al 380, SS 316 and SAE 8620 materials when processed with TiAlN-coated carbide tools. The TiAlN coating was obtained by Physical Vapor Deposition (PVD) technique. Response surface methodology was used to prepare predictive models. Cutting speed (from 140 to 340 m/min), feed (from 0.08 to 0.24 mm/rev) and depth of cut (from 0.6 to 1 mm) were used as input parameters to measure the characteristics of all materials in terms of surface roughness and cutting temperature. The tool-work thermocouple principle was used to measure the temperature at the chip-tool interface. Novel Calibration Setup was developed to establish the relationship between the Electromotive Force (EMF) generated during machining and the cutting temperature. Results and Discussion. It is observed that the energy required for mechanical processing was largely converted into heat. The highest cutting temperature is recorded with SS 316, followed by SAE 8620 and EN 8. However, low temperature was reported during machining of Al 380 and it was mainly governed by the thermal conductivity of the material. The lowest surface roughness is observed for SAE 8620, EN 8, followed by SS 316 and Al 380. The semi-empirical method and regression model equations are in good agreement with each other. Statistical analysis of the nonlinear evaluation reveals that cutting speed, feed rate, and material density have a greater influence on the surface roughness, whereas depth of cut has a greater influence on the temperature change. The study will be very useful for predicting industrial performance when machining EN 8, Al 380, SS 316 and SAE 8620 materials with TiAlN-coated carbide tools.

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List of symbols

Symbol	Description					
f	Feed (mm)					
V _c	Cutting speed (m/min)					
doc	Depth of cut (mm)					
R _a	Surface roughness (µm)					
MRR	Material removal rate (mm ³ /rev)					
HSM	High speed machining					
F _c	Cutting forces (N)					
ρ	Density (kg/m ³)					
C _p	Specific Heat (J/kg k)					
K	Thermal Conductivity(W/mk)					
σ	Yield strength (N/m ²)					
α	Coefficient of thermal expansion (m/mk)					
θ	Temperature (°C)					
SS 316	Stainless steel SS 316					
SAE 8620	Low alloy case hardening steel SAE 8620					
EN 8	Engineering steel EN 8					
Al 380	Aluminium alloy <i>Al 380</i>					
Ø	Buckingham's π theorem constant					
$a_1 a_2 a_3 a_4 a_5$	Power indices					
$\boldsymbol{b}_1 \boldsymbol{b}_2 \boldsymbol{b}_3 \boldsymbol{b}_4 \boldsymbol{b}_5$	Power indices					
ΜLΤΘ	Dimensions					
CBN	Cubic boron nitride					
RSM	Response surface methodology					
ССД	Central composite design					
ANN	Artificial neural network					
LM	Levenberg-Marquardt					

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Introduction

Surface finish is critical to quality because it directly affects the appearance, functionality, and performance of machined components. Precision machining is essential, especially in aerospace and medical applications where specified surface finish is required to reduce friction, improve wear resistance, or improve corrosion resistance. The influence of surface finish on tribological parameters such as friction and lubrication is crucial to achieve maximum performance and durability. Increased temperatures during machining have a significant impact on tool wear, material integrity and dimensional accuracy. Temperature control is critical for extending tool life and maintaining the structural integrity of machined parts. Predictive modelling optimizes processes by identifying optimal parameters for cost savings through increasing tool life, reducing scrap rates and increasing efficiency. The use of cutting fluid in hard turning is not recommended, since at elevated temperatures when processing materials with a hardness of 48 to 68 HRC, the coolant in the cutting zone begins to boil. This boiling phenomenon promotes thermal deformation, thereby reducing both R_{a} (surface roughness) and the service life of the cutting tool [1]. In case of machining different materials, its machinability was evaluated using various process parameters such as tool life, material removal rate (MRR), cutting force (F_c), energy consumption, chip morphology and machined surface roughness (R_a). Using high speed machining (HSM) while maintaining surface integrity and maintaining tolerance limits requires optimal coordination of factors such as cutting force (F_c) , process and machine parameters. The right combination of these elements is critical to increasing the efficiency of HSM without compromising the quality of the machined surfaces or exceeding specified tolerance limits. This balance ensures that machining can proceed without compromising accuracy and surface quality, contributing to the overall success of high-speed machining operations [2].

Zhao et al., [3] measured the cutting temperature of *Inconel 718* using a two-color infrared thermometer with a ceramic whisker-reinforced tool, and concluded that the large amount of heat generated during machining deteriorates the surface quality of the machined material. Due to the increase in temperature in the cutting zone during machining, the surface quality deteriorated [4]. High tool wear and temperature increase during machining of hardened *AISI 4340* steel can be eliminated using bio-cutting fluid [5]. Postmachining operations are required to improve the surface quality of superalloys, [6]. *Kumar et al.* [7] compared a *RSM* model with an *ANN* model to analyze the turning performance of *AISI D2* steel and concluded that that the *RSM*-based prediction model is more accurate than the *ANN* model for predicting surface quality and cutting temperature. *Gosai* and *Bhavsar* [8] used mathematical models and equations generated by *CCD*-based *RSM* to predict cutting temperature.

The material removal rate during the turning process was higher compared to other traditional machining processes. Abhang et al. [9] experimentally measured the temperature of the EN 31 alloy during turning with tungsten carbide inserts using the natural thermocouple technique. F has a significant effect on the surface roughness: as the f increases, the roughness increases, and as the V_c increases, the roughness decreases [10– 12]. Bhopal et al. [13] used RSM with CCD for turning austenized high-strength cast iron with a carbide tool and found that V_c has a more significant effect on surface roughness. Aouici et al. [14] used a CBN tool for turning AISI H11 steel, as well as a mathematical model based on RSM for Ra and F_c , however, when processing materials reinforced with particles, the surface morphology was changed. Longbottom and Lanham [15] conducted a review of temperature measuring devices and found that the measured temperature varied in different places. Korkut et al. [16] compared the ANN model and the RA model and found that the training ANN model with the LM algorithm demonstrated a higher prediction rate and was useful in measuring the cutting temperature when tested by a qualified RA method during machining. Dhar and Kamruzzaman [17] found that an increase in temperature significantly affects tool wear and surface roughness, and the use of cryogenic cooling gives good results. Patil and Brahmankar [18] developed a model for surface roughness that takes into account the input parameters, material properties, size of ceramic particles and its volume fraction, and found that the volume fraction and particle size significantly affect the output parameters, as well as that the presence of ceramic particles affects the surface roughness. Patel and Kiran [19] used a linear regression model to analyze the assessment of the roughness of the surface





when processing AISI 1040 steel. Patel and Gandhi [20] machined AISI D2 steel with an CBN tool and developed a mathematical model based on the simultaneous action of f, V_c and the nose radius, and is in good agreement with the experimental values. But none of them used more than one material for experiments, with the exception of Rodriguez et al., [21] who used SS 304, 316L and 420 materials for turning and developed a model of cutting temperature taking into account thermal conductivity and maximum strength. According to the literature reviewed, the cutting parameters, in particular the cutting speed and feed, have a significant effect on the temperature of the chip-tool contact surface. Various predictive models have been developed, but each model predicted results in a specific parameter range. In addition, several studies have been reported on the effect of TiAlN cutting modes and coating parameters on cutting temperature and surface roughness when turning EN 8, Al 380, SS 316 and SAE 8620 materials. In this work, the simplest and most economical method for measuring temperature is developed, involving the use of a tool-work thermocouple. Further, response surface models were developed for the cutting temperature and roughness of these materials, the influence of technological parameters are studied, and a semi-empirical model is developed to predict the cutting temperature and surface roughness.

Materials and methods

The experimental results were obtained on a CNC lathe machine. V_c , f and doc were the three adjustable factors in turning operation. In the present work, workpieces made of four materials were used, namely mild steel (EN 8) with a diameter of 75 mm, aluminum alloy (Al 380) with a diameter of 50 mm, stainless steel (SS 316) with a diameter of 75 mm and low alloy steel (SAE 8620) with a diameter of 75 mm. The length of each workpiece was 300 mm and each of it was machined. To determine the chemical composition of the above materials, spectroscopic analysis was carried out, the results of which are presented in Table 1. Since the literature indicates that TiAlN-coated carbide tools have minimal R_a and tool wear, Sandvik PVD (TiAlN) coated carbide inserts CNMG-120408 MS PR1310 (0.8 mm nose radius) with eight cutting edges were used in this work for 20 tests under dry conditions. The contact point between the tool and the workpiece was hot during machining, while the carbon brush touching the workpiece remained cold. The workpiece was mounted in a three-jaw chuck, and insulation was provided between the workpiece and the chuck. The experimental setup, temperature calibration setup, and workpiece material are shown in figure 1, a, b and c respectively. The cutting parameters used for machining are given in Table 2.

Table 1

		<u>^</u>		
Element, %	SS 316	EN 8	SAE 8620	Al 380
С	0.07	0.39	0.22	_
Mn	0.16	0.87	0.8	0.5
Si	0.9	0.22	0.28	8.5
Р	0.05	0.04	0.031	—
S	0.02	0.05	0.04	_
Cr	18.50	_	0.49	_
Мо	2.25	—	0.22	—
Ni	12.23	—	0.52	0.5
Mg	—	—	—	0.1
Си	—	—	—	3.6
Sn	—	—	_	0.35
Zn	—	_	—	3
Fe	balance	balance	balance	1.3
Al	_	_	_	balance

Chemical composition of work material

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Fig. 1. Machining Setup (a), Temperature calibration setup (b), Work materials (c)

Table 2

C

Parameters/Levels	L 1	L 2	L 3	L 4	L 5
V_c (m/min)	140	190	240	290	340
f(mm/rev)	0.08	0.12	0.16	0.20	0.24
doc (mm)	0.6	0.7	0.8	0.9	1.0

Process parameters and experimental levels

Results and discussion

The central composite design of the response surface method was used for the main experiments. Table 3 shows the experimental results. The objective of the experimental analysis was to identify the significant factor that has a greater influence on the response variables and to develop a generalized empirical model to predict surface roughness and generated temperature using *Buckingham's* π theorem. Statistical analysis of surface roughness and temperature rise was carried out using RSM.

The main objective of this paper is to develop semi-empirical formulae using the Levenberg-Marquardt method to predict the surface roughness and temperature of various materials. Using the values from Table 2, individual regression equations were constructed and the full factorial values were extracted from the regression. These full factorial values are used to derive the semi-empirical formula.

The regression equations for surface roughness of materials are given below.

$$SSR_{a} = 0.60 + 0.00018V_{c} + 2.7f - 1.37d - 0.000003V_{c}^{2} + 19.03f + 0.79d^{2} - 0.0050V_{c}xf + 0.00050V_{c}xd + 1.87fxd;$$
(I)

$$SAER_a = 0.31 - 0.00202V_c + 10.01f - 1.20d - 0.00005V_c^2 + 31c61f^2 - 0.00005V_c^2 + 0.00005V_$$

$$0.11d^2 - 0.2604V_c xf + 0.00908V_c xd - 5.1fxd;$$
(II)

$$ENR_a = 3.135 - 0.01331V_c - 9.76f - 1.09d + 0.000023V_c^2 + 59.66f^2 +$$

 $0.670d^2 - 0.00312V_c xf + 0.00125V_c xf + 0.31f xd;$ (III)

$$AIR_{a} = 14.32 - 0.0478V_{c} - 12.4f - 12.97d + 0.000093V_{c}^{2} + 53.7df^{2} + 7.97d^{2} - 0.0444V_{c}xf - 0.0027V_{c}xd + 16.6fxd.$$
(IV)

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Table 3

Experimental data of Ra and temperature for SS 316, EN 8, SAE 8620 and Al 380 materials

r											
Run	Speed,	Feed,	doc,	SS	EN	SAE	AL	SS	EN	SAE	Al
No	V_c ,	<i>f</i> , (mm/	<i>d</i> ,	316	8	8620	380	316	8	8620	380
INU.	(m/min)	rev)	(mm)	R_a	R_a	R_a	R_a	Temp	Temp	Temp	Temp
1	190	0.12	0.7	0.73	0.84	0.63	2.88	635	636	629	243
2	290	0.12	0.7	0.56	0.66	0.50	1.73	812	657	733	264
3	190	0.2	0.7	1.39	1.54	1.60	3.56	643	654	648	247
4	290	0.2	0.7	1.22	1.31	1.25	2.24	997	672	741	318
5	190	0.12	0.9	0.74	0.92	0.55	2.95	782	647	675	236
6	290	0.12	0.9	0.62	0.74	0.59	1.93	1082	665	782	271
7	190	0.2	0.9	1.47	1.6	1.42	4.08	815	664	735	274
8	290	0.2	0.9	1.27	1.42	1.27	2.52	1157	679	818	334
9	140	0.16	0.8	1.08	1.32	1.12	4.25	732	644	595	229
10	340	0.16	0.8	0.78	1.03	0.80	1.86	1243	689	837	323
11	240	0.08	0.8	0.3	0.59	0.47	2.01	619	629	625	216
12	240	0.24	0.8	1.86	2.06	1.96	2.92	883	666	718	306
13	240	0.16	0.6	0.91	0.92	0.98	2	646	644	693	289
14	240	0.16	1	1.07	1.02	1.04	2.88	1082	653	791	310
15	240	0.16	0.8	1.01	0.95	0.99	2.12	805	649	704	283
16	240	0.16	0.8	0.92	1	0.96	2.24	766	642	694	291
17	240	0.16	0.8	0.93	0.94	1.00	2.31	775	644	699	293
18	240	0.16	0.8	0.99	0.94	1.00	2.09	764	645	701	296
19	240	0.16	0.8	0.96	0.94	1.00	2.1	769	644	703	298
20	240	0.16	0.8	0.98	0.95	1.00	2.08	765	643	701	297

The regression equations for material temperature are given below.

$$SSTemp = 3,517 - 2.74V_c + 696f - 8645d + 0.01054V_c^2 + 3,963f^2 + 699d^2 + 6.6V_c xf - 1.57V_c xd - 3,281fxd;$$
(V)

$$SAETemp = 1,073 + 0.57V_c + 457f - 1,899d + 0.00210V_c^2 - 3,672f^2 + 1,175d^2 - 2.14V_c xf - 0.175V_c xd + 2,156fxd;$$
(VI)

$$ENTemp = 748 - 0.787V_c + 87f - 175d + 0.002436V_c^2 + 838d^2 + 159f^2 - 0.375V_c xf - 0.150V_c xd - 63fxd;$$
(VII)

$$AlTemp = 239 + 0.579V_c + 39f - 353d - 0.001918V_c^2 - 5341d^2 + 108f^2 + 4.69V_cxf - 0.075V_cxd + 1,344fxd.$$
 (VIII)

Buckingham's π theorem

This study uses the dimensional homogeneity principle of *Buckingham's* π theorem [22]. Table 4 shows the mechanical properties of the materials.

X7	T.T., 14	Sym-	Dimen- sions	Workpiece properties			
variable	Unit	bol		SS 316	EN 8	SAE 8620	A1 380
Feed	mm	f	L	_	_	_	_
Speed	m/min	V _c	LT^{-1}	_	_	_	_
Depth of cut	mm	d	L	_	_	_	_
Surface roughness	μm	R _a	L	_	_	_	_
Density	kg/m ³	Р	ML^{-3}	8,000	7,850	7,845	2,760
Specific Heat	J/kg k	C_p	$L^2 T^2 \Theta^{-1}$	0.5	0.475	1.6	0.963
Thermal Conductivity	W/mk	K	$MLT^{-3}\Theta^{-1}$	16.3	46.6	27	109
Yield Strength	N/m ²	σ	$M^{-1} T^{-2}$	240	560	450	159
Coeff. Of Thermal Exp.	m/m×K	α	$L \Theta^{-1}$	16.18×10^{-10}	12.2×10^{-6}	11.6×10 ⁻⁶	12.1×10 ⁻⁶
Temperature	°C	θ	θ	1.371	2.600	1.400	650

Units, dimensions and properties of the machined materials

Quantities of different nature cannot be homogeneous. Applying dimensional analysis, surface roughness can be given by an equation of the following form,

$$R_a = f(F, V, D, \theta, \sigma, K, C_p, \rho, \alpha), \qquad (1)$$

where the fundamental dimensions are ρ , *L*, *T* and Θ . Therefore, since the total number of variables is ten, there are four fundamental dimensions.

The number of dependent and independent variables is n = 10, and the number of repeated variables is m = 4. Therefore, none of the π terms in the present study will be n-m = 6. Thus,

$$f(\pi_1, \pi_2, \pi_3, \pi_4, \pi_5, \pi_6) = 0.$$
⁽²⁾

Note that equation (2) can also be written as:

$$\pi_1 = f(\pi_1, \pi_2, \pi_3, \pi_4, \pi_5, \pi_6); \qquad (3)$$

$$\pi_1 = R_a / F ; \tag{4}$$

$$\pi_2 = \left(\frac{C_p \theta}{V^2}\right)^{a_1}; \tag{5}$$

$$\pi_3 = \left(\frac{K\theta}{FV^3\rho}\right)^{a_2};\tag{6}$$

$$\pi_4 = \left(\frac{\alpha\theta}{F}\right)^{a_3};\tag{7}$$

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$$\pi_5 = \left(\frac{\sigma}{V^2 \rho}\right)^{a_4};\tag{8}$$

$$\pi_6 = \left(\frac{D}{F}\right)^{a_5}.\tag{9}$$

Therefore, the final form of the equations can be written as

$$R_a = \emptyset \cdot F\left(\frac{C_p \theta}{V^2}\right)^{a_1} \left(\frac{K\theta}{FV^3 \rho}\right)^{a_2} \left(\frac{\alpha \theta}{F}\right)^{a_3} \left(\frac{\sigma}{V^2 \rho}\right)^{a_4} \left(\frac{D}{F}\right)^{a_5}.$$
 (10)

Similarly, the temperature increase (T) can be given by an equation of the following form:

$$\theta = f(F, V, \sigma, K, C_p, \rho, \alpha); \tag{11}$$

$$\pi_1 = \alpha \theta / F \,; \tag{12}$$

$$\pi_2 = \left(\frac{FC_p}{\alpha V^2}\right)^{b_1};\tag{13}$$

$$\pi_3 = \left(\frac{K}{\alpha V^2 \rho}\right)^{b_2};\tag{14}$$

$$\pi_4 = \left(\frac{R_a}{F}\right)^{b_3};\tag{15}$$

$$\pi_5 = \left(\frac{\sigma}{V^2 \rho}\right)^{b_4}; \tag{16}$$

$$\pi_6 = \left(\frac{D}{F}\right)^{b_5}.\tag{17}$$

Thus, the final form of the equation can be written as

$$\theta = F/a \cdot \mathcal{O}\left(\frac{FC_p}{\alpha V^2}\right)^{b_1} \left(\frac{K}{\alpha V^3 \rho}\right)^{b_2} \left(\frac{Ra}{F}\right)^{b_3} \left(\frac{\sigma}{V^2 \rho}\right)^{b_4} \left(\frac{D}{F}\right)^{b_5}.$$
(18)

Although α appears repeatedly, its' influence on R_a appears to be quite significant. In this work, energy indicators are determined using the *Levenberg-Marquardt* method (see Table 5). The adequacy of the model is further analyzed by comparing the regression of R_a and the predicted values of the semi-empirical model.

Table 5

Energy indicators	Surface roughness	Energy indicators	Temperature
Ø	1.687688	Ø	0.098376
<i>a</i> ₁	0.118057	<i>b</i> ₁	-0.186434
a2	0.322659	b ₂	-0.384552
<i>a</i> ₃	-0.591654	b ₃	-0.177437
<i>a</i> ₄	-0.272547	<i>b</i> ₄	0.407445
<i>a</i> ₅	0.548434	<i>b</i> ₅	0.660121

Coefficients and energy indicators of R_a and temperature model

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Comparison of surface roughness of SS 316, EN 8, SAE 8620 and Al 380

To obtain a complete understanding of the influence of input parameters on surface roughness, threedimensional (3D) surface diagrams were constructed for all cutting materials by varying process parameters. These visual representations use empirically derived equations to ensure accuracy. Figure 2 shows threedimensional surface diagrams illustrating the surface roughness changes during turning of *SS 316*, *EN 8*, *SAE 8620* and *Al 380* with *PVD*-coated (*TiAlN*) tools generated using Eqs. (I)–(IV).

From figure 2 it becomes clear that the surface roughness is primarily affected by the feed. However, this effect can be considered to be more significant for *Al 380* and *SS 316*. During the processing of aluminum alloys, built-up edges are formed due to the adhesion of chips to the cutting tool, which leads to an increase in surface roughness. In the case of *SS 316*, there is a tendency for the formation of drain chips that spin around the work material, damaging the new surface, and this may be the cause of poor surface finish. *EN 8* and *SAE 8620* materials seem well suited for machining, mainly due to their low hot hardness and easy machinability. Therefore, the roughness of these materials is higher compared to others. It was also observed that as cutting speed increases, there is a tendency for surface roughness to improve for all materials. The literature reports that at high cutting speeds, the tool-chip contact length is reduced, thereby minimizing cutting tool vibration and improving surface roughness. In addition, at higher speeds, the cutting temperature increases; this contributes to the softening of the material. This in turn helps reduce cutting forces, thereby minimizing vibration and improving surface finish.

Figure 3, *a* shows the effect of f on R_a at $V_c = 140$ m/min and doc = 0.6 mm for both regression and semiempirical values. Aluminum material has poor surface finish because aluminum produces more continuous chips than other materials. In addition, this continuous chip damages the finished parts [23].

Figure 3, b shows the effect of f on R_a at $V_c = 190$ m/min and doc = 0.7 mm. As f increases, R_a increases compared to other materials, the thermal conductivity of SS 316 is lower, due to the increase in temperature,



Fig. 2. Surface roughness 3D-plots for SS 316 (a), EN 8 (b), SAE 8620 (c) and Al 380 (d)



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Fig. 3. Effect of feed rate on surface roughness at different cutting speed and depth of cut for all materials using *TiAlN*-coated tool

the material becomes more ductile when cutting and a smoother cut is possible, which leads to better surface quality [11]. The minimum R_a is achieved by increasing V_c from 240 m/min to 340 m/min and *doc* from 0.8 mm to 1 mm, as shown in figure 3 c–e, since at higher V_c the strain rate in the shear zone is expected to be high, which will lead to an increase in temperature [2]. As V_c and f increase, the temperature increases because the heat dissipation time decreases and the larger chip-tool contact area increases friction. V_c and *doc* are significant factors in increasing tool temperature for SS 316 and SAE 8620. R_a decreases due to increasing strain rate [24].

Figure 4, a-e clearly shows that higher Vc provides good surface roughness for almost all materials. However, as f and doc increase, the surface roughness increases first for SS 316 and then for Al 380. EN 8 shows even better results due to low heat generation in the cutting zone, which maintains tool shape stability. Since the thermal conductivity of SS 316 is lower compared to other materials, it becomes more ductile during cutting due to increased temperature, and a smoother cut is possible due to better surface quality [2]. R_a was found to be the worst when machining Al 380 and was superior to SS 316 and SAE 8620. The sticking of Al 380 material results in a rough surface. Built-up edge occurs because the material easily adheres to the cutting edge, which ultimately changes the geometry of the tool and Ra increases [12].

Figure 5 a-e shows the effect of *doc* on various materials. It is observed that *doc* does not have a significant effect on R_a . This may be due to the increase in strain volume with increasing *doc*. Thus, severe deformation of the workpiece results in more surface irregularities and hence poor surface quality. *Zou et al.* [25] also obtained similar results. *Doc* is less significant for *Ra* than *Vc* and *f*[11]. At higher values of technological parameters, the thermal wear of the tool and surface roughness increase [3].

Comparison of cutting temperatures of SS 316, EN 8, SAE 8620 and Al 380

To obtain a comprehensive understanding of the influence of input parameters on cutting temperature, three-dimensional (3-D) surface plots are constructed by varying the process parameters for all cutting materials. These visual representations use empirically derived equations to ensure accuracy. Figure 6 shows three-dimensional diagrams illustrating the cutting temperature changes during turning of *SS 316, EN 8, SAE 8620* and *Al 380* stainless steels for *PVD*-coated (*TiAlN*) tools obtained using equations (V)–(VIII).



Fig. 4. Effect of cutting speed on surface roughness at different feed rate and depth of cut for all materials using *TiAlN*-coated tool



Fig. 5. Effect of depth of cut on surface roughness at different feed rate and cutting speed for all materials using *TiAlN*-coated tool

In the case of cutting temperature, f does not have a significant effect (see figure 6, a-d). Compared to other materials, Al 380 exhibits a less significant temperature increase. In other materials such as SS 316, SAE 380 and EN 8, the temperature rise is linear, low thermal conductivity and specific heat capacity are responsible for the large variations in temperature rise in SS 316. Consequently, the temperature during





Fig. 6. Cutting temperature 3D-plots for SS 316 (a), EN 8 (b), SAE 8620 (c) and Al 380 (d)

processing of SS 316 increases as the process parameters increase. High-speed, high-temperature processing results were obtained with increasing V_c . Most of the heat is carried away by the chips, and little heat is lost into the workpiece. It can be seen that f affects the temperature slightly, but gradually the temperature continues to increase as f increases. The same result was obtained by *Dessoly et al.* [26] using a *FEM* model and an *IR* camera. Figure 7 a, b shows that with increasing f the temperature increases, since a larger surface area of the workpiece and the tool is in contact. Aluminum has the lowest yield strength, so heat generation in aluminum is less compared to other materials.

Figure 7, c-e shows how the temperature increases with increasing f, doc, and V_c increases. Increasing f increases the temperature due to greater chip-tool contact and associated friction [27]. In aluminum, the temperature rises to a lesser extent because due to higher thermal conductivity, heat transfer occurs faster, so the material remains in the same state throughout, the material does not become more ductile, and the friction between the workpiece and the cutting tool is reduced [12]. As the process parameters increase, the temperature increases. *Kitagawa et al.* [28] used ceramic tools to turn *Inconel 718* and found that the cutting temperature continued to increase with increasing process parameters as the workpiece material was deformed into chips by the cutting tools. Deformation of the workpiece, cohesion or friction of the chips on the rake surface of the tool leads to strong heating [3].

As V_c increases, the temperature continues to rise. As a result, the surface quality decreases and tool wear increases [1]. In figure 8 cutting temperature is directly proportional to cutting speed. However, it also depends on other factors such as f, *doc*, cutting width, machine operating conditions [27]. Figure 8, *a–e* shows the effect of *doc* on cutting temperature. The temperature continues to increase with increasing *doc* because at maximum feed and *doc*, large frictional heat is generated due to the friction between the work material and the cutting tool, which leads to thermal softening of the material [29]. According to semi-empirical and regression results, *doc* is a more significant temperature parameter than *f* and V_c [1].



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Fig. 7. Effect of feed rate on cutting temperature at different cutting speed and depth of cut for all materials using *TiAlN*-coated tool



Fig. 8. Effect of cutting speed on cutting temperature at different feed rate and depth of cut for all material using *TiAlN*-coated tool

In figure 9, a-e the workpiece or tool is enlarged due to the heat generated. Cutting temperature greatly influences the mechanical properties of the workpiece and the forces acting on the workpiece and tool [30]. Most of the total heat is transferred to the chip, and this total heat in the chip flow is due to shear and friction at the chip-tool interface. Changing *doc* has a greater impact on the cutting temperature compared to *f* and *V*_c [8].

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Fig. 9. Effect of depth of cut on cutting temperature at different feed rate and cutting speed for all material using *TiAlN*-coated tool

All figures show the results of regression values taken from the empirical model and experimental *RSM* values for temperature and surface roughness, which were found to be comparable. All values of the *RSM* output parameters and the empirical model values are in good agreement with each other. Therefore, Equations (10) and (18) can be used to determine the theoretical value of R_a and temperature at different cutting parameters for different work materials with *TiAlN*-coated carbide tool inserts.

Conclusions

A semi-empirical method, taking into account the dimensions of material properties, is proposed for estimating cutting temperature and surface roughness when turning *SS 316*, *SAE 8620*, *EN 8* and *Al 380* workpieces with *PVD*-coated carbide (*TiAlN*) inserts. In addition, a multilinear regression analysis was carried out and based on the analysis of the results of the regression and semi-empirical model, the following conclusions were drawn:

• At higher feed rates, low surface roughness is observed for all materials. However, as feed and depth of cut increase, surface roughness tends to increase more in *SS 316*, then *Al 380*. *EN 8* shows better results due to low heat generation in the cutting zone, which maintains tool shape stability.

• The rapid work hardening of the chips in the case of SS 316, the toughness of the chip and built-up material, the stability of the tool shape in the case of EN 8 and SAE 8620 are the main reasons for the surface roughness quality.

• Higher cutting temperature is obtained when turning *SS 316* and lower cutting temperature is obtained when turning *Al 380*. This is due to the significant difference in thermal conductivity of these materials.

- When machining EN 8 and SAE 8620, the cutting temperature range is found to be moderate.
- Surface roughness is found to be worst for Al 380 and best for SS 316 and SAE 8620.

• In addition, using a dimensional analysis model, a generalized empirical formula is developed to predict the surface roughness and temperature encountered during metal cutting. These models are found to fit well with regression equations derived from experimental values.

• The proposed method for measuring surface roughness and temperature can be conveniently used. This is a useful way to cost-effectively evaluate heat generation and surface roughness when turning various materials with *TiAlN*-coated carbide tools.



См

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Conflicts of Interest

The authors declare no conflict of interest.

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