

Modeling and multi-objective optimization of surface roughness and productivity in dry turning of AISI 52100 steel using (TiCN-TiN) coating cermet tools

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ABSTRACT

The present work concerns an experimental study of turning with coated cermet tools with TiCN-TiN coating layer of AISI 52100 bearing steel. The main objectives are firstly focused on the effect of cutting parameters and coating material on the performances of cutting tools. Secondly, to perform a Multi-objective optimization for minimizing surface roughness (R_a) and maximizing material removal rate by desirability approach. A mathematical model was developed based on the Response Surface Methodology (RSM). ANOVA method was used to quantify the cutting parameters effects on the machining surface quality and the material removal rate. The results analysis shows that the feed rate has the most effect on the surface quality. The effect of coating layers on the surface quality is also studied. It is observed that a lower surface roughness is obtained when using PVD (TiCN-TiN) coated insert when compared with uncoated tool. The values of root mean square deviation and coefficient of correlation between the theoretical and experimental data are also given in this work where the maximum calculated error is 2.65 %.

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Nomenclature

V	cutting speed (m/min)	DF	degrees of freedom
f	feed rate (mm/rev)	SS	sum of squares
d	depth of cut (mm)	MS	mean squares
F_a	feed force (N)	PC	Percentage contribution ratio (%)
F_r	thrust force (N)	R^2	determination coefficient
F_t	tangential cutting force (N)	α	clearance angle (°)
R_a	arithmetic average of absolute roughness (μm)	χ_r	cutting edge angle (°)
X_i	coded machining parameters	γ	rake angle (°)
a_j	coefficients of linear terms	λ	cutting edge inclination angle (°)
a_{ii}	quadratic terms	p	tested parameters
a_{ij}	cross-product terms	j	level number of tested parameters
BUE	built-up-edge	t	repetition of tested parameters
$ANOVA$	analysis of variance	r	correlation coefficient
RMS	response surface method	e	root mean square of percentage deviation

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1. Introduction

Coated cemented carbide cutting tools are extensively used in Material cutting. Several research works were investigated in order to improve the performance of these cutting tools with regards to the wear resistance, thermal shock resistance and surface quality of workpiece (Berkani et al. 2015; Tebassi et al. 2016). The coating material of cutting tools contributes efficiently in improving the cutting conditions, where, its effects on the performance of cutting tools was studied in prior investigations Grzesik and Nieslony (2004). It protects the wear substrate, improves the crack resistance and created a thermal barrier (Das et al. (2015). Among coating layers used there are the titanium carbide (TiC), titanium nitride (TiN), titanium carbonitride (TiCN), titanium aluminum nitride (TiAlN), and aluminum oxide (Al_2O_3) are the most selected materials Sahoo and Sahoo (2012). These layers are mostly deposited with Chemical Vapor Deposition (CVD) or by Physical Vapor Deposition (PVD) technique Aurich et al. (2012). The investigations conducted by Grzesik (1998) highlights the crucial effect of different types of coating at dry machining of carbon-based steel AISI 1045 and AISI 304 stainless steel. He used three type of coating layers: TiC, TiC/TiN, and TiC/ Al_2O_3 /TiN. He found that the contact area tool-chip, of the couple tool-material TiC/ Al_2O_3 /TiN–AISI 1045 gives lower values of friction factor and the contact pressure compared to the couple TiC/TiN-AISI 1045 and TiC-AISI 1045. The tribological phenomenon at tool-chip interface was investigated by Zhang et al. (2015) in dry cutting of hardened steel AISI 1045 using two types of coated cemented carbide. The first one is a nano-scale surface textured on the rake face of the cutting tool followed by the deposition of Ti55Al45N hard-coatings. In the second insert, the order sequence of coating layers was inverted. A significant decrease in cutting forces; cutting temperature and friction coefficient at the tool–chip interface were found. Where, the textured tool reduces the contact area and then the friction at the tool–chip interface. Sahoo and Sahoo (2013) conducted an experimental investigation on the flank wear behavior in hard turning using multilayer coated carbide (TiN/TiCN/ Al_2O_3 /TiN) insert. They found that the flank wear value increases with increasing cutting speed, feed rate and depth of cut. They also observed that both abrasion and diffusion wear mechanisms are predominantly at extreme cutting conditions. A comparative study has been found by Cakir et al. (2009) using PVD coated (TiAlN) insert and CVD coated (TiCN/ Al_2O_3 /TiN) insert. They observed that the lower surface roughness a value is obtained by employing coated (TiAlN) insert.

In machining process, the Designs and method such as factorial designs, response surface methodology (RSM) and Taguchi method are frequently used for analyzing experimental data, which allowed to reduce the consuming time and cost. In order to study the performance of coated carbide inserts, ceramic and cubic boron nitride (CBN) tools, recently a various investigation have been carried out. Azizi et al. (2012) conducted an experimental investigation to determine the significant parameters affecting the surface roughness and cutting forces, in finish hard turning of AISI 52100 steel, by using the response surface methodology (RSM). They applied an optimization study of machining conditions to produce the lowest surface roughness with minimal cutting force components using desirability function approach for multiple response factors optimization. They concluded that the feed rate, workpiece hardness and cutting speed have significant effects in reducing the surface roughness; whereas the depth of cut, workpiece hardness and feed rate are observed to have a statistically significant impact on the cutting force components than the cutting speed. Chinchankar and Choudhury (2013) investigated the effect of work materials hardness and cutting parameters on cutting forces, surface roughness and tool life, when turning hardened AISI 4340 steel at two levels of hardness (35 and 45 HRC), using coated tungsten based cemented carbide inserts. Mathematical models were developed for surface roughness and cutting force components using the response surface methodology (RSM). The results concluded that the Cutting forces get affected mainly by depth of cut followed by feed; Surface roughness gets affected significantly at higher feed and depth of cut Cutting speed followed by depth of cut was found to be the most influencing factors on tool life, especially when turning harder work piece. Hessainia et al. (2013) Developed mathematical prediction model of the surface roughness for turning operations of 42CrMo4 hardened steel by Al_2O_3 /TiC mixed ceramic cutting insert, based on cutting parameters and tool vibration in radial and in main cutting force directions. They reported according to their experimental

results that the feed rate is the dominant factor affecting the tow criteria of surface roughness (R_a and R_t), whereas vibrations in two directions have a low effect on the surface roughness. Elbah et al. (2013) had compared between the surface roughness criteria (R_a , R_z and R_t) of the wiper inserts with conventional inserts during hard turning of AISI 4140 hardened steel (60 HRC), The analysis of results reveals that the feed rate and depth of cut have significant effects in reducing the surface roughness. Asiltürk and Akkus (2011) conducted an optimization study by machining a hardened AISI 4140 (51HRC) steel with coated carbide cutting tools using the Taguchi method to obtain optimum cutting parameters. They applied the statistical methods of signal to noise ratio (SNR) and the analysis of variance (ANOVA) to investigate effects of cutting speed, feed rate and depth of cut on surface roughness. They reported that the feed rate has the most significant effect on R_a and R_z , In addition, the effects of two factor interactions of the feed rate-cutting speed and depth of cut-cutting speed appear to be important

In this work, an experimental investigation was found in order to study the effect of cutting parameters (namely the cutting speed (V_c), fed rate (f) and the depth of cut (d)) on surface roughness and material removal rate and to perform a Multi-objective optimization intended for minimize roughness R_a and maximize material removal rate by desirability approach. A comparison was performed between uncoated and coated cemented carbide tools with (TiCN/TiN) coating layer. The insets have identical substrate composition and geometry. The effect of tool wear was neglected by using for each experiment a fresh cutting edge. In addition, a mathematical model was developed based on the Response Surface Methodology (RSM). ANOVA method was used to quantify the cutting parameters effects on the machining surface quality and the material removal rate. The analysis shows that among the studied cutting parameters, the feed rate has the most important effect on the surface quality with a contribution rate above 85%. The effect of coating layers on the surface quality is also studied. It is observed that a lower surface roughness value is achieved using PVD (TiCN/TiN) coated insert when compared to the uncoated tool.

2. Experimental setup and design

2.1. Experimental design

This work is an experimental study focused on the effect of cutting parameters on surface roughness, and material removal rate with, developing a correlation between them. The experimental design involves variation of three factors at three levels (low, medium and high), including cutting speed (v), feed rate (f) and depth of cut (d) as indicated in Table 1. The selected experimental design requires 27 runs with 26 degrees of freedom (DF). For each run, one test was performed (no replications). Also a random order was determined for running the tests.

Table 1
Operating conditions of cutting parameters

Level	Cutting speed, v (m/min)	Feed rate, f (mm/rev)	Cutting depth, d (mm)
Low	150	0.08	0.15
Medium	200	0.12	0.30
High	250	0.16	0.45

The analysis of variance (ANOVA) method was used to quantify the impact of the cutting speed (v), feed rate (f) and depth of cut (d) and their interactions on the surface roughness. The obtained results were analyzed using Design-Expert® Software, statistical analysis software, which is widely used in engineering optimizations. In addition, the response surfaces methodology (RSM) was used in order to determine the relationship between the independent process parameters with the desired response and exploring the effect of these parameters on responses, including six steps Myers & Montgomery, (2002).

These are, in the following order, (1) define the independent input variables and the desired responses with the design constants, (2) adopt an experimental design plan, (3) perform regression analysis with the quadratic model of RSM, (4) calculate the statistical analysis of variance (ANOVA) for the independent input variables in order to find which parameter significantly affects the desired response, then, (5) determine the situation of the quadratic model of RSM and decide whether the model of RSM needs screening variables or not and finally, (6) Optimize and conduct confirmation experiment and verify the predicted performance characteristics. In the present work, the RSM-based second order mathematical model is given by the following Eq. (1):

$$Y = a_0 + \sum_{i=1}^n a_{ii} X_i + \sum_{i=1}^n a_{ii} X_i^2 + \sum_{i \neq j}^n a_{ij} X_i X_j \quad (1)$$

Where, Y: is the estimated response (surface roughness or cutting forces). a_0 , a_i , a_{ij} and a_{ii} are the adopted mathematical model coefficients. They are not identified and must be calculated from the experimental results. x_i , are variables or factors influences the response (Y), corresponding to the studied cutting condition parameters such as cutting speed (V), feed rate (f) and depth of cut (d) and their interactions.

2.2. Work piece material and Cutting inserts

The aim of this experimental work is to study the effect of cutting parameters on surface roughness and material removal rate and to develop a correlation between them. In this study a turning runs are carried out in dry conditions using a universal lathe SN 40C type with 6.6 kW spindle power. The work piece material is AISI 52100 steel in the form of round bars with 66 mm of diameter and 380 mm cutting length. This material is widely used in manufacturing of automotive components regarding to their properties like high tensile strength, shock resistance and Brinell hardness about 230 HB. The chemical composition of AISI 52100 steel is given in Table 1. Coating effect on the cutting performances is investigated using two type of cutting inserts. The first type is GC 1525 coated insert (PVD with TiCN/TiN) layer sequence, which is an ISO class P15 grade, with a total thickness of 3 μm . The main coating layer includes titanium carbonitride (TiCN) and a thin layer of titanium nitride (TiN). The second insert is CT5015 uncoated cermet. The two insets (GC 1525 and CT5015) have an identical substrate with the same geometry designation as CNMG 120408. For each experiment a fresh cutting edge was used. A right hand style tool holder designated by ISO as CSBNR2525M12 was used for mounting the inserts. It is characterized by the following angles: $\chi_r = 75^\circ$, $\lambda = -6^\circ$, $\gamma = -6^\circ$, $\alpha = +6^\circ$. The measurement of the arithmetic surface roughness (Ra) is obtained from a SurfTest 201 Mitutoyo roughness-meter, as showing in experimental design flowchart Fig.1. The machined length is 24 mm with a basic span of 3 mm. The measurements were repeated at three equally spaced locations around the circumference of the workpiece at 120° and the result is an average of these values. The surface roughness is directly measured on the workpiece, without dismounting from the lathe, in order to reduce error measurements. Finally, the material removal rate (MRR) is calculated using Eq. (2)

$$\text{MRR} = V * f * d \text{ (cm}^3/\text{min)} \quad (2)$$

Table 2

Chemical composition of AISI 52100 steel

C	Si	Mn	Cr	Ni	Mo	V	Cu
1.09	0.256	0.35	1.382	0.077	0.017	0.005	0.150

3. Analysis and discussion of experimental results

In this work a comparative assessment of coated and uncoated cutting tool was made in order to quantify the impact of selected turning parameters on surface roughness (Ra) and material removal rate (MRR), their experimental results are showing in Table 3.

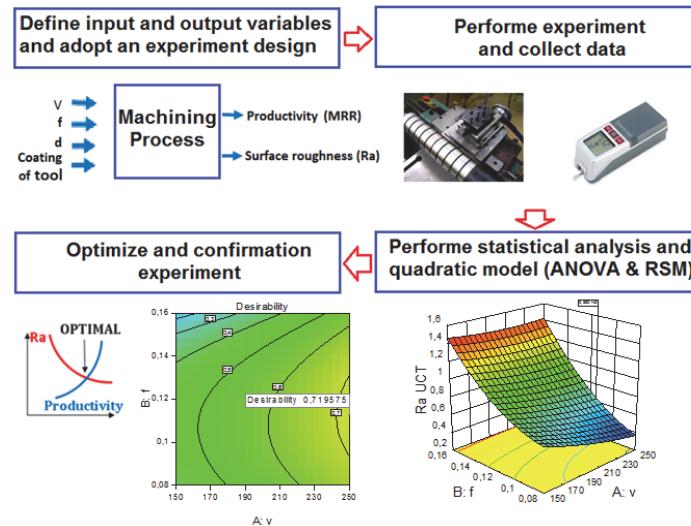


Fig.1. Experimental design flowchart

Table 3

Experimental results for surface roughness and material removal rate

machining parameters			Coated inserts	Uncoated inserts	Material Removal Rate
V (m/min)	f (mm/rev)	d (mm)	Ra (μm)	Ra (μm)	Q (cm^3/min)
150	0,08	0,15	0,47	0,96	1,8
150	0,08	0,3	0,45	0,8	3,6
150	0,08	0,45	0,51	0,6	5,4
150	0,12	0,15	0,78	0,77	2,7
150	0,12	0,3	0,89	0,9	5,4
150	0,12	0,45	0,84	0,96	8,1
150	0,16	0,15	1,15	1,39	3,6
150	0,16	0,3	1,18	1,45	7,2
150	0,16	0,45	1,42	1,46	10,8
200	0,08	0,15	0,52	0,4	2,4
200	0,08	0,3	0,4	0,42	4,8
200	0,08	0,45	0,51	0,47	7,2
200	0,12	0,15	0,7	0,72	3,6
200	0,12	0,3	0,82	0,74	7,2
200	0,12	0,45	0,7	0,93	10,8
200	0,16	0,15	1,06	1,37	4,8
200	0,16	0,3	1,06	1,37	9,6
200	0,16	0,45	1,3	1,44	14,4
250	0,08	0,15	0,49	0,32	3
250	0,08	0,3	0,39	0,37	6
250	0,08	0,45	0,5	0,37	9
250	0,12	0,15	0,46	0,57	4,5
250	0,12	0,3	0,75	0,72	9
250	0,12	0,45	0,58	0,83	13,5
250	0,16	0,15	1,02	1,26	6
250	0,16	0,3	0,9	1,27	12
250	0,16	0,45	1,15	1,3	18

3.1. ANOVA results and effects of machining parameters

The results of variance analysis for surface roughness criteria (Ra) of coated and uncoated cermet inserts are given in Tables 4 and 5. These tables also show the degrees of freedom (DF), sum of squares (SSp), p-values and the percentage of contribution (PC) (Lin, 2002; Aouici et al. 2016). The p-value is used to determine the significance of each factor if its value is lower than 0.05. The characterization of the machined surface quality was limited to the criteria of arithmetic mean roughness (Ra). From Table 4 It can be concluded that the feed rate is the most important parameter affecting surface finish compared to

the other factors and their interaction and its contribution is 93.97% and 82.09% for coated and uncoated inserts respectively. The second factor affecting the surface quality is the cutting speed which contributes by 2.32% and 7.41% for coated and uncoated inserts respectively. The cutting depth gives lower influences on the surface roughness where its contribution is less than 2%. This result can be explained by the phenomenon of the grooving helicoidally furrows on the finish machining surface caused by the rising of feed rate combined with tool-workpiece movement as shown in another study Bouzid et al. (2014), Bouchelaghem et al. (2007), on the other hand, this phenomenon is explained by the reduction of feed rate caused low cutting forces, which resulting less vibration after that provide better surface finish. Similar investigations have been made in the same context, for example, Yalçın et al. (2009) and Hessainia et al. (2015), where they found that the improvement of surface roughness (R_a) is caused by decreasing in the cutting forces at high cutting speeds accordingly influences the machining system stability. Fig. 2 and Fig. 3 show the main effect of the studied factors on the surface roughness. It is clear that the surface roughness is proportional to the increase of the feed rate and cutting depth and inversely proportional to the cutting speed.

Table 4

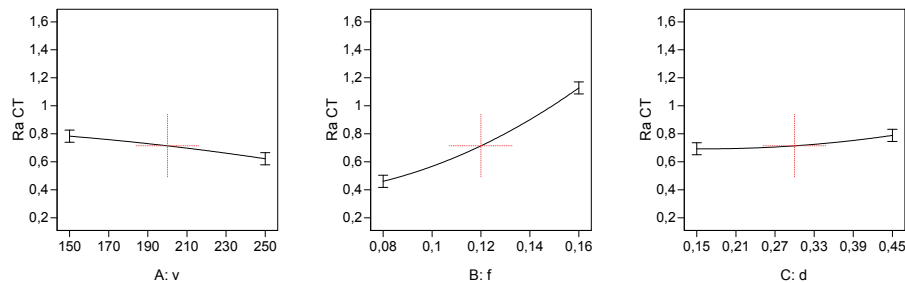
Analysis of variance results of surface roughness R_a using Coated inserts

Source	Sum of Squares	df	Mean Square	F Value	p-value	Cont %	Remarks
Model	4,26	9	0,47	212,47	< 0,0001	99,06	significant
A-v	0,1	1	0,1	45,43	< 0,0001	2,327	significant
B-f	4,04	1	4,04	1813,69	< 0,0001	93,954	significant
C-d	0,055	1	0,055	24,6	0,0001	1,275	significant
AB	8,33E-04	1	8,33E-04	0,37	0,549	0,019	not significant
AC	2,08E-04	1	2,08E-04	0,093	0,7635	0,004	not significant
BC	3,33E-05	1	3,33E-05	0,015	0,9041	0,007	not significant
A ²	2,18E-03	1	2,18E-03	0,98	0,3362	0,057	not significant
B ²	0,06	1	0,06	26,82	< 0,0001	1,395	not significant
C ²	5,14E-04	1	5,14E-04	0,23	0,6371	0,011	significant
Residual	0,038	17	2,23E-03	-	-	0,883	-
Total	4,3	26	-	-	-	-	-

Table 5

Analysis of variance results of surface roughness R_a using Uncoated inserts

Source	Sum of Squares	df	Mean Square	F Value	p-value	Cont %	Remarks
Model	3,75	9	0,42	45,4	< 0,0001	95,90	significant
A-v	0,29	1	0,29	31,45	< 0,0001	7,416	significant
B-f	3,21	1	3,21	349,43	< 0,0001	82,09	significant
C-d	0,02	1	0,02	2,18	0,1583	0,511	not significant
AB	0,057	1	0,057	6,25	0,0229	1,457	significant
AC	0,017	1	0,017	1,84	0,193	0,434	not significant
BC	0,015	1	0,015	1,6	0,2229	0,383	not significant
A ²	6,23E-03	1	6,23E-03	0,68	0,4216	0,159	not significant
B ²	0,14	1	0,14	15,14	0,0012	3,580	significant
C ²	2,96E-05	1	2,96E-05	3,23E-03	0,9554	0,000	not significant
Residual	0,16	17	9,18E-03	-	-	4,092	-
Total	3,91	26	-	-	-	-	-

**Fig. 2.** Main effect of factor coated inserts

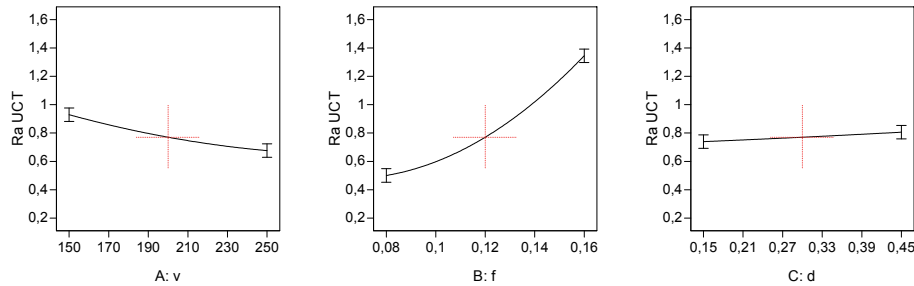


Fig. 3. Main effect of factor Uncoated inserts

3.2. Regression equations

The quadratic models for surface roughness (Ra) were obtained from the experimental data. The values of the coefficients involved in Eq. (1) have been calculated by (RSM) using Design-Expert® Software. Mathematical equations of the fitted models for the arithmetic mean roughness (Ra) models of inserts with and without coating are given below in Eq. (3) and Eq.(4) respectively.

$$Ra_{coated} = 0.178 + 0.004 v - 0.917f - 1.148 d - 4.667e-006 v^2 - 0.026vf - 0.002v.d + 50 f^2 + 8.333f.d + 1.185 d^2 \tag{3}$$

$$Ra_{uncoated} = 3.178 - 0.013v - 20.94f - 1.537d + 1.289 e-005 v^2 + 0.0346 v.f + 0.005 v.d + 95.139 f.f + 5.833f.d + 0.098 d^2 \tag{4}$$

3.3. Adequacy test

The adequacy of the obtained models is tested using (ANOVA) method, according to its results in Table 6 and Table 7. The quadratic models of surface roughness with and without coating insert, indicated that the relationship between the response variable and the predictors factors is significant within the confidence limit, which is illustrated by the (p-value = 0.000) at what time are less than alpha value (0.05). The coefficient of determination (R^2) was estimated to check the accuracy of the fit. For the surface roughness with and without coating, the (R^2) values are 98.65% and 94.39, respectively. The predicted coefficient of determination (R^2_{pred}) for surface roughness using quadratic model with coated and uncoated insert is 97.7% and 88.11% respectively. Where they in excellent agreement; which indicate the fitness of the developed regression models.

Table 6
Fit Summary of surface roughness Ra using Uncoated inserts

Source	Sequential p-value	Adjusted R-Squared	Predicted R-Squared	Remarks
Linear	< 0.0001	0,8871	0,8571	
Interaction	0,1513	0,8997	0,8382	
<u>Quadratic</u>	<u>0,0094</u>	<u>0,9439</u>	<u>0,8811</u>	<u>Suggested</u>
Cubic	0,0028	0,9831	0,9456	Aliased

Table 7
Fit Summary of surface roughness Ra using Coated inserts

Source	Sequential p-value	Adjusted R-Squared	Predicted R-Squared	Remarks
Linear	< 0,0001	0,9733	0,9685	-
Interaction	0,9746	0,9697	0,9591	-
<u>Quadratic</u>	<u>0,0007</u>	<u>0,9865</u>	<u>0,977</u>	<u>Suggested</u>
Cubic	0,0261	0,9939	0,9845	Aliased

Likewise, a test of the normality may possibly be made by constructing the normal probability plot of the residuals. If the underlying error distribution is normal, this plot will be similar to a straight line this plot is illustrated in Fig.4. This can be indicated that the transformation of the surface roughness results provides a good analysis, and the proposed models are adequate.

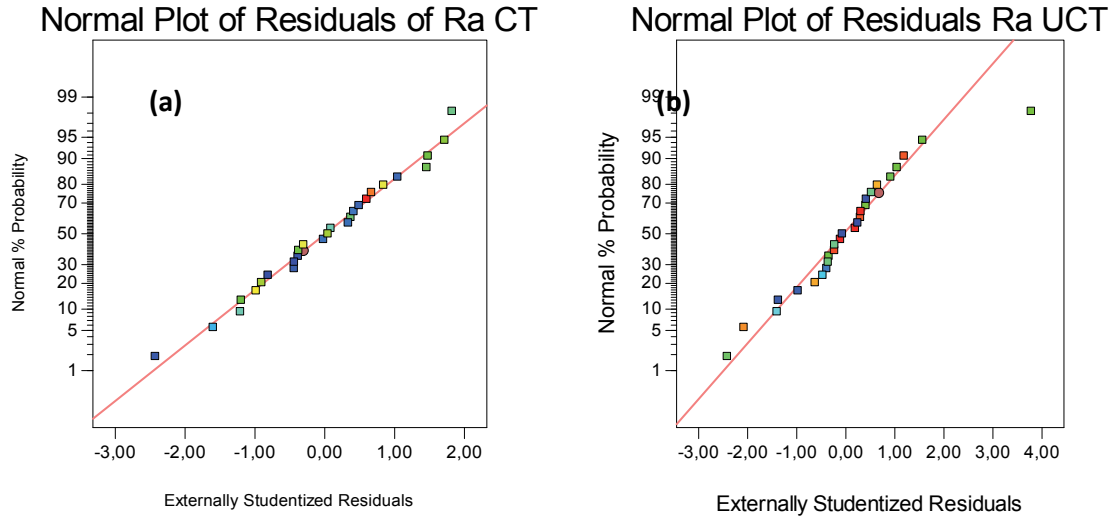


Fig. 4. Normal plot of residuals (a) for coated insert and (b) for uncoated insert

In addition, the accuracy in prediction may be found by determining the correlation coefficient (r) and root mean square of percentage deviation (e) between the theoretical and experimental values in Fig. 5 the correlation coefficient (r) for “ N ” number of observations is evaluated as

$$r = \frac{N \sum X_i Y_i - (\sum X_i)(\sum Y_i)}{\sqrt{N \sum X_i^2 - (\sum X_i)^2} \sqrt{N \sum Y_i^2 - (\sum Y_i)^2}} \quad (5)$$

The percentage deviation (e_i) is expressed as

$$e_i = \frac{X_i - Y_i}{X_i} \times 100 \quad (6)$$

Also, the root mean square of percentage deviation (e) is expressed as

$$e = \sqrt{\frac{\sum (e_i)^2}{N}} \quad (7)$$

The coefficient of correlation value (r) for all models is 0.97, showing a good agreement between the experimental and theoretical results. For the surface roughness with and without coating inserts, the root mean square is 2.06% and 2%, respectively.

3.4. 3D surface plots

Figs. 5 (a) and (b) shows the effect of cutting speed (V_c) and feed rate (f) on the surface roughness parameters (R_a) with and without coating cutting tool, whereas the depth of cut (d) is kept at the middle level. In each graph of the 3D response surface, the four vertices represent the four combinations of the two considered factors. It is clearly observed that the feed rate (f) affects strongly the surface roughness parameters. Feed rate (f) has an increasing effect while the cutting speed has an important and decreasing effect. Indeed, the Surface roughness is improved by increasing the cutting speed, producing a better surface finish at higher cutting speed is a well known phenomenon in metal cutting; the conventional

explanations are related to the built-up-edge (BUE) yaltese et al. (2009). That is to say, the formation of BUE is favoured in a certain cutting speed range. By increasing cutting speed beyond this region, BUE is eliminated and as a result, the surface quality is improved. Also, Fig. 6 (a) and (b) shows the impact of the feed rate and depth of cut on the surface roughness Ra using coated and uncoated insert, the analysis of this plot shows that the variation of surface roughness parameter Ra is proportional to the feed rate and depth of cut. These 3D illustrations clearly confirm the effects of interactions that were presented with the interaction diagrams.

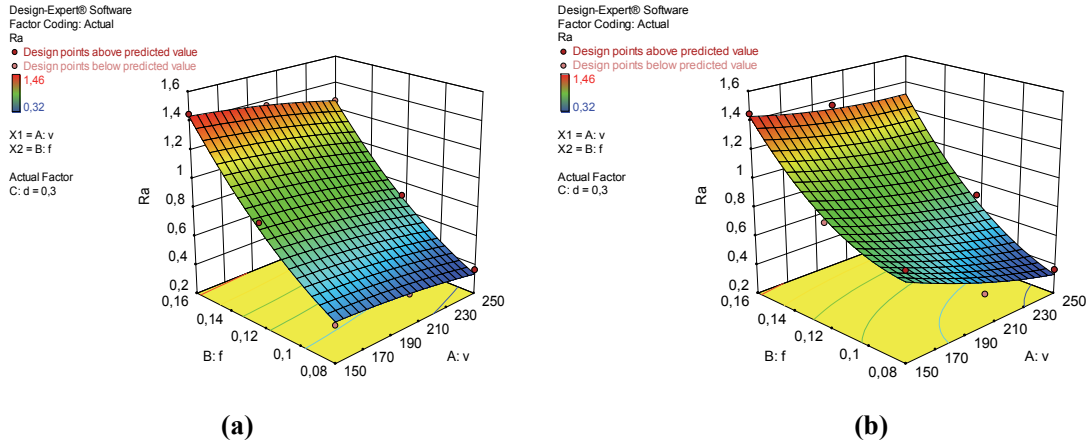


Fig. 5. Effect of cutting speed and feed rate on surface roughness Ra (a) Of coated tool and (b) Of uncoated tool

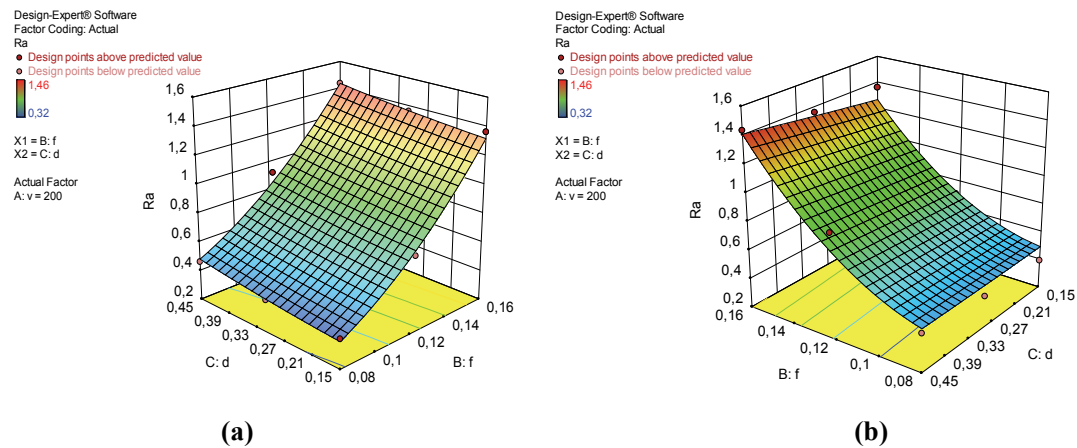


Fig. 6. Effect of feed rate and depth of cut on surface roughness Ra (a) Of coated tool and (b) of uncoated tool

3.5. Optimization of response

The aim of the present experimental study is to perform a Multi-objective optimization intended for minimize roughness Ra and maximize material removal rate by desirability approach, using Design Expert Software. During the optimization process the aim was to find the optimal values of machining parameters in order to produce the lowest surface roughness (Ra) with maximum material removal rate. To resolve this type of parameter design problem, an objective function, $F(x)$, is defined as follows Myers and Montgomery, (2002).

$$\begin{cases} DF = \left(\prod_{i=1}^n d_i^{w_i} \right)^{\frac{1}{\sum_{j=1}^n w_j}} \\ F(X) = -DF \end{cases} \quad (8)$$

where d_i is the desirability defined for the (i^{th}) targeted output and w_i is the weighting of d_i . For various goals of each targeted output, the desirability, d_i , is defined in different forms. If a goal is to reach a specific value of T_i , the desirability d_i is:

$$\begin{aligned}
 d_i &= 0 \quad \text{if } Y_i \leq \text{Low}_i \\
 d_i &= \left[\frac{Y_i - \text{Low}_i}{T_i - \text{Low}_i} \right] \quad \text{if } \text{Low}_i \leq Y_i \leq T_i \\
 d_i &= \left[\frac{Y_i - \text{High}_i}{T_i - \text{High}_i} \right] \quad \text{if } T_i \leq Y_i \leq \text{High}_i \\
 d_i &= 0 \quad \text{if } Y_i \geq \text{High}_i
 \end{aligned} \tag{9}$$

For a goal to find a maximum, the desirability is shown as follows:

$$\begin{aligned}
 d_i &= 0 \quad \text{if } Y_i \leq \text{Low}_i \\
 d_i &= \left[\frac{Y_i - \text{Low}_i}{\text{High}_i - \text{Low}_i} \right] \quad \text{if } \text{Low}_i \leq Y_i \leq \text{High}_i \\
 d_i &= 1 \quad \text{if } Y_i \geq \text{High}_i
 \end{aligned} \tag{10}$$

For a goal to search for a minimum, the desirability can be defined by the following formulas:

$$\begin{aligned}
 d_i &= 1 \quad \text{if } Y_i \leq \text{Low}_i \\
 d_i &= \left[\frac{\text{High}_i - Y_i}{\text{High}_i - \text{Low}_i} \right] \quad \text{if } \text{Low}_i \leq Y_i \leq \text{High}_i \\
 d_i &= 0 \quad \text{if } Y_i \geq \text{High}_i
 \end{aligned} \tag{11}$$

where, Y_i is the found value of the (i^{th}) output during optimization processes. Low_i and High_i are respectively, the minimum and the maximum values of the experimental data for the (i^{th}) output. In Eq. (8), w_i is set to one since the d_i is important in this study. DF is a combined desirability function Myers & Montgomery, (2002), and the objective is to choose an optimal setting that maximizes a combined Desirability Function (DF) minimizes $F(x)$. Table 7 shows the goals, parameter ranges and importance degrees for the output parameters. The optimum cutting parameters obtained for *Ra* and *MRR* using coated cermet are chosen in terms of the highest desirability value with cutting speed (v) of 250m/mn, feed rate (f) of 0.113 mm/rev and depth of cut (d) of 0.45 mm, where the optimal solutions are reported in Table 8 in order to decrease the desirability level. On the other hand, for uncoated cermet, the optimal cutting parameters are cutting speed (v) of 250m/mn, feed rate (f) of 0.102 mm/rev and depth of cut (d) of 0.45 mm, as indicated in Table 8. The predicted responses of coated insert are $Ra=0.63\mu\text{m}$ and 12.75 cm^3/min for *MRR*, with desirability value of 0.72, by against to uncoated insert the values of predicted responses are $Ra=0.54\mu\text{m}$ and 11.629 cm^3/min for *MRR*, with desirability value of 0.698. The Fig. (7-9) presents the contour plot for Desirability of multi-objective optimization, the contour plot for surface roughness (*Ra*) and the contour plot for productivity (*MRR*) using coated insert, in relation to the cutting speed (V_c) feed rate (f) and depth of cut (d). For each plot, the variables not represented are held at the optimal value. Besides, the estimated contour plot using uncoated tool are depicted in Fig.(10-12) as a function of cutting speed, feed rate and depth of cut.

Table 7
Constraints for optimization of machining parameters

Name	Goal	Lower Limit	Upper Limit	Lower Weight	Upper Weight	Importance
V (m/min)	in range	120	250	1	1	3
f (mm/rev)	in range	0,08	0,16	1	1	3
d (mm)	in range	0,15	0,45	1	1	3
Ra coated tool (μm)	minimize	0.39	1.42	1	1	3
Ra uncoated tool (μm)	minimize	0,32	1,46	1	1	3
Q coated tool (cm ³ /min)	maximize	1.8	18	1	1	3

Table 8
Optimal solutions

Number	Cutting tool	v	f	d	Ra	Q	Desirability	remarks
1	Coated tool	250	0.113	0.45	0.631	12.75	0.720	suggested
	Uncoated tool	250	0.102	0.45	0.545	11.629	0.698	suggested
2	Coated tool	250	0.113	0.45	0.634	12.75	0.720	-
	Uncoated tool	250	0.102	0.45	0.542	11.602	0.698	-
3	Coated tool	250	0.113	0.45	0.628	12.75	0.720	-
	Uncoated tool	250	0.102	0.45	0.549	11.673	0.698	-

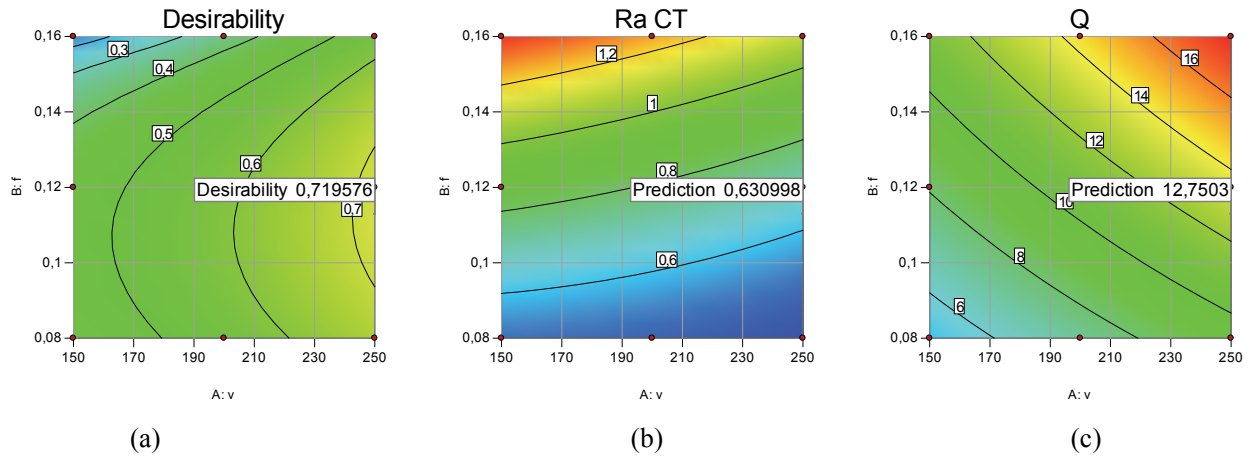


Fig.7. (a) Contour plot for Desirability of multi-objective optimization
(b) Contour plot for Ra of coated tool, (c) Contour plot for MRR of coated tool,
While (d) kept at the optimal value

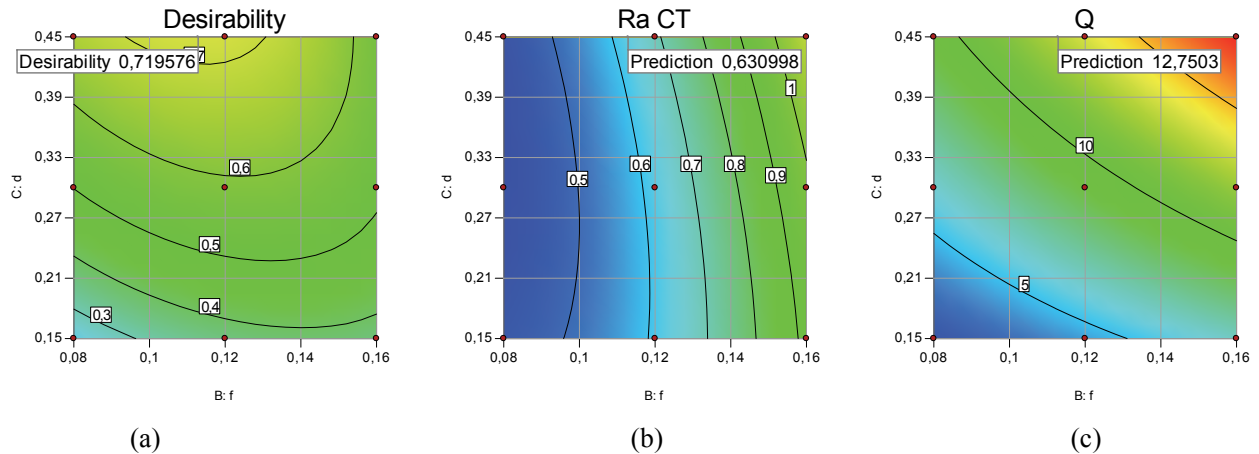


Fig.8. (a) Contour plot for Desirability of multi-objective optimization
(b) Contour plot for Ra of coated tool, (c) Contour plot for MRR of coated tool
while (V) kept at the optimal value

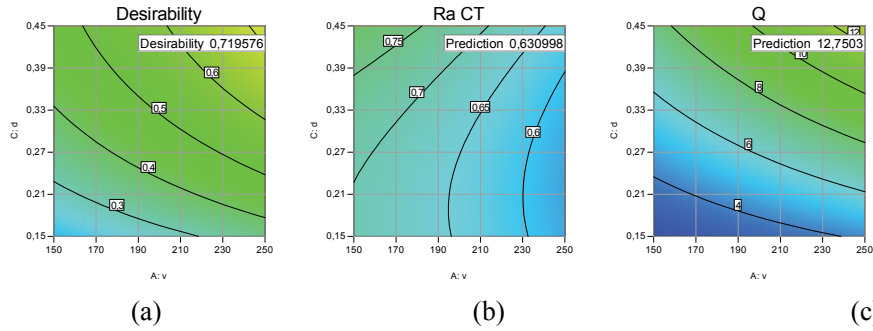


Fig. 9. (a) Contour plot for Desirability of multi-objective optimization (b) Contour plot for Ra of coated tool, (c) Contour plot for MRR of coated tool, while (f) kept at the optimal value

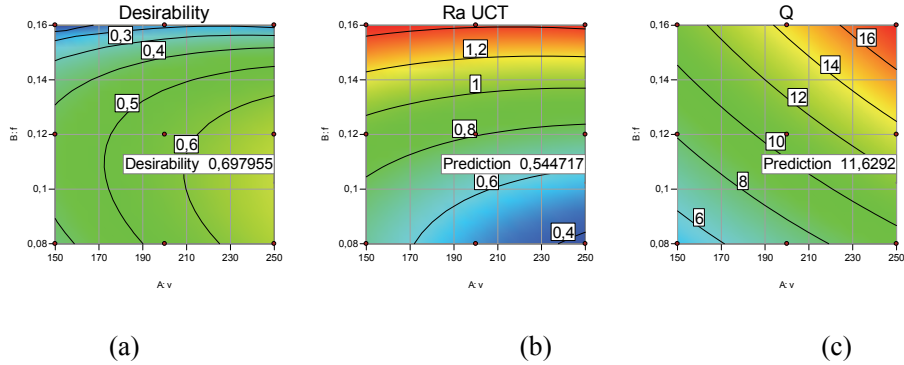


Fig.10. (a) Contour plot for Desirability of multi-objective optimization (b) Contour plot for Ra of uncoated tool, (c) Contour plot for MRR of uncoated tool, while (d) kept at the optimal value

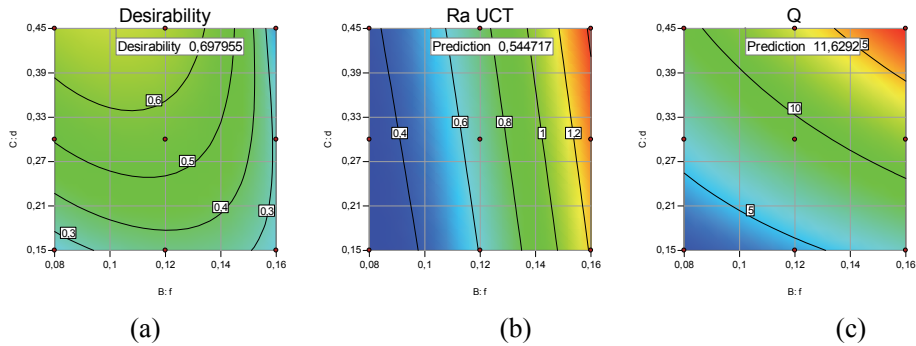


Fig. 11. (a) Contour plot for Desirability of multi-objective optimization (b) Contour plot for Ra of uncoated tool, (c) Contour plot for MRR of uncoated tool while (V) kept at the optimal value

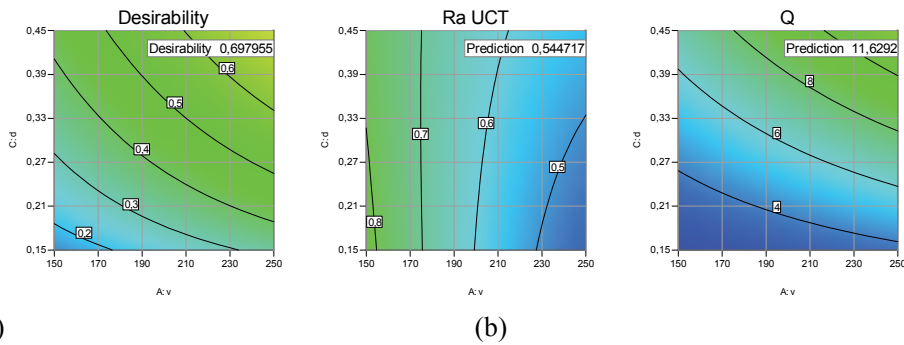


Fig. 12. (a) Contour plot for Desirability of multi-objective optimization (b) Contour plot for Ra of uncoated tool, (c) Contour plot for MRR of uncoated tool while (f) kept at the optimal value

4. Conclusions

In this work, the impact of coating material of cutting tool and cutting parameters on surface roughness are analyzed and optimized, in dry turning of AISI 52100 steel. The important findings are recapitulated as follows:

- 1) ANOVA result using coated cermet of surface roughness shows that feed rate is the most significant parameter that contributes toward the output responses by 93.95%. The effect of cutting speed is only 2.32 %, followed by depth of cut by 1.27%, which are negligible. Versus, to uncoated insert the effect of feed rate is the most significant parameter with the percent contribution of 82.09%. The effect of cutting speed is only of 7.41 %, followed by depth of cut by 0.51%.
- 2) High surface quality are obtained when employing PVD (TiCN-TiN) coated insert. The coating effect on the surface roughness is improved from 0.77 for the uncoated insert to 0.89 for the coated one.
- 3) The coefficient of correlation (r) was of 0.97 and the root mean square of percent deviation values varies in the range of 2.64% to 2%. Consequently, a good agreement was found between the proposed theoretical models and the experimental data.
- 4) Desirability of multi-objective optimization results using coated insert shows that the optimum cutting parameters obtained, are cutting speed (v) of 250m/mn, feed rate (f) of 0.113 mm/rev and depth of cut (d) of 0.45 mm, which are chosen in terms of the highest desirability value with 0.72, which gives the predicted responses are $R_a = 0.63\mu\text{m}$, and 12.75 cm³/min for MRR.
- 5) Furthermore, for uncoated cermet, the optimal cutting parameters are cutting speed (v) of 250m/mn, feed rate (f) of 0.102 mm/rev and depth of cut (d) of 0.45 mm. The values of predicted responses are $R_a = 0.54\mu\text{m}$, and 11.629 cm³/min for MRR, with desirability value of 0.698, In this study, the using optimization methodology is a powerful approach and can present to scientific works as well machining process a helpful optimization procedure.

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References

- Aouici, H., Fnides, B., Elbah, M., Benlahmidi, S., Bensouilah, H., & Yallese, M. A. (2016). Surface roughness evaluation of various cutting materials in hard turning of AISI H11. *International Journal of Industrial Engineering Computations*, 7(2), 339.
- Asiltürk, I., & Akkuş, H. (2011). Determining the effect of cutting parameters on surface roughness in hard turning using the Taguchi method. *Measurement*, 44(9), 1697-1704.
- Aurich, J. C., Eyrisch, T., & Zimmermann, M. (2012). Effect of the coating system on the tool performance when turning heat treated AISI 4140. *Procedia CIRP*, 1, 214-219.
- Azizi, M. W., Belhadi, S., Yallese, M. A., Mabrouki, T., & Rigal, J. F. (2012). Surface roughness and cutting forces modeling for optimization of machining condition in finish hard turning of AISI 52100 steel. *Journal of Mechanical Science and Technology*, 26(12), 4105-4114.
- Berkani, S., Yallese, M., Boulanouar, L., & Mabrouki, T. (2015). Statistical analysis of AISI304 austenitic stainless steel machining using Ti (C, N)/Al₂O₃/TiN CVD coated carbide tool. *International Journal of Industrial Engineering Computations*, 6(4), 539-552.

- Bouchelaghem, H., Yallese, M. A., Amirat, A., & Belhadi, S. (2007). Wear behaviour of CBN tool when turning hardened AISI D3 steel. *Mechanika*, 65(3), 57-65.
- Bouزيد, L., Boutabba, S., Yallese, M. A., Belhadi, S., & Girardin, F. (2014). Simultaneous optimization of surface roughness and material removal rate for turning of X20Cr13 stainless steel. *The International Journal of Advanced Manufacturing Technology*, 74(5-8), 879-891.
- Cakir, M. C., Ensarioglu, C., & Demirayak, I. (2009). Mathematical modeling of surface roughness for evaluating the effects of cutting parameters and coating material. *Journal of Materials Processing Technology*, 209(1), 102-109.
- Chinchanikar, S., & Choudhury, S. K. (2013). Effect of work material hardness and cutting parameters on performance of coated carbide tool when turning hardened steel: An optimization approach. *Measurement*, 46(4), 1572-1584.
- Das, S. R., Dhupal, D., & Kumar, A. (2015). Experimental investigation into machinability of hardened AISI 4140 steel using TiN coated ceramic tool. *Measurement*, 62, 108-126.
- Elbah, M., Yallese, M. A., Aouici, H., Mabrouki, T., & Rigal, J. F. (2013). Comparative assessment of wiper and conventional ceramic tools on surface roughness in hard turning AISI 4140 steel. *Measurement*, 46(9), 3041-3056.
- Grzesik, W. (1998). The role of coatings in controlling the cutting process when turning with coated indexable inserts. *Journal of Materials Processing Technology*, 79(1), 133-143.
- Grzesik, W., & Nieslony, P. (2004). Prediction of friction and heat flow in machining incorporating thermophysical properties of the coating–chip interface. *Wear*, 256(1), 108-117.
- Hessainia, Z., Belbah, A., Yallese, M. A., Mabrouki, T., & Rigal, J. F. (2013). On the prediction of surface roughness in the hard turning based on cutting parameters and tool vibrations. *Measurement*, 46(5), 1671-1681.
- Lin, T. R. (2002). Experimental design and performance analysis of TiN-coated carbide tool in face milling stainless steel. *Journal of Materials Processing Technology*, 127(1), 1-7.
- Myers, R.H., & Montgomery, D.C. (2002). *Response surface methodology: process and product optimization using designed experiments*. 2nd ed. John Wiley and Sons, Inc.: New York,
- Sahoo, A. K., & Sahoo, B. (2012). Experimental investigations on machinability aspects in finish hard turning of AISI 4340 steel using uncoated and multilayer coated carbide inserts. *Measurement*, 45(8), 2153-2165.
- Sahoo, A. K., & Sahoo, B. (2013). Performance studies of multilayer hard surface coatings (TiN/TiCN/Al₂O₃/TiN) of indexable carbide inserts in hard machining: Part-I (An experimental approach). *Measurement*, 46(8), 2854-2867.
- Tebassi, H., Yallese, M., Khettabi, R., Belhadi, S., Meddour, I., & Girardin, F. (2016). Multi-objective optimization of surface roughness, cutting forces, productivity and Power consumption when turning of Inconel 718. *International Journal of Industrial Engineering Computations*, 7(1), 111-134.
- Yallese, M. A., Chaoui, K., Zeghib, N., Boulanouar, L., & Rigal, J. F. (2009). Hard machining of hardened bearing steel using cubic boron nitride tool. *Journal of Materials Processing Technology*, 209(2), 1092-1104.
- Zahia, H., Athmane, Y., Lakhdar, B., & Tarek, M. (2015). On the application of response surface methodology for predicting and optimizing surface roughness and cutting forces in hard turning by PVD coated insert. *International Journal of Industrial Engineering Computations*, 6(2), 267-284.
- Zhang, K., Deng, J., Meng, R., Gao, P., & Yue, H. (2015). Effect of nano-scale textures on cutting performance of WC/Co-based Ti 55 Al 45 N coated tools in dry cutting. *International Journal of Refractory Metals and Hard Materials*, 51, 35-49.

