THE TURNING FORCE ANALYSIS USING THE "MINITAB" SOFTWARE

Maria NEAGU¹

¹ Department of Manufacturing Engineering, "Dunărea de Jos" University of Galați, România

ABSTRACT

This paper presents a new way to analyze the experimental results of the cutting force of the cylindrical turning process. Making use of the modern technology and the Minitab software, this paper shows the way to analyze the influence of the parameters (the thickness and the feed rate) on the cutting force and to conduct the ANOVA analysis for this particular manufacturing process. The determination of the capability of this technological process is the last step of this analysis. This work leads the way of the modern analysis of other manufacturing processes and experimental results.

Keywords: turning, Minitab, ANOVA, DOE, cutting force.

1. INTRODUCTION

The experimental data analysis plays a very important part for the complete understanding of an experimental process and for the future predictions of the process characteristics.

The process used as an example in this paper is the cylindrical turning, while the analyzed data are gathered through an experimental set-up aimed to determine the force developed during this engineering process. This experiment constitutes a classical laboratory that the undergraduate mechanical engineering students are attending in order to understand the turning process. The paper analyses the second part of the laboratory: the data analysis.

The experimental and industrial analysis of the data using a modern software is accepted nowadays. Minitab is one of these softwares [1]. In industry, educational process and research, this software is used in a variety of domains: turning [3, 6, 11, 16, $21\div23$], vehicle design/testing [4, 13, 20, 24], non-conventional machining [7, 8, 10, 17, 25], milling [2], extruding [15], casting [5, 9, 12, 18, 19], etc.

Here, the Minitab software is used to examine the dependence of the cutting force of the cylindrical turning process on the process thickness and feed rate as well as the capability of the process as follows:

- The Taguchi method is used to design the experiment, to create the matrix of the experiment and to analyze the results in the steps 1÷3 presented bellow;
- ANOVA method is used to establish the statistical influence of the parameters in the step 4;

- the nonlinear regression analysis gives us the algebraic formula of the dependence of the cutting force on the process parameters in the step 5;
- the capability of the process to give certain results is calculated and presented numerically and graphically by the step 6.

2. MODEL CONSTRUCTION

We are considering the case of a cylindrical turning process with the parameters presented by Table 1. The two parameters, thickness ("t" [mm]) and feed rate ("s" [mm/rot]) have four levels of variation. We design the experiments having in view Table 1 and Taguchi's Design Of Experiments (DOE) theory that indicates the necessity of $L_4^2 = 16$ experiments.

Table 1. Factors and levels of the analysis.

Factor	Unit	Level			
		1	2	3	4
t	mm	0.5	1.0	1.5	2.0
S	mm/rot	0.063	0.128	0.204	0.317

The modern technology helps us to organize, develop and analyze in a better way the experiment. Following the steps defined by Neagu [14], using the "Minitab" software, the steps of the model construction are the following:

- the determination of the experimental matrix (Step 1);
- the construction of the experimental matrix (Step 2).

Step 1. The determination of the experimental matrix

The "Stat \rightarrow DOE \rightarrow Taguchi \rightarrow Create Taguchi Design" sequence allows us to choose the number of factors (2) and the number of levels (4) as Fig. 1 indicates.

Taguchi Design			×		
Type of Design (2 to 31 factors) 2-Level Design (2 to 31 factors) 3-Level Design (2 to 13 factors) 4-Level Design (2 to 5 factors) 5-Level Design (2 to 6 factors) Mixed Level Design (2 to 26 factors)					
Number of factors: 2 💌	-	Display Availa	ble Designs		
·		Designs	Factors		
		Options			
Help		ОК	Cancel		

Fig. 1. The DOE set-up.

Modify	Factors			×
Factor	Name	Level Val	ues	Levels
Α	t [mm]	0.40.8 1.0 1.5		4
В	s [mm/rot]	0.096 0.151 0.208 0.250		4
B	s [mm/rot]	0.096 0.151 0.208 0.250		

Fig. 2. The definition of the factors name and the level values.

U Worksheet 1 ***						
Ŧ	C1	C2	C3			
	t [mm]	s [mm/rot]	Fz [daN]			
1	0.4	0.096	10.8			
2	0.4	0.151	13.5			
3	0.4	0.208	16.2			
4	0.4	0.250	18.9			
5	0.8	0.096	16.2			
6	0.8	0.151	18.9			
7	0.8	0.208	24.3			
8	0.8	0.250	27.0			
9	1.0	0.096	21.6			
10	1.0	0.151	24.3			
11	1.0	0.208	35.1			
12	1.0	0.250	37.8			
13	1.5	0.096	29.7			
14	1.5	0.151	37.8			
15	1.5	0.208	40.5			
16	1.5	0.250	45.9			

Fig. 3. The working sheet of the experiment.

The "Stat \rightarrow DOE \rightarrow Modify Design" selection gives us the possibility to define the name and the level values as Table 1 indicates. Figure 2 reflects this process, while Fig. 3 shows the new form of the worksheet.

Step 2. The construction of the experimental matrix

The results of the experiment are placed on the third column " C_3 " of the worksheet and they have the values indicated by Fig. 3.

3. MODEL ANALYSIS

This analysis contains the following elements:

- the analysis of the Taguchi model (Step 3);
- ANOVA analysis of the experimental results (Step 4);
- the nonlinear regression analysis of the cutting force (Step 5);
- the capability analysis of the technological process (step 6).

Step 3. The analysis of the Taguchi model.

The "Stat \rightarrow DOE \rightarrow Taguchi \rightarrow Define Custom Taguchi Design" sequence allows us to define the factors (Fig. 4) while the "Stat \rightarrow DOE \rightarrow Taguchi \rightarrow Analyze Taguchi Design" selection allows us to define the response of the analysis (Fig. 5). In the window presented by Fig. 5, we choose:

Define Custom Taguchi Design			
C1 t [mm] C2 s [mm/rot] C3 Fz [daN]	Factors: ['t [mm]' 's [mm/rot]'	< >	
	Signal Factor ⓒ No signal factor ⓒ Specify by column:		
Select			
Help	OK Cano	:el	

Fig. 4. The definition of the factors.

Analyze Taguchi Design			×
C3 Fz [daN]	Response data are	in:	
Select	Graphs Analysis Graphs	Analysis Options	Terms Storage
Help		ОК	Cancel

Fig. 5. The analysis response definition.

■ "Graphs", "Analysis" and "Storage" that allows us to choose, further: "Signal to Noise ratios" and "Means".

■ "Terms" that allows us to define the analysis factors (Fig. 6)

■ "Options" is followed by the selection "Smaller is better" (Fig. 7) because we are looking for a smaller cutting force.

Analyze Taguchi Design: Terms				
Available Terms:	Selected Term A:t [mm] B:s [mm/rot] <<	ns:		
A:t [mm] B:s [mm/rot] Help	ОК	ancel		

Fig. 6. The terms selection.

×					
Use adjusted formula for nominal is best					
Use In(s) for all standard deviation output					
Cancel					

Fig. 7. The options of the Taguchi design.

Taguchi Analysis: Fz [daN] versus t [mm], s [mm/rot]

Response Table for Signal to Noise Ratios

Smaller is better

Level	t [mm]	s [mm/rot]
1	-23.25	-25.25
2	-26.51	-26.85
3	-29.21	-28.74
4	-31.60	-29.74
Delta	8.35	4.48
Rank	1	2

Response Table for Means

Level	t [mm]	s [mm/rot]
1	14.85	19.57
2	21.60	23.63
3	29.70	29.02
4	38.48	32.40
Delta	23.63	12.82
Rank	1	2

Fig. 8. The Taguchi analysis: F_z [daN] versus t[mm], s[mm/rot].



Fig. 9. Cutting force variation.



Fig. 10. SN variation.

As a result of these selections, in the session window, we obtain the following results:

• the mean cutting force and the Signal/Noise ratios for each level (Fig. 8).

• the cutting force variation (Fig. 9) as well as the SN (Signal to Noise) variation (Fig. 10) as a function of the two parameters.

Figure 8 reveals that the depth, "t", has the maximum influence on the cutting force as its rank is 1, while the rank of the feed rate, "s", is 2. This result is valid for the mean values analysis as well as the signal to noise ratios analysis.

The increase of the mean values with both parameters ("t" and "s") is revealed graphically by Fig. 9 which shows that a smaller depth and feed rate is desired in order to have a smaller cutting force.

Figure 10 reveals the SN, signal to noise ratio, variation on the process parameters. As we desire a high value of SN, smaller process parameters are desired. We regain the same conclusion obtained from Fig. 9.



Fig. 11. The ANOVA selection.

Step 4. ANOVA analysis of the experimental results. The "Stat \rightarrow ANOVA \rightarrow One Way" selection and setting successively "t" (Fig. 11) and "s" as the factors, we obtain the ANOVA analysis for each factor.

Figure 12 and Fig. 13 present the ANOVA analysis results for the "t" factor and the "s" factor, respectively. They show that both parameters have a significant influence on the cutting force as the "F-value" (11.49 for the depth, "t", and 1.19 for the feed rate, "s") is bigger than the threshold. We notice, again, that the depth has a higher influence than the feed rate.

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
t [mm]	3	1251.6	417.20	11.49	0.001
Error	12	435.6	36.30		
Total	15	1687.2			

Fig. 12. ANOVA analysis results for "t".

Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
s [mm/rot]	3	387.7	129.2	1.19	0.354
Error	12	1299.4	108.3		
Total	15	1687.2			
Fig 13 ANOVA analysis results for "s"					

Step 5. The nonlinear regression analysis of the turning force.

The "Stat \rightarrow Regression \rightarrow Nonlinear Regression" selection allows us to search for a formula of the vertical force of the turning process:

$$Fz = c \cdot t^a \cdot s^b \tag{1}$$

Nonlinear Re	gression		
[C1 t [mm] C2 s [mm/r C3 Fz [daN	ot]]	Response: Expectation Use Ca Edit direct	'Fz [daN]' Function talog Use Ca dy: '^a* 's [mm/rot]'^b
		P <u>a</u> rameters	. Optio <u>n</u> s
. Sele	ct	<u>G</u> raphs	<u>R</u> esults
Help			

(Fig. 14), where "a", "b" and "c" are real constants.

Fig. 14. The nonlinear regression.

In the iteration process that find these unknowns, the starting values for the "a", "b" -and "c" parameters as well as their lower and upper values are given. The case considered here is presented by Fig. 15.

Nonlinear Regressi	on: Parameters					
Function						
c*'t [mm]'^a* 's [mm/rot]'^b						
Required starting values:						
Parameter	Values	Locked				
С		50				
a	0.2					
b	0.1					
Optional <u>c</u> onstraints:						
rarallieter	Lower bound	opper bound				
C	0	100				
a	0	1				
b	0	1				

Fig. 15. The parameters of the nonlinear regression.

For this particular situation, the coefficients that fit the best the experimental values as well as the formula of the cutting force is given by Fig. 16:

Equation

Fz [daN] = 71.4272 * 't [mm]' ^ 0.755525 * 's [mm/rot]' ^ 0.525815

Fig. 16. The cutting force (F_z) formula.

Step 6. The capability analysis of the technological process.

The "Stat \rightarrow Quality Tools \rightarrow Capability Analysis \rightarrow Normal" selection leads us to the software window presented by Fig. 17. We specify: "Data are arranged in" : "F_z[daN]"; "Lower Specification Limit" (LSL): 5; "Upper Specification Limit" (USL): 50.

CL t[mm] C3 x[mm/ml] C3 Fz[deA]	Data are arranged as 4 Single calumn: Tro Inte	10	Tanfen.
	Sdgnesm js		Edware .
	(are a constant or an ID or	Storage	
	ritine. Hec.	5	F Bundary
	Hattinical wears		(letim)
the l	Herein a standard de cation	-	- Instand

Fig. 17. The capability analysis.

The capability analysis graph presented by Fig. 18 reveals a C_p value of 0.7 and a C_{pk} value of 0.65 which indicates the capability of the process to deliver the results in the required quality domain.



Fig. 18. The process capability.

4. CONCLUSIONS

This paper presents a new way to study the cutting force of the turning process, the influence of the parameters and the process capability.

Concentrating on the second part of the laboratory, the data analysis, this work makes use of the modern technology, the Minitab software, to establish the degree of influence of the parameters on the cutting force, the ANOVA statistical analysis of the influence of these parameters, the cutting force formula as well as the turning process capability.

These results can be an example of the way in which the modern technology can be used to analyze the experimental data and to understand the technological process.

REFERENCES

[1] http://www.minitab.com, *Minitab, User's guide 2: data analysis and quality tools*, Version 13.2. USA: Minitab Inc.; 2000.

[2] Abou-El-Hossein, K. A., Kadirgama, K., Hamdi, M., Benyounis, K. Y., *Prediction of cutting force in end-milling operation of modified AISI P20 tool steel*, Journal of Materials Processing Technology 182, 2007, pag. 241-247.

[3] Ahmed, G. M. S., Quadri, S. S. H., Mohiuddin, Md S., *Optimization of Feed and Radial Force in Turning Process by using Taguchi Design Approach*, Materials Today: Proceedings 2, 2015, pag. 3277-3285.

[4] Baloni, B. D., Pathak, Y., Channiwala, S. A., *Centrifugal blower volute optimization based on Taguchi method*, Computers & Fluids 112, 2015, pag. 72-78.

[5] Baradeswaran, A. Elayaperumal, A., Issac, R. F., *A Statistical Analysis of Optimization of Wear Behavior of Al-Al2O3 Composites Using Taguchi Technique*, Procedia Engineering 64, 2013, pag. 973-982.

[6] Bilga, P. S., Singh, S., Kumar, R., Optimization of energy consumption response parameters for turning operation using Taguchi method, Journal of Cleaner Production 137, 2016, pag. 1406-1417.

[7] Chandramouli, S., Eswaraiah, K., Optimization of EDM Process parameters in Machining of 17-4 PH Steel using Taguchi Method, Materials Today: Proceedings 4, 2017, pag. 2040-2047.

[8] Chaudhury, P., Samantaray, S., Sahu, S., Multi Response Optimization of Powder Additive Mixed Electrical Discharge Machining by Taguchi Analysis, Materials Today Proceedings 4, 2017, pag. 2231-2241.

[9] Dabade, U. A., Bhedasgaonkar, R. C., Casting Defect Analysis using Design of Experiments (DoE) and Computer Aided Casting Simulation Technique, Procedia CIRP 7, 2013, pag. 616-621.

[10] Hanizam, Soufhwee, A. R. H., Anuar, A. R. K., Nizam, A. R. Md., Mohamad, N., *The Effect of Pulse DC and DC Substrate Bias during In Situ Cleaning PVD Process on Surface Roughness*, Procedia Engineering 53, 2013, pag. 562-568.
[11] Kumar, R., Bilga, P. S., Singh, S., *Multi objective*

[11] Kumar, R., Bilga, P. S., Singh, S., Multi objective optimization using different methods of assigning weights to energy consumption responses, surface roughness and material removal rate during rough turning operation, Journal of Cleaner Production 164, 2017, pag. 45-57.

[12] Mhalla, M. M., Bahloul, A., Bouraoui, Ch., Analytical models for predicting tensile strength and acoustic emission count of a glass fiber reinforced polyamide using response surface method, Journal of Alloys and Compounds 695, 2017, pag. 2356-2364.

[13] Mitra, A. C., Soni, T., Khan, K. G. R., S., Banerjee, N., Experimental Design and Optimization of Vehicle Suspension System, Materials Today: Proceedings 2, 2015, pag. 2453-2462.

[14] Neagu, M., Inginerie Concurențială. Lucrări de laborator (Concurrent Engineering. Laboratories), Tehnopress, Iași, Romania, 2017.

[15] Pouvafar, V., Sadough, S. A., Hosseinj, F., Rahmani, M.R., Design of experiments for determination of influence of different parameters on mechanical properties of semi-solid extruded parts, Trans. Nonferrous Met. Soc. China 20, 2010, pag. 794-797.

[16] Qasim, A., Nisar, S., Shah, A., Khalid, M. S., Sheikh, M. A., Optimization of process parameters for machining of AISI-1045 steel using Taguchi design and ANOVA, Simulation Modeling Practice and Theory 59, 2015, pag. 36-51.

[17] Raju, P., Sarcar, M. M. M., Satyanarayana, B., Optimization of wire electric discharge machining parameters for surface roughness on 316 L stainless steel using full factorial experimental design, Procedia Materials Science 5, 2014, pag. 1670-1676.

[18] Rezayati-charani, R., Mohammadi-Rovshandeh, J., Effect of pulping variables with dimethyl formamide on the characteristics of bagasse-fiber, Bioresource Technology 96, 2005, 1658-1669.

[19] Rostamiyan, Y., Fereidoon, A., Ghalebahman, A. G., Mashhadzadeh, A. H., Salmankhani, A., Experimental study and optimization of damping properties of epoxy-based nanocomposite: Effect of using nanosilica and high-impact polystyrene by mixture design approach, Materials and Design 65, 2015, pag. 1236-1244.

[20] Salleh, N. A. M., Kasolang, S., Jaffar, A., Green Lean Total Quality Information Management in Malaysian Automotive Companies, Procedia Engineering 41, 2012, pag. 1708-1713.

[21] Sankar, B. R., Rao, P. U., Analysis of Forces during Hard Turning of AISI 52100 Steel Using Taguchi Method, Materials Today: Proceedings 4, 2017, pag. 2114-2118.

[22] Sharma, P., Sidhu, B. S., Sharma, J., Investigation of effects of nanofluids on turning of AISI D2 steel using minimum quantity lubrication, Journal of Cleaner Production 108, 2015, pag. 72-79.

[23] Surya, M. S., Shalini, M., Sridhar, A., Multi-Response Optimization On En19 Steel Using Grey Relational Analysis Through Dry & Wet Machining, Materials Today: Proceedings 4, 2017, pag. 2157-2166.

[24] Tan, K. S., Wong, S. W., Umar, R. S. R., Hamouda, A. M. S., Gupta, N. K., *An experimental study of deformation behavior of motorcycle front wheel-tire assembly under frontal impact loading*, International Journal of Impact Engineering 32, 2006, pag. 1554-1572.

[25] Tiwari, A., Mandal, A., Kumar, K., Optimization of Overcut in Electrochemical Machining for EN 19 Tool Steel using Taguchi Approach, Materials Today: Proceedings 2, 2015, pag. 2337-2345.