

Optimization of CNC Turning Parameters & Prediction of Surface Roughness using Machine Learning Algorithm

Vinay P. Mohite¹,

¹UG Student, Mechanical Engineering Department,
K. K. Wagh Institute of Engineering Education and Research, Nashik -422003, India.

Shiphan R. Pathan²,

²UGStudent, Mechanical Engineering Department,
K. K. Wagh Institute of Engineering Education and Research, Nashik -422003, India.

Adarsh R Patkal³,

³UGStudent, Mechanical Engineering Department,
K. K. Wagh Institute of Engineering Education and Research, Nashik -422003, India.

Divya V. Raut⁴

⁴UGStudent, Mechanical Engineering Department,
K. K. Wagh Institute of Engineering Education and Research, Nashik -422003, India.

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Abstract—In industries, Machining process is gaining importance & more attention in order to manufacture high-quality precision robust products with great performance, which qualifies all stages of its application. In order to achieve it within less time, with less machining cost & maintenance, the need for optimization of machining processes arises. Optimization is a tool in engineering practice for continuous improvement & to get the most excellent results from the available set of resources. Among different machining processes, Turning is most widely used in order to gain the desired dimension and precise surface finish. The project's aim is to optimize various machining parameters of Turning operations i.e Spindle speed, Depth of cut, and Feed rate to gain a good surface finish by practicing optimization techniques of Design Of Experiments (DOE) & employing Taguchi methodology with the least number of experimental runs by developing a mathematical model based on statistical regression and machine learning algorithm for predicting the surface roughness values within the selected specified range.

Keywords: Design of Experiments (DOE), CNC Turning, Surface finish, Optimization, Machine-Learning, Experimentation, Decision-Tree, Flask, Prediction, Python.

1. INTRODUCTION

Optimization is the science of getting the most excellent results from the available resource constraints. There is a lot of scope for optimizing various manufacturing functions such as design, planning, operation (process), quality control, maintenance, and so forth. Nowadays in modern manufacturing processes, optimization techniques in metal cutting are compulsory for manufacturing concerns to respond effectively to severe competitiveness and growing demand for quality products in the market. Optimization methods in metal cutting processes is considered to be viable tool for continuous improvement of the output quality in products and process include modeling of input-output and in process parameter relationship and determination of optimal cutting condition for getting robust product or process. A robust product or process is one that is insensitive to noise. Metal cutting process, generally understood by developing an explicit mathematical model, which can be of two types; mechanistic and empirical, the mechanistic model are inadequate and unacceptable to metal cutting process, but the empirical models are commonly used.

Among all the cutting processes, turning is the most widely used. Turning is the removal of metal from the outer diameter of a rotating cylindrical work piece. Turning is used to reduce the diameter of the work piece, usually to a specified dimension, and to produce a smooth finish on the metal. Often the work piece will be turned so that adjacent sections have different diameters. Turning is the machining operation that produces cylindrical parts. Today's fast changing manufacturing environment requires the application of optimization techniques in metal cutting processes. The machine tool operators always face the problem in turning process. Machine tool condition, job clamping, tool and work piece geometry and cutting parameters are the major reasons for occurrence of the problem. In machining, there

has been recently and intensive computation focusing on surface roughness at international level. This computation can be observed in turning processes especially in plane and automotive industry by increasing the alternative solutions for obtaining more proper surface roughness. So it becomes important to study the effect of machining parameters on multiple quality characteristics like surface roughness and vibration etc.

In present industrial age, the growing competition call for all the effort to be directed towards the economical manufacture of machined best quality part and all these is made possible by use of high-end technology machines i.e CNC lathe. Generally cutting parameters are selected on the basis of experience of operator or from the handbook. Mostly it does not ensure selection of optimal or near optimal parameters. To select the cutting parameters properly, several mathematical models based on statistical regression techniques or neural computing have been constructed to establish a relationship between the response and machining parameters. For getting these relations large number of experiments are to be conducted. In order to reduce the number of experiments of design of experiment (D.O.E) techniques (Taguchi) have been used. D.O.E is a tool in engineering application to reduce the no of experiments without comprising the quality.

2. PROBLEM STATEMENT

To optimize the CNC parameters in turning operations in order to reduce surface roughness of the desired workpiece.

ELABORATION:

Accomplishing a smooth surface finish is significant while machining metal, as it influences both the usefulness and aesthetics of the end product. The determination of cutting parameters, for example, spindle speed, feed rate, and depth of cut, alongside the kind and state of the cutting device, assume a huge part in accomplishing the ideal surface completion. it may, even with proper choice of these parameters, there might be examples where the surface completion isn't great. In such cases, rework is frequently required, which includes extra machining activities, sanding, or other surface planning strategies. This can be a tedious and work-concentrated process, bringing about diminished productivity and expanded costs.

To limit the expense and time related to accomplishing the ideal surface completion and stay away from the requirement for modification, it is fundamental to improve the cutting parameters and utilize suitable cutting procedures. This can be achieved by utilizing measurable strategies, like Design of Experiments (DOE), to recognize the ideal blend of cutting parameters that will bring about the ideal surface finish efficiently. Moreover, high-level machining advancements like CNC Lathe can improve the precision and consistency of machining activities, decreasing the probability of surface roughness issues.

3. OBJECTIVES

1. To study the influence of different machining parameters i.e. speed, feed, depth of cut etc. on surface roughness of machined work piece and predict the best performance characteristics using ANOVA analysis.
2. Determination of optimized values of cutting parameters for minimization /maximization of desired responses using the DOE optimization technique with a minimum number of experimentations.
3. Analysis and prediction of response variable i.e. (Ra, Vibrations) by establishing the empirical equation of in-process parameters & responses using the Minitab statistical Tool.
4. Validation of the predicted results by performing confirmation experiments and checking its accuracy.

4. LITERATURE REVIEW

PrateekVaishnav (Research Scholar, Aravali Institute of Technical Studies, Udaipur, Rajasthan, India.)GauravPurohit (Asst. Prof. & Head, Aravali Institute of Technical Studies, Udaipur, Rajasthan, India) is The goal of the study is to improve the cutting conditions when milling the 8011 aluminium alloy using Computer Numeric Controlled (CNC) equipment in order to achieve the proper surface roughness and enhanced material removal rate (MRR). 27 experimental runs of the full factorial design of experiments technique were used to examine the performance traits under various cutting conditions. The results were analysed using an orthogonal array, analysis of variance (ANOVA), and signal to noise ratio. According to the study's findings, it is possible to accurately determine the quality of surface

finish and MRR by taking into account independent criteria that have a major impact. The results of this study can help several fields of study and academics working on this topic.

Shrikant S. Jachak ,Vinay R. Pandey: In the present disquisition an attempt is made to estimate the effect of certain slice variables on face roughness in plain turning of medium carbon sword AISI 1055 under cutting condition. Cutting speed, depth of cut, feed and slice inflow rate are named as the impacting parameters. The trials are conducted by factorial trial medium carbon sword AISI 1055 was machined using tenacious clicked tool and compared the performance with brazed tool.The cutting condition of turning parameters was determined by Design of experiment method to find the optimal levels and to analyse the effect on the turning parameters.

Jacqueline K. Telford: In an experimental design, deliberate modifications to the input variables or their processing are made, and the consequences on the response variables are then observed. It is an effective way to expand knowledge while requiring less data to be collected. Factorial experimental designs can evaluate sensitivity to individual components and combined effects while simultaneously examining the impacts of numerous factors. It has been used to increase data with fewer simulation runs in ballistic missile defence sensitivity tests. In competitive sectors, experimental design is crucial for testing and evaluation.

5. METHODOLOGY

This chapter sets forth the methodology and work flow followed in this project in brief. Provides in-depth insights on workflow and research methodology followed in this project.



Fig.5.1 ResearchWorkflow

6. EXPERIMENTATION

In order to satisfy the desired objectives, the experimental layout has been set up and trials have been performed to test the developed algorithm. This chapter covers in-depth details of the experimental setup, procedure and results observed. Different trials have been performed on different objects of different materials.

6.1 Experimental Setup

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6.2 CNC Turning Specification

In order to achieve the goal of this experimental work the cutting tests were carried out in a Sinumerik 828 DB Basic CNC Lathe (JYOTI DX-200 4B). The CNC turning center has 9kw / 13.5 kw spindle motor power and a maximal machining diameter of 150mm, maximal spindle speed of 4000rpm, spindle speed range 40 to 3500 rpm and maximal turning length 400mm.

CNC TURNING JYOTI DX200 4B SPECIFICATIONS	
CAPACITY	
Swing Over Bed	500 mm
Maximum Turning Length	400 mm
Standard Turning Ø	200 mm
Max. Turning Ø	265 mm
SLIDES	
Cross (X-Axis) Travel	150 mm
Longitudinal (Z-Axis) Travel Rapid Feed (X & Z Axis)	400 mm 24 m/min
Ball Screw (X-Axis) & (Z-Axis)	Ø32x10 mm pitch
MAIN SPINDLE	
Spindle Motor Power	9/13.5 kw
Spindle Bore	(S1/S6-40% Rating) 70 mm
Spindle Nose Max. Bar capacity	A2-6 52 mm
Speed Range	50-4000rpm
Full Power Speed Range	750-2250rpm
TURRET	
No. of Stations	8
Max Boring Bar Ø	40 mm
Tool Size (Cross-Sectional)	25x25mm
ACCURACY (As per VDI/DGQ3441)	
Positioning	0.007mm
Repeatability	0.005mm
TAIL STOCK	
Quill Diameter, Quill Stroke	85mm, 120mm
Thrust (Adjustable) (Max.)	300 (500) Kgf
MACHINE WEIGHT APPROX.	
3500 Kg	
DIMENSIONS (L X W X H)	
2420 X 1500 X 1710 mm	

Fig. 6.1. CNC Turning Specifications

6.3 Work Piece Material:

In this experimental investigations, EN 8 steel rods have been selected and the work piece were prepared from the raw materials dimension of length 70mm and diameter 33mm. The chemical composition of the work piece is shown in below table.

Table No. 6.3. Chemical composition of En8 Material

Elements	C	Si	Mn	S	P
Percentage	0.36-0.44	0.10-0.40	0.60-1.00	0.50(max)	0.50

TableNo. 6.4.MechanicalPropertiesofEn8

Maxstress	700-850n/mm2
Yieldstress	465n/mm2min
0.2% proofstress	450n/mm2min
Elongation	16% min
ImpactKCV	28Joulesmin
Hardness	201-255Brinell

KeyFeatures:

1. Unalloyedmediumcarbonsteel
2. Reasonabletensilestrength
3. Canbeflameorinductionhardened
4. Readilymachinable
5. Moderatewearresistance ifheattreated.

6.4 Selectionof cuttingtool:

Cuttingtoolsmusthavethreepropertiesinordertogeneratehigh-qualityproducts:hardness,toughness,andwear resistance. Thecarbide-coated toolis used tocuttheselectedworkpiece.

TableNo.6.5.CuttingToolSpecificationsTNMGInserts

Tooltype	Manufacturer	Designation	Code	Geometry	Grade
Carbidetool	KORLOY	TNMG	160404	VF	CC115

TableNo.6.6.InsertSpecification

Inserttype	Code	Length of the insertin mm	Thickness of insertin mm	Radiusofinserts
TNMG is a triangularinsert	160404	16	04	0.4



Fig.6. .2.TNMGinsert

6.5 Machinetool:

On the CNC machine tool, the turning operation is performed on the specified work piece EN8 steel bar.The specifications of the CNC machine tool are of Jyoti-DX-200 4B (Sinumerix) CNC system, the numberofworkstations is 8, thechuck sizeis 169 mm,and the spindlespeedrangeis 45-4500 rpm.



Fig.6.3.CNCLatheMachine&itsControllerTable

No.6.7.CNC machinespecifications

S.No	Description	Specifications
1.	Maximum Swing over carriage	150mm
2.	Distancebetweencenters	220mm
3.	Heightofcenter	165mm
4.	Spindleborediameter	70mm
5.	Spindlespeed	50-4000rpm

6.6 Surface Roughness Ra Measurement Apparatus used:

Roughness is usually a good predictor of the performance of mechanical parts, because surface irregularities may form cracks or corroded nucleation sites. Roughness is a measure of surface texture. It is quantified by the vertical deviation of the actual surface from its ideal form. In the trials, a digital surface tester was utilized to measure surface roughness Ra. SURFTTEST, S.J.210, sr.no: 178-561-02A, MITUTOYO MECH, an experimental device for the surface roughness tester of JAPAN.



Fig.6.4. Mitutoyo SJ-210 Surface roughness

tester Table No.6.8. Specification Of Surface Roughness Tester

Brand Name	Mitutoyo
Height	10.92 centimeters
Weight	2.27 kilograms
Length	40.13 centimeters
Width	34.29 centimeters
Manufacturer Series Number	178-561-02A

6.7 Orthogonal array and experimental factors:

The Taguchi technique begins with the selection of a suitable orthogonal array. The L9 (3 3) orthogonal array, which is a standardized experimental design based on Taguchi. The basic design uses up to three control factors, of which three levels are used for cutting speed C_0 , cutting depth D and feed F . A total of 9 experimental runs are performed, using the control factors Horizontal combination. The dependent variable is called the response variable.

The purpose of the experiment is to examine how turning factors affect EN8 steel surface roughness. Taguchi's orthogonal array was utilized to create the experiment, lowering the number of trials. The experiment is carried out using the three factors of the L9 orthogonal array's three levels.



Fig.6.5. Testspecimenafter machining

6.8 DOEforCNC Turning:

TableNo.6.9.Factorsandtheirlevels inCNCStraightTurning

ProcessParameter	Level1	Level2	Level3
Spindlespeed(m/min)	1500	2000	2500
Feedrate(mm/rev)	0.160	0.175	0.190
Depthofcut(mm)	0.2	0.4	0.6

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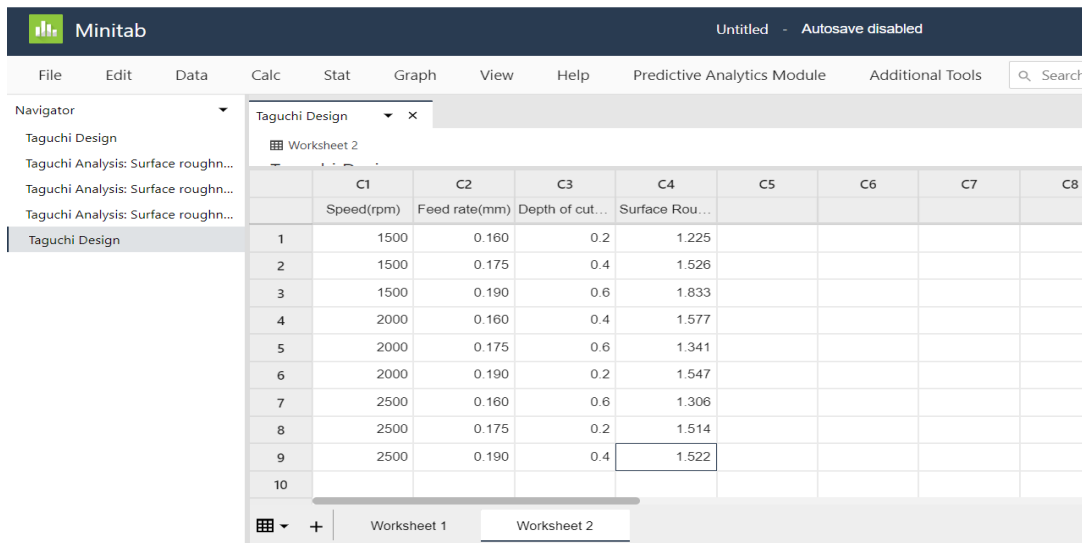


Fig.6.6.MinitabTaguchiDesign Plan

7. CONCLUSION

Taguchi methods have been very successful in designing high-quality products and processes of many different fields. In this work an approach is proposed for selecting the most preferred set of parameters for optimal operation of lathe machining. The selection criteria for parameters are based on physical measured outputs of machining i.e. surface roughness.

For our project, we have selected three input turning parameters for optimization i.e. Cutting speed, feed rate, depth of cut. For each parameter, we selected three levels: low, medium and high.

Experiments were conducted using L9 orthogonal array. For each experiment, surface roughness was measured, recorded and analysis using Taguchi S/N ratios. These ratios were calculated with consideration of performance characteristic: Lower-the-

Better, as surface roughness is requested to be low. The optimum level of parameters for minimizing the surface roughness were determined from the response table for Signal-to-Noise ratios. The best combination was obtained with:

- 1) Cutting speed
- 2) Feed rate
- 3) Depth of cut
- 4) Confirmation experiment,
- 5) Development of regression model with interactions between parameters.

From the response for a signal to noise ratio it is observed that Depth of cut is impacting more on the surface roughness followed by feed rate and spindle speed.

Design of experiment was done using Taguchi method and we obtain optimized level of speed 2500 and feed rate was 0.4 and depth of cut 0.160.

Machine learning was used for predicting surface roughness was decision tree algorithm was used.

8. FUTURE SCOPE

Very few attempts have been made to investigate the challenges faced by this technology and discover the scope for optimization and refinement. This chapter provides the future scope for the researcher to work in this domain.

Optimization of parameters is important in every machining process results affected by each parameter focus is to produce good quality within a minimum cost. CNC turning process is the most common machining process that is used in now days. Present work has been done, Tool vibration and surface roughness are two important parameters which affect quality of component and tool life which indirectly affect component cost. To optimize the machining parameters, for material EN8, depth of cut, Feed rate, Speed. MINITAB software is used for formulating matrix and for the analysis of Taguchi methodology for response.

The Machine learning field has encountered a turning point, which allows machines to learn, improve, and perform a specific task through data without being explicitly programmed. Machine learning can be utilized with machining processes to improve product quality levels and productivity rates, to monitor the health of systems, and to optimize design and process parameters. The domain of the study is to find the best suitable model of machine learning algorithm to find accuracy and predict response variable. Data is most important part of machine learning without data we can't train any model and dataset will be used in this study will be data obtained from experiments.

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