



Review of Surface Roughness Prediction in Cylindrical Grinding process by using RSM and ANN

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Abstract— In recent years, one of the innovative technologies which enable automatic computerized devices for the purpose of evaluation, inspection, analyzing of various static and dynamic images is Machine Vision System. MVS provides various benefits in various fields like Surveillance cameras for Security, MAT LAB for analysis of Engineering designs, RSM and ANN for analysis of various parameters for optimizing variables, etc. MVS has some various innovative techniques like automatic capturing of images, evaluation and processing capabilities. This paper is about optimization of various parameters such as depth of cut, feed rate, cutting speed, dressing conditions, etc. For improving the accuracy obtained in Cylindrical (Surface) grinding process. The main objective of our idea is to identify the surface roughness of work piece obtained in Cylindrical grinding process in according to various input parameters like feed rate, depth of cut, dressing speed, etc. The method used here is RSM and ANN which are used to optimize the response of input parameters on output parameters and to provide algorithm for pre-trained model for visualization along with simulation of neural networks.

Keywords— Surface Roughness, Response Surface Methodology, Artificial Neural Network, Work piece Materials, Methodology.

I. INTRODUCTION

The determination of surface roughness of cylindrical job has been made easy from the following researches. The researches we used for analysis are Marius Winter, Wen Li (2013), determining the optimal process parameters to increase eco-efficiency of grinding process using the input parameters like cutting depth, cutting speed, dressing speed, work piece speed. These features results in fine surface finish along with geometrical accuracy using the algorithm of empirical model and geometric programming. M.Janardhan (2015), an integrated evaluation approach for modeling and optimizing of surface grinding process parameters which as Wheel speed and Depth of cut using Response Surface Methodology.

N.Sudheer Kumar Varma, S.Rajesh(2016), Neural network and fuzzy logic based prediction of surface roughness and MRR in cylindrical grinding process using input parameters such as work speed, depth of cut, feed rate based on the principle Neural Network Modeling and ANFIS on AISI1040 STEEL. Jae-SeobKwak(2005) application of Taguchi and response surface methodology for geometric error in surface grinding process by input parameters as Wheel speed, Depth of cut, on

the principle RSM and Taguchi. Jae-SeobKwak(2006), an analysis of grinding power and surface roughness of hardened steel using RSM by optimizing Depth of cut, Work piece speed.

Kuldip Singh Sangwan(2015), optimization of machining parameters such as Cutting speed, Feed ,depth of cut to minimize surface roughness using Integrated ANN-GA approach. Kiran A patel(2016) A study of RSM and ANN model for surface roughness using RSM and ANN by thee input parameters of Spindle Speed, Interference, Feed and Number of tool passes.

II. LITERATURE SURVEY

S.No	Author &Year	Title	Input Parameters	Work piece Material	Methodology/ Algorithm	Result / Output Parameter
1.	Marius Winter(2013)	Determining optimal process parameters to increase the eco-efficiency of grinding process.	Cutting depth, Cutting speed, Dressing speed, Work piece speed.	Hardened Carbon Steel 1.3505 [DIN100G6]	Empirical Model and Geometric Programming.	Great Geometrical accuracy and Surface Roughness.
2.	M. Janardhan (2015)	An integrated evaluation approach for modeling and optimization of Surface grinding process parameters.	Wheel speed, Depth of cut, Table speed, Dressing condition.	AISI4 340 STEEL	Response surface methodology and NSGA-II.	Good Surface roughness and desired surface finish.
3.	Jae-Seob Kwak (2006)	An analysis of grinding power and surface roughness in external cylindrical grinding process.	Depth of cut, Work piece speed, Traverse speed, Chip thickness, Contact length.	Hardened SCM440 STEEL	Response Surface Methodology.	Surface roughness and grinding power.

4	N Saran Kumar, N Kaleeswaran, B Radha Krishnan	Review on optimization parametrs in Abrasive Jet Machining process	Nozzle Diameter, Standoff Distance, Particle Size	steel	ANN, RSM	Surface roughness
5.	Kiran A Patel(2016)	A Study of RSM and NN model for surface roughness.	Spindle speed, Interference, Feed and number of tool passes.	Aluminium Alloy 6061	RSM and ANN.	Surface roughness.
6.	Kuldip Singh Sangwan (2015)	Optimization of machining parameters to minimize surface roughness.	Cutting speed, Feed, Depth of cut.	Ti-6AL-4v Titanium alloy	Integrated ANN -GA.	Surface roughness.
7.	Jae-Seob Kwak(2005)	Application of RSM and Taguchi for geometric error in Surface grinding.	Wheel speed, Depth of cut, Table speed, Dressing condition.	High speed tool STEEL (SKH51)	Taguchi and RSM.	Surface roughness and geometric error.
8.	N. Sudheer Kumar Varma(2016)	NN and ANFIS based prediction of Surface roughness and MRR in cylindrical grinding process.	Work speed, Depth of cut, Feed rate.	AISI1040 STEEL	Neural Network Modeling and ANFIS.	Surface roughness and MRR.
9.	Rogelio L. Hecker(2003)	Predicative modeling of surface roughness.	Chip thickness, Depth of cut, Speed ratio.	Hardened Steel 52100 with 62HRC	Analysis of grooves left by grains.	Surface roughness.
10.	Grynal D Mello (2017)	Surface roughness modeling in high speed	Cutting speed, Feed rate, Depth of cut, tool	Ti-6Al-4V Titanium alloy	ANN	Surface roughness (R_a , R_t)

		turning	flank wear.			
11	B.Radha Krishnan, M. Ajith, .Ajith kumar, P. Bala, G.Gwalbert Maurice	Determination of Surface Roughness in AA6063 Using Response Surface Methodology	Cutting speed, Feed rate, Depth of cut	AA6063	Response Surface Methodology	Surface roughness (R_a)
12.	Yanni Chen (2016)	A nested ANN Prediction model for Surface roughness.	Cutting forces and Tool vibration.	Ti-6Al-4V Titanium alloy	Nested –ANN and RSM	Surface Roughness (R_a)
13.	M.N.Dhavlikar (2002)	Combined taguchi and dual response method for optimization of grinding operation	Feed speed, Depth of cut, Angle of regulating wheel	Steel	Taguchi and dual response method	Roundness error
14.	Mei-Ling Huang (2016)	Validation of a method using Taguchi, RSM, Neural network and genetic algorithm	Cutting depth, Type of cutting fluid, Spindle speed.	Carbon Steel	Taguchi, RSM, Back-propagation, Genetic algorithm.	Process yield.
15.	Sathish T, Tamizharasan N (2017)	Optimisation of Surface Roughness in CNC Milling Process Using RSM	Cutting speed, Feed rate, Depth of cut.	Aluminium 6063	Response Surface Methodology	Surface roughness, R_z
16.	Ravinder Kumar (2015)	Study of Surface roughness using RSM and ANN.	Cutting speed, Feed rate.	Al 7075 hard ceramic and hybrid composites.	RSM and ANN	Surface roughness.
17.	B.Radhakrishnan, S.TharunKumar (2017)	Optimization of CNC machining parameters for surface roughness in	Cutting speed, Feed rate, Depth of cut.	Aluminium 6063 T6	Response Surface Methodology	Surface roughness, R_z

		turning of aluminium 6063 T6 with response surface methodology				
18.	Kuldip Singh Sangwan (2017)	Optimization of machining parameters for improving energy efficiency	Cutting speed, Feed rate, Depth of cut.	AISI 1045 Steel	RSM and Genetic algorithm.	Power Consumption
19.	B Radha Krishnan, R Aravindh (2018)	Prediction of Surface Roughness (AISI 4140 Steel) in Cylindrical Grinding Operation by RSM	Cutting speed, Feed rate, Depth of cut.	AISI 4140 Steel	Response Surface Methodology	Surface roughness.

III. EXPERIMENTAL SETUP

A. Work piece Material

We have selected EN 19(V 320) STEEL for our experiment. It has the following Chemical composition and heat treatment specification respectively.

Table 1. Work piece chemical Composition

<u>Chemical</u>	<u>Composition (%)</u>
Carbon	0.35 TO 0.45
Silicon	0.10 TO 0.30
Manganese	0.5 TO 0.80
Chromium	0.90 TO 1.40
Molybdenum	0.20 TO 0.40
Hardening Temperature	900-950 °C
Quenching Medium	OIL
Tempering Temperature	200-225 °C
Brinell Rockwell Hardness	45-55 RC

B. Equipment

Cylindrical Grinding Machine has been used to carry out our experiment It has the following specifications: (All dimensions are in mm)

Table 2. Grinding Machine Specification

Size	355
Maximum length of work-piece	355
Maximum diameter of work-piece	76
Height of centre	130
Number of speed	3
Spindle speed range	4/58-130rpm
Maximum table swivel	7 degrees
Power required	2 HP
Approx. weight	1400 kgs



Fig .1.Cylindrical Grinding Machine

C. Camera

For image capturing and processing, we have selected a Canon EOS 600D With the following specification:

- 18-megapixel CMOS sensor
- Scene Intelligent Auto mode
- Full-HD EOS Movie
- On-screen Feature Guide
- Up to 3.7fps continuous shooting

- Wide-area 9-point AF
- 1,040k-dot vari-angle 7.7cm (3.0”) screen
- Basic+ and Creative Filters
- Built-in wireless flash control

IV. CONCLUSION

From the review papers, it is clear that the prediction of surface roughness of AISI 4040 STEEL in cylindrical grinding process is efficient by the principle of ANN. The optimum machining parameters observed for best surface roughness are 400 rpm work speed with 50 microns depth of cut and a feed rate of 1600 mm/min result efficiency of 91%. The material for our experiment is AISI 4140 STEEL [C%:0.35 TO 0.45, Si%:0.10 TO 0.30, Mn%:0.5 TO 0.80, Cr%:0.90 TO 1.40, Mo%:0.20 TO 0.40] using Response Surface Methodology and Artificial Neural Network with machining parameters such as Cutting speed, Depth of cut, Feed rate. The material AISI 4140 is used in components of mediums & large cross section, requiring high tensile strength & toughness such as in automobile engineering, gear & engine construction. Our conclusion is that the obtained efficiency in our experiment is about 95% using RSM and ANN.

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