

# Optimization Of Process Parameters Of Wirecut Electric Discharge Machine Processed En 31 Steel By Grey Relational Analysis

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## Abstract:

WIRE EDM has become an important non-traditional machining process as it provides an effective solution for producing components made of difficult to machine materials like titanium, zirconium etc. and intricate shapes which are not possible by conventional machining methods. Due to large number of process parameters and responses lots of researchers have attempted to model this process. This paper reviews the research trends in WEDM on relation between different process parameters like pulse on time(TON), pulse off time(TOFF), servo voltage(sv) on different process responses like Material Removal Rate, Surface Roughness. In addition, the paper highlights different modelling and optimization methods. The final part of the paper includes some recommendations about the trends for future wire Edm researches.

The main objectives of this study investigate and evaluate the effect of different input process parameters (pulse on time, pulse off time, servo voltage) on material removal rate, surface roughness as response parameters have been considered for each Experiment. Experimentation was planned as per Taguchi's L27 Orthogonal array during machining of EN 31 work material which is High Carbon High Chromium Die Steel (HCHCR). Brass wire electrode with 0.25mm Diameter was used as tool in the Experiments. The results are analyzed using analysis of variance (ANOVA) method. This parametric analysis (ANOVA) shows the percentage contribution of parameters individually. Grey relational analysis are applied to determine the suitable selection of machining parameters for wire cut EDM process. A grey relational grade obtained from the grey relational analysis is used to optimize the process parameters. By analyzing the Grey relational grade we find the optimum parameters. Confirmation test has been

conducted to optimize the process parameter. Further mathematical models are developed using Box-Behnken design of experiments of response surface methodology to optimize the process parameters using state of art optimization techniques for future studies.

## 1. INTRODUCTION:

### 1.1 Material Selection :

The material selected for this process is EN 31 steel, which is an high carbon- chromium steel. This steel offers high measure of hardness with compressive strength and abrasion resistance. These steels retain their hardness up to a temperature of 4500c. It has increased wear resistance.

Element	Content (%)
C	1.50
Mn	0.52
Si	1.3
Cr	0.22
Ni	0.33
Mo	0.06
P	0.05

### 1.1.1 CHEMICAL COMPOSITION OF EN-31 STEEL

ELEMENT	METRIC
Tensile Strength	750 N/mm <sup>2</sup>
Yields stress	450 N/mm <sup>2</sup>
Reduction of Area	0.45

Elongation	0.3
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### 1.1.1 Mechanical Properties of EN-31

Thermal Conductivity(w/m-k)	Density (Kg/m <sup>3</sup> )	Melting point ( 0c )
46.6	7810	1540

### 1.1.2 Physical Properties of EN-31

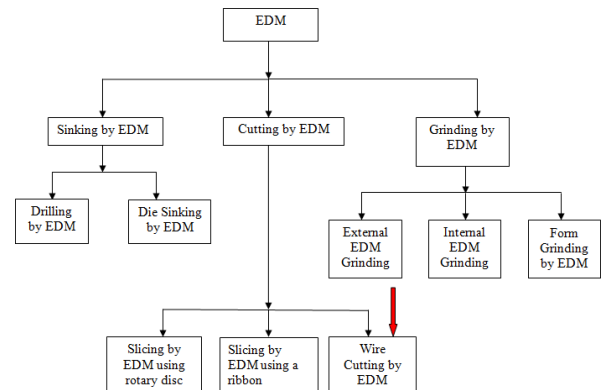
## 1.2 Applications of EN-31:

Typical applications in stamping dies, metal cutting tools or any other industries because of its high strength and heavy weight.

- Aerospace industry
- Punching dies
- Ball and Roller Bearings
- Spinning tools
- By its character this type of steel has high resisting nature against wear and can be used for components which are subjected to severe abrasion, wear or high surface loading.

### 1.3 Classification of EDM

There are several EDM processes such as Wire Electrical Discharge Machining, Electrical Discharge Milling, Electrical Discharge Grinding (EDG), Electrical Discharge Dressing (EDD), Ultrasonic Aided EDM (UEDM), Abrasive Electrical Discharge Grinding (AEDG), Micro Electrical Discharge Machining (MEDM), Micro Wire EDM (MWEDM), Mole EDM, and Double Rotating Electrodes EDM [3]. Pandey and Shan [7] classified EDM processes into three main categories as shown in Fig. 1.2.



### 1.4 Wire Electric Discharge Machining (WEDM)

WEDM is considered as a unique adoption of the conventional EDM process which comprises of a main worktable, wire drive mechanism, a CNC controller, working fluid tank and attachments. The work piece is placed on the fixture table and fixed securely by clamps and bolts. The table moves along X and Y-axis and it is driven by the DC servo motors. Wire electrode usually made of thin copper, brass, molybdenum or tungsten of diameter 0.05-0.30 mm, which transforms electrical energy to thermal energy, is used for cutting materials. The wire is stored and wound on a wire drum which can rotate at 1500rpm. The wire is continuously fed from wire drum which moves through the work piece and is supported under tension between a pair of wire guides located at the opposite sides of the work piece. During the WEDM process, the material is eroded ahead of the wire and there is no direct contact between the work piece and the wire, eliminating the mechanical stresses during machining. Also, the work piece and the wire electrode (tool) are separated by a thin film of dielectric fluid that is continuously fed to the machining zone to flush away the eroded particles. The movement of table is controlled numerically to achieve the desired three-dimensional shape and accuracy of the workpiece.

### Application



**Fig. 1.8 various shapes cut with wire cut EDM**

Wire electrical discharge machining is mainly used to cut contour shapes and design into hard metals, which are otherwise difficult to machine. Contour parts can be easily cut with help of wire cut EDM. Fig. 1.8 shows application of WEDM for the complex parts

- Tooling elements
- Automobile parts
- Plastic Molding.
- Dies, Hobs Blanking.
- Shear Blades.
- Hot Shearing Tools.
- Hardened Rolls.
- Thread Rolling Dies.
- Blade Cutters for Wire Nails.
- Dies for Nut Manufacturers etc.

## 2. LITERATURE REVIEW:

1. Kruth, et. al. of Katholieke University, Belgium studied and experimentally tested several compositions of wires, with high tensile core and several coatings. They have found that, while cutting with prototype wires, a significant rise in accuracy is obtained, especially in corner cutting, while the cutting rate is at a comparable level as commercial reference wire.

2. Prohaszka et al (1996) proposed the requirements of the materials that can be used as WEDM electrodes and will lead to the improvement of WEDM performance. He discussed the material requirements for fabricating WEDM electrodes for improving WEDM performance. Experiments were carried out regarding the choice of suitable wire electrode materials, the

effects of the material properties on the machinability of WEDM. He evaluated the influence of the various materials used for the fabrication of wire electrodes on the machinability of WEDM. A series of experiments have been conducted on a standard EDM unit. Negative polarity rods of pure magnesium, tin and zinc, of a diameter of 5.0 mm were used as the tool electrodes.

3. Tarng et al. [12] formulated a neural network model and simulated annealing algorithm in order to predict and optimize the surface roughness and cutting velocity of the WEDM process when machining SUS-304 stainless steel materials.

4. Gokler et al. [17] investigated under various experimental conditions the surface roughness achievable for 1040, 2379, and 2738 steel materials and the relative machining parameters for the WEDM process.

5. Puri et al. [18] investigated the variation of geometrical inaccuracy caused due to wire lag with various machine control parameters.

6. Ramakrishnan et al. [19] developed a mathematical model using the response surface methodology. A good amount of research has already been done in the area of WEDM technology. To the best of the knowledge of the authors of this work, there is not found any published paper for optimizing multiple performance characteristics of the WEDM process using the Taguchi method. Keeping this consideration in view, this paper describes optimization of multiple performance characteristics using the robust parametric design approach, for achieving a better material removal rate, surface finish, and wire wear ratio.

## 3. DESIGN OF EXPERIMENT:

### 3.1 WEDM Process Parameters and Response Variables

The main goals of WEDM are to achieve a better stability and higher productivity. As newer and more exotic materials are developed, and more complex shapes are required, conventional machining operation will continue to reach their limitations and the increased use of WEDM in manufacturing will continue to grow at

an accelerated rate [26]. However, due to a large number of variables in WEDM, it is difficult to achieve the optimal performance of WEDM processes and the effective way of solving this problem is to establish the relationship between the response variables of the process and its controllable input parameters.

### 3.2 Process Parameters

Throughout this dissertation work input parameters considered for Wire cut EDM are machining parameters like as pulse on time, pulse off time, servo voltage and output parameters are surface roughness and material removal rate.

#### 3.3.2 Steps to perform Taguchi design of the experiment

1. Identify the main function, side effects, and failure mode.
2. Identify the noise factors, testing conditions, and quality characteristics.
3. Identify the objective function to be optimized.
4. Identify the control factors and their levels.
5. Select the orthogonal array matrix experiment.
6. Conduct the matrix experiment.
7. Analyze the data; predict the optimum levels and performance.
8. Perform the verification experiment and plan the future action. Identify the main function, side effects, and failure mode.

#### 3.3.3 Key terms used in DOE

##### 1. Factors

These are variables that have direct influence on the performance of the product or process under investigation.

##### Factors are of two types:

- (a) **Discrete** - Assumes known values or status for the level. Example: Container, Vendor, Type of materials, etc.
- (b) **Continuous** - Can assume any workable value for the factor Levels. Example: Temperature, Pressure, Thickness, etc.

##### 2. Levels

This is the values or descriptions that define the condition of the factor held while performing the experiments.

Sr. No.	Machining process parameter	Leve 11	Leve 12	Leve 13
1	Pulse on Time ( $\mu$ s)	110	120	130
2	Pulse Off Time ( $\mu$ s)	45	55	65
3	Servo Voltage (volt)	20	30	40

#### 3.3 Factors with levels value

Sr. No.	Fixed Parameters	Set Value
1	Wire material	Brass (0.25mm)
2	Peak current (A)	230
3	Pulse peak voltage(V)	2
4	Wire feed rate (m/min)	4
5	Flushing Pressure(kgf/cm <sup>2</sup> )	12
6	Wire Tension (kgf)	9

### 3.6 Experimental design

As per table, L27 orthogonal array of “Taguchi method” has been selected for the experiments design in MINITAB 17. Box-Behnken design of experiments in Response Surface Methodology for developing mathematical models for responses in MINITAB 17.

## CHAPTER 4

### EXPERIMENT WORK AND MEASUREMENT

The experimental setup and the experiment is designed and carried out at the Vinayaka wire cut which is placed at old Airport Road Hyderabad. The primary goal of the dissertation work is to predict the Material Removal Rate and surface roughness. The work is carried out in sprint cut wire cut electro discharge machine of EN 31 HCHCR material by varying machining parameters.

**Fig. 4.1 Sprint cut Wire Cut EDM**



**Fig. 4.3 Control Cabinet**

### WORK TABLE

<b>Design</b>	<b>Fixed column,</b>
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	<b>moving table</b>
<b>Table size</b>	<b>440 x 650 x 300 mm</b>

### MAX.WORK PIECE DIMENSION

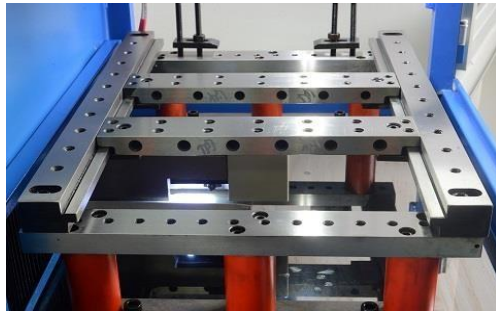
<b>Max. work piece height</b>	<b>200 mm</b>
<b>Max. work piece weight</b>	<b>500 kg</b>
<b>Main table traverse (X,Y)</b>	<b>300 x 400 mm</b>
<b>Aux. table traverse (U,V)</b>	<b>80 x 80 mm</b>
<b>Wire electrode diameter</b>	<b>0.25 mm (std.) 0.15, 0.20 mm (opt.)</b>

### PULSE GENERATOR

<b>Pulse Generator</b>	<b>ELPULS-40 A DLX</b>
<b>Pulse peak voltage</b>	<b>1 Step</b>
<b>CNC Controller</b>	<b>EMT 100W-5</b>
<b>Input power supply</b>	<b>8.3 phase, AC, 415 V , 50 Hz</b>
<b>Connected load</b>	<b>10 kVA</b>
<b>Average power consumption</b>	<b>6 to 7 kVA</b>

### DIELECTRIC SYSTEM

<b>Dielectric Unit</b>	<b>DL 25 P</b>
<b>Dielectric fluid</b>	<b>Deionized water</b>
<b>Tank capacity</b>	<b>250 Liters</b>
<b>Cooling system</b>	<b>1700 k Cal</b>



**Fig. 4.4 Work Table**

## CHAPTER 5

### ANOVA ANALYSIS:

#### 5.1 Introduction

The analysis of variance is the statistical treatment most commonly applied to the results of the experiment to determine the percent contribution of each factors. Study of ANOVA table for a given analysis helps to determine which of the factors need control and which do not. Once the optimum condition is determined, it is usually good practice to run a confirmation experiment. In case of fractional factorial only some of the tests of full factorial are conducted. The analysis of the partial experiment must include an analysis of confidence that can be placed in the results. So, analysis of variance is used to provide a measure of confidence. The technique does not directly analyse the data, but rather determines the variability (variance) of the data. Analysis provides the variance of controllable and noise factors. By understanding the source and magnitude of variance, robust operating conditions can be predicted.

EXP NO	T on ( $\mu$ s)	T off ( $\mu$ s)	SV(V)	MRR ( $\text{mm}^3/\text{min}$ )
1	110	45	20	0.0320
2	110	45	30	0.0325
3	110	45	40	0.0299
4	110	55	20	0.0339
5	110	55	30	0.0330
6	110	55	40	0.0325
7	110	65	20	0.0345
8	110	65	30	0.0338
9	110	65	40	0.0331
10	120	45	20	0.0540
11	120	45	30	0.0579
12	120	45	40	0.0669
13	120	55	20	0.0488
14	120	55	30	0.0475
15	120	55	40	0.0520
16	120	65	20	0.0472
17	120	65	30	0.0445
18	120	65	40	0.0535
19	130	45	20	0.0640
20	130	45	30	0.0683
21	130	45	40	0.0655
22	130	55	20	0.0618
23	130	55	30	0.0698
24	130	55	40	0.0668
25	130	65	20	0.0575
26	130	65	30	0.0670
27	130	65	40	0.0645

Total no of runs = n = 27

**Table 5.1 Control parameter**

#### 5.1.1 Table for factor with level

Total degree of freedom =  $f_T = n - 1 = 26$

## CHAPTER 6

### MULTI RESPONSE OPTIMIZATION

#### 6.1 Grey relational analysis for multi objective optimization

In grey relational analysis, the function of factors is neglected in situations where the range of the sequence is large or the standard value is enormous. However, this analysis might produce incorrect results if the factors, goal and directions are different. Therefore, one has to preprocess the data which are related to a group of sequence, which is called "grey relational generation". "Data preprocessing is a process of transferring the original sequence to a comparable sequence for this purpose the experimental results are normalized in the range between zero and one; the normalization can be done from three different approaches.

Fig. 6.1 Graph of Grey relational grades

**Table No: 6.5 Response table for means of Grey relational grade**

Machining parameters	Average Grey relational grade by factor level			Main effect [Maximum]	Rank
	Level 1	Level 2	Level 3		
Pulse On Time, A	0.372461	0.609508	0.882459*	0.509938	1
Pulse Off Time, B	0.650298*	0.618361	0.595769	0.054529	2
Servo Voltage, C	0.591199	0.627601	0.645629*	0.054430	3

**Fig. 6.2 Graph of main effects plot for means for Grey relational grade**

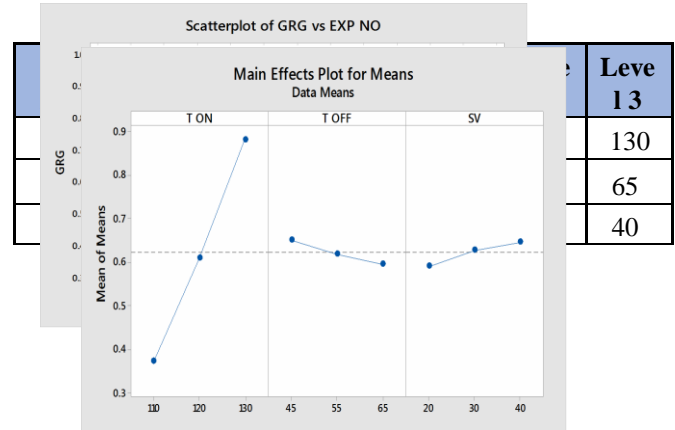


Table No: 6.5 shows average grey relational grade by factor level. From this table, one has concluded optimum parameter levels which are indicated by "\*". In this table, higher grey relational grade from each level of factor indicates the optimum level. From this table, it is concluded that the optimum parameter level for Pulse on time, Pulse off time and Servo voltage is (110  $\mu$ s), (36  $\mu$ s) and (20volts) respectively.

#### 6.6 Confirmation test

	Grey theory prediction Design	Experimental values	
LEVEL	A1B2C3	A1B2C3	A1B2C3
Material Removal Rate(mm /min)	-	0.0698	0.0783
Surface Roughness ( $\mu$ m)	-	3.4925	3.5014
Grey Relational Grade	0.9353	0.9987	-

**Table No: 6.6 Table for confirmation experiment**

- ANOVA for GRG of brass wire electrode shows Toff and SV are most significant parameters and contributes with 30.9894% and 29.9966% respectively, SV is insignificant parameter and contributes 24.9449%.
- The error between Grey theory prediction design using software and experimental values using Grey relational analysis for GRG is 0.14999% which shows that model developed is significant.

## CHAPTER 7

### RESULT AND DISCUSSION:

After performing the experiment for all 27 runs and measuring the output parameters like material removal rate and surface roughness for wire cut EDM of EN 31 HCHCR, whatever results generated are discussed in this chapter

#### 7.1 Main Effect Plots for Input Parameters V/S Output Parameters

**Fig. 7.1 Graph of input parameters v/s material removal rate**

**Fig. 7.2 Graph of input parameters v/s surface roughness**

## CHAPTER 8

### DEVELOPMENT OF MATHEMATICAL MODELS

The experimental results are used to obtain the



mathematical relationship between process parameters and machining outputs. The coefficients of mathematical models are computed using method of multiple regressions. In this study, SPSS, Minitab17 (Software Package for Statistical Solutions), for regression analysis custom made software created by the author was used for the regression analysis. This software is used to test several models, viz., linear, exponential, power series (user-defined). Out of all models tested, the model that has high coefficient of multiple determination ( $R^2$ ) value is chosen. The relationship between response variable(s) and process parameters can be expressed as

EXP.N	Ton	Toff	SV	MRR	SR
1	110	55	20	0.0339	3,1970
2	130	55	40	0.0668	3.4692
3	110	65	30	0.0338	3.0250
4	120	45	40	0.0669	3.3808
5	120	45	20	0.0540	3.3575
6	120	65	20	0.0472	3.2907
7	120	55	30	0.0475	3.2907
8	130	65	30	0.0670	3.4890
9	130	55	20	0.0618	3.4578
10	120	55	30	0.0475	3.2907
11	120	55	30	0.0475	3.2907
12	120	65	40	0.0535	3.3496
13	110	45	30	0.0365	3.1919
14	110	55	40	0.0325	3.0145



15	130	45	30	0.0683	3.4416
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**Table No: 8.1 BOX-BEHNKEN design of experiments**

## CHAPTER 9

### CONCLUSION:

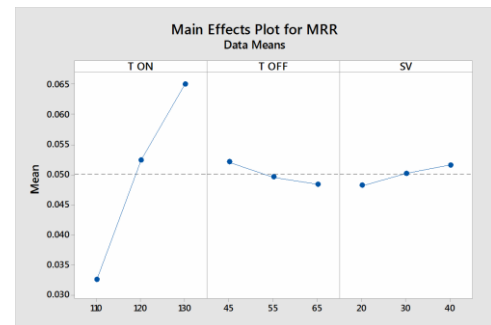
In the presented work, experiments are carried out for material removal rate and surface roughness with variables as pulse on time, pulse off time and servo voltage. There are 27 experimental readings taken for all variables to conduct the parametric study.

Finally, it can be concluded that:

- The ANOVA analysis is conducted to know the percentage contribution of the input parameters on output parameters. ANOVA analysis results that the percentage contribution of pulse on time is 87.2%, pulse off time is 1.81% and servo voltage is 1.81% for material removal rate, which shows that the influence of Pulse off time is very less compare to other parameters. The percentage contribution of Pulse on time is 74.2%, Pulse off time is 1.01% and Servo voltage is 15.9% for Surface Roughness which shows that the influence of Pulse off time is very less compare to other parameters. The error and percentage contribution of interaction terms found in ANOVA analysis is 9.18% for material removal rate and 8.89% for surface roughness.
- Grey relational analysis is done to find out optimal parameter levels. After grey relational analysis, it is found that optimal parameter levels are pulse on time at level 1 (110  $\mu$ s), pulse off time at level 2 (55  $\mu$ s), servo voltage at level 3 (40 volts). The results of optimum parameters are material removal rate of 0.0698 mm<sup>3</sup>/min, surface roughness of 3.4925  $\mu$ m .
- Process parameters do not have some little effect for every response. Significant parameters and its percentage contribution changes as per the behaviour of the parameter

with objective response.

- Increase of Pulse on time generates more spark energy as the length of time that electricity supply increases.



MRR and SR all response increasing with pulse on time. Pulse on time found most significant parameter in all response. Surface roughness also increases with increase of pulse on time because the increases of pulse on time produce crater with broader and deeper characteristic.

- Pulse off time has opposite effect to pulse on time. MRR decrease with increase of pulse off time, while surface roughness first increases and slightly reduces further when level of pulse on time increases. During off time removed material flushed away. More the off time better the flushing. Pulse off time found least significant parameter in all response.
- Servo voltage has little effect on SR and MRR. MRR and Surface roughness reduces with increase of servo voltage.
- Mathematical models for responses which are developed are fitted into the data are further optimized using state of art techniques like Teacher Learner Based Optimization (TLBO) and Particle Swarm Optimization (PSO).

## CHAPTER 10

### FUTURE SCOPE

- The mathematical model can be developed with different work piece and electrode materials for EDM and WEDM processes.



- Responses like Kerf width, roundness, circularity, cylindricity, machining cost etc. are to be considered in further research.
- The standard optimization procedure can be developed and the optimal results are to be validated with different Multi criteria decision making method.
- The weightage can be finding with different techniques such as AHP, Weighting approach etc.

The developed mathematical model can be further optimized using state of art techniques like Teacher Learner Based Optimization (TLBO) and Particle Swarm Optimization (PSO).

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