

#### 3 Days Online Workshop on Design of Experiments-An Engineering Perspective

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## TAGUCHI'S TECHNIQUE FOR OPTIMIZATION

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#### WHAT IS MEANT BY OPTIMIZATION

- 1. Optimization is the act of obtaining the best result under given circumstances.
- 2. Optimization can be defined as the process of finding the conditions that give the maximum or minimum value of a function.



Mathematical programming or optimization techniques	Stochastic process techniques	Statistical methods
Calculus methods Calculus of variations Nonlinear programming Geometric programming Quadratic programming Linear programming	Statistical decision theory Markov processes Queueing theory Renewal theory Simulation methods Reliability theory	Regression analysis Cluster analysis, pattern recognition Design of experiments Discriminate analysis (factor analysis)
Integer programming Stochastic programming Separable programming Multiobjective programming Network methods: CPM and PERT Game theory	Methods of	f Operation Research

Modern or nontraditional optimization techniques

Genetic algorithms Simulated annealing Ant colony optimization Particle swarm optimization Neural networks Fuzzy optimization



#### **INTRODUCTION**

- Genichi Taguchi, (January 1, 1924 June 2, 2012) was an engineer and statistician.
- Dr. Genechi Taguchi is a Japanese scientist who spent much of his professional life researching ways to improve the quality of manufactured products.
- 3. Taguchi methods have been controversial among some conventional Western statisticians, but others have accepted many of the concepts introduced by him as valid extensions to the body of knowledge.



### INTRODUCTION

- After World Wat· II, the Japanese telephone system was badly damaged and dysfunctional, subsequently he was appointed as head of Japan's newly formed Electrical Communications Laboratories (ECL) of Nippon Telephone and Telegraph Company.
- 5. Much of his research at ECL involved developing a comprehensive quality improvement methodology that included use of the DOE technique.
- 6. Taguchi's concept was adopted by many companies, including
  Toyota, Nippon Denso, Fuji Film and other Japanese firms.
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### **INTRODUCTION**

- 7. Here are some of the major contributions that Taguchi has made to the quality improvement world:
  - 1. The Quality Loss Function
  - 2. Orthogonal Arrays and Linear Graphs
  - 3. Robustness



### WHY TAGUCHI'S DOE?

- 1. The quality engineering method that Taguchi proposed is commonly known as the Taguchi method or Taguchi approach.
- 2. His approach is a new experimental strategy in which he utilizes a modified and standardized form of DOE.
- 3. Taguchi approach is a form of DOE with special application principles.
- For most experiments practiced in the industry today, the difference between DOE and the Taguchi approach is in the method of application.

#### WHAT IS NOVEL IN TAGUCHI'S DOE?

- Taguchi showed that DOE could be used not only to improve quality, but also to quantify the improvements made in terms of savings in economy aspect.
- 2. To make the technique easier and friendlier to apply, Taguchi standardized the application method.
- 3. For laying out experiments, he created a number of special orthogonal arrays, each of which is used for a number of experimental situations.

#### WHAT IS NOVEL IN TAGUCHI'S DOE?

- 4. To analyze the results of the experiments for the purpose of determining the design solution that produces the best quality.
- 5. He introduced signal-to-noise ratio (S/N) for analysis of repeated results helps experimenters easily assure a design is robust to the influence of uncontrollable factors.



#### NEED FOR TAGUCHI'S DoE

- 1. Optimisation of processes and formulations
- 2. Reduction in the variation of products and processes i.e. improved robustness
- 3. Investigation of process variables
- 4. Investigation of noise variables
- 5. Reduced costs and improved efficiency

- In order to draw statistically sound conclusions from the experiment, it is necessary to integrate simple and powerful statistical methods into the experimental design methodology.
- 2. In the context of DOE in manufacturing, one may come across two types of process variables or factors: qualitative and quantitative.
- For quantitative factors, one must decide on the range of settings and how they are to be measured and controlled during the experiment.
   For example, in injection moulding process, screw speed, mould temperature, etc. are examples of quantitative factors. 12

- 5. Qualitative factors are discrete in nature. Type of raw material, type of catalyst, type of supplier, etc. are examples of qualitative factors.
- A qualitative factor generally requires more levels when compared to a quantitative factor. Here the term 'level' refers to a specified value or setting of the factor being examined in the experiment.
- 7. In the DOE terminology, a trial or run is a certain combination of factor levels whose effect on the output (or performance characteristic) is of interest.

- 8. In the context of statistics, the term 'degrees of freedom' is the number of independent and fair comparisons that can be made in a set of data.
- 9. For example, consider the heights of two students, say John and Kevin. If the height of John is  $H_J$  and that of Kevin is  $H_K$ , then we can make only one fair comparison  $(H_I-H_K)$ .
- 10. The number of degrees of freedom associated with each factor is equal to unity or 1 (i.e. 2-1=1).

 $\therefore$  Degrees of freedom for a main effect = Number of levels -1 14

- The number of degrees of freedom for the entire experiment is equal to one less than the total number of data points or observations.
- 12. The degrees of freedom for an interaction is equal to the product of the degrees of freedom associated with each factor involved in that particular interaction effect.
- 13. The term 'confounding' refers to the combining influences of two or more factor effects in one measured effect.
- 14. Design resolution (R) is a summary characteristic of aliasing or confounding patterns.
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- 15. For designed experiments, designs of resolution III, IV and V are particularly important.
- 16. Design resolution identifies for a specific design the order of confounding of the main effects and their interactions.
- 17. Resolution III designs: These are designs in which no main effects are confounded with any other main effect, but main effects are confounded with two- factor interactions and two-factor interactions may be confounded with each other.

- 18. Resolution IV designs: These are designs in which no main effects are confounded with any other main effect or with any two-factor interaction effects, but two-factor interaction effects are confounded with each other.
- 19. Resolution V designs: These are designs in which main effects are not confounded with other main effects, two-factor interactions or three-factor interactions, but two-factor interactions are confounded with three-factor interactions.

#### **TAGUCHI'S DESIGN OF EXPERIMENT**

- Dr. Taguchi suggested the use of Orthogonal Arrays (OA) for designing the experiments based on Latin square.
- 2. The difference between the number of experiments conducted between full factorial experiments and Taguchi's OA is:

Number of	Number of	Number of exp	periments
factors	levels	Full factorial	Taguchi
3	2	$2^{k}$ 8	4
7	2	128	8
15	2	32,768	16
4	3	81	9
13	3	1,594,323	27

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#### WHAT IS LATIN SQUARE

 A Latin square is an *n x n* table filled with *n* different symbols in such a way that each symbol occurs exactly once in each row and exactly once in each column.

$$\begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 1 \\ 3 & 1 & 2 \end{bmatrix} \qquad \begin{bmatrix} a & b & d & c \\ b & c & a & d \\ c & d & b & a \\ d & a & c & b \end{bmatrix}$$

#### LATIN SQUARE vs. OA

1. Two Latin square of orders *n* are said to be orthogonal to each other if when one is superimposed on the other, the ordered pairs (i, j) of corresponding entries consists of all possible  $n^2$  pairs.

Example of orthogonal Latin squares of order 3:

$$L_1 = \begin{bmatrix} 1 & 2 & 3 \\ 3 & 1 & 2 \\ 2 & 3 & 1 \end{bmatrix} \quad L_2 = \begin{bmatrix} 1 & 3 & 2 \\ 3 & 2 & 1 \\ 2 & 1 & 3 \end{bmatrix}$$

(1,1)	(2,3)	(3,2)
(3,3)	(1,2)	(2,1)
(2,2)	(3,1)	(1,3)

#### **TAGUCHI'S ORTHOGONAL ARRAY**

#### Nomenclature of arrays

L = Latin square  $L_a(b^c) \quad a = \text{number of rows}$  b = number of levelsc = number of columns (factors)

Degrees of freedom associated with the OA = a - 1

Two-level series	Three-level series	Four-level series	Mixed-level series
$L_4(2^3)$	$L_{9}(3^{4})$	$L_{15}$ (4 <sup>5</sup> )	$L_{18} (2^1, 3^7)^{\dagger}$
$L_8(2^7)$	$L_{27}$ (3 <sup>13</sup> )	$L_{64}$ (4 <sup>21</sup> )	$L_{36}$ (2 <sup>11</sup> , 3 <sup>12</sup> )
$L_{16}$ (2 <sup>15</sup> )	$L_{81}$ (3 <sup>40</sup> )		
$L_{32}$ (2 <sup>31</sup> )			
$L_{12} (2^{11})^*$			

\* Interactions cannot be studied

† Can study one interaction between the 2-level factor and one 3-level factor

### **STEPS IN TAGUCHI'S DOE**

- 1. State the problem
- 2. Determine the objective
- 3. Determine the response and its measurement
- 4. Identify factors influencing the performance characteristics
- 5. Separate the factors into control and noise factors
- 6. Determine the number of levels and their values for all factors
- 7. Identify control factors that may interact
- 8. Select the orthogonal array
- 9. Assign factors and interactions to the columns of OA

#### **STEPS IN TAGUCHI'S DOE**

- 10. Conduct the experiment
- 11. Analyze the data
- 12. Interpret the results
- 13. Select the optimal levels of the significant factors
- 14. Predict the expected results
- 15. Run a confirmation experiment

#### **DOE TERMINOLOGY**

- EXPERIMENT It is defined as a study in which certain independent variables are manipulated, their effect on one or more dependent variables is determined.
- PURPOSE To discover something about a particular process (or) to compare the effect of several factors on some phenomena.
- 3. **LEVELS OF A FACTOR** The variation of independent variable under each factor (or) number of different possible values of a factor.
- EFFECT OF A FACTOR Defined as the change of response produced by a change in the level of that factor.
- 5. **TREATMENT** It means the factor.

#### **DOE TERMINOLOGY**

#### 6. **DEPENDENT VARIABLE**

- 1. It is an outcome or response of an experiment.
- 2. It is also called as response variable.
- 3. Criterion used is also a dependent variable.

#### 7. INDEPENDENT VARIABLE (OR) FACTORS

- 1. Variables, which are varied in the experiment.
- 2. Can be controlled at fixed levels.
- 3. Can be varied or set at levels of our interest.
- 4. Can be qualitative or quantitative and can be randomized.



#### **ORTHOGONAL ARRAY SELECTOR**

	[														Numb	er of P	aramet	ers (P)													
		2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30	31
els	2	L4	L4	L8	L8	L8	L8	L12	L12	L12	L12	L16	L16	L16	L16	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32	L32
of Lev	3	L9	L9	L9	L18	L18	L18	L18	L27	L27	L27	L27	L27	L36	L36	L36	L36	L36	L38	L36	L36	L36	L36								
mber	4	L'16	L'16	L'16	L'16	L'32	L'32	L'32	L'32	L'32																					
Z	5	L25	L25	L25	L25	L25	L50	L50	L50	L50	L50	L50																			
-															Л																

#### **ORTHOGONAL ARRAY 2 Level Designs**

12

2

2

L

L

4

#### L4 (2^3)

Run	Columns									
Kull	1	2	3							
1	1	1	1							
2	1	2	2							
3	2	1	2							
4	2	2	1							

#### L8 (2^7)

Dun		Columns												
Kull	1	2	3	4	5	6	7							
1	1	1	1	1	1	1	1							
2	1	1	1	2	2	2	2							
3	1	2	2	1	1	2	2							
4	1	2	2	2	2	1	1							
5	2	1	2	1	2	1	2							
6	2	1	2	2	1	2	1							
7	2	2	1	1	2	2	1							
8	2	2	1	2	1	1	2							

Dun					C	olum	ns				
Kun	1	2	3	4	5	6	7	8	9	10	11
1	1	1	1	1	1	1	1	1	1	1	1
2	1	1	1	1	1	2	2	2	2	2	2
3	1	1	2	2	2	1	1	1	2	2	2
4	1	2	1	2	2	1	2	2	1	1	2
5	1	2	2	1	2	2	1	2	1	2	1
6	1	2	2	2	1	2	2	1	2	1	1
7	2	1	2	2	1	1	2	2	1	2	1
8	2	1	2	1	2	2	2	1	1	1	2
9	2	1	1	2	2	2	1	2	2	1	1
10	2	2	2	1	1	1	1	2	2	1	2
11	2	2	1	2	1	2	1	1	1	2	2
10	2				2		2		2		

2

L

1

2 |

2

1 12 (2411)

#### **ORTHOGONAL ARRAY 2 Level Designs**

#### L16 (2^15)

Dun							C	olum	ns						
Kun	1	2	3	4	5	6	7	8	-9	10	11	12	13	14	15
1	1	1	1	1	1	1	1	1	1	1	1	1	1	1	1
2	1	1	1	1	1	1	1	2	2	2	2	2	2	2	2
3	1	1	1	2	2	2	2	1	1	1	1	2	2	2	2
4	1	1	1	2	2	2	2	2	2	2	2	1	1	1	1
5	1	2	2	1	1	2	2	1	1	2	2	1	1	2	2
6	1	2	2	1	1	2	2	2	2	1	1	2	2	1	1
7	1	2	2	2	2	1	1	1	1	2	2	2	2	1	1
8	1	2	2	2	2	1	1	2	2	1	1	1	1	2	2
9	2	1	2	1	2	1	2	1	2	1	2	1	2	1	2
10	2	1	2	1	2	1	2	2	1	2	1	2	1	2	1
11	2	1	2	2	1	2	1	1	2	1	2	2	1	2	1
12	2	1	2	2	1	2	1	2	1	2	1	1	2	1	2
13	2	2	1	1	2	2	1	1	2	2	1	1	2	2	1
14	2	2	1	1	2	2	1	2	1	1	2	2	1	1	2
15	2	2	1	2	1	1	2	1	2	2	1	2	1	1	2
16	2	2	1	2	1	1	2	2	1	1	2	1	2	2	1

#### **ORTHOGONAL ARRAY 3 Level Designs**

#### L9 (3^4)

		-	-	
Run		Colu	imns	
Kun	1	2	3	4
1	1	1	1	1
2	1	2	2	2
3	1	3	3	3
4	2	1	2	3
5	2	2	3	1
6	2	3	1	2
7	3	1	3	2
8	3	2	1	3
9	3	3	2	1

#### L27 (3^13)

Dun			Columns											
Kull	1	2	3	4	5	6	7	8	9	10	11	12	13	
1	1	1	1	-1	1	1	1	1	1	1	1	1	- 1 -	
2	1	1	1	1	2	2	2	2	2	2	2	2	2	
3	1	1	1	1	3	-3	3	3	3	3	3	3	3	
4	1	2	2	2	1	-1	1	2	2	2	3	3	3	
5	1	2	2	2	2	2	2	3	3	3	1	1	1	
6	1	2	2	2	3	3	3	1	1	1	2	2	2	
7	1	3	3	-3	1	1	1	3	3	3	2	2	2	
8	1	3	3	-3	2	2	2	1	1	1	3	3	3	
9	1	3	3	- 3	- 3	- 3	-3	2	2	2	1	1	1	
10	2	1	2	3	1	2	3	1	2	3	1	2	3	
11	2	1	2	3	2	- 3	1	2	-3	1	2	3	1	
12	2	1	2	3	3	1	2	3	1	2	-3	1	2	
13	2	2	3	1	1	2	3	2	3	1	-3	1	2	
14	2	2	3	1	2	3	1	3	1	2	1	2	3	
15	2	2	3	1	3	1	2	1	2	3	2	3	1	
16	2	3	1	2	1	2	3	3	1	2	2	3	1	
17	2	3	1	2	2	3	1	1	2	3	3	1	2	
18	2	3	1	2	3	1	2	2	3	1	1	2	3	
19	3	1	3	2	1	3	2	1	3	2	1	3	2	
20	3	1	3	2	2	1	3	2	1	3	2	1	3	
21	3	1	3	2	3	2	1	3	2	1	3	2	1	
22	3	2	1	3	1	3	2	2	1	3	3	2	1	
23	3	2	1	3	2	1	3	3	2	1	1	3	2	
24	3	2	1	3	3	2	1	1	- 3	2	2	1	3	
25	3	3	2	1	1	3	2	3	2	1	2	1	3	
26	3	3	2	1	2	1	3	1	3	2	3	2	1	
27	3	3	2	1	3	2	1	2	1	3	1	3	2	

## ORTHOGONAL ARRAY 3 Level

Designs

#### **ORTHOGONAL ARRAY Mixed Level**

#### Designs

#### L8 (2^4 4^1)

Rup		C	olum	ns	
Kull	1	2	3	4	5
1	1	1	1	1	1
2	2	2	2	2	1
3	1	1	2	2	2
4	2	2	1	1	2
5	1	2	1	2	3
6	2	1	2	1	3
7	1	2	2	1	4
8	2	1	1	2	4

#### **ORTHOGONAL ARRAY Mixed Level**

#### Designs

L16 (2^12 4^1)

Dum		Columns											
Kun	1	2	3	4	5	6	7	8	9	10	11	12	13
1	1	1	1	1	1	1	1	1	1	1	1	1	1
2	1	1	1	1	2	2	2	2	2	2	2	2	1
3	2	2	2	2	1	1	1	1	2	2	2	2	1
4	2	2	2	2	2	2	2	2	1	1	1	1	1
5	1	1	2	2	1	1	2	2	1	1	2	2	2
6	1	1	2	2	2	2	1	1	2	2	1	1	2
7	2	2	1	1	1	1	2	2	2	2	1	1	2
8	2	2	1	1	2	2	1	1	1	1	2	2	2
9	1	2	1	2	1	2	1	2	1	2	1	2	3
10	1	2	1	2	2	1	2	1	2	1	2	1	3
11	2	1	2	1	1	2	1	2	2	1	2	1	3
12	2	1	2	1	2	1	2	1	1	2	1	2	3
13	1	2	2	1	1	2	2	1	1	2	2	1	4
14	1	2	2	1	2	1	1	2	2	1	1	2	4
15	2	1	1	2	1	2	2	1	2	1	1	2	4
16	2	1	1	2	2	1	1	2	1	2	2	1	4

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## **TAGUCHI'S FORMULA FOR**

- For analysis, there are 3 categories of performance characteristics, (i.e.) Smaller-the-better, Larger-the-better and Nominal-the-better.
- Smaller-is-the better (Minimize):  $S / N = -10 \log \left( \frac{1}{n} \sum_{i=1}^{n} y_i^2 \right)$
- Larger-is-the better (Maximize):  $S/N = -10\log\left(\frac{1}{n}\sum_{i=1}^{n}\frac{1}{y_i^2}\right)$
- Nominal is the best:

$$S / N = 10 \log\left(\frac{\overline{y}}{s_y^2}\right)$$

## TAGUCHI'S FORMULA FOR OPTIMIZATION

The predicted optimum response ( $\mu_{pred}$ ) is given by

$$\mu_{\text{pred}} = \overline{Y} + (\overline{A}_2 - \overline{Y}) + (\overline{C}_2 - \overline{Y}) + (\overline{E}_1 - \overline{Y})$$
$$= \overline{A}_2 + \overline{C}_2 + \overline{E}_1 - 2\overline{Y}$$



### **EXAMPLE FOR TAGUCHI'S**

## **OPTIMIZATION**

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#### **INPUT CONTROL PARAMETERS**

The turning features such as depth of cut, feed rate and cutting speed were analyzed to obtain an optimal cutting conditions.

<b>Parameter / Level</b>	Level 1	Level 2	Level 3
Cutting Speed (m/min)	120	150	180
Feed Rate (mm/rev)	0.05	0.07	0.09
Depth of Cut (mm)	0.15	0.30	0.45

	Input values							
No	Cutting Speed (m/min)	Feed Rate (mm/rev)	Depth of Cut (mm)					
1	120	0.05	0.15					
2	120	0.07	0.30					
3	120	0.09	0.45					
4	150	0.05	0.30					
5	150	0.07	0.45					
6	150	0.09	0.15					
7	180	0.05	0.45					
8	180	0.07	0.15					
9	180	0.09	0.30					

	Uncut chip	Chip	Chip thickness	S/N ratio
Exp. No	thickness (t)	thickness (t <sub>c</sub> )	ratio $(r_t = t/t_c)$	(LTB)
1	0.36	0.82	0.44	7.131
2	0.39	0.79	0.49	6.196
3	0.45	0.85	0.53	5.514
4	0.37	0.78	0.47	6.558
5	0.42	0.8	0.52	5.680
6	0.33	0.76	0.43	7.331
7	0.30	0.72	0.42	7.535
8	0.26	0.75	0.35	9.119
9	0.29	0.77	0.38	8.404

Factors	Level 1	Level 2	Level 3	Max - Min	Rank
<b>Cutting Speed</b>	6.281	6.523	8.353	2.072	1
Feed Rate	7.075	6.998	7.083	0.085	3
<b>Depth of Cut</b>	7.860	7.053	6.243	1.617	2



#### Interaction Plot for Chip thickness ratio Data Means



Source	DF	Adj. SS	Adj. MS	<b>F-Value</b>	P-Value	% Contribution
Cutting Speed	2	0.018956	0.009478	853.00	0.001	64.13
Feed rate	2	0.000156	0.000078	7.00	0.125	0.53
Depth of Cut	2	0.010422	0.005211	469.00	0.002	35.26
Error	2	0.000022	0.000011			0.07
Total	8	0.029556				

Source	DF	Adj. SS	Adj. MS	<b>F-Value</b>	P-Value	% Contribution
Cutting Speed	2	0.018956	0.009478	853.00	0.001	64.13
Feed rate	2	0.000156	0.000078	7.00	0.125	0.53
Depth of Cut	2	0.010422	0.005211	469.00	0.002	35.26
Error	2	0.000022	0.000011			0.07
Total	8	0.029556				

### **TAGUCHI BASED ANALYSIS – Prediction**

### of Output S/N ratio

For authenticating the confirmed result, the assessed mean of Taguchi

S/N ratio is calculated as, A3B2C1.

$$SNR_{em} = V_{3m} + F_{3m} + D_{1m} - 2SNR_{m}$$

SNR<sub>em</sub>

= 8.353 + 7.083 + 7.860 - (2\*7.052)= 9.192.

## TAGUCHI – MULTIOBJECTIVE OPTIMIZATION



<b>Parameter / Level</b>	Symbol	Level I	Level 2	Level 3
<b>Cutting Tool Shape</b>	А	C (80°)	D(55°)	<b>S(90°</b> )
<b>Relief Angle (°)</b>	В	0	3	7
Nose Radius (mm)	С	0.4	0.8	1.2

Trial	<b>Cutting Insert</b>	<b>Relief Angle</b>	Nose Radius	ISO Insert
No	Shape	(°)	(mm)	Designation
1	С	0	0.4	CNMG 12 04 04
2	С	3	0.8	CAMG 12 04 08
3	С	7	1.2	CCMG 12 04 12
4	D	0	0.8	DNMG 15 04 08
5	D	3	1.2	DAMG 15 04 12
6	D	7	0.4	DCMG 15 04 04
7	S	0	1.2	SNMG 12 04 12
8	S	3	0.4	SAMG 12 04 04
9	S	7	0.8	SCMG 12 04 08

Tricl No	Flank wear	Surface Roughness	Material Removal Rate
Iriai no	(mm)	(µm)	(gm./min)
1	0.143	0.687	0.026
2	0.112	1.898	0.031
3	0.483	1.702	0.042
4	0.118	1.560	0.050
5	0.134	0.814	0.047
6	0.033	1.007	0.047
7	0.168	2.504	0.046
8	0.136	1.245	0.045
9	0.502	1.689	0.047

Trial	Flank	Surface	Material Removal	<b>Combined Objective</b>
No	wear (A)	Roughness (B)	Rate (C)	(A*0.5+B*0.3+C*0.2)
1	16.8933	3.26087	-31.7005	3.085
2	19.0156	-5.56592	-30.1728	1.803
3	6.3211	-4.61919	-27.5350	-3.732
4	18.5624	-3.86249	-26.0206	2.918
5	17.4579	1.78751	-26.5580	3.954
6	29.6297	-0.06059	-26.5580	9.485
7	15.4938	-7.97269	-26.7448	0.006
8	17.3292	-1.90339	-26.9357	2.706
9	5.9859	-4.55259	-26.5580	-3.684

### **OPTIMIZATION**

**Main Effects Plot for Combined Objective** 

Data Means



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## TAGUCHI – MULTIOBJECTIVE OPTIMIZATION

**Interaction Plot for Combined Objective** 



**Nose Radius** 

Source	DF	Seq SS	Adj MS	F	Р	% Contribution
Insert Shape	2	59.544	29.772	3.49	0.223	45.27 %
<b>Relief Angle</b>	2	6.937	3.468	0.41	0.711	5.28 %
Nose Radius	2	47.761	23.88	2.80	0.263	36.37 %
Residual Error	2	17.063	8.531			12.99 %
Total	8	131.304				100 %

## TAGUCHI – MULTIOBJECTIVE OPTIMIZATION



# HOW TO PERFORM TAGUCHI'S DESIGN OF EXPERIMENTS AND ANALYSIS USING MINITAB – Hands

on Training



## QUERIES



# THANK YOU

