

Application of Taguchi Method for Optimization of Wave Soldering Process Parameters

Dr. Indu Uprety

Abstract - This paper highlights the application of Taguchi's orthogonal array (OA) experimentation for arriving at the optimum process parameter combination to keep the number of solder defects in printed circuit board (PCB) found after wave soldering, at a minimum level. Wave soldering machine provides technically a better substitute over the hand soldering process. The process parameters viz. solder temperature, contact time, preheat temperature setting, board thickness and wet flux amount need to be maintained at the optimum levels to get best results. Optimization helps in finding out and isolating the "effect of uncontrollable factors" and exploring the possible measures to counter them. The Design of Experiments method enables us to plan an experiment that simultaneously alters a number of variables in an experimental set-up to see how they interact with each other and affect responses. An experiment consists of testing combinations of different values (termed levels) of factors is likely to influence the characteristic (called response) of interest. If the factors are not independent, their interaction may also be considered. For experimentation, the effects on the five factors -- Solder Temperature (A), Contact Time (B), Preheat temperature (C), Board Thickness (D) and Wet Flux Amount (E) at three levels each, are simultaneously investigated by using the Taguchi Orthogonal Array experiments and analysis of variance (ANOVA). Some of the interactions are also to be considered. Hence, $L_{27} (3^5)$ orthogonal array design is chosen. The design has 5 factors at 3 levels each. Finally, signal-to-noise ratios, analysis of variance and response surface regression is carried out for analyzing the results and for obtaining insight into the process as a whole.

Key Words: Optimization, Design of Experiments (DOE), Orthogonal Arrays (OA), Signal-to-Noise Ratio, Analysis of Variance (ANOVA), Response Surface Regression.

1. Introduction

To improve the quality of products design of experiment (DOE) methods have been widely used by industries. A DOE experiment can simultaneously alter a number of variables in an experimental system to see how they affect the responses [1]. Although the design of experiments concept was introduced by Fisher in the early 1920s, the most research on this topic was carried out in the academic environment [2]. The usefulness of DOE concept in agricultural experiments was demonstrated by Fisher [3] and he analyzed the optimum water, rain, sunshine, fertilizer, and soil conditions needed to produce the best crop. Taguchi [4] went further with the design of experiment concept by introducing his approach in 1986. The Taguchi approach is a special form of DOE with applications in manufacturing industries. The overall objective of the experiments is to make comparisons between the effects of different factors and then determine the best setting for each factor.

The wave soldered PCBs undergo various stages of testing or inspection and generally it is observed that no PCB is passed without some rework for soldering defects. Wu and Hamada [5] discussed a two-level factorial experiment to study the number of defects in a wave soldering process. Although there are many defects found in wave soldering process, however, Insufficient hole fill is seen as a major problem that occurs on PCBs with pre-drilled holes for components to be set or mounted onto the board. Normally, insufficient hole fill occurs when an inadequate amount of solder has covered the holes drilled for the components, meaning solder won't stick to the circuit board once it cools down. This problem may appear as a result of a temperature variations in the process setting and board problem. Wave soldering process has following variables that control the level of defects:

A = Solder Temperature ($^{\circ}\text{C}$); B = Contact time (s); C = Preheat Temperature ($^{\circ}\text{C}$);

D = Board Thickness (inches); E = Wet Flux Amount (mg/dm^2)

The method proposed in this study uses an approach from robust design. Taguchi’s design of experiment method using orthogonal arrays is used to determine the “optimal settings” of the discrete design parameters. In this paper, the optimal process parameters for a wave soldering process have been obtained using Taguchi Methods.

2. Literature Review

DOE and Taguchi’s Orthogonal Arrays in Optimization:

A full factorial experiment combines the levels of each factor with each of the levels of all the remaining factors. It is possible to estimate interactions between factors using the full factorial design. However, the no. of experiments in this design become very large as the number of factors and levels increases.

The Taguchi method enables to plan experiments using a specially designed “Orthogonal Array” table to affect the design process, such that quality is built into a product during its design stage. For a multi-factor process, it is a very good technique for designing and executing experiments to investigate the processes without tediously and uneconomically running the process with all possible combinations of values. For most experiments carried out in the industry, the difference between the DOE and Taguchi approach is in the method of application [6]. Taguchi method reduces the number of experimental runs to maximum extent in terms of cost and time by making use of orthogonal arrays [7]. Due to systematically chosen combinations of variables it is possible to separate their individual effects. Taguchi method tests pairs of combinations instead of testing all possible combinations like the factorial design. The Factorial analysis can be used in order to find the best values for parameters to be used in the manufacturing process [8].

Orthogonal Array:

An orthogonal array is an experimental design constructed to allow a mathematically independent assessment of the effect of the different factors affecting the experiment. OA is the matrix of numbers arranged in columns and rows [9]. These arrays allow for the maximum number of main effects to be estimated in an unbiased (orthogonal) manner, with a minimum number of runs in the experiment. Orthogonal arrays are used to design experiment and describe trial conditions.

The most suitable orthogonal array for experimentation in this case is $L_{27}(3^5)$ array as shown in Table 1. Therefore, a total twenty seven experiments are to be carried out. As depicted in the Table 1, the L-27 experiment consists of 27 rows and 5 columns where each row corresponds to a particular trial and each column identifies settings of experimental factors. In the first trial, for example, the five experimental factors are set at their low level (level = 1). In the second trial, the first 4 factors are set at level 1 and the remaining one factor is set to level 1, and so on.

Run	A	B	C	D	E
1	1	1	1	1	1
2	1	1	1	1	2
3	1	1	1	1	3
4	1	2	2	2	1
5	1	2	2	2	2
6	1	2	2	2	3
7	1	3	3	3	1
8	1	3	3	3	2
9	1	3	3	3	3
10	2	1	2	3	1
11	2	1	2	3	2
12	2	1	2	3	3
13	2	2	3	1	1
14	2	2	3	1	2
15	2	2	3	1	3
16	2	3	1	2	1
17	2	3	1	2	2
18	2	3	1	2	3
19	3	1	3	2	1

20	3	1	3	2	2
21	3	1	3	2	3
22	3	2	1	3	1
23	3	2	1	3	2
24	3	2	1	3	3
25	3	3	2	1	1
26	3	3	2	1	2
27	3	3	2	1	3

Table 1. Standard L27 (3⁵) orthogonal array

The Taguchi method is used whenever the settings of interest parameters are necessary for manufacturing processes. Therefore, the Taguchi approach is used in many domains such as: environmental sciences [10, 11], agricultural sciences [12], physics [13], chemistry [14], statistics [15], management and business [16], medicine [17]. The challenge is to choose the proper orthogonal array suitable for the given problem of interest. The best use of Taguchi method comes with approximately an intermediate number of variables (3 to 50), with few interactions between variables, and when only a few variables contribute significantly and hence Taguchi is the preferable method among statistical experimental design methods [18].

3.1 Objectives of the Study

The objectives of the study are as under:

- Identifying the significant wave soldering parameters that most affect the process.
- Effect of these parameters on the quality of wave soldering with particular emphasis on insufficient solder fill and filling time.
- Optimum level of wave soldering process control parameters in order for setting-up of process & recommending the same for future production.

3.2 Preparation of Taguchi Experiment

The following needs to be established before starting Taguchi experiments - clear statements of problems, objectives, desired output characteristics and the method of measurement used to design the appropriate experiment. Afterwards, all the process parameters are identified and relevant factors affecting the outcome are defined. The two such important factors are:

- **Controllable factors**

These are set factors, but are not or can not be measured in-process so their variance is assumed to be zero or, if they are measured in-process, their variance is so small that it has a little effect on process performance. The mean of these controllable variables in-process should ideally be the same as their input set points and analysis of empirical data can be used to establish the nature of the functional relationship. The controllable factors in the study are as under:

A = Solder Temperature (C); B= Contact time (s); C = Preheat Temperature (C); and D = Board Thickness (inches); E= Wet Flux Amount (mg/dm²)

- **Noise Factors**

Noise factors are various process variables that affect variation, but are not identifiable nor controllable. Changes in environment temperature, humidity, dust etc., during the production/experiment, are some of the examples of noise. No 'noise' element was factorized into the experiment for practical reasons.

3.3 Choice of Experimental Factors and Levels

The factors most affecting the outcome is due to Wave Soldering Machine Parameters. Accordingly, out of a number of possible parameters only four machine parameters were selected for experimentation: Solder Temperature(A), Contact Time(B), Preheat Process Temperature(C), Board Thickness (D) and Wet Flux Amount (E). Some of the interactions AXB, AXC and BXC are also studied under response surface regression. The material and environment related factors can be maintained at the most desirable conditions or can be treated as fixed.

3.4 Experimental Layout and Design

According to Resit et al. [19], quality and cost are the two main ingredients in Taguchi’s approach to design optimization. In order to create and analyze the experimental runs, signal to noise ratio and orthogonal arrays are two major tools used in this robust design. This study makes use of both the tools. The experimental design is shown in Table-1. Experiments were conducted using Taguchi’s method with five factors at three levels each. The factors and their levels are shown in Table 2.

Factors	Level 1	Level 2	Level 3
A. Solder temperature (°C)	250	265	275
B. Contact time(s)	2	4	7
C. Preheat Temperature (°C)	90	110	130
D. Board Thickness(inches)	0.063	0.122	0.236
E. Wet Flux Amount (mg/dm ²)	350	470	580

Table 2. Factors and levels

It is planned to study the effect of five main factors and their interactions. The $L_{27}(3^5)$ orthogonal arrays which provides the required number of degrees of freedom is selected. This array consists of 27 rows each representing an experiment with 5 factors at three levels.

Main effect is the change in the average response of a factor. Randomization and replication principles are to be followed in experimental trials to eliminate the effect of systematic changes.

It is possible to investigate the main and interaction effects of the factors and their levels using Taguchi experiments and response surface regression. These experiments are useful because they require much fewer runs, though they do not allow the separation of main effects from high-order interactions.

3.5. Analysis of Variance for Orthogonal Array Analysis

For the test, 27 experimental runs were conducted to obtain the data needed for orthogonal array analysis to achieve the following objectives:

- To estimate the significance of individual quality influencing factors.
- To obtain the optimum settings for process.
- To determine the variation each control parameter has contributed.

A commonly applied statistical technique- The Analysis of Variance (ANOVA) is used to analyze the results of the OA experiment in product/process design, and to determine how much variation each process parameter has contributed. The main effects for each of the factors projects the general trend in the influencing factors. In a particular quality influencing factor, a lower or a higher value produces the preferred result. Thus, the optimum levels of influencing factors to produce the best outcome can be predicted.

ANOVA has been used to determine the influence of the main factors and their various levels.

4. Results and Discussion

The analysis of variance (ANOVA) and S/N ratio were evaluated to determine the effect of each selected factor on the optimization criteria. For these analyses, the Minitab14 software was utilized. The results of the ANOVA for SN ratios for determination of significant factors are shown in Section 4.1.3. In general, the F values would indicate the importance of these variables in the wave soldering process.

4.1 Linear Model Analysis: SN ratios versus Solder Temperature, Contact Time, Preheat Temperature, Board Thickness and Wet Flux Amount

4.1.1 Estimated Model Coefficients for SN ratios

Term	SE		T	P
	Coef	Coef		
Constant	- 0.01965	- 0.000		
	35.4757	1805.129		

Solder T 250	0.5773	0.02779	20.770	0.000
Solder T 265	0.0901	0.02779	3.241	0.005
Contact 2	0.2001	0.02779	7.198	0.000
Contact 4	-0.0003	0.02779	-0.010	0.992
Preheat 90	-0.0873	0.02779	-3.141	0.006
Preheat 110	0.0435	0.02779	1.564	0.137
Board Th 0.063	-0.0646	0.02779	-2.325	0.034
Board Th 0.122	0.0174	0.02779	0.625	0.541
Wet Flux 350	0.1057	0.02779	3.804	0.002
Wet Flux 470	-0.0633	0.02779	-2.279	0.037

4.1.2 Model Summary

S	R-Sq	R-Sq	Sq(adj)
0.1021	97.99%	96.73%	

4.1.3 Analysis of Variance for SN ratios

Source	DF	Seq SS	Adj SS	Adj MS	F	P
Solder Temperature(C)	2	7.08043	7.08043	3.54021	339.48	0.000
Contact Time(s)	2	0.71940	0.71940	0.35970	34.49	0.000
Preheat Temperature(C)	2	0.10292	0.10292	0.05146	4.93	0.021
Board Thickness(inches)	2	0.06039	0.06039	0.03019	2.90	0.084
Wet Flux Amount(mg/dm ²)	2	0.15288	0.15288	0.07644	7.33	0.005
Residual Error	16	0.16685	0.16685	0.01043		
Total	26	8.28287				

4.2 Linear Model Analysis: Means versus Solder Temperature(C), Contact Time(s), Preheat Temperature(C), Board Thickness(inches), Wet Flux Amount(mg/dm²):

4.2.1 Estimated Model Coefficients for Means

Term	Coef	SE Coef	T	P
------	------	---------	---	---

Constant	43.5417	0.095	454.616	0.000
		78		
Solder T	-2.5272	0.135	-18.658	0.000
250		45		
Solder T	-0.5383	0.135	-3.974	0.001
265		45		
Contact 2	-0.9533	0.135	-7.038	0.000
		45		
Contact 4	0.0000	0.135	0.000	1.000
		45		
Preheat 90	0.3911	0.135	2.888	0.011
		45		
Preheat	-0.1517	0.135	-1.120	0.279
110		45		
Board Th	0.3294	0.135	2.432	0.027
0.063		45		
Board Th	-0.1294	0.135	-0.956	0.353
0.122		45		
Wet Flux	-0.4856	0.135	-3.585	0.002
350		45		
Wet Flux	0.3011	0.135	2.223	0.041
470		45		

4.2.2 Model Summary

	R-S	R-Sq	Sq(adj)
	0.497	97.69	96.25%
	7	%	

4.2.3 Analysis of Variance for Means

Source	DF	Seq SS	Adj SS	Adj MS	F	P
Solder Temperature(C)	2	144.66	144.6	72.334	292.0	0.00
Contact Time(s)	9	16.359	16.35	8.1796	33.03	0.00
Preheat Temperature(C)	2	2.100	2.100	1.0499	4.24	0.03
Board Thickness(inches)	2	1.488	1.488	0.7438	3.00	0.07
Wet Flux Amount(mg/dm ²)	2	3.244	3.244	1.6220	6.55	0.00
Residual Error	16	3.963	3.963	0.2477		
Total	26	171.82				

Discussion:

In the variance analysis for S/N ratios (section 4.1.3), SS stands for sum of squares (factors) between groups and the sum of squares in the group (error). MS stands for sum of squares divided by the number of degrees of freedom of the mean square, F stands for test statistics, p stands for significance level. If p value is smaller, the influence is more significant. When the general P value is less than 0.05, it just means the factor has a significant effect. Thus, the factors which are less than 0.05 in the analysis of variance (for S/N ratios) are significant. Similar conclusion can be drawn with regard to section 4.2.3, where the factors solder temperature, contact time, preheat temperature and wet flux amount are significant.

4.3 Signal to Noise Ratios:

Response Table for Signal to Noise Ratios

Nominal is best ($-10 \times \text{Log}_{10}(s^2)$)

Level	Contact		Preheat	Board	Wet Flux
	Solder Temperature (C)	Time (s)			
1	-34.90	-35.28	-35.56	-35.54	-35.37
2	-35.39	-35.48	-35.43	-35.46	-35.54
3	-36.14	-35.68	-35.43	-35.43	-35.52
Delta	1.24	0.40	0.13	0.11	0.17
Rank	1	2	4	5	3

Response Table for Means

Level	Contact		Preheat	Board	Wet Flux
	Solder Temperature (C)	Time (s)			
1	41.01	42.59	43.93	43.87	43.06
2	43.00	43.54	43.39	43.41	43.84
3	46.61	44.49	43.30	43.34	43.73
Delta	5.59	1.91	0.63	0.53	0.79
Rank	1	2	4	5	3

Discussion:

The ranks in a response table identify which factors have the largest effect. The factor with the largest delta value is given rank 1, the factor with the second largest delta is given rank 2, and so on. Thus, in the response table for Signal to Noise Ratios and for Means, solder temperature is ranked 1, contact time ranked 2, followed by factors 3, 4 and 5.

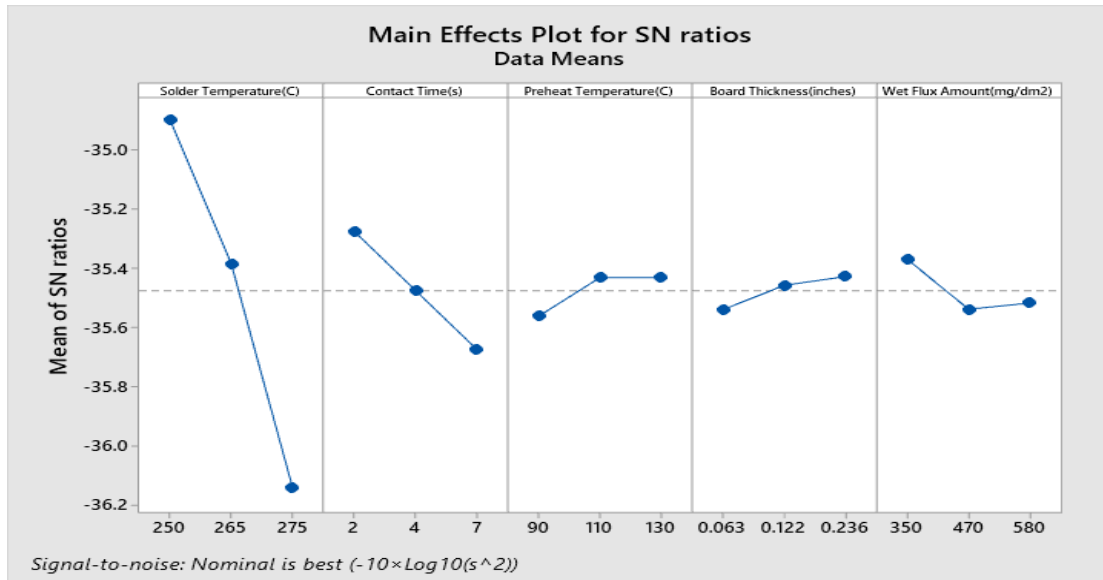


Figure-1 Main Effects Plots for SN Ratios

The main effects of independent variables (A, B, C, D and E) on the responses R_1 and R_2 , of using Taguchi method, are shown in Figure 1. Figure 1 illustrates the average of S/N ratios for each variable at the three different levels and the corresponding response variables.

In case of response variables R_1 and R_2 , as illustrated in Figure 1, the variables are highly dependent on solder temperature, contact time, preheat setting and wet flux amount, i.e. changing one of these settings will have the most dramatic effect on output.

5.1 Response Surface Regression with R_1 :

Coded Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VI
Constant	85.06	0.340	250.01	0.000	8
Solder Temperature(C)	6.726	0.508	13.23	0.000	1.8
Contact Time(s)	2.020	0.731	2.76	0.015	3.8
Preheat Temperature(C)	-1.673	0.680	-2.46	0.027	3.3
Board Thickness(inches)	-2.15	1.06	-2.03	0.062	8.2
Wet Flux Amount(mg/dm ²)	0.763	0.379	2.01	0.064	1.0
Solder Temperature(C)*Contact Time(s)	-2.34	1.18	-1.98	0.068	6.9
Solder Temperature(C)*Preheat Temperature(C)	-0.09	1.26	-0.07	0.946	7.7
Solder Temperature(C)*Wet Flux	-0.455	1.06	-1.06	0.309	1.0

Amount(mg/dm ²)	0.480				1
Contact Time(s)*Preheat Temperature(C)	2.452	0.769	3.19	0.007	2.88
Contact Time(s)*Wet Flux Amount(mg/dm ²)	0.340	0.455	0.75	0.466	1.01
Preheat Temperature(C)*Wet Flux Amount(mg/dm ²)	0.266	0.458	0.58	0.571	1.00
Board Thickness(inches)*Wet Flux Amount(mg/dm ²)	0.113	0.450	0.25	0.805	1.02

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
1.585	95.23	91.14%	81.66%
52	%		

5.1.1 Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Model	12	702.582	58.549	23.29	0.000
Linear	5	455.872	91.174	36.27	0.000
Solder Temperature(C)	1	440.037	440.03	175.04	0.000
Contact Time(s)	1	19.185	19.185	7.63	0.015
Preheat Temperature(C)	1	15.217	15.217	6.05	0.027
Board Thickness(inches)	1	10.384	10.384	4.13	0.062
Wet Flux Amount(mg/dm ²)	1	10.178	10.178	4.05	0.064
2-Way Interaction	7	32.875	4.696	1.87	0.151
Solder Temperature(C)*Contact Time(s)	1	9.845	9.845	3.92	0.068
Solder Temperature(C)*Preheat Temperature(C)	1	0.012	0.012	0.00	0.946
Solder Temperature(C)*Wet Flux Amount(mg/dm ²)	1	2.799	2.799	1.11	0.309
Contact Time(s)*Preheat Temperature(C)	1	25.534	25.534	10.16	0.007
Contact Time(s)*Wet Flux	1	1.411	1.411	0.56	0.466

Amount(mg/dm ²)					
Preheat Temperature(C)*Wet Flux Amount(mg/dm ²)	1	0.847	0.847	0.34	0.571
Board Thickness(inches)*Wet Flux Amount(mg/dm ²)	1	0.159	0.159	0.06	0.805
Error	14	35.19	2.514		
			4		
Total	26	737.7			
			76		

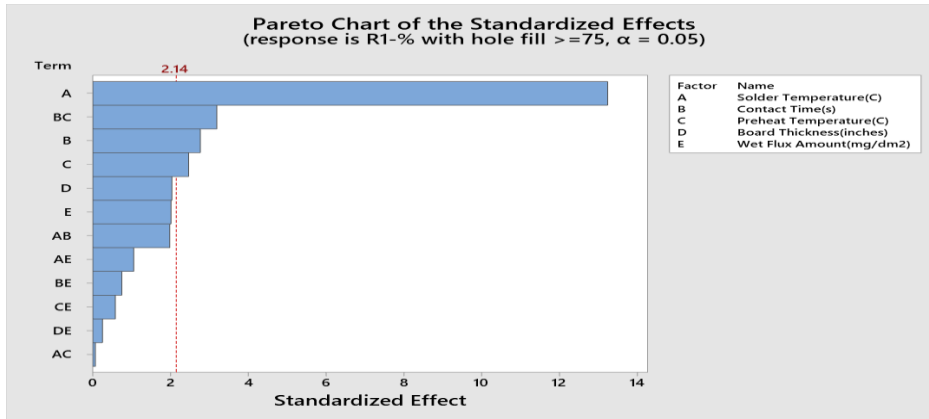


Figure-2 Pareto Chart of the Standardized Effects

Discussion:

For Response Surface Regression results, the factors solder temperature, contact time, preheat setting and interaction between contact time and preheat temperature are significant at 5% level of significance. Also, the model explains 91.14% of the variation in the output.

In the pareto chart of the standardized effects shown above, the interaction between contact time and preheat temperature is significant. In addition, the largest effect is due to solder temperature because it extends the farthest. Solder Temperature*Preheat Temperature (AC) is the smallest because it extends the least.

5.2 Response Surface Regression with R₂:

Coded Coefficients

Term	Coef	SE Coef	T-Value	P-Value	VI F
Constant	1.4738	0.0030	480.01	0.000	
	9	7			
Solder Temperature(C)	-0.0045	0.2285	-49.81	0.000	1.8
	1	9			8
Contact Time(s)	-0.0066	0.0218	-3.31	0.005	3.8
	0	3			8
Preheat Temperature(C)	0.0452	0.0061	7.38	0.000	3.3
	7	4			1
Board Thickness(inches)	0.0568	0.0095	5.96	0.000	8.2
	5	4			8

Wet Flux Amount(mg/dm ²)	- 0.0034	-3.62	0.003	1.0
	0.0123	2		3
		8		
Solder Temperature(C)*Contact Time(s)	0.0424	0.0106	3.98	0.001 6.9
				1
Solder Temperature(C)*Preheat Temperature(C)	0.0250	0.0114	2.19	0.046 7.7
				7
Solder Temperature(C)*Wet Flux Amount(mg/ dm ²)	- 0.0041	-1.10	0.289	1.0
	0.0045	0		1
		2		
Contact Time(s)*Preheat Temperature(C)	0.0081	0.0069	1.17	0.262 2.8
		2	4	8
Contact Time(s)*Wet Flux Amount(mg/ dm ²)	0.0043	0.0041	1.06	0.309 1.0
		3	0	1
Preheat Temperature(C)*Wet Flux Amount(mg/ dm ²)	0.0042	0.0041	1.02	0.326 1.0
		0	3	0
Board Thickness(inches)*Wet Flux Amount(mg/ dm ²)	0.0036	0.0040	0.91	0.379 1.0
		9	6	2

Model Summary

S	R-sq	R-sq(adj)	R-sq(pred)
0.01430	99.73	99.50%	98.80%
81	%		

5.2.1 Analysis of Variance

Source	DF	Adj SS	Adj MS	F-Value	P-Value
Model	12	1.060	0.0883	431.79	0.000
	76		97		
Linear	5	0.523	0.1046	511.21	0.000
	27		55		
Solder Temperature(C)	1	0.507	0.5079	2481.0	0.000
		91	14	0	
Contact Time(s)	1	0.002	0.0022	10.94	0.005
		24	40		
Preheat Temperature(C)	1	0.011	0.0111	54.42	0.000
		14	41		
Board Thickness(inches)	1	0.007	0.0072	35.48	0.000
		26	64		
Wet Flux Amount(mg/ dm ²)	1	0.002	0.0026	13.09	0.003
		68	80		
2-Way Interaction	7	0.007	0.0010	5.25	0.004
		52	74		

Solder Temperature(C)*Contact Time(s)	1	0.003	0.0032	15.82	0.001
		24	39		
Solder Temperature(C)*Preheat Temperature(C)	1	0.000	0.0009	4.80	0.046
		98	83		
Solder Temperature(C)*Wet Flux Amount(mg/dm ²)	1	0.000	0.0002	1.22	0.289
		25	49		
Contact Time(s)*Preheat Temperature(C)	1	0.000	0.0002	1.37	0.262
		28	80		
Contact Time(s)*Wet Flux Amount(mg/dm ²)	1	0.000	0.0002	1.12	0.309
		23	28		
Preheat Temperature(C)*Wet Flux Amount(mg/dm ²)	1	0.000	0.0002	1.03	0.326
		21	12		
Board Thickness(inches)*Wet Flux Amount(mg/dm ²)	1	0.000	0.0001	0.83	0.379
		17	69		
Error	14	0.002	0.0002		
		87	05		
Total	26	1.063			
		63			

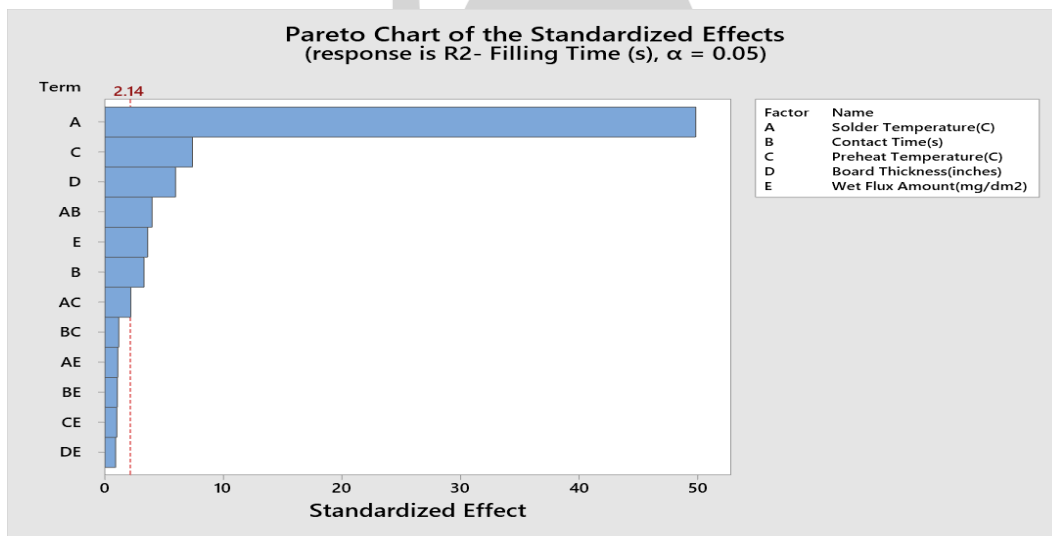


Figure-3 Pareto Chart of the Standardized Effects

Discussion:

For Response Surface Regression results, the factors solder temperature, contact time, preheat setting and interaction between contact time, preheat temperature, board thickness and wet flux amount are significant at 5% level of significance. Interactions solder temperature*contact time and solder temperature*preheat temperature is also significant. The adjusted R-square values in the response surface regression are observed to be 91.14% and 99.5%, respectively indicating that the model parameters can explain variation in the response variables R₁ (% with hole fill ≥ 75) and R₂ (filling time (s)) very well.

In the pareto chart of the standardized effects shown above, the interaction between contact time and solder temperature is significant. In addition, the largest effect is due to solder temperature because it extends the farthest. Board Thickness*Wet Flux Amount (DE) is the smallest because it extends the least.

5.3 Prediction with R₁ -% with hole fill ≥ 75 %:

Setting-1

Variable	Setting
Solder Temperature(C)	250
Contact Time(s)	2
Preheat Temperature(C)	90
Board Thickness(inches)	0.063
Wet Flux Amount(mg/dm ²)	350

Prediction

Fit	SE Fit	95% CI	95% PI
79.65	1.333	(76.7913, 82.5120)	(75.2081, 84.0952)
16	61		

Setting-2

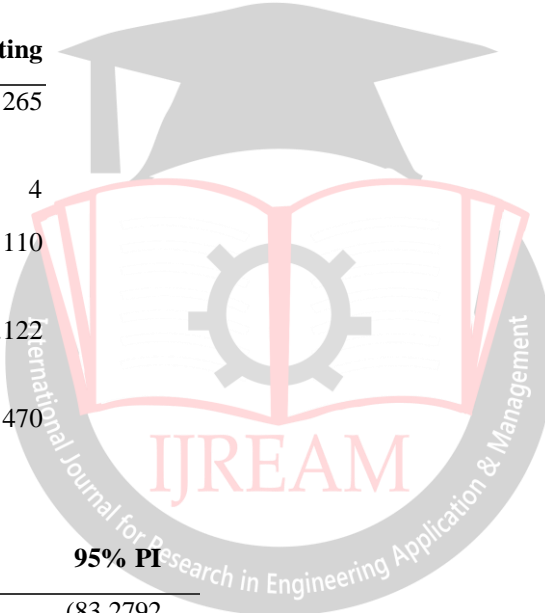
Variable	Setting
Solder Temperature(C)	265
Contact Time(s)	4
Preheat Temperature(C)	110
Board Thickness(inches)	0.122
Wet Flux Amount(mg/dm ²)	470

Prediction

Fit	SE Fit	95% CI	95% PI
86.81	0.4428	(85.8602, 87.7597)	(83.2792, 90.3407)
00	23		

Setting-3

Variable	Setting
Solder Temperature(C)	275
Contact Time(s)	7
Preheat Temperature(C)	130
Board Thickness(inches)	0.236
Wet Flux Amount(mg/dm ²)	580



Prediction

Fit	SE Fit	95% CI	95% PI
91.02	3.901	(82.6569,	(81.9923,
44	33	99.3919)	100.057)

5.4 Prediction with R₂- Filling Time (s):

Setting-1

Variable	Setting
Solder Temperature(C)	250
Contact Time(s)	2
Preheat Temperature(C)	90
Board Thickness(inches)	0.063
Wet Flux Amount(mg/dm ²)	350

Prediction

Fit	SE Fit	95% CI	95% PI
1.717	0.01203	(1.69181,	(1.67752,
62	48	1.74344)	1.75772)

Setting-2

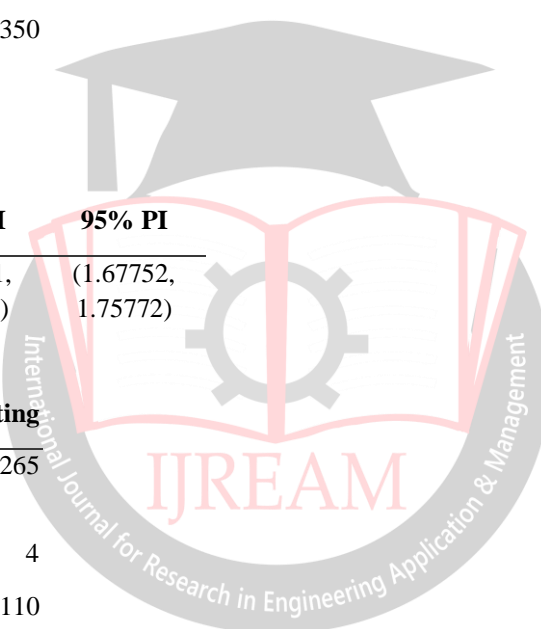
Variable	Setting
Solder Temperature(C)	265
Contact Time(s)	4
Preheat Temperature(C)	110
Board Thickness(inches)	0.122
Wet Flux Amount(mg/dm ²)	470

Prediction

Fit	SE Fit	95% CI	95% PI
1.412	0.00399	(1.40355,	(1.38026,
12	61	1.42069)	1.44398)

Setting-3

Variable	Setting
Solder Temperature(C)	275



Contact Time(s)	7
Preheat Temperature(C)	130
Board Thickness(inches)	0.236
Wet Flux Amount(mg/dm ²)	580

Prediction

Fit	SE Fit	95% CI	95% PI
1.396	0.03520	(1.32093, 1.47195)	(1.31493, 1.47795)
44	65		

Discussion:

After conducting the experiment, the overall optimum settings for the outputs based on experimental data were determined to be A2, B2, C2, and E2 because the combination of these settings gives the smallest standard error fit. The optimum settings for the process are as under:

- A. Solder temperature= 265⁰C ; B. Contact time = 4s; C. Preheat temperature = 110⁰C;
- D. Wet flux volume = 470 mg/dm².

It is important to mention here that the Solder temperature of 265⁰C causes least damage (in terms of thermal shock) to components and board material.

6. Findings and Suggestions

In this study, an orthogonal Taguchi L27 (3⁵) array was employed for optimization of wave soldering process parameters. For this purpose, five variables including solder temperature, contact time, preheat temperature, board thickness and wet flux amount were examined at three levels. According to the ANOVA results, solder temperature and contact time were the most significant variables among others and preheat temperature was the least significant variable.

The coefficient of determination, the adjusted R-square values in the response surface regression were observed to be 91.14% and 99.5% respectively indicating that the model parameters can explain variation in the response variables R₁(% with hole fill >=75) and R₂ (filling time (s)) very well. Therefore, the model has good practical significance.

The optimum conditions for solder temperature, contact time, preheat temperature and wet flux amount were equal to 265⁰C, 4s, 110⁰C and 470 mg/dm², respectively. Thus, Taguchi’s design of experiments would enable the process engineer to conduct the study with only a small number of test-runs to achieve minimum variation in the newly-developed process.

7. Conclusion

As a result of this study, the factors solder temperature, contact time, preheat temperature and wet flux amount are found significant. The aim of the completed optimization was to decrease the number of defects. The interactions solder temperature*contact time and solder temperature*preheat temperature and contact time*preheat temperature have also been found significant. It was observed that for reducing insufficient solder fill defects and filling time, a higher solder temperature was better. Moderate contact times and preheat temperature with a setting of 110⁰C yielded better results for this experiment.

The present study would enable the process engineers to reach a practical understanding of what will be required in their own specific application with only a small number of experimental runs. Many of the above points can be

applied in situations where hole fill is not adequate and can be used to determine the best set of parameters to achieve maximum wetting and hole fill.

References

- [1]. Ranjit, K.R. *Design of Experiments Using the Taguchi Approach: 16 Steps to Product and Process Improvement*; John Wiley & Sons: Hoboken, NJ, 2001.
- [2]. Fisher, R.A. *Statistical Methods for researcher Workers*; Oliver and Boyd: London, UK, 1925.
- [3]. Fisher, R.A. The arrangement of field experiments. *Jour. Min. Agr. Engl.* 1926, 33, 503-513.
- [4]. Taguchi, G. *Introduction to Quality Engineering: Designing Quality into Products and Processes*; Asian Productivity Organization/UNIPUB, White Plain, Unipub/Kraus, NY, 1986.
- [5]. Wu, C. F. J., and Hamada, M.; *Experiments: Planning, Analysis, and Parameter Design Optimization*, New York: Wiley, 2000.
- [6]. R.K. Roy, *Design of experiments using the Taguchi approach*, John Willey & Sons. Inc., New York, 2001.
- [7]. Taguchi, G.; Jugulum, R.; Taguchi, S.; *Computer-based Robust Engineering: Essentials for DFSS*; ASQ Quality Press: Milwaukee, WI, 2004.
- [8]. Montgomery, D.C.; Runger, G.C. *Applied Statistics and Probability for Engineers*; John Wiley & Sons: Hoboken, NJ, 2006.
- [9]. P. Sharma, A. Verma, R.K. Sidhu, O.P. Pandey; Process parameter selection for strontium ferrite sintered magnets using Taguchi L9 orthogonal design, *Journal of Materials Processing Technology* 168 (2005) 147-151.
- [10]. Daneshvar, N.; Khataee, A.R.; Rasoulifard, M.H.; Pourhassan, M. Biodegradation of dye solution containing Malachite Green: Optimization of effective parameters using Taguchi method. *Journal of Hazardous Materials* 2007, 143(1-2), 214-219.
- [11]. du Plessis, B.J.; de Villiers, G.H. The application of the Taguchi method in the evaluation of mechanical flotation in waste activated sludge thickening. *Resources, Conservation and Recycling* 2007, 50(2), 202-210.
- [12]. Tasirin, S.M.; Kamarudin, S.K.; Ghani, J.A.; Lee, K.F. Optimization of drying parameters of bird's eye chilli in a fluidized bed dryer. *Journal of Food Engineering* 2007, 80(2), 695-700.
- [13]. Wu, C.-H.; Chen, W.S.; Injection molding and injection compression molding of three-beam grating of DVD pickup lens. *Sensors and Actuators A: Physical* 2006, 125(2), 367-375.
- [14]. Houn, J.-Y.; Liao, J.-H.; Wu, J.-Y.; Shen, S.-C.; Hsu, H.-F. Enhancement of asymmetric bio reduction of ethyl 4-chloro acetoacetate by the design of composition of culture medium and reaction conditions. *Process Biochemistry* 2007, 42(1), 1-7.
- [15]. Romero-Villafranca, R.; Zúnica, L.; Romero-Zúnica, R. Ds-optimal experimental plans for robust parameter design. *Journal of Statistical Planning and Inference* 2007, 137(4), 1488-1495.
- [16]. Elshennawy, A.K. Quality in the new age and the body of knowledge for quality engineers. *Total Quality Management and Business Excellence* 2004, 15(5-6), 603-614.
- [17]. Ng, E.Y.K.; Ng, W.K. Parametric study of the biopotential equation for breast tumour identification using ANOVA and Taguchi method. *Medical and Biological Engineering and Computing* 2006, 44(1-2), 131-139.
- [18]. Keles, O. ; "An optimization study on the cementation of silver with copper in nitrate solutions by Taguchi design", *Hydrometallurgy*, 2009, 95(3-4), 333-336.
- [19]. Unal, R. and Dean, E.B. ; Taguchi approach to design optimization for quality and cost: an overview. *Proceedings of the International Society of Parametric Analysts*, 1991, pp. 21-24.