

ID: 2016-ISFT-321

Optimization of Welding Parameters for Weld Dilution in GMAW using Genetic Algorithm

Pratibha Kumari¹, R.P. Singh²

¹Mechanical Engineering Department, Krishna Institute of Engineering & Technology, Ghaziabad ²GLA University, Mathura karwalpratibha@gmail.com

Abstract: Hardfacing is a technique in which deposition of material on the surface of similar/dissimilar material is done to improve its wear resistant properties. The benefits also include the minimization of downtime needed to replace worn components and reduction of spare part inventory and finally saves money. So, it is essential to select welding process parameters carefully, to achieve a quality bead which is defect free. In order to achieve the above objective, a set of mathematical models has been developed for the prediction of weld bead dilution using 5factor, 2 – levels Factorial design for 140MXC Nano structured wire with IS2062 substrate. The developed model was checked for its adequacy. The main and the interaction effects of the process parameters on weld bead dilution are presented in graphical form. Moreover, Genetic Algorithm computational model was used for the optimization of welding parameters to achieve desired optimum weld dilution.

Keywords: Harfacing, Gas Metal Arc Welding; Factorial Design Approach, Genetic Algorithm.

1. INTRODUCTION

The welding process, due to its complexity, has relied on empirical and experimental data to determine its welding conditions. However, trial-and-error methods to determine optimal conditions incur considerable time and cost, as stated by Kim et al. [1]. In order to overcome these problems optimization of welding input parameters, for making high quality weldment can be envisaged as a vital area of research. Weld bead shape plays an important role in determining the quality and mechanical strength of the weldment. Weld bead geometry is dependent on number of input parameters such as wire feed rate, welding speed, voltage, nozzle to plate distance, gas flow rate etc. as stated by Kim et al. [2].To predict the effect of various welding parameters on weld bead geometry various researchers used variety of techniques. (Benyounis et al. [3]; Kannan and Yoganandh, [4]; Singla et al. [5];Sudhakaran et al.[6]).Statistical approach is one way to increase the amount of information-rich data gathered. Numerous studies validated that efficient use of statistical design of

experiment techniques, works as an excellent tool for the development of a mathematical model for the prediction of weld bead geometry, (Murugan and Parmer[7], Subramaniam[8], Allen et al.[9], Kim et al.[10].

So, to achieve the above mentioned objective statistical designing technique was employed to depict weld bead geometry of surfaced steel. Experiments were conducted based on fractional factorial designing i.e., 2 level 5 factors. Process control variables considered were wire feed rate, welding speed, welding voltage, nozzle to plate distance and torch angle.

To ascertain the acceptability of weldment, control of dilution is very important in hardfacing, where low dilution is typically desirable. For the present research, it was planned to use GA computational technique for the optimization of welding parameters to minimize dilution, as it owns high reliability, robustness and accuracy, Palaniswamy et al. [12].

2. MATERIALS AND METHODS

2.1 MATERIALS

The experiments were carried out by the deposition of 140MXC Nano-structured wire of 1.6 mm diameter using Miller Migmatic 273 welding machine onIS2062 substrate. The chemical composition of substrate and filler wire are given in table 1. Shielding gas used was pure Argon with flow rate kept constant at 25 l/min. Bead- on – plate technique was used to make the weld runs.

TABLE 1: Chemic	al composition	of base plate	and filler wire
-----------------	----------------	---------------	-----------------

IS 2062													
Eleı	nent	C Si Mn S P		Mn S P Fe		Fe							
%		0.22	27	0.1	61	0.5	0) 0.05		0.023		Bal	ance
TAI	TAFA 140MXC Nano-structured wire												
	Elen	nent	Cı	•	С		Μ	lo	В	W	Si	Nb	Fe
	%		20).8	2.8	4	12	2.1	0.64	9.79	0.54	0.8	Balance

2.2 EXPERIMENTAL DESIGN

Factorial design can be written in the form of a design matrix where the rows correspond to different trials and the columns correspond to the levels of the process parameters. The design matrix developed to conduct the sixteen trials of 2^n fractional factorial design is given in table-2. To determine the two levels of each control variables, pilot study was carried out by varying the control variables in all possible combination. The experiments so carried, gave the limits, which is shown in table- 3.

Welding parameters were coded as (+1) and (-1), corresponding to the high and low levels for the ease of recording and processing of the data using equation (1).

$$X_{j} = (X_{jn} - X_{jo}) / J_{j}$$

$$\tag{1}$$

Where, X_{ji} is the coded value of the parameter, X_{jn} is the natural value of the parameter, X_{jo} is the natural value of the basic level, J_{ji} is the variation interval and j is the number of parameter.

Experiment	W	S	V	Ν	Т
No.	1	2	3	4	5= -(1234)
1	-1	-1	-1	-1	-1
2	1	-1	-1	-1	1
3	-1	1	-1	-1	1
4	1	1	-1	-1	-1
5	-1	-1	1	-1	1
6	1	-1	1	-1	-1
7	-1	1	1	-1	-1
8	1	1	1	-1	1
9	-1	-1	-1	1	1
10	1	-1	-1	1	-1
11	-1	1	-1	1	-1
12	1	1	-1	1	1
13	-1	-1	1	1	-1
14	1	-1	1	1	1
15	-1	1	1	1	1
16	1	1	1	1	-1

TABLE 2: Design matrix

 TABLE 3: Levels of the welding parameters

Parameters	Units	Symbols	Lower Limit -1	Upper Limit +1
WireFeed Rate	m/min	W	7.62	9.04
Welding Speed	cm/min	S	27	36
Voltage	Volts	V	26	32
Nozzle to Plate Distance	mm	N	15	20
Torch Angle	Degree	Т	80	100

2.3 WELDING PROCEDURE AND DILUTION MEASUREMENT

From design of experiments, it was clear that 16 trials were needed to conduct the experiments, keeping the values set in the design matrix. The weld runs were then performed on each of these strips throughout the length of the strip using bead- on – plate technique.

To get the dimensions of the bead geometry all the samples were cut and scanned. The scanned images were then transferred to Corel Draw X5, using the ruler tool in the software. Calculated values of %dilution are tabulated in Table 4.

TABLE 4: %Dilution

Experime nt No.	1	2	3	4	5	6	7	8
Dilution,	40.0	33.3	34.	36.9	44.2	33.	38.2	50.8
%D	1	4	7	5	5	9	5	6
Experime nt No.	9	10	11	12	13	14	15	16
Dilution,	24.5	39.6	32.	42.2	52.5	41.	53.4	36.6
%D		1	6	9	6	6	6	9

3. SELECTION AND DEVELOPMENT OF MATHEMATICAL MODEL

The response function representing % dilution, can be represented as:

$$\% Dilution = f(W, S, V, N, T)$$
(2)

Where, W, S, V, N and T represent the wire feed rate, welding speed, welding voltage, nozzle to plate distance and torch angle respectively.

The above expression, being a second degree response surface can be expressed as:

 $y = b_0+b_1W+b_2S+b_3V+b_4N+b_5T+b_6WS+b_7WV+b_8WN+b_9WT+b_{10}SV+b_{11}SN+b_{12}ST+b_{13}VN+b_{14}VT+b_{15}(3)$

Where, b_{0} , b_{1} , b_{2} , b_{15} are the coefficients of the polynomial equation

The value of regression coefficients was calculated using equation 4,

$$bij = \frac{\sum_{i=1}^{M} X_{ji} Y_i}{M}, j = 0, 1, 2, 3, \dots, k$$
(4)

Where, X_{ji} is the value of a parameter or interaction in coded form, Y_i is the average value of the response parameters, is the number of observations and k is the

number of coefficients of the model. These models for predicting different responses mentioned above are given below:

%D= 39.72 - 0.32W + 1.0S + 4.22V + 0.69N + 0.9T + 1.29WS -2.87WV- 0.05WN +1.72WT - 0.13SV - 0.16SN + 3.7ST + 1.44VN+ 2.69VT - 0.85NT (5)

4. CHECKING THE ADEQUACY OF THE MODELS

The adequacy of the model was tested using Analysis of Variance (ANOVA) as explained byKumariet. al. [13] and Singla et al.[14]. The scatter diagram for observed vs. estimated values of % dilution for different rows of the design matrix is given in figure 1.

Resp-	DOF		Var. response	Std. dev.	Var. adeq.	'F' model	'F' table	F _m <f<sub>t</f<sub>
onse	S ² _Y	S ² _{ad}	S ² _Y	S _{bj}	S ² _{ad}	S_{ad}^2/S_Y^2	(10,16,0.05)	
%D	16	10	0.594	0.192	0.059	0.10	2.49	Yes

TABLE 5: Analysis of variance (ANOVA)



Fig. 1.Scatter plot for weld bead dilution model

The proposed model developed for %dilution after dropping insignificant coefficients is given in equation 6. Based on these models, graphs have been drawn for interpretation of the direct and interaction effects of the process parameters on %dilution.

%D= 39.72 -0.32W + 1.0 S + 4.22V -2.87 WV +1.72 WT -0.13 SV + 2.69 VT (6)

5. OPTIMIZATION

Minimization of dilution was the main purpose of the present study while keeping in mind other important bead parameters with their constraints in the equation form. GA search ranges for delivering the optimal responses are given in Table 6. The experiments were carried out with the conditions as mentioned in Table 7.

Direct, indirect and response parameters for the selected set of problem are mentioned in Table 8.

TABLE 6: GA search ranges

Parameters	Units	Range
WireFeed Rate	m/min	7.62 - 9.04

function [f,g]=fp(x)

Welding Speed	cm/min	27 - 36
Voltage	Volts	26 - 32
Nozzle to Plate Distance	mm	15 - 20
Torch Angle	Degree	80 - 100

TABLE 7.Experimental condition

Base Metal	IS 2062 steel of 10mm thickness
Deposited Alloy	TAFA 140MXC Nano-structured wire of 1.6 mm dia.
Welding conditions	Flat
Technique	Bead on plate
Process	Semiautomatic GMAW

 TABLE 8. Direct, indirect and response parameters for the selected set of problem

Direct parameters	Indirect par	rameters	Response parameters
WireFeed Rate	Shielding gas	Argon	Penetration, p
Welding Speed	Power source	Constant potential	Width, w
Voltage	Welding current	D.C	Reinforcement height, h
Nozzle to Plate Distance	VI characteristic	Flat	%Dilution, D
Torch Angle	Electrode polarity	Positive	

GA optimization tool available in MATLAB R2009a, was employed for fulfilling the purpose of optimization. M-file was created which is shown below: $\begin{aligned} f(1) &= 39.72 \cdot 0.3872 * x(1) + 1.0 * x(2) + 4.22 * x(3) \cdot 2.87 * x(1) * x(3) + 1.72 * x(1) * x(5) \cdot 0.13 * x(2) * x(3) + 2.69 * x(3) * x(5); \\ g(1) &= 8.51 + 0.62 * x(1) \cdot 1.01 * x(2) + 1.54 * x(3) + 0.33 * x(1) * x(2) + 0.59 * x(1) * x(5) \cdot 0.54 * x(2) * x(3) + 0.54 * x(3) * x(5) \cdot 0.24 * x(4) * x(5); \\ g(2) &= 5.5 \cdot 8.51 + 0.62 * x(1) \cdot 1.01 * x(2) + 1.54 * x(3) + 0.33 * x(1) * x(2) + 0.59 * x(1) * x(5) \cdot 0.54 * x(2) * x(3) + 0.54 * x(3) * x(5) \cdot 0.24 * x(4) * x(5); \\ g(3) &= 4.01 \cdot 0.32 * x(2) \cdot 0.39 * x(3); \\ g(4) &= 3.0 \cdot 4.01 \cdot 0.32 * x(2) \cdot 0.39 * x(3); \\ g(5) &= 3.91 + 0.18 * x(1) \cdot 0.34 * x(2) + 0.19 * x(3) \cdot 0.35 * x(1) * x(3) + 0.38 * x(2) * x(5) + 0.21 * x(3) * x(4) + 0.30 * x(3) * x(5); \\ g(6) &= 2.4 \cdot 3.91 + 0.18 * x(1) \cdot 0.34 * x(2) + 0.19 * x(3) \cdot 0.35 * x(1) * x(3) + 0.38 * x(2) * x(5) + 0.21 * x(3) * x(4) + 0.30 * x(3) * x(5); \\ g(7) &= f \cdot 53.46; \\ g(8) &= 24.5 \cdot f; \\ end \end{aligned}$

Parameters for GA computation and setting of GA graphical interface are given in table 9 and figure 2 respectively, Yoganandhet al.[15].

T/	ABI	LE 9	: Par	ameters	for	GA	computation

Population type	Double vector	Crossover function	Intermediate
Population size	100	Migration	Forward
Number of generations	52	Lower boundary limits	-1, -1, -1, -1, -1.
Reproduction elite count	2	Upper boundary limits	1, 1, 1, 1, 1.

Problem Setup and Results					Options									
					Population									
Problem					Population type: Double Vector									
Fitness function:	@fp				Population size: 🔘 Use default: 20	-								
Number of variables	umber of variables: 5				Specify: 100									
Constraints:					Creation function: Uniform									
Linear inequalities:	A:		b:											
Linear equalities:	Aeq:		beq:		Initial population: Use default: [] 									
Bounds:	Lower:	-1	Upper:	1	Specify:									
Nonlinear constraint	function:				Initial scores: O Use default: []									
Run solver and view results					Specify:									
Use random states from previous run					Initial range: 💿 Use default: [0;1]									
Start Pause	St	op			Specify: [-1,-1,-1,-1,-1;1,1,	1,1,1]								
Current iteration: 52	1115		Cle	ar Results	🕀 Fitness scaling									
Optimization running.				•	☑ Selection]								
Warning: You are using 'mutationuniform' mutation function for constrained minimization. Solution may be infeasible; use '@mutationadaptfeasible' function for constrained minimization. Optimization terminated.					Reproduction Mutation Crossover									
								Objective function value: 28.326781810440702 Optimization terminated: maximum number of generations					⊞ Migration ■	
													🗄 Algorithm settings	
rinai point:					Hybrid function									
1 - 2	3	0.002	4	5	🗉 Stopping criteria]								
-0.030	-0'303	-0.992	-Ų.	502 0],								

Fig. 2.GA algorithm tool



6. RESULT AND DISCUSSION

- 1. Increase in voltage, speed and torch angle increases the % dilution from 34.3 to 49.94,43.42 to 46.9and 32.8 to 50.44 respectively.
- 2. Increase in wire feed rate decreases the % dilution from 50.44 to 44.56.
- 3. % dilution remains constant at 46.03 with increase in nozzle to plate distance
- 4. Different coded values obtained after running the Mfile are:

- x(1) = Wire feed rate (W) = -0.836
- x(2) = Welding speed (S) =-0.969

x(3) = Voltage(V) = -0.992

- x(4) = Nozzle to plate distance (N) = -0.362
- x(5) = Torch angle (T) = 0.866

Different plots for the developed code are shown in figure 3. The values of the direct parameter for the coded results and optimized response parameters of the hardfacedweldment are given in table 10.



Fig. 3. GA output plots

TABLE 10. Values of optimized parameters

Direct parameter	Value	Response parameters	Value
WireFeed Rate	8.03322 m/min		
Welding Speed	29.31976 cm/min	%Dilution, D	28.3288
Voltage	27.512 V		
Nozzle to Plate Distance	17.0475 mm		
Torch Angle	94.33 °		

7. CONCLUSIONS

- 1. Response Surface Methodology (RSM) approach is an excellent and effective tool, for the development of mathematical model for the prediction of response.
- 2. Scatter diagram ensures the accuracy of the model.
- 3. Genetic Algorithm was found to be very effective and powerful technique for optimization of welding parameters.
- 4. Genetic Algorithm graphical user interface toolbox of MATLAB was used for minimization of dilution.
- 5. The optimum dilution obtained was 28.3288.
- 6. GMAW optimal process parameters were determined:

Wire Feed Rate = 8.03322 m/min; Welding Speed = 29.31976 cm/min; Voltage = 27.512V; Nozzle to Plate Distance = 17.0475 mm; Torch Angle = 94.33° .

REFERENCES

- Kim, D.; Rhee, S.; Park, H.Modelling and optimization of a GMA welding process by genetic algorithm and response surface methodology. International Journal of Production Research, 2002, 40, 7, 1699-1711.
- [2] Kim, I.S.; Son, J.S.; Kim, I.G.; Kim, J.Y.; Kim, O.S. A study on relationship between process variables and bead penetration for robotic CO2 arc welding. Journal of Materials Processing Technology, 2003, 136, 139-145.
- [3] Benyounis, K. Y.; Olabi, A. G.; Hashmi, M.S.J. Effect of laser welding parameters on the heat input and weld-bead profile. J. Mater Process Tech, 2005, 164–165.
- [4] Kannan, T.;Yoganandh,J. Effect of process parameters on clad bead geometry and its shape relationships of stainless steel claddings deposited by GMAW. Int. J Adv Manuf. Technol. 2010, 47, 1083–1095.
- [5] Singla, M.; Singh, D.; Dharmpal, D. Parametric optimization of gas metal arc welding processes by using factorial design approach. Journal of Minerals & Materials Characterization & Engineering, 2010, 9, 4, 353-363.
- [6] Sudhakaran, R.;Murugan, V.;Senthil, K. M.;Jayaram, R.;Pushparaj, A.; Praveen, C.; Venkat, P. N. Effect

of welding process parameters on weld bead geometry and optimization of process parameters to maximize depth to width ratio for stainless steel gas tungsten arc welded plates using genetic algorithm. European Journal of Scientific Research, 2011, 62, 1, 76-94.

- [7] Murugan, N.; Parmar, R.S.Stainless steel cladding deposited by automatic gas metal arc welding. Weld J. 1997, 76, 10, 391-s-402-s.
- [8] Subramaniam, S.; White, D. R.; Jones, J. E.; Lyons, D. W. Experimental approach to selection of pulsing parameters in pulsed GMAW, AWS. Weld J. 1999, 78, 5, 166-s-172-s.
- [9] Allen, T.T.; Richardson, R.W.; Tagliabue, D.P.; Maul, G.P. Statistical process design for robotic GMA welding of sheet metal. Weld J.2002, 81, 5, 69–76.
- [10] Kim, I.S.; Son, J.S.; Kim, I.G.; Kim, J.Y.; Kim, O.S. A study on relationship between process variables and bead penetration for robotic CO2 arc welding. Journal of Materials Processing Technology, 2003, 136, 139-145.
- [11] Lakshminarayanan, A.K.; Balasubramanian, V.;Varahamoorthy, R.;Babu, S. Predicting the dilution of plasma transferred arc hardfacing of stellite on carbon steel using response surface methodology. Metals and Materials International, 2008, 14, 6, 779-789.
- [12] Palaniswamy, P.; Rajendran I.; Shanmuga Sundram S. Optimization of machining parameters using genetic algorithm and experimental validation for end milling operations. Int J AdvManuf Technology, 2007, 32, 644 – 655.
- [13] Singla, M.; Singh, D.; Dharmpal, D.Parametric optimization of gas metal arc welding processes by using factorial design approach. Journal of Minerals & Materials Characterization & Engineering, 2010, 9, 4, 353-363.
- [14] Kumari, P.; Archana, A.; Parmar, S.Effect of Welding parameters on weld bead geometry in MIG welding of low carbon steel. International Journal of Applied Engineering Research, 2011, 6, 2, 249–258.
- [15] Yoganandh, J.; Kannan,T.;Kumaresh, S.P.; Natarajan, S. Optimization of GMAW process parameters in austenitic stainless steel cladding using genetic algorithm based computational models, experimental techniques. Society for Experimental Mechanics, 2010, 34, 01-11.