

39 properties are primarily controlled by the consumable composition of the shielding gas which
40 can directly influence the strength and ductility of a weld. Its paramount to understand the
41 statistical influence of process parameters in mild steel cladding weld bead geometry (Murugan
42 and Gunaraj, 2005).

43 This study investigates the application of statistics to investigate and to analyze the influence
44 of cladding weld metal geometry in mild steel using response surface method with the
45 application of TIG welding method. **Therefore**, the main objective of the study is to determine and
46 to evaluate the statistical solutions and its influences of the impact strength in mild steel cladding weld
47 metal bead geometry.

48 **2. PROCESSES OF WELDING**

49 These processes use a welding power supply to create and maintain an electric arc between an
50 electrode and the base material to melt metals at the welding point. They can use either direct
51 current or alternating current and consumable or non-consumable electrodes. The welding region
52 is sometimes protected by some type of inert or semi-inert gas, known as a shielding gas, and
53 filler material is sometimes used as well (Lincoln, 2014).

54 **2.1 Review of Related Literature under Study**

55 Palani and Murugan (2006a) expressed the mechanical and corrosion-resistant properties of the
56 coated components depend on the geometries of the coated beads, which in turn are controlled by
57 the process parameters. Therefore, it is essential to study the effect of the process parameters on
58 the cord geometry to allow effective control of these parameters. The above objective can be
59 easily achieved by developing equations to predict the dimensions of the weld bead in terms of
60 process parameters. The models developed were reviewed for their suitability. Confirmation
61 experiments were also performed and the results show that the developed models can predict the
62 geometries and the dilution of the beads with reasonable precision. It was observed from the
63 research that the interactive effect of the parameters of the process in the geometry of the account
64 is significant and cannot be neglected. Eutimio et al (2013), shows that most of statistical tools
65 currently applied in the bioprocess area were classified. The main three categories were: fair
66 comparison of results, mathematical modeling for little studied systems and taking advantage of
67 large volume of data for enhance robustness and efficiency. However, a chart was constructed

68 for guiding researchers to select the correct statistical technique according to the specific
69 bioprocess problem.

70 Achebo (2016) describes the process of developing a model that relates the shear stresses in a
71 gas welded aluminum alloy weldment with the corresponding flux constituent elements that
72 make up the flux composition. The weldments made from the 13 flux compositions were
73 subjected to evaluation by some professional welders whose judgments about the quality of the
74 weldments were evaluated by using the rank correlation coefficient method. Stefano et al (2009)
75 present the results of a research through the design of an experimental technique on the influence
76 of temperature, the residence time and the pressure of the bar in the resistance to heat sealing of
77 oriented polypropylene films coated with a thin layer of gelatin. This chemo-metric approach
78 allowed to achieve a complete understanding of the effect of each independent factor in the two
79 different responses considered as a measure of the force required to break the link through the
80 sealed interface.

81 Marko et al (2017), express that the process of laser cladding has become more important
82 during recent years because of its broad application for cladding, repair or additive
83 manufacturing. For high quality and reliability of the repaired components, it is necessary to
84 adjust the weld bead geometry to the specific repair task. The bead geometry influences the
85 metallurgical bonding and the degree of dilution as well as the formation of defects like pores or
86 cracks. The results show, the essential effects are detected with a full factorial test plan as well as
87 with a central composite design. Merely the effect strength could not always be specified
88 unambiguously. Mastanaiah et al (2014) described the Prediction of weld bead geometry is
89 always an interesting and challenging research as it involves understanding of complex multi
90 input and multi output system. The weld bead geometry has a profound impact on the load
91 bearing capability of a weld joint. The results of investigation suggests the effective thickness of
92 weld, a geometric parameter of weld bead has the most significant influence on tensile breaking
93 load of dissimilar weld joint. The observations on bead geometry and the mechanical are
94 correlated with detailed metallurgical analysis. Xu et al (2014) described the oscillating arc
95 narrow gap all-position gas metal arc welding process was developed to improve efficiency and
96 quality in the welding of thick-walled pipes. The developed models were checked for their

97 adequacy and significance by ANOVA, and the effects of wire feed rate, travel speed, dwell
 98 time, oscillating amplitude and welding position on weld bead dimension were studied. Finally,
 99 the optimal welding parameters at welding positions of 0° to 180° were obtained by numerical
 100 optimization using RSM. Nuri et al (2013) study is aimed at obtaining a relationship between the
 101 values defining bead geometry and the welding parameters and also to select optimum welding
 102 parameters. The welding process parameters that have the most effect on bead geometry are
 103 considered and the other parameters are held as constant. Then, the relationship between the
 104 welding parameters is modeled by using artificial neural network and neurofuzzy system
 105 approach. The models developed are compared with regard to accuracy and the appropriate
 106 welding parameters values can be easily selected when the models improve.

107 **3. DESIGN OF EXPERIMENT**

108 Design of experiment is a scientific approach of combining input parameters optimally so as to
 109 optimize a target response and this can be achieved by using computer software like design
 110 expert. For proper polynomial approximation, experimental designs are used to collect the data.
 111 In this research, central composite design in response surface method was used to generate the
 112 experimental runs. Furthermore, response surface method was used to evaluate, model and
 113 analyze the data statistically which generates the statistical results.

114 **3.1 Identifying the Range of Input Parameters**

115 The key parameters considered in this work are welding current, gas flow rate, welding speed
 116 and voltage. The range of process parameters obtain from literature is shown in the table below.

117 **TABLE 1 Process parameters at Low and High Levels**

Parameter	Units	Symbol	Low	High
Current	Amp	A	180	240
Gas flow rate	Lit/min	F	10	16
Voltage	Volt	V	18	24
Welding speed	Mm/s	S	90	145

118 Impact testing machine is a machine used for the impact testing analysis. It is used to test the
 119 impact strength of the materials to determine the energy or strength of the materials at a specific
 120 location of the material basically at the weldment and other specified locations the researcher
 121 wished to determine the strength in that location. It measures the unit of the material strength in
 122 Joules.



Fig. 1: Impact Testing Setup

In the fabrication industry materials standard and specification plays a very vital role in achieving good weld quality. The welding parameter specification is shown in the table below.

3.5 Method of Data Collection

The central composite design matrix was developed using the design expert software, producing 20 experimental runs. The input parameters and output parameters make up the experimental matrix and the responses recorded from the weld samples was used as the data. The input process factors are welding current, welding voltage, welding speed and gas flow rate. The output process response is impact strength of the weldment. The input and output parameters were analyzed statistically modeled and optimized.

3.6 Method of Data Analysis

Response Surface Methodology (RSM) Engineers often search for the conditions that would investigate the process of interest. RSM is one of the techniques currently in widespread usage to describe the performance of the welding process and find the statistical investigation of the responses of interest. RSM is a set of mathematical and statistical techniques that are useful for modeling and predicting the response of interest affected by several input variables with the aim of optimizing this response.

4.1 Modeling and Statistical evaluation using Response Surface Technique

In this paper, the researcher revealed a mathematical relationship between selected process factors, namely; current, speed, gas flow rate and voltage to the response variable. The response variable of interest is impact strength of the material.

Source	Sum of Squares	df	Mean Square	F Value	p-value Prob > F	
Model	2718.36	10	271.84	3.31	0.0429	significant
A-Gas flow vt	04.50	1	04.50	1.03	0.3370	
D-Welding sp	351.13	1	351.13	4.27	0.0006	
C-Welding vt	7.54	1	7.54	0.092	0.7688	
D-Current	40.50	1	40.50	0.49	0.5003	
AB	25.30	1	25.30	0.31	0.5924	
AC	1809.01	1	1809.01	22.02	0.0011	
AD	1.70	1	1.70	0.021	0.8888	
BC	0.36	1	0.36	4.398E-003	0.9486	
BD	49.82	1	49.82	0.61	0.4361	
CD	14.85	1	14.85	0.18	0.6807	
Residual	739.30	9	82.14			
Lack of Fit	224.38	5	44.88	0.33	0.8609 not significant	
Pure Error	514.91	4	128.73			
Cor Total	3457.66	19				

Fig. 2: Model Significance of the Impact Strength using ANOVA

145
 146
 147 Analysis of the model standard error was employed to assess the suitability of process factor and
 148 response variables using the central composite design-value model in response surface to optimize the
 149 impact strength on the weldment. The computed ANOVA of design responses was presented in
 150 figure 2. From the results of figure 2, the Model has ten (10) degree of freedom, with the model
 151 F-value of 3.31 which implies that the model is significant. There is only a 4.29% chance that an
 152 F-Value this large could occur due to noise. Values of "Prob > F" less than 0.0500 indicate
 153 model terms are significant. In this case A, D, AC are significant model terms. Values greater
 154 than 0.1000 indicate the model terms are not significant. The "Lack of Fit F-value" of 0.35
 155 implies the Lack of Fit is not significant relative to the pure error. There is 86.09% chance that a
 156 "Lack of Fit F-value" this large could occur due to noise. Non-significant lack of fit is good for
 157 the model fitness.

Std. Dev.	6.91	R-Squared	0.8971
Mean	91.78	Adj R-Squared	0.7827
C.V. %	7.52	Pred R-Square	N/A
PRESS	N/A	Adeq Precisor	8.724
-2 Log Likeliho	118.08	BIC	151.03
		AICc	173.08
Case(s) with leverage of 1.0000: Pred R-Squared and PRESS statistic not defined			
"Adeq Precision" measures the signal to noise ratio. A ratio greater than 4 is desirable. Your ratio of 8.724 indicates an adequate signal. This model can be used to navigate the design space.			

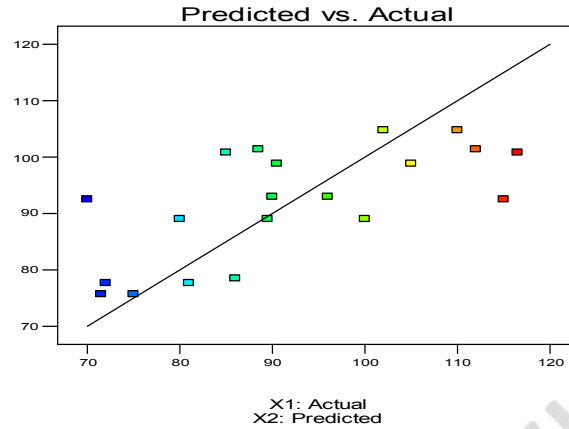
Fig. 3 Model Summary Analysis for validating Model Significance in Impact Strength

160 To validate the adequacy of the model based on its ability to maximize the impact strength,
 161 the goodness of fit statistics was presented in figure 3. From the result of figure 3, it was
 162 observed that the "Predicted R-Squared" value of null is obtained. In case(s) where leverage of
 163 1.0000 is obtained, Predicted R-Squared and PRESS statistic are not defined. However, the R-
 164 Squared value of the model is 0.8971 while the Adjusted R-Squared value of the model is
 165 0.7827. "Adequate Precision" measures the signal to noise ratio. A ratio greater than 4 is
 166 desirable. The computed ratio of 8.724 as observed in figure 3 indicates an adequate signal.
 167 This model can be used to navigate the design space. Variance inflation factor (VIF) less than
 168 10.00 calculated for all the terms in the design indicate a significant model in which the variables
 169 are correlated with the response.

Run Order	Actual Value	Predicted Value	Residual	Leverage	Internally Studentized Residual	Externally Studentized Residual	Cook's Distance	Influence on Fitted Value (DFFITS)	Standard Order
1	80.00	75.42	4.58	0.553	0.991	0.990	0.110	1.101	15
2	75.00	74.54	0.46	0.076	0.190	0.179	0.023	0.477	8
3	110.00	109.54	0.46	0.876	0.190	0.179	0.023	0.477	4
4	90.50	88.68	1.82	0.876	0.748	0.729	0.359	1.935	7
5	112.00	111.54	0.46	0.876	0.190	0.179	0.023	0.477	5
6	72.00	70.18	1.82	0.876	0.748	0.729	0.359	1.935	6
7	70.00	70.00	0.000	1.000					9
8	85.00	90.81	5.81	0.260	0.977	0.974	0.030	0.577	13
9	90.00	92.92	-2.92	0.053	-0.435	-0.414	0.001	-0.098	18
10	100.00	98.67	1.33	0.553	0.288	0.273	0.009	0.303	12
11	88.50	87.17	1.33	0.553	0.288	0.273	0.009	0.303	11
12	96.00	92.92	3.08	0.053	0.458	0.437	0.001	0.103	19
13	71.50	71.04	0.46	0.076	0.190	0.179	0.023	0.477	2
14	81.00	92.92	-11.92	0.053	-1.774	-2.073	0.016	-0.489	21
15	115.00	110.42	4.58	0.553	0.991	0.990	0.110	1.101	16
16	102.00	92.92	9.08	0.053	1.351	1.426	0.009	0.336	17
17	105.00	103.18	1.82	0.876	0.748	0.729	0.359	1.935	1
18	116.50	114.68	1.82	0.876	0.748	0.729	0.359	1.935	3
19	89.50	92.92	-3.42	0.053	-0.509	-0.487	0.001	-0.115	20
20	86.00	95.04	-9.04	0.260	-1.521	-1.664	0.074	-0.985	14

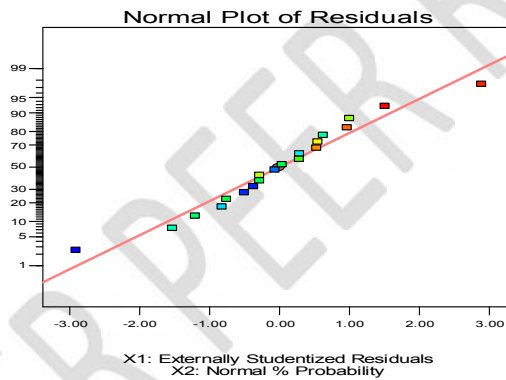
170 **Fig. 4: Diagnostics Statistics Report of Impact Strength (J)**
 171 The diagnostics case statistics report which shows the observed values of each response variable
 172 (impact strength) against their predicted values is presented in figure 4. The diagnostic case
 173 statistics actually give insight into the model strength and the adequacy of the optimal equation
 174 in terms of actual factors. To accept any model, its satisfactoriness must first be checked by an
 175 appropriate statistical analysis output.
 176

Design-Expert® Software
Impact Strength
Color points by value of
Impact Strength :
116.5
70

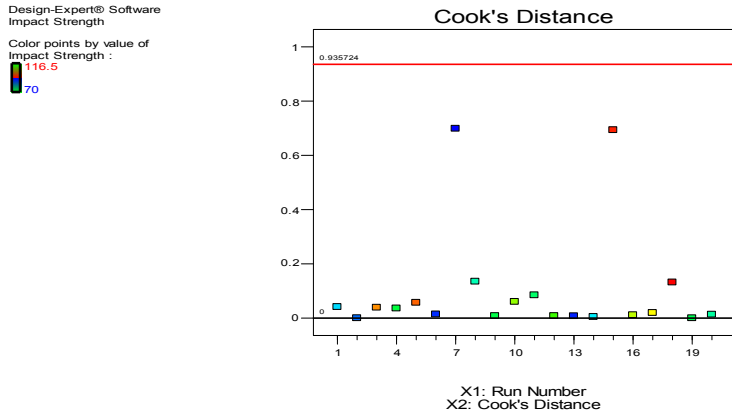


177
178 **Fig. 5 Statistical Investigation of the Predicted versus Actual Residuals**
179 Figure 5 shows the statistical plot of the predicted versus the the actual data in the response
180 parameter. It reveals the variations in the predicted and the actual data using linear fitted line, to
181 understand the differences between the predicted and actual response parameter variations.

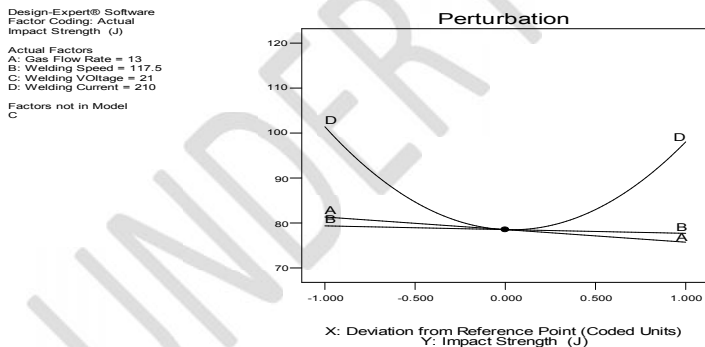
Design-Expert® Software
Impact Strength
Color points by value of
Impact Strength :
116.5
70



182
183 **Fig. 6 Normal Probability Plot of Residuals for Impact Strength**
184 To diagnose the statistical properties of the input factor design, the normal probability plot of
185 residual for impact strength is presented in figure 6. The normal probability plot of studentized
186 residuals was employed to assess the normality of the calculated residuals. The normal
187 probability plot of residuals which is the number of standard deviations of actual values based on
188 the predicted values was employed to ascertain if the residuals (observed – predicted) follows a
189 normal distribution. It is the most significant assumption for checking the sufficiency of a
190 statistical model. Result of figure 6 revealed that the computed residuals are approximately
191 normally distributed which is an indication that the model developed is satisfactory.

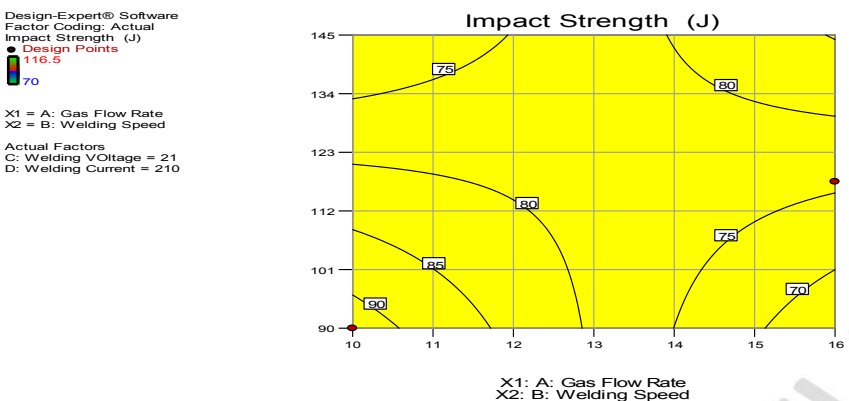


192
193 **Fig. 7 Generated Cook Distance-Versus-Impact Strength**
194 To determine the presence of a possible outlier in the experimental data, the cook's distance plot
195 was generated for the different responses. The cook's distance is a measure of how much the
196 regression would change if the outlier is omitted from the analysis. A point that has a very high
197 distance value relative to the other points may be an outlier and should be investigated. The
198 generated cook's distance is presented in figure 7. The cook's distance plot has an upper bound
199 of 1 and a lower bound of 0. Experimental values smaller than the lower bound or greater than
200 the upper bounds are considered as outliers and must be properly investigated. Result of figure 7
201 indicates that the data used for this analysis are devoid of any possible outliers thus revealing the
202 adequacy of the experimental data.

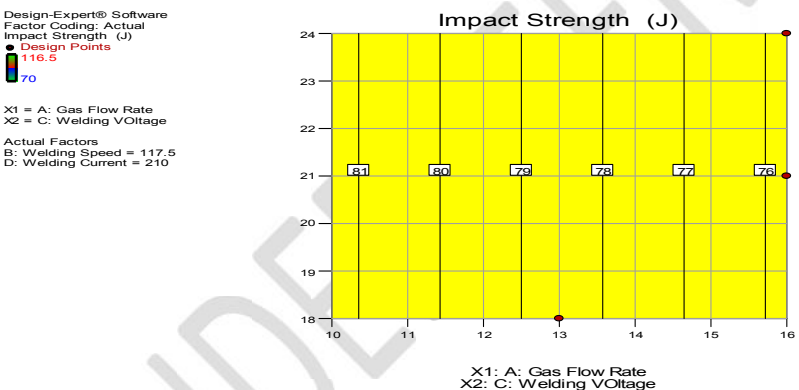


203
204 **Fig. 8: Perturbation Analysis of the Impact Strength**
205 To ascertain the influence of the alterations of process factors to the response variable,
206 perturbation analysis were employed as shown in figure 8. From the results of figure 8, it shows
207 that the disturbances in the response factors by the process factors, and the alterations of the
208 function of the external or internal means of the process factors in the response variables does
209 not make any of the responses to deviate from its reference points. This shows that the deviation

210 of the process factors does not disengage the responses from obtaining a good model and
 211 adequate optimization results.

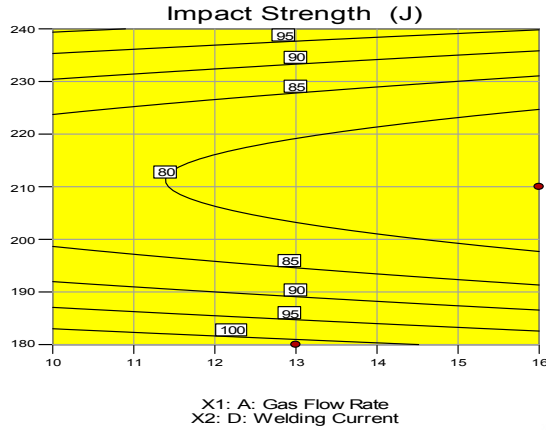


212
 213 **Fig. 9 Contour Plot of Impact Strength Influenced by Gas Flow Rate and Speed**
 214 From the results, the analyses in figure 9 express the influence of the input factors in the
 215 responses from the minimum bounded region of the response to the maximum bounded region of
 216 the response. It expressed that decrease in gas flow rate and welding speed will increase the
 217 impact strength.



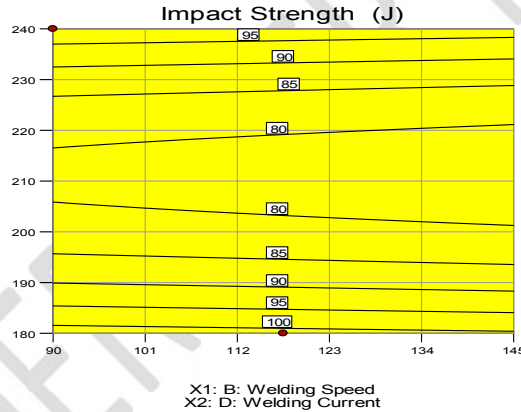
218
 219 **Fig. 10 Contour Plot of Impact Strength Influenced by Gas Flow Rate and Voltage**
 220 From the results, the analyses in figure 10 express the influence of the input factors in the
 221 responses from the minimum bounded region of the response to the maximum bounded region of
 222 the response. It expressed that decrease in gas flow rate increase the impact strength while
 223 voltage has no influence in the increase or decrease of the impact strength.

Design-Expert® Software
 Factor Coding: Actual
 Impact Strength (J)
 ● Design Points
 116.5
 70
 X1 = A: Gas Flow Rate
 X2 = D: Welding Current
 Actual Factors
 B: Welding Speed = 117.5
 C: Welding Voltage = 21

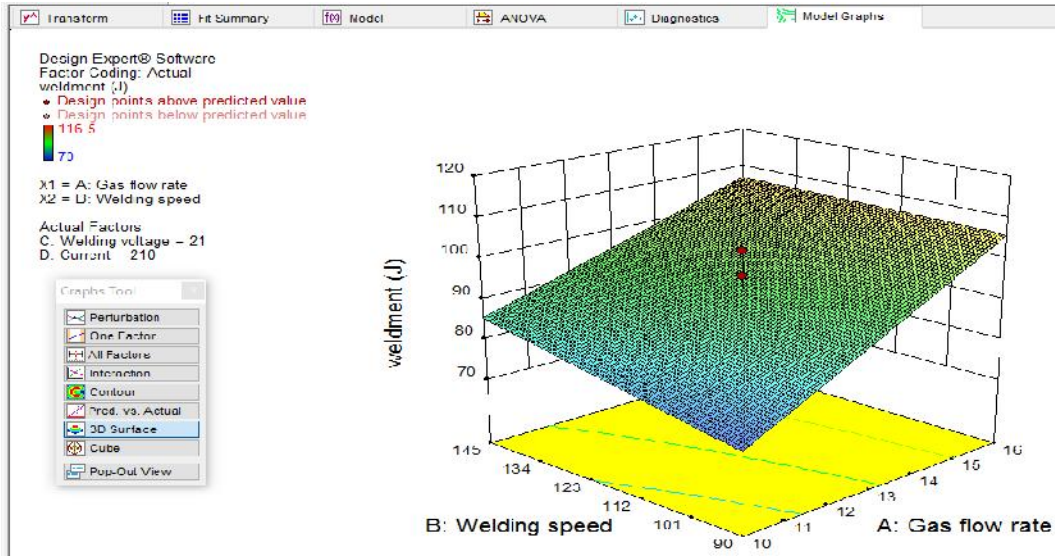


224
 225 **Fig. 11 Contour Plot of Impact Strength Influenced by Gas Flow Rate and Current**
 226 From the results, figure 11 indicates that an increase in gas flow rate increases the impact
 227 strength while current from its initial decrease the impact strength and at a point starts to increase
 228 the impact strength. This shows that the selection of the current will be carefully done due to its
 229 effects to impact strength.

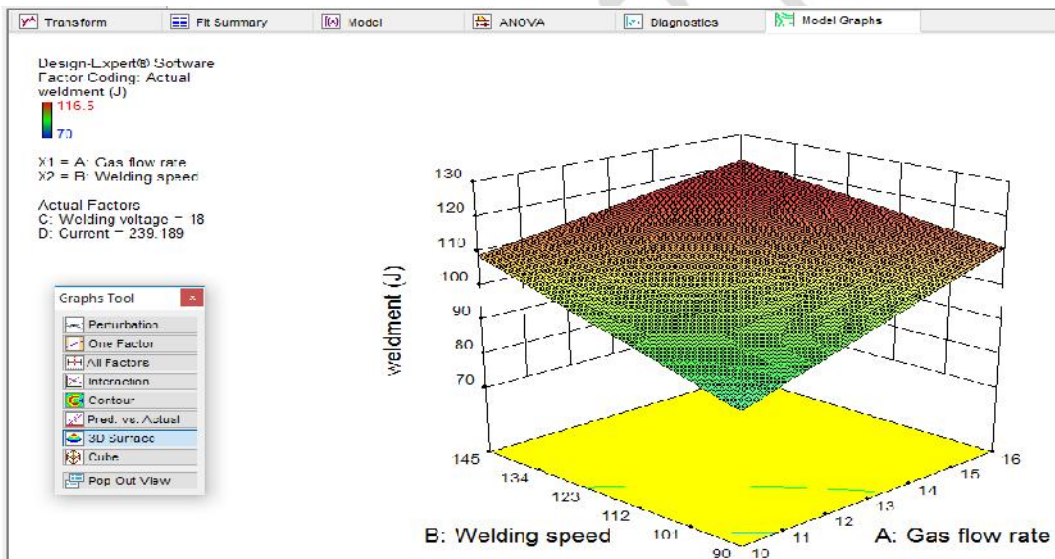
Design-Expert® Software
 Factor Coding: Actual
 Impact Strength (J)
 ● Design Points
 116.5
 70
 X1 = B: Welding Speed
 X2 = D: Welding Current
 Actual Factors
 A: Gas Flow Rate = 13
 C: Welding Voltage = 21



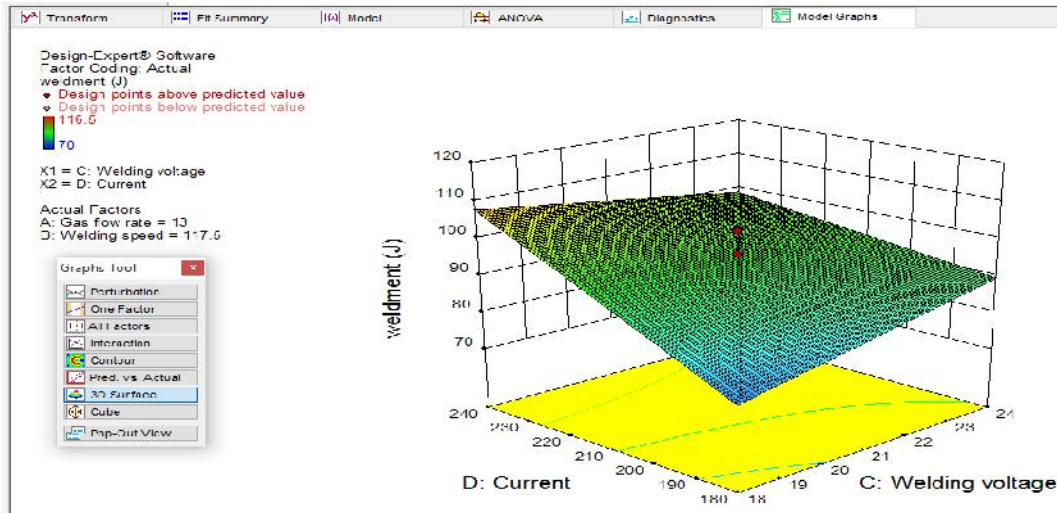
230
 231 **Fig. 12 Impact Strength Contour Plot Influenced by Speed and Current**
 232 From the results, the analyses in figure 12 expressed that increase in gas flow rate increase the
 233 impact strength while current from its initial decrease the impact strength and at a point starts to
 234 increase the impact strength. This shows that the selection of the current will be carefully done
 235 due to its effects to impact strength. However, the decrease in welding speed will increase the
 236 impact strength.



237
 238 **Fig. 13 Effects of Process Factors (with CD factors ratio of 50:50) on the Impact Strength**
 239 To study the effect of process factors with welding voltage and welding current at its average,
 240 figure 13 was presented.



241
 242 **Fig. 14 Effects of Process Factors (with CD factors ratio of 10:90) on the Impact Strength**
 243 To study the effect of process factors with welding voltage and welding current at its ratio of
 244 10:90, figure 14 was presented.



245
 246 **Fig. 15 Effects of Process Factors (with AB Factors Ratio of 50:50) on the Impact Strength**
 247

248 Figures 13-15 express the 3-dimensional (3D) response surface plots of impact strength on heat
 249 zone and its significant effects on process factors.

250

251 4. Discussion of Results

252 In this study, central composite design was employed owing to its simplicity and flexibility to
 253 variable adjustment and analysis of process interaction relating to process factors combination.
 254 The design and analysis was executed with the aid of statistical tool. For this particular problem,
 255 Design Expert 10.0.1 was employed. However, using response surface method, the results of the
 256 statistical evaluation for the selected process parameters and response parameter were observed.
 257 Analysis of the model standard error was employed to assess the suitability of process factor and
 258 response variables using the central composite design model in response surface to analyze
 259 statistically, the impact strength on the weldment. The computed ANOVA of design responses
 260 was presented in figure 2. From the results, the model F-value of 3.31 implies that the model is
 261 significant. There is only a 4.29% chance that an F-Value this large could occur due to noise.
 262 Values of "Prob > F" less than 0.0500 indicate model terms are significant. The "Lack of Fit F-
 263 value" of 0.35 implies the Lack of Fit is not significant relative to the pure error. There is 86.09%
 264 chance that a "Lack of Fit F-value" this large could occur due to noise. Non-significant lack of fit
 265 is good for the model fitness.

266 From the result of figure 3, it was observed that the "Predicted R-Squared" value of null is
267 obtained. In case(s) where leverage of 1.0000 is obtained, Predicted R-Squared and PRESS
268 statistic are not defined. However, the R-Squared value of the model is 0.8971 while the
269 Adjusted R-Squared value of the model is 0.7827. "Adequate Precision" measures the signal to
270 noise ratio. A ratio greater than 4 is desirable. The computed ratio of 8.724 as observed in
271 figure 3 indicates an adequate signal. This model can be used to navigate the design space.
272 Variance inflation factor (VIF) less than 10.00 calculated for all the terms in the design indicate a
273 significant model in which the variables are correlated with the response.

274 Experimental values smaller than the lower bound or greater than the upper bounds are
275 considered as outliers and must be properly investigated. Result of figure 7 indicates that the data
276 used for this analysis are devoid of any possible outliers thus revealing the adequacy of the
277 experimental data.

278 **Figure 13 shows** the process factors ratio of 50 to 50 (in current and voltage). It was observed
279 that increase in response (impact strength) increases welding speed (B) and gas flow rate (A).
280 This shows that increase or decrease on the process factors affect the response variable. In Figure
281 14, gas flow rate (A) and welding speed (B) were hold at a mix ratio of 50 to 50 or at its mean
282 which was used to determine the influence of other process factors to the response. It was
283 observed that increase in current (D), will increase the response(impact strength on weldment).
284 In addition the geometry of the surface was observed to be concave.

285 In Figure 15 a ratio of 10 to 90 in welding voltage (C) and welding current (D) was used. It
286 was observed that increase in welding speed (B) and gas flow rate (A) process factors, increases
287 the response(impact strength on weldment). This shows the lower the welding voltage (C) and
288 higher the welding current (D) will increase the impact strength on weldment which will
289 influence and enhance the increase on welding speed and gas flow rate of the process factors to
290 its response. The 3D surface plot as observed in figures 13-15, show the relationship between the
291 process factors (current, gas flow rate, speed and voltage), against the response variable (impact
292 strength). It is a 3-dimensional surface plot which was employed to give a clearer concept of the
293 surface. Although not as useful as the contour plot for establishing coordinates, this view
294 provides a clearer picture of the surface. It was observed from Figures 13-15 that the input

295 factors has significant influence on the surface geometry and the overall contributions towards
296 the response variable (impact strength).

297 **5. CONCLUSIONS**

298 A close examination of the mild steel cladding weld metal was experimented with the input
299 parameters of current, voltage, speed and gas flow rate to predict and to analyze the mild steel
300 cladding weld metal response parameter (impact strength) using response surface method.
301 Welding parameters were carefully selected.

302 The results of the statistical investigation revealed the model F-value of 3.31 is significant.
303 There is only a 4.29% chance that an F-Value this large could occur due to noise. Values of
304 "Prob > F" less than **0.05** indicate model terms are significant. The "Lack of Fit F-value" of 0.35
305 implies the Lack of Fit is not significant relative to the pure error. There is 86.09% chance that a
306 "Lack of Fit F-value" this large could occur due to noise. Non-significant lack of fit is good for
307 the model fitness. It was observed that the R-Squared value of the model is 0.8971 while the
308 Adjusted R-Squared value of the model is 0.7827. Adequate precision measures the signal to
309 noise ratio and a ratio greater than 4 is desirable. The computed ratio of 8.724 as observed
310 which indicates an adequate signal. This model can be used to navigate the design space.
311 Variance inflation factor (VIF) less than **10** calculated for all the terms in the design indicate a
312 significant model in which the variables are correlated with the response. In response surface
313 plots and contour plots, the process parameters influence the impact strength except voltage,
314 which has no effect on the response parameter.

315 The performed experiment will appraise the knowledge of mild steel cladding weld
316 formulation and composition in tungsten inert gas (TIG) welding system and also in
317 industrialization. The experimental analysis and its statistical evaluation will help in decision
318 making systematically **in the industrialization where** the product is more utilized.

319

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