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Optimization of the Stirring Parameters of AZ91Mg-based Stir Casted Hybrid Composites using Taguchi and ANN

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In the current research, the fabricated Mg hybrid composites amalgamate with the vacuum-based squeezed stir casting process having TiC and Al_2O_3 reinforcement's in which AZ91Mg-alloy is the base material. The study includes the processing parameters of the vacuum-based squeezed stir casting method (such as stirring time, stirrer depth, and stirring speed) that have been varied and then investigates their influential effect by using the L_{16} orthogonal Taguchi approach. By considering these parameters, the tribo-mechanical properties are also optimized like porosity, ultimate strength, and rate of wear loss of Mg-hybrid composites. The result reveals a significant association between processing parameters and optimized tribo-mechanical properties. The results signify that the processing parameters are best optimized at processing parameters of 500 rpm of stirring speed, 5 min of stirring time, and 50 mm of stirrer depth and the optimized tribo-mechanical properties are 0.4% porosity, 245 MPa of ultimate strength, and 0.0011 mm^3 per min is the loss of wear rate. However, the ANOVA results contribute that the stirrer depth has the most significant processing parameter compared to other optimized parameters. This research also includes an artificially designed networking model which validates the designated set of empirical data. Thus due to their suitability for abrasion wear AZ91-hybrid composite gives a new divergence towards the designing parts of minuscule airplanes used in surveillance applications such as ailerons and wing flaps.

Keywords: Artificial neural network, Mg-Hybrid composites, Stirrer optimize, VSSC technique, Taguchi approach

Introduction

Nowadays, during fabrication casting, surface deformities such as gas holes, white inclusions have been observed in Mg/Al matrix composites which minimize the potential ability of hard reinforcements in the fabricated composites. These defects also depreciate the strengthening properties of fabricated composites.^{1,2} Therefore, the better alternative for this problem is the use of hybrid composite materials because their mechanical and tribological properties are customized as per the requirements.³ The most frequently counted reinforcing materials in Mg-matrix are reviewed as boron carbide, silicon carbide, silicon nitride alumina, and TiC, etc.⁴

The stir casting method is a commonly used fabrication technique to fabricate Al/Mg matrix composites. This method provides homogeneous distribution in Al/Mg matrix composites with the least defects as compared to other processing techniques.⁵⁻⁷ Singh and Chauhan compared the mechanical properties of Mg/B₄C composites using electromagnetic and squeezed-pressure stir casting processes.⁸ Their study

reveals that Mg-composites fabricated by squeezed stir-casting technique shows improved mechanical results as compared to the electromagnetic stir-casting technique with the least defects. Singh et al.⁹ fabricate Mg/B₄C hybrid composites by employing the same fabrication method and using the statistical Taguchi approach to evaluate the mechanical properties. Their study evaluates the improved hardness and tensile properties of Mg/B4C hybrid composites when compared with Mg-matrix base material. Muhammad et al.¹⁰ studied Al-hybrid composites using a simulated statistical approach with the same fabrication technique and considering the parameters like stirring diameter and stirring speed. Their study accomplishes significant results without depreciating their microstructural properties. Then, another study performed by Deepak et al.¹¹ illustrated the homogeneous dispersion of aluminum oxide (up to 6 wt.%) in Mg-based composites and leads to minimal deformities by using the same stir-casting fabrication method.

In the literature study, researchers also propose different mathematical tools like neural networks and regression models.^{12–14} These models design a suitable association of input-output values by using different mathematical tools. With the help of different multi-

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response optimization tools, input parameter values are identified. During fabrication, the processing parameters reflect the service conditions of the fabricated composite as input parameters. The output factors help to determine the actual turnout values without repeating experimental fabrication trial runs. However, the output factors are mechanical and tribological parameters of the fabricated composites like hardness, modulus of elasticity, friction coefficient, and wear rate.^{15–17}

Aydin et al.¹⁸ worked on regression analysis of fabricated Mg/TiC-based composites by the stir casting process. And their study investigated the influence of different parameters like applied load, reinforcement percentage, sliding speed, and distance. And based on these parameters, wear loss rate values of Mg/TiC-based composites have been evaluated and optimized results had been obtained. Jeyaprakasam et al.¹⁹ evaluated the wear parameters of Al/TiC/Gr hybrid composites by using the surface method. According to their results, the sliding distance was the dominant factor that majorly affects the wear loss rate of Al-hybrid composites. Moreover, Prasad et al.²⁰ demonstrated the influence of TiC reinforcement with Mg/Si composites and examined the dry sliding wear conditions by using a Taguchi-based optimization method. Similarly, Samanta et al.²¹ applied an orthogonal array approach to Mg-based composites and analyzed the significant optimized value of reinforcement (TiC) and their wear parameters. Rajmohan et al.²² used the grey fuzzy (hybridized) method to forecast the optimal composition of wear rate and friction coefficient. Afzal et al.23 demonstrate an artificial neural network design model as an effective approach to predicting the tribological behavior of Mg/Al₂O₃ composites. Furthermore, Arif et al.24 examined the wear loss of Al/SiC hybrid composites by employing the ANN model i.e. feedforward back propagation method. Niranjan et al.²⁵ worked on optimizing the mechanical properties of Al-hybrid composites (alumina as reinforcing particles) by using the ANN model and considering processing temperature as one of the key parameters. Madhu et al.²⁶ determined the optimized wear parameters of Al/B4C/Gr composites by considering the processing parameters such as processing temperature and applied pressure by using artificial networking models. However, Anand et al.²⁷ designed the ANN model using non-linear regression parameters and illustrate the regression ANN model is more convenient than the basic ANN model design.

Thus, the above literature study stated that the Taguchi approaches and neural network models are conventionally discovered as the best-optimizing method for Mg-based composite materials but least work has been reported on Mg-hybrid composites using TiC and alumina as reinforcement fabricated through the Vacuum-Based Squeezed Stir Casting Method (VSSC).

Thus, the current study has been formulated by selecting two objectives by using VSSC processing (fabrication) parameters. The first objective is to examine the effect of stirring (fabrication) parameters of AZ91Mg hybrid composite by using the L_{16} orthogonal Taguchi approach. Secondly objective is to formulate the tribo-mechanical properties (like porosity, ultimate strength, and loss of wear rate) of fabricated AZ91Mg-composites. Also, an ANN-based regression mathematical model is designed by using these parameters through this design; a limit of experimentation has been performed.

Methodologies

Material Selection

A Hybrid Mg-based composite has been fabricated using VSSC process. The optimized amounts of reinforcing particles were included in the hybrid Mg-composites which was previously evaluated in the literature.²⁸ The reinforcing particles, titanium carbide (98.5% purity, 150-300 microns), and aluminum oxide (99.9% purity, 100-300 microns) were procured from SAC (Sigma Aldrich Chemicals) Pvt. Ltd, Bangalore, and CDH (Chemical Drug House) Pvt. Ltd, Delhi respectively. Local vendors provide Mg-alloy ingot from India-mart Intermesh Ltd., Noida, UP, India having 98.9% purity. However, the illustration of the metallographic images of the matrix, reinforcements materials, and the fabricated composite has been represented in Fig. 1. The TiC and Al₂O₃ reinforcements show irregular block and tubular shapes respectively, as shown in Fig. 1(a) and (b). Moreover, Fig. 1(c) presents the AZ91Mg alloy microscopic image having an erratic particle surface. The chemical distribution of AZ91Mg-alloy is 9.7 Al, 1.0 Zn, 0.5 Mn, 0.1Si, 0.005 Fe with a balance of pure Mg.

Fabrication Method

The Mg-hybrid composites (85 wt.% AZ91, 9 wt.% TiC, and 6 wt.% Al_2O_3) are fabricated using squeezed vacuum-based stir casting method as shown in Fig. 2. Initially, to remove the moisture contents

from TiC and Al_2O_3 reinforcing powders have been put in an electric oven at 300°C separately. Then under constant stirring, the AZ91 matrix (ingot) is placed in a vacuum-preheated furnace attached with Ar + SF6 cover gas to control from the fire. Then TiC $+ Al_2O_3$ have been dropped into the AZ91 melt with constant stirring. About 800–850°C of stirring temperature and 455 rpm of stirring speed have been given to the fabricating setup to achieve uniform composition. Further, the stirring parameters set by the design of the experiment as mentioned in Table 1.



Fig. 1 — Microscopic images of (a) AZ91Mg-alloy, (b) Titanium carbide, (c) Alumina, and (d) AZ91/TiC/Al $_2O_3$ hybrid composite



Fig. 2 — Vacuum-based squeezed stir casting setup

After 15 minutes of stirring, the hybridized melt has been dropped to the MS mold through the bottom pouring passage. It is a passage act as a pre-heater having a narrow inclined track (250°C) integrated at the bottom of the furnace to preserve the hybridized compo-melt. Then the compo-melt is released to the MS mold and immediately squeezed pressure (250 MPa) is provided by the hydraulic press for 10 minutes in order to eliminate the residual deformities. After solidification under atmospheric temperature, fabricated specimens moved out from MS mold and then cut out as per experimenting dimensions.

Orthogonal Taguchi Approach

According to Taguchi's orthogonal approach, the count of empirical trials has been calculated as 16 related to examining 3 factors at 4 distinct levels between the functioning ranges. The numbers of absolute variables for this experimental work are illustrated in Table 1 for 4 sets of levels. Each of the experiments was calculated according to the tabulated L_{16} orthogonal array. The optimized results have been investigated as output responses as represent in Table 2.

	Table	l — Proc	essing pa	arameters	and their	level	s	
Factors				Factor Levels				
				1	2	3	4	
A, Stirring Speed (rpm)				400	450	500	550	
B, Stirring Time (min)				5	10	15	20	
C, Stirring Depth (mm)			30	40	50	60		
Table	e 2 — Th	e experir	nental des	sign matri	x and their	outp	ut result	
			val	ues				
Sp No	Α	В	С	Р	US		WR	
-	(rpm)	(min)	(mm)	(%)	(MPa)	(m	m ³ /min)	
S_1	400	5	30	3.2	189	(0.0077	
S_2	400	10	40	2.7	210	(0.0062	
S_3	400	15	50	0.5	250	(0.0021	
S_4	400	20	60	1.8	227	().0039	
S_5	450	5	40	2.5	222	(0.0050	
S_6	450	10	30	2.8	202	(0.0070	
S_7	450	15	60	1.4	233	(0.0031	
S_8	450	20	50	0.2	241	(0.0013	
S_9	500	5	50	0.4	245	(0.0011	
S_{10}	500	10	60	1.6	229	(0.0035	
S_{11}	500	15	30	3.0	179	(0.0080	
S_{12}	500	20	40	2.6	214	(0.0055	
S_{13}	550	5	60	1.9	226	(0.0044	
S_{14}	550	10	50	0.6	237	(0.0026	
S_{15}	550	15	40	2.9	193	(0.0075	
S16	550	20	30	3.1	184	(0.0081	

Sp No = Specimen No, A = Stirring speed (rpm), B = Stirring time (min), C = Stirring depth (mm), P = Porosity (%), US = Ultimate strength (MPa), WR = Wear rate (mm³/min)

The strengthening (tensile) test has been tested on a tensometer testing machine, JMI, Delhi by following ASTM E8 standard at room temperature. The shape of the tensile specimen is shaped in a dog-bone type. The dry-sliding wear tests have been investigated on the cylindrical specimen of Mg-hybrid composite by following the ASTM G99 standard.²⁹ The specimen dimensions of the wear test are 5 mm in diameter and 50 mm in length. The dry sliding wear test has been performed on pin-on-disc computerized tribometer, Workshop, DTU, Delhi with MS disc materials having a surface roughness of 1.5 Ra. The sliding speed and applied load during the tribometer test have been set as 5 m/s and 30 N, respectively. The constant sliding distance has been maintained at 1 km. Since the work was conducted on optimized processing parameters of stirring, the conditions of the wear experiment are to be fixed for all experiments. Before and after, the wear mass loss of Mg-composites has been measured by using an electronic weighing setup following ASTM G99 standards.

The output responses have been analyzed by using the mean values of the Main Effect Plots (MEP). Fig. 3 (a & c) calculate the rise of stirrer speed from the initial level (450 rpm) to the second position (500 rpm). This signifies the proper blending of AZ91Mg alloy and reinforcing particles yields low porosity and minimum wear rate in the fabrication of Mg-hybrid composite. However, when an increment in the stirring speed induced the penetration of gas in the hybridized melt because of the initiation of pressure variation between the outer and inner surfaces of the compo-melt resulted in to increase in porosity and wear rate.³⁰ Also, the increment of stirring time leads to the evolution of oxides in the fabricated composite. This causes the declination of the tensile property i.e. ultimate strength of the Mg-hybrid composite. This can be examined from the evaluated mean effect data plots for the tensile strength of the Mg-hybrid composite as represented in Fig. 3(b).

Results and Discussion

Microstructure and Porosity

The metallographic studies of the fabricated Mgcomposites have been studied using an electron microscope as shown in Fig. 4. The porosity of Mghybrid composites has been evaluated using the theoretical density calculated through Archimedes principle and compared with the base matrix as shown in Table 2 and Fig. 5.

Specimens S_1 and S_{16} have noted that 4% increment in the porosity as illustrated in Fig. 4 (a & b) and Fig. 5. However, Fig. 4 (c–e) represents the metallographic images of specimens S_6 , S_5 , and S_2 respectively. As, these images show an increment in the pores of 3.6%, 3.7%, and 3.8% when compared to the base matrix. It is noted that specimens S_8 and S_9 represent only 1.5% of pores in comparison to the base matrix. The reduction in porosity can be validated from the micrographs as presented in Fig. 4 (f & g).



Fig. 3 — Mean main effect Plots for (a) Porosity, (b) Tensile strength, and (c) Loss of wear rate based on the parameters (i) Stirring speed (ii) Stirring time and (iii) stirring depth



Fig. 4 — Microscopic Images of Mg-based composite specimens (a) S₁, (b) S₂, (c) S₅, (d) S₆, (e) S₈, (f) S₉, and (g) S₁₆



Fig. 5 — Comparison of actual density and fabricated Mg-composite specimens

Desirable Functional Approach

Different multi-objective optimization mechanisms are involved in various literature works like the study of grey relational technique, utility method, desirability approach, TOPSIS, etc. In the current research study, the desirability approach technique has been used for the optimization of multi-objective functions. As per this technique, the output responses of each desirability value after all the experimental runs have been evaluated and estimated by using two criteria i.e. (a) 'larger is better' or 'nominal is better' and (b) 'smaller is better'.

In this research, porosity, and wear loss rate is considered as output parameters under the rule of 'smaller is better' whereas tensile strength has been analyzed with the rule of 'larger is better' criteria. The output response of the desirability value is fixed as '0' if its values occur above the suitable range whereas the desirability value is fixed as '1' if its data values are closer to the permissible range. Despite this, these acceptable ranges are determined as the upper and lower limits of the desirability output response values from the experimentation. Specific desirability output response values have been calculated by using the given formula in Eqs. (1 & 2).⁽³¹⁾

$$d_{i} = \begin{cases} 1 & y_{i} \leq L_{L} \\ \left(\frac{U_{L} - y_{I}}{U_{L} - L_{L}}\right)^{r} & L_{L} \leq y_{i} \leq U_{L} \\ 0 & y_{i} \geq U_{L} \end{cases} \dots (1)$$

('Smaller is better' criteria)

$$d_{i} = \begin{cases} 1 & U_{L} \leq y_{i} \\ \left(\frac{y_{i} - L_{L}}{U_{L} - L_{L}}\right)^{r} & L_{L} \leq y_{i} \leq U_{L} \\ 0 & L_{L} \geq y_{i} \end{cases} \dots (2)$$

('Larger is better' criteria)

where, y_i: output (response) value,

L_L: output lowermost (response) values,

UL: output uppermost (response) values and

r: shape constant which represents the rank order of the estimated model of the output value (i.e. specific weight-age provided for output results)

In current work, each of the output outcomes is treated as equally significant, and thus 'r' and '1' values are considered for analysis. The desirability value of the Mg-hybrid composite (overall) is the key parameter that shows the combined effect of multioutput responses. Then by the virtue of another Eq. (3), the desirability values of Mg-hybrid composite (d_c) for each experimental run are calculated.³²

$$d_{\mathcal{C}} = (d_1 \times d_2 \dots \dots \times d_n)^{\frac{1}{n}} \qquad \dots (3)$$

where, d_1, d_2, \dots, d_n is the specific desirability output response of each value and n is the estimation of output results studied for this work.

The consolidation of measured input includes the highest absolute desirability value have been determined by the optimal solution. The assigned solution of measured inputs gives the optimal data values for each of the output (response) variables. The measured values of each desirability value of the Mg-hybrid composite are listed in Table 3. Based on the investigated results, the A3B1C3 combination has been determined as the fixed optimal factor. The evaluated values for ultimate strength, porosity, and wear loss rate are calculated as 245 MPa, 0.4%, and 0.0011 mm³/min respectively with the significant solution of each input parameter. The main effect plots include the SNR (signal-to-noise) ratio and mean ratio are equivalent to the overall desirable values evaluated by using Minitab 18 software and represented in Fig. 6(a & b). 0.5734 is the computed comprehensive mean output result value of all the Mg-hybrid composite desirable data values. Then, the layout of a regression design is formulated for the Mg-composite desirable value in contrast to the input values (A, B & C), with 0.9935 R^2 parameters as shown in Eq. (4).

Mg-hybrid composite desirability value

 $= -4.80674 + (0.014266 \times A) + (0.42134 \times B)$ $+ (0.048708 \times C) - (5.45304 \times 10^{-4} \times A \times B)$ $+ (1.82485 \times 10^{-6} \times A \times C) - (9.5446 \times 10^{-5} \times B \times C)$ $- (1.12412 \times 10^{-5} \times A^{2})$

 $-(0.012407 \times B^{2}) - (6.73845 \times 10^{-4} \times C^{2}) \qquad \dots (4)$

Table 3 — Composite and individual desirable values of response values

G . M	Individual desirability values							
Sp No	Р	US	WR	CDV	Rank			
S_1	0	0.14	0.057	0	14			
S_2	0.16	0.43	0.257	0.265	11			
S_3	0.9	1	0.857	0.917	3			
S_4	0.46	0.67	0.585	0.569	7			
S_5	0.23	0.60	0.428	0.392	9			
S_6	0.13	0.32	0.157	0.189	12			
S_7	0.6	0.76	0.714	0.688	5			
S_8	1	0.87	0.971	0.946	2			
S_9	0.93	0.92	1	0.953	1			
S_{10}	0.53	0.70	0.642	0.622	6			
S_{11}	0.06	0	0.001	0	14			
S_{12}	0.2	0.50	0.342	0.326	10			
S ₁₃	0.43	0.66	0.514	0.528	8			
S_{14}	0.86	0.81	0.771	0.817	4			
S_{15}	0.1	0.19	0.085	0.119	13			
S_{16}	0.03	0.07	0	0	14			

Sp No = Specimen Number, P = Porosity 'smaller is better', US = Ultimate strength 'larger is better', WR = Wear rate 'smaller is better', CDV = Composite desirable values

The residuary data values for the derived regression designed formulation are conferred in Fig. 6. The residuary values represent each of the residual data outcomes shown within the range -0.02 and +0.02 values, which confirms the adequacy of the regression formulation. The dual peaks of the histogram represent the homogeneous dispersion of the residual plots in contrast to the data results in Fig. 7.

Validation test

The validation test is executed empirically with the average Mg-composite desirable values to represent the divergence of all desirability values. The absolute sequence of the leading variable was obtained as (A4B3C2) which is ordered as 13 in desirability analysis and then in comparison with the optimal solution of input variables (A3B1C3). With the help of Eq. (5) confirmation test has been carried out.

Predicted Mg-Composite Desirability

Value =
$$\gamma_m + \sum_{i=1}^{l} (\gamma_i^m - \gamma_m)$$
 ... (5)

where, γ_m are the absolute mean Mg-composite desirable values and γ_i^m is the average desirable data points at each independent solution.³³

The actual and evaluated Mg-composite desirable values have been correlated and represents in Table 3. 0.834465664 is the calculated variation of the Mg-composite desirable values (optimal and initial



Fig. 6 — Mean main effect plots of (a) Composite desirable values (b) Signal to noise ratio of composite desirable values



Fig. 7 — Plotting results of derived regression design formulation: (a) Normal probability plot, (b) Versus fits, (c) Histogram, (d) Versus order

	Table 4 — Resul	ts of the validation	ion test				
			Optimal com	Optimal combination			
	Initial varia	Initial variable setting A4 B3 C2			Direct approach base		
Level	A4 B3 C2						
Porosity (%)	1.8		0.4				
Ultimate strength (MPa)	226	226					
Wear Rate (mm ³ /min)	0.0044	0.0044					
Composite desirable values	0.1191171	0.119117115			0.953763679		
	Table 5 — Anal	ysis of Variance	results				
Input Parameters	DOF	SS	MS	F-Ratio	Contrib (%)		
A, Stirring speed (rpm)	3	0.05530	0.018334	11.41	5.5387		
B, Stirring time (min)	3	0.00304	0.001143	0.96	0.3432		
C, Stirring depth (mm)	3	0.92989	0.309949	186.22	93.6260		
Error	6	0.00449	0.001674		0.4921		
Total	15	0.99272			100.00		
DOF = Degree of freedom, SS = Sum of	squares. MS = Mean of s	quare. Contrib :	= Contribution				

solution) by using the direct path. The concluded results in Table 4, it is observed that the evaluated value of optimal Mg-composite desirable value has been improvised in contrast with the selected prime set by 0.834465664. It is also found that porosity percentages for the significant solution are obtained as 2.8% and 1.3%. However, analogous tensile strengthening values are noted as 226 MPa and 245 MPa. Also, the wear rate data values are recorded to be 0.0044 mm³/min and 0.0011 mm³/min.

ANOVA

The influence of input factors for evaluating the optimum output outcome is investigated using ANOVA (analysis of variance) as shown in Table 5. It is noticed from the outcomes observed in Table 5 that the stirring

depth is found as the most significant input parameter/factor that leads to 93.272% of output responses. At similar time duration, the stirring time is found as the least affecting factor compared to other input factors considered.

Formation of Neural Network Structure

ANN is an effective modelling tool employed to identify the relation in-between the significant factors and the outcome results in different fields. The artificial networking method emulates the functions of the human intellect that includes the actions such as collecting the facts from the data, receiving the information, and evaluating the relation of the data variations.³⁴

In this research work, a neural network is considered for the evaluation of the outcomes inbetween the input parameter ranges that are used in the experiments. By using MATLAB software, the ANN model of the current study is designed. The neuron-like points are showing in the input layers show the list of input factors like stirring time, stirring speed, and stirring depth. The neuron-like points in the output slab are similar to the 3 input parameters studied in this research, namely ultimate strength, porosity, and loss of wear rate. The dormant (hidden) slabs, linked through output and input layers, are boosted up in the ANN model to enhance the accuracy of the responses. One dormant slab is observed to be optimum subsequently performing the number of runs. The dormant layer comprises ten neuron-like points considered in the ANN model as represented in Fig. 8. The designed artificial network is competent and includes 80% of the accessible data and performed 20 percent of the left-out data by considering the back-propagation algorithm. 80% of the nourished input-out data sets (thirteen numbers), about 70% of competent data are considered for preparation and 15% of the data are values for confirmation as well as 15% of left-out data are employed for verification. The experimental outcomes of specimens S16, S9, and S1 are considered for the experiment of the artificial neural design, whereas all other data results are selected for the training scheme.

 (\mathbf{R}^2) regression The responses for the experimenting and confirmed results as compared to actual outcomes are presented in Fig. 9 along with the correlation plot. These plots confirm the applicability of the derived artificial neural model. The results plotting of the derived artificial neutral design is illustrated in Fig. 10. The neuron-like points approve the data accurately at zero periods (epoch) itself, and further focalize by operating the trials up to 1000 periods (epochs). After the experimentation of the neuron-like points, the derived network has been performed by using the experimental outcomes (i.e. S1, S9, and S16). However, this derived network helps to evaluate the competence of the design to investigate the required output results.



Fig. 8 — The Structure of modelled artificial neural network

Affirmation of Artificial Predicting Model

The estimated outcomes (Porosity, Tensile Strength, and Loss of Wear Rate) of considered experimental artificial neural design are correlated by using the actual experimenting data series and shown in Fig. 11(a–c). The relation data values represent a very minimal difference between the experimenting data series. It validates the suitability of the artificial neural design for the current study.

The microscopic images of the tensile test shows fracture in form of fine dimple flocks with tear ridges are represented in Fig. 12 (a & b). This represents the ductile failure of specimen S_{10} . However, in S_{15} showing coarse dimple flocks ridges confirms the ductile failure. Although, in a few places pores have been observed which act as the initiation of crack nucleation growth. Moreover, if this nucleation growth is successive then fails and considered as a tensile failure mechanism.



Fig. 10 — Performance curve of generated neural network (best validation performance is 28.1686 at epoch 148)



Fig. 11 — Compared results of experimenting and networking model of (a) Porosity, (b) Tensile strength and (c) Loss of wear rate



Fig. 12 — Microscopic image of tensile fracture (a & b) and Worn surfaces (c & d) of Mg-based S_{10} and S_{15} specimens

The microscopic images of the wear results of specimens S_{15} and S_{10} are represented in Fig. 12 (c & d). Both Mg-composites ($S_{10} \& S_{15}$) show adhesive wear having white regions in form of oxide surfaces. The origin of oxidation behavior is because of the high interfacial temperature developed throughout the wear process.³⁵ In specimen S_{15} , slight delamination of the surface has been observed. From mellow to severe wear transition region, adhesive wear and delamination have been observed which confirms the wearing of the Mg-composite specimens become stern.

Conclusions

Hybrid AZ91Mg-composite has been successfully fabricated by using the vacuum-based squeezed stir cast process technique. The consecutive outcomes have been deduced that within the selected range of processing parameters and derived significant optimized results simulates high porosity and the formation of clusters in the hybrid composite melt. The stirring depth represents a uniform distribution of reinforcing particles in the Mg-hybrid composites. A non-linear regression design of Mg-hybrid composite propagates the desirable value i.e. R2 data value of 0.9935 and confirms the validity test. However, ANN design develops the given series of experimental data values. The metallographic images of fracture surfaces of Mg-hybrid composite signify ductile fracture transition. The metallographic worn images of Mg-hybrid composite specimens show the delamination and adhesive behavior under different circumstances.

Thus these results give new divergence towards the designing of small-size surveillance airplanes parts such as ailerons, and wing flaps due to their suitability for abrasion wear of Mg-based composites under variable load and speed conditions. But during actual modelling analysis of such design parts, conditions may vary due to its validity as per requirements.

Declaration

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