

# Experimental Investigation of Jet Erosion Wear for Copper IS 191 due Solid-liquid Mixture

Sanket A. Patil<sup>1</sup>, E. R. Deore<sup>2</sup>, Purushottam S. Desale<sup>3</sup>

PG Student, Mechanical Engineering Department, S.S.V.P.S.'s, B.S.D. College of Engineering, Dhule, India<sup>1</sup>

HOD, Mechanical Engineering Department, S.S.V.P.S.'s, B.S.D. College of Engineering, Dhule, India<sup>2</sup>

Associate. Prof., Mechanical Engineering Department, S.S.V.P.S.'s, B.S.D. College of Engineering, Dhule, India<sup>3</sup>

**Abstract:** Erosion wear is a serious problem for slurry transportation and slurry handling equipment's. It plays an important role in design and operational conditions of slurry transportation system. This paper shows the experimental investigation of weight loss in copper material due to solid-liquid (slurry) impact by using different operating conditions and the validation of the readings using regression and Adaptive Neuro-Fuzzy Inference system (ANFIS). For the experimental investigation, Jet Erosion pot tester is fabricated by using propeller for uniform distribution. Special arrangement is made for angular movement of test carrying fixture in which it can move from 15° to 90°. Different experiments are carried on copper material for repeatability. For experimental study, different parameters like solid-liquid (slurry) concentration, particle size, variable angles (from 15° to 90°) and time variable are used. These experimental repeatability readings are then validated by using Regression analysis and Adaptive Neuro Fuzzy Inference system (ANFIS). During this, it is observed that the erosion wear of copper (Ductile material) is maximum at 30° and also the percentage of accuracy in ANFIS is better than regression analysis.

**Keywords:** Erosion wear, Slurry containing Pot tester, Slurry flow, Repeatability, Adaptive Neuro-Fuzzy Inference System, Regression.

## I. INTRODUCTION

Slurry erosion is happen because of the interaction of solid particles suspended in a liquid and a surface which experiences loss of mass by the repeated impacts of particles. It is one of the major sources of mass loss and failure of several slurry equipment and hydraulic components used in many industrial applications. Therefore, there is need to solve or minimize effect of mass loss. The slurry erosion is a difficult phenomenon and it is not fully understood because it is influenced by different parameters, which act simultaneously. These factors include flow field parameters, target material properties and erodent particle characteristics and its size distribution, velocity of slurry.

Among these parameters, the impingement angle and microstructure of the target material play an important role on the material removal process [1]. Erosive wear is the dominant process which can be say as the removal of material from a solid surface. It is due to mechanical interaction between the surface and the impinging particles in a liquid stream. In Erosion process there is a transfer of kinetic energy to the surface. With the increase in kinetic energy of the particles impacting at the target surface, it leads to increase in the material loss due to erosion. [2-3].

## II. LITERATURE REVIEW

Erosion wear is classified due to solid particle, liquid impingement and cavitations erosion. This classified erosion effect on slurry containing equipment, pumps,

fittings, pipes etc. Where Slurry erosion is a serious problem may lead due to the impingement of solid-liquid (slurry) mixture on the surface of target material. This slurry erosion may vary due to its impingement factors such as velocity of slurry, solid-liquid concentration, particle size, material properties of target surface, different operating angles, viscosity of slurry. During this experiments with different percentage of solid-liquid concentration(10%, 30%), angle of inclination(15°, 30°, 45°, 60°,75°, 90°), particle size(455 μm, 650 μm) and time dependency, it seem that the erosion in the form of weight loss for copper (Ductile) material is maximum at 30° and weight loss reduces, if angle of degree increases.

Many researchers have been trying to reduce the wear through various techniques but it has been difficult to find out the common cause and remedy of this problem due to its variation and dependency on large number of parameters. G.R. Desale[1], studied the erosion-corrosion wear of aluminium alloy composites. The LM13 alloy and LM13-Sic composites were taken to study the effect of sand concentration on the wear behavior in acidic and marine environment for a traversed distance of 763km and a rotational speed of 900rpm. They observed that the wear rate increase with increasing sand concentration irrespective of the material. This is because as the concentration of sand in the slurry increases the severity of erosion /abrasive attack increase because a greater number of particles are impinging on the surface. On the other hand, the slurry of corrosion attack may decrease because the effective volume of the corodent decreases. The alloy



Base Stand: Main function of the base stand is to hold and support the hole assembly of jet erosion tester [Fig4].

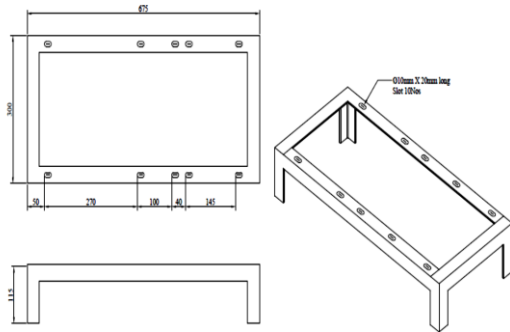


Fig.4 Base Stand

Motor Holder: Purpose of motor holder is to hold and support the stirrer motor in vertical direction and exactly at the centre of the pot where propeller blade turbine is situated [Fig5].

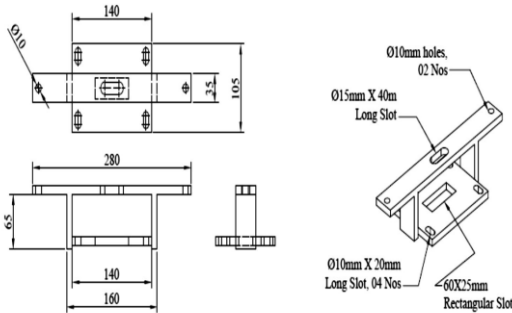


Fig.5 Motor Holder

Specimen Holder: This is the most important part of the test rig it has four holes drilled on it to clamp the specimen plate. It is made up of mild steel, one rectangular flat pate is welded to the bar which has threads at one end for tightening purpose and also marking on it, which coincides with the lines mark on a bar at different working angles [Fig6].

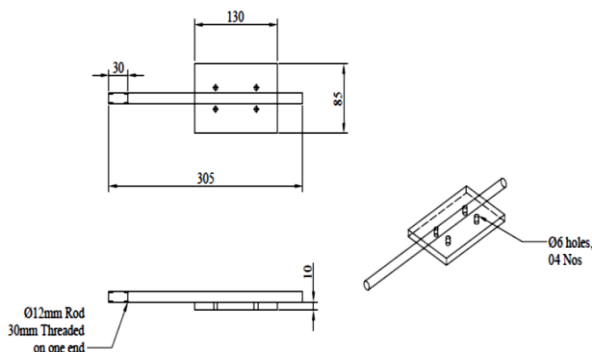


Fig.6 Specimen Holder

Particle size distribution:

Measurement of particle size distribution is essential to establish the variation in the particles in the solid sample and the percentage of particles present in different size ranges. For the coarser particles, sieve analysis can be

used to determine the particle size distribution. This distribution has been obtained by dry sieve analysis method. A representative sample of the solid particle is taken and sieving is done with a set of sieves. Special care is taken to ensure that the sample is properly dried. The sample retained on each sieve is collected and its percentage is calculated following the standard procedure. The particles of IS sand is selected as solid material for the present investigation and its physical properties are given in Table 1. It is not possible to collect identical size particles of the solid material. The particles are, therefore, sieved using successive sieves sizes and the particles collected between two successive sieves are designated by the mean sieve size.

TABLE I. PHYSICAL PROPERTIES OF ERODENT USED

Solid Particle	Chemical Formula	Colour	Sp. Gravity (Kg/m <sup>3</sup> )	Hardness VHN	Particle Shape
Quartz (IS Sand)	SiO <sub>2</sub>	Whitish	2652	1100	Sub Angular

TABLE II ELEMENTAL COMPOSITION OF TARGET MATERIAL USED

Target Material	Element Composition (Wt. %)					
	Cu%	Zn%	Pb %	Mn %	P%	Si%
Copper IS 191	99.974	0.000	0.003	0.000	0.013	0.010

TABLE III RANGE OF PARAMETERS FOR THE INVESTIGATION ON EROSION BEHAVIOUR OF DUCTILE MATERIAL COPPER IS 191

Target Material	Impact angles	Particle Size (µm)	Solid particle	Solid Concentration % by wt.	Time
Copper IS 191	15°, 30°, 45°, 60°, 75°, 90°	450	Quartz (IS sand)	10%	60 min

VI. EXPERIMENTAL PROCEDURE

The procedure has to be followed on erosion tester to calculate the erosion wear of different materials is as follow:

1. Firstly the specimen is cleaned properly.
2. Drying, if required.
3. Weighing the specimen (initial weight).
4. Clamp the specimen in fixture provided in test rig.

5. Setting the holder at required angle.
6. Weight the required sand as per concentration of slurry.
7. Mixing the proper amount of water and sand in tank.
8. Start the pump.
9. Adjust the flow rate to obtain desired value of mass flow rate and running the test for required time interval.
10. Removing the specimen from the fixture.
11. Cleaning and drying the specimen.
12. Weighing the specimen after erosion to measure the mass loss.
13. Repeat the steps.

## VII. REGRESSION

Since multiple regression is used to determine the correlation between a criterion variable and a combination of predictor variables, the statistical multiple regression method is applied. It can be used to analyse data from any of the major quantitative research designs such as causal-comparative, correctional, and experimental. This method is also able to handle interval, ordinal, or categorical data and provide estimates both of the magnitude and statistical significance of the relationships between variables. Therefore, multiple regression analysis will be useful to predict the criterion variable of material erosion rate via predictor variables such as Angle, Time, Particle size, and solid concentration.

In which experiment 48 test reading are conducted with variable time, concentration, angle, and size.  $E_a = 0.00118 X (C^{0.042}) X (T^{0.037}) X (S^{0.038}) X (A^{0.27})$  Where,  $E_a$ - Erosion rate of selected material, C- Solid concentration in percentage- Time in minute, S- Particle size in micron and A- Angle in degree. New empirical model Eq. (1) is developed for prediction of material erosion rate using Angle, Concentration, Time, particle size. It is observed that the proposed equation establishes relation among input variables and response variable. The average deviation observed in measured value and regression predicted value is 18.34 % at confidence level of 81.65%. These regression prediction models compare with ANFIS prediction model values to verify the accuracy of prediction models.

## VIII. ADAPTIVE NEURO FUZZY INFERENCE SYSTEM

Adaptive Neuro Fuzzy Inference System (ANFIS) is a fuzzy inference system implemented in the frame work of adaptive networks. Using a given input/output data set, the ANFIS method constructs a fuzzy inference system (FIS) whose membership function parameters are tuned (adjusted) using a back propagation gradient descent and a least-squares type of method.

This allows fuzzy systems to learn from the data they are modelling. FIS structure is a network-type structure, which maps inputs through input membership functions and associated parameters, and then through output membership functions and associated parameters to

outputs. ANFIS applies two techniques in updating parameters. For premise parameters that define membership functions, ANFIS employs gradient descent to fine-tune them. For consequent parameters that define the coefficients of each output equations, ANFIS uses the least-squares method to identify them. This approach is thus called hybrid learning method since it combines the gradient descent method and the least-squares method.

ANFIS modelling process starts by obtaining a data set (input-output data pairs) and dividing it into training and testing data sets.

The training data set is used to find the initial premise parameters for the membership functions by equally spacing each of the membership functions. A threshold value for the error between the actual and desired output is determined. The consequent parameters are found using the least-squares method. Then an error for each data pair is found. If this error is larger than the threshold value, update the premise parameters using the gradient decent method. The process is terminated when the error becomes less than the threshold value. Then the testing data set is used to compare the model with actual system. A lower threshold value is used if the model does not represent the system Jang (1993).

The adaptive network in the proposed fuzzy inference system was feed forward type. The membership function was Sugeno type triangular shape membership function. There are three input membership function for each input and 27 learning rules. The learning weight was set to 1 with linear input / output values and for the defuzzification that is to produce crisp output of the consequent part weighted average scheme was used. For simplicity, we assume the fuzzy inference system under consideration has two inputs X and Y and one output Z.

For the first order Sugeno fuzzy model a typical rule with two fuzzy if-then rules can be expressed as: Rule 1: If (X is A1) and (Y is B1) then ( $Z_1 = p_1X + q_1Y + r_1$ ), Rule 2: If (X is A2) and (Y is B2) then ( $Z_2 = p_2X + q_2Y + r_2$ ), Where, X and Y are the inputs,  $A_i$  and  $B_i$  are the fuzzy sets,  $Z_i$  ( $i = 1,2$ ) are the output within the fuzzy region specified by the fuzzy rules,  $p_i$ ,  $q_i$  and  $r_i$  are the design parameters that are determined during the training process.

## XI. RESULTS AND DISCUSSION

In this experiment we can conduct 48 readings. The fuzzy inference system was trained up to 100 epochs. The experiment was conducted using Sugeno type fuzzy inference system with 27 rule, triangular membership function of each input parameter, hybrid of least square-gradient descent learning algorithm and linear output.

The analysis of experimental data was conducted on Mat Lab 7.6.0 workstation. The first four columns were treated as input and the last column as output data for the Fuzzy inference system (FIS). In which two tables are shown in

below. First table shows the experimental readings validates with regression analysis and second table shows the experimental readings validates with Adaptive Neuro Fuzzy Inference System.

**TABLE 1**

It shows the experimental readings validates with Regression Analysis. And in between error is shown in percentage.

Sr	Input Parameter				Predicted Erosion		
	Concentration	Time	Size	Angle	Erosion	Regression	Error
No	%	Min	µm	Degree	Experimental	Regression	%
1	10	60	45	15	0.057	0.070	22.01
2	10	60	45	30	0.090	0.058	35.91
3	10	60	45	45	0.065	0.052	20.46
4	10	60	45	60	0.050	0.048	4.33
5	10	60	45	75	0.042	0.045	7.23
6	10	60	45	90	0.032	0.043	33.98
7	10	90	45	15	0.065	0.081	24.31
8	10	90	45	30	0.095	0.067	29.45
9	10	90	45	45	0.075	0.060	19.91
10	10	90	45	60	0.065	0.056	14.49
11	10	90	45	75	0.057	0.052	8.19
12	10	90	45	90	0.040	0.050	24.53
13	10	60	65	15	0.06	0.080	32.74
14	10	60	65	30	0.082	0.066	19.44
15	10	60	65	45	0.068	0.059	12.93
16	10	60	65	60	0.055	0.055	0.4
17	10	60	65	75	0.045	0.052	14.61
18	10	60	65	90	0.04	0.049	22.74
19	10	90	65	15	0.064	0.093	44.58
20	10	90	65	30	0.098	0.077	21.69
21	10	90	65	45	0.074	0.069	7.04

1			0				
2			65				
2	10	90	0	60	0.068	0.064	6.4
2			65				
3	10	90	0	75	0.06	0.060	0.12
2			65				
4	10	90	0	90	0.054	0.057	5.63
2			45				
5	30	60	5	15	0.088	0.110	25.37
2			45				
6	30	60	5	30	0.15	0.091	39
2			45				
7	30	60	5	45	0.107	0.082	23.35
2			45				
8	30	60	5	60	0.09	0.076	15.68
2			45				
9	30	60	5	75	0.062	0.071	15.23
3			45				
0	30	60	5	90	0.05	0.068	36.02
3			45				
1	30	90	5	15	0.098	0.128	30.8
3			45				
2	30	90	5	30	0.131	0.106	18.84
3			45				
3	30	90	5	45	0.098	0.095	2.77
3			45				
4	30	90	5	60	0.081	0.088	8.84
3			45				
5	30	90	5	75	0.068	0.083	22.07
3			45				
6	30	90	5	90	0.06	0.079	31.7
3			65				
7	30	60	0	15	0.098	0.126	28.92
3			65				
8	30	60	0	30	0.132	0.105	20.62
3			65				
9	30	60	0	45	0.102	0.094	7.92
4			65				
0	30	60	0	60	0.092	0.087	5.54
4			65				
1	30	60	0	75	0.084	0.082	2.6
4			65				
2	30	60	0	90	0.07	0.078	11.26
4			65				
3	30	90	0	15	0.122	0.147	20.32
4			65				
4	30	90	0	30	0.168	0.122	27.53
4			65				
5	30	90	0	45	0.144	0.109	24.22
4			65				
6	30	90	0	60	0.116	0.101	12.96
4			65				
7	30	90	0	75	0.098	0.095	3
4			65				
8	30	90	0	90	0.08	0.090	13.11
Average							18.34 7



TABLE 2

IT SHOWS THE EXPERIMENTAL READINGS VALIDATES WITH ADAPTIVE NEURO FUZZY INFERENCE SYSTEM (ANFIS) AND IN BETWEEN ERROR SHOWN IN PERCENTAGE.

Sr	Input Parameter				Predicted Erosion			
	Concentration	Time	Size	Angle	Erosion		Error	
No	%	Min	µm	Degree	Experimental	ANFIS	%	
1	10	60	45	75	0.042	0.044	5.71	
2	10	90	45	45	0.075	0.082	9.33	
3	10	60	65	90	0.04	0.035	12	
4	10	90	65	30	0.098	0.088	10	
5	30	60	45	75	0.062	0.064	3.8	
6	30	90	45	60	0.081	0.084	4.19	
7	30	60	65	90	0.07	0.065	7.14	
8	30	90	65	60	0.116	0.119	2.58	
Average							6.843	7

**CONCLUSIONS**

Arrangements in the Jet Erosion Tester have been made to evaluate the effect of impact angle, concentration, velocity, particle size etc. on erosion wear. This design of jet erosion tester is intend to conduct wear tests at moderate solid concentrations and actual flow of velocities to simulate the wear conditions for pipeline, pipe and fittings, bend, pump etc. and may provide more realistic results. Finally, the contributions of jet erosion wear is find out in the form of weight loss for the series of different impact angles such as 150,300,450,600,750,900at 10% solid concentration of sand for 60 minutes for Ductile materials.

For different angular positions, it is observe that the weight loss decreases as the angle is increases from 30° to 900 the maximum weight loss is occur at 300 angle of impact and it decreases as angle is increases from 300 to 900respectively.

By completing the validation with ANFIS and Regression analysis, it is observed that the erosion wear of copper material by ANFIS with triangular membership function is 93.156%, with the error of 6.843% and by using Regression analysis is 81.65%, with the error of 18.34%.

**REFERANCES**

- [1] Desale, G. R., Gandhi, B. K., and Jain, S. C., 2005B, "Improvement in the design of a pot tester to simulate erosion wear due to solid-liquid mixture", *Wear*, Vol. 259, pp. 196-202.
- [2] Gandhi, B. K., Singh, S. N., and Seshadri, V., 1999, "Study of the parametric dependence of erosion wear for the parallel flow of solid-liquid mixtures", *Tribology International*, Vol. 32, pp. 275-282.
- [3] Desale, G. R., Gandhi, B. K., and Jain, S. C., 2005A, "Effect of physical properties of solid particle on erosion wear of ductile materials", *Proceedings of 2005 ASME World Tribology Congress III*, Washington, D.C., USA.
- [4] Purushottam S. Desale and Ramchandra S. Jahagirdar., 2013, "Modelling the effect of variable work piece hardness on surface roughness in an end milling using multiple regression and adaptive Neuro fuzzy inference system" *International Journal of Industrial Engineering Computations* 5 (2014) 265–272.
- [5] Abbade, N. P., and Crnkovic, S. J., 2000, "Sand-water slurry erosion of API 5L X65 pipe steel as quenched fromintercritical temperature", *Tribology International*, Vol. 33, pp. 811-816.
- [6] Bree, S. E. M. de, Rosenbrand, W. F., and Gee, A. W. J. de, 1982, "On the erosion resistance in water-sand mixtures of steels for application in slurry pipelines", *Hydrotransport 8*, BHRA Fluid Engineering, Johannesburg, (S.A.), Paper C3.
- [7] Bitter, J. G. A. 1963A and B, "A study of erosion phenomena Part I", *Wear*, Vol. 6, pp. 5-21, pp. 169-190.
- [8] Desale G.R., Paul C.P., Gandhi B.K., Jain S.C., 2009, " Erosion wear behavior of laser clad surfaces of low carbon austenitic steel", *Wear*, Vol. 266, pp. 975-987.
- [9] Desale G.R., Paul C.P., Gandhi B. K., 2009, "Slurry Erosion Wear Properties of Laser Cladding", *Kiran*, Vol. 20., No. 2., pp.26-33.
- [10] Jang, J. S. (1993). ANFIS: adaptive-network-based fuzzy inference system. *Systems, Man and Cybernetics*, *IEEE Transactions on*, 23(3), 665-685.
- [11] Al-Bukhaiti. M.A., S.M. Ahmedb, F.M.F. Badran b, K.M. Emarab, 2007, "Effect of impingement angle on slurry erosion behaviour and mechanisms of 1017 steel and high-chromium white cast iron", *Wear*, Vol. 262, pp. 1187- 1198.
- [12] Lin, F. Y., and Shao, H. S., 1991A, "Effect of impact velocity on slurry erosion and a new design of a slurry erosion tester", *Wear*, Vol. 143, pp. 231-240.
- [13] Lin, F. Y., and Shao, H., 1991B, "The effect of impingement angle on slurry erosion", *Wear*, Vol. 141, pp. 279-289.
- [14] Levy, A. V., 1981, "The solid particle erosion behaviour of steel as a function of microstructure", *Wear*, Vol. 68, pp. 269-287.
- [15] Gupta, R., Singh, S. N. and Seshadri, V. 1995, "Prediction of uneven wear in a slurry pipeline on the basis of measurements in a pot tester", *Wear*, Vol. 184, pp. 169-178.