Evaluation of Tribological Properties and Optimization of Electroless Ni-P-W Coating under Dry Condition using Grey Fuzzy Analysis

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Keywords:
- Electroless nickel
- Ni – P – W coating
- Coefficient of Friction
- Wear Depth
- Grey – Fuzzy Logic

ABSTRACT

Electroless Ni-P-W coating has enhanced tribological properties compared to the conventional binary alloys. Hence, electroless Ni-W-P, which is a ternary alloy, is synthesized on mild steel substrates. The coatings are heat treated at 400 °C for 1 hr to obtain a high hardness. Their tribological behavior is investigated on a pin–on–disc tribometer under dry condition. Three test parameters which include the applied normal load (L), sliding speed (S) and sliding time (T) are varied and their effects on wear depth and coefficient of friction are analyzed. An attempt is made to ascertain the optimum combination of the parameters that would minimize the coefficient of friction and wear depth of the deposits simultaneously using Grey Fuzzy logic. Optimum tribological behavior of the deposits is obtained at an applied normal load of 10 N, sliding speed of 80 rpm and sliding duration of 5 min. Load and time are seen to be the most significant factors in influencing the friction and wear of the deposits. The variation of coefficient of friction and wear depth considering the combined effect of the aforesaid parameters is analyzed from three dimensional surface plots. The wear mechanism is found to be predominantly abrasive in nature.

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1. INTRODUCTION

One of the major concerns associated with any industry is the reduction of friction and wear of mating components, which would otherwise limit their use and render them useless after a certain period. To enhance the tribological properties of interacting surfaces and consequently improve their performance capabilities, hard coatings are applied. Brenner and Riddell [1], in the year 1946 brought about a revolution in the field of surface coating technology by discovering the electroless method of deposition of coating. With the passage of time, the electroless deposition technique evolved into a mature subject of research and thereby gaining popularity and becoming a substitute to hard chromium plating [2]. Though electroplating is a more straightforward process, electroless has several advantages over the conventional method of plating due to the high hardness, wear
Electroless Ni-P [5-7] and Ni-B [8-12] coatings possess excellent physical, mechanical and tribological properties and finds wide industrial usage. The wear resistance of electroless nickel (EN) coatings shows a remarkable improvement on addition of hard particles such as diamond, Al2O3 and TiO2 [13-15]. The addition of a third element to the binary Ni-P and Ni-B coating leads to the formation of a ternary alloy such as Ni-P-W and Ni-P-Cu [16-23], which again leads to an enhanced wear resistance. Softer particles incorporated with EN coatings such as PTFE, MoS2, graphite and CNTs [24-27] promote self lubricity and finds their applications in the reduction of friction. The first instance of electroless Ni-P-W coating was reported by Pearlstein and Weightman [28] in 1963 and many investigations were reported henceforth. Several research works has been carried out by Balaraju and Rajam [29] as well as Balaraju et al. [30-33] to investigate the deposition mechanism, surface morphology, microstructure, phase transformation and corrosion behavior. Electroless Ni-P-W coatings exhibit enhanced tribological behavior. The wear resistance of electroless nickel coating depends largely on its hardness value. The hardness of Ni-P-W coating increases with the increase in tungsten content in the deposit [16]. The same depends invariably on the heat treatment temperature of the coatings and a maximum value of hardness was observed at 500°C for 1 hour. This happens due to the precipitation of phosphides and solid solution strengthening of the nickel matrix due to dissolved tungsten [17]. The wear resistance also depends on the tribological testing conditions such as the applied normal load, test speed, sliding time as well as test environment [18-21]. With the increase in applied normal load, the specific wear rate of heat treated Ni-P-W coating increases though it is lower than Ni-P deposits or Ni-P-W coating in its as-plated condition [16]. The optimum combination of tribo-testing parameters for minimum friction and wear of Ni-P-W coating using Grey-Taguchi method and WPCA was reported by Roy and Sahoo [18,19]. Laser treatment increases the wear resistance of electroless Ni-P-W coating further, than obtained by furnace annealing [34]. The wear mechanism of the deposits is observed to be adhesive and in some cases to be abrasive depending on the test conditions and sliding environment [16,18-21,34].

Exhaustive review of the literature reveals that most of the studies related to electroless Ni-P-W coating are directed mainly towards effect of heat/mechanical treatments of the deposits on its tribological properties, corrosion resistance and coating characterization under different deposition conditions [16-21, 29-34]. Due to the broad industrial usage of EN coatings in industries, their investigation and optimization is gaining immense importance [35-38]. The use of experimental design coupled with sophisticated optimization techniques for the investigation as well as minimization of friction and wear of electroless Ni-P-W coating can be explored further. Design of experiments is a powerful statistical technique that can be used to determine the complex relationship that exist between the tribo – testing parameters and the tribological responses (friction and wear) of Ni-P-W coatings by varying the former in a pre – planned and organized manner such that their effects on the latter can be obtained using the least number of trials. Hence, the present study is directed towards the same. Electroless Ni-P-W coating is deposited on AISI 1040 steel substrates. Taguchi’s orthogonal array (OA) along with a hybrid grey – fuzzy reasoning analysis has been employed to obtain the optimum combination of tribo – testing parameters (applied normal load, sliding speed and sliding time) for minimum friction and wear of Ni-P-W coating simultaneously. The grey fuzzy logic has proven capabilities in solving multi – objective optimization problems [9,10]. Moreover, the fuzzy logic reduces the uncertainties introduced into the grey relational grade and obtains a better system performance containing multiple variables [39,40]. Therefore, this method has been adopted in the present study since it is a case of multi – response optimization involving a complex relationship.
between the tribo–testing parameters and coefficient of friction and wear depth of the deposits. Analysis of variance (ANOVA) is carried out on the multiple performance index which is the grey fuzzy grade to assess the contribution of each of the process parameters and their interactions in controlling the friction and wear of Ni-P-W coating. Coating characterization is done using energy dispersive X-ray analyzer (EDX), X-Ray diffraction analyzer (XRD) and scanning electron microscopy (SEM). Three dimensional surface and contour plots are generated to analyze the trends in variation of coefficient of friction (COF) and wear depth with the process parameters. An attempt is made to relate interaction effect of the process parameters as well as the coating characteristics with the tribological performance of the deposits under dry sliding condition. The synergistic effects of the applied normal load, sliding speed, sliding duration and the coating morphology on the tribological behavior of the coatings would help in assessing their viability for the reduction of friction and wear of tribological contact applications working under dry condition. Finally, worn out Ni-P-W coated specimen is observed under SEM to determine the prevalent wear mechanism.

2. THE INTEGRATED GREY – FUZZY LOGIC

The grey relational analysis was first introduced by Deng [41] in 1982 as a part of the grey system. The grey system can deal with uncertain and imprecise information. In the grey system, white means complete information while black means no information. The grey data resembles information which lies in-between the two. Since the interrelationships between the tribo–testing parameters and the COF and wear depth of Ni-P-W coating is a complex multivariate system; it can be supposed to be a grey system. The grey relational analysis is an approximate measure of correlation between sequences and is very effective in analyzing relationships between sequences with less data [8,18,40]. The capability of grey relational analysis can be further improved by the use of fuzzy reasoning technique which is an artificial intelligence tool efficient in dealing with data represented in terms of “linguistic variables”. The fuzzy logic approach to decision making was introduced by Zadeh [42]. In the conventional theory of sets, an object can have membership value 0 or 1. But in the fuzzy set theory, an object is mapped onto the unit interval. A fuzzy set contains numerous membership functions which assign membership values to the objects between 0 and 1 [9,10,39,40]. The uncertainties which are introduced into the grey relational grade due to higher–the–better, lower–the–better or nominal–the–best characteristics of multiple responses can be handled by fuzzy logic approach. This integrated approach to solving complex optimization problems is the proposed grey fuzzy logic, and has been used in the present study. A fuzzy system involves fuzzification of the input variables, inference operation which invokes a certain number of rules in the expert system, fuzzy output and finally defuzzification of the output to obtain a crisp value. The input and output space consists of several membership functions which can be triangular, trapezoidal, Gaussian, etc. [43], the choice of which depends on the complexity of the problem in hand. The inference engine consists of a set of pre–defined IF (antecedent) – THEN (consequent) rule base derived from expert knowledge. After a set of fuzzy inputs is received, a certain number of rules are invoked and a set of fuzzy outputs is obtained using Mamdani’s max–min implication. The defuzzification can be done by any of the popular known methods such as max–membership method, centroid method, weighted average method, mean-max membership method, etc [43]. The centroid method is used in this study to obtain the multiple performance index which is referred here as the grey fuzzy grade (GFG). The final algorithm of grey fuzzy logic is summarised as follows:

**Normalization of sequences between 0 and 1:** This is done to account for the variation in range and units of the sequences. The sequences are normalized based on their characteristics. For “higher–the–better” normalization characteristics the formula is:

$$x_i^*(k) = \frac{x_i(k) - \min x_i(k)}{\max x_i(k) - \min x_i(k)}$$ (1)

For “lower–the–better” characteristics, the formula is:

$$x_i^*(k) = \frac{\max x_i(k) - x_i(k)}{\max x_i(k) - \min x_i(k)}$$ (2)
For "nominal-the-best" characteristics, the formula is:

$$x_i^*(k) = 1 - \frac{|x_i(k) - x_k^*(i)|}{\max x_i(k) - \min x_i(k)}$$  \hspace{1cm} (3)

where \(\min x_i(k)\) and \(\max x_i(k)\) are respectively the smallest and largest values of \(x_i(k)\) for the \(k^{th}\) response. \(x_i(k)\) is the target of \(x_i(k)\).

**Calculation of grey relational coefficient (GRC) of the sequences**: The GRCs \((\xi)\) compare the relationship between the ideal \((\text{best} = 1)\) and the experimental sequences. They are calculated as follows:

$$\xi_i(k) = \frac{\Delta_{\min} + r \Delta_{\max}}{\Delta_{\max}}$$  \hspace{1cm} (4)

where \(\Delta_{\theta} = |x_i(k) - x_i^*(k)|\) = difference of the absolute value between \(x_i(k)\) and \(x_i^*(k)\), \(\Delta_{\min}\) and \(\Delta_{\max}\) are respectively the minimum and maximum values of the absolute differences \(\Delta_{\theta}\) of all comparing sequences. In Eqn. (4) the term \(r\) is the distinguishing coefficient which is used to adjust the difference of the relational coefficient, usually \(r \in \{0, 1\}\) [41]. The value of \(r\) controls the effect of \(\Delta_{\max}\). Lower the value of \(r\), higher is the distinguishing ability. The value of \(r\) for the present analysis is taken to be 0.5 [9,10]. In this study, the grey relational coefficients for wear depth \((\xi_1)\) and COF \((\xi_2)\) are calculated.

**Fuzzification process**: Fuzzification of the GRCs of the sequences, application of the IF – THEN rules and obtaining fuzzy outputs using max – min implication.

**Defuzzification process**: Defuzzification to obtain a crisp value of the GFG. The centroid method of defuzzification is employed in the present work.

**Optimization**: Obtain optimal setting of tribo – testing parameters from main effects plot.

### 3. EXPERIMENTAL DETAILS

#### 3.1 Deposition of electroless Ni-P-W coating

AISI 1040 steel specimen of dimension Ø6 × 30 mm is used as the substrate for the deposition of coating. The specimens are selected very carefully and the ones having average surface roughness values within a variation of ±1 % are chosen since EN coatings follow the substrate roughness. Prior to deposition, the substrates are freed from any corrosion products or surface layers which may have formed, by wiping in de-ionized water and acid pickling in dilute (50 %) HCl following which methanol cleaning is done. Finally, the specimens are again rinsed in de-ionized water. For the deposition of coating, the chemicals given in Table 1 are mixed appropriately and in proper sequence. The bath composition is decided after a lot of trials and literature review and is kept constant so that all the specimens are uniformly coated [20,21]. The role of each chemical and operating conditions are also enlisted in Table 1.

A high precision weighing balance (Afcoset, India, Model ER182A, Maximum Range 180 gm, VACC 0.01 gm, Class-II) is used for the measurement of chemicals to ensure accurate composition of the coating bath. The substrates are activated by dipping in palladium chloride (of about 0.01 % strength) kept at 55 °C for a few seconds and finally submerged into the electroless bath. Activation is done to kick – start the deposition process, obtain better adhesion of the deposits and obtain a significant deposit thickness [9, 10]. After 3 hours of deposition, the coated specimens are rinsed in de-ionized water and heat treated in a box furnace to 400 °C for 1 hour. Since heat treatment significantly affects the hardness and consequently the wear resistance of EN coatings, this step has been incorporated in the present investigation. Microhardness of the deposits is measured in their as-deposited condition and after heat treatment using a UHL Vicker's microhardness tester (VMHTOT, Technische Mikroskopie) at an indentation load of 500 gf, 15 s dwell time and an indentation speed of 25 µm/s. An average of six microhardness values is recorded.
3.2 Friction and wear tests

The friction and wear tests are carried out on a pin-on-disc arrangement tribotester (TR-20LE-CHM-400, Ducom, India) under non lubricated condition at ambient temperature by varying the applied normal load (N), sliding speed (rpm) and sliding time (min) at three different levels as given in Table 2. The responses viz. COF and wear depth (µm) are recorded in this study. The track diameter is kept fixed at 80 mm and the coated specimen pin of dimension Ø6 × 30 mm is held stationary against a rotating counterface disc of EN 31 specification having a hardness value of 60 HRC. The aforesaid material is chosen to ensure that wear of only the coated pin takes place and not on the counterface disc. The frictional force and wear depth are recorded by a computer attached to the tribo-tester and are displayed on the monitor in real time. Wear is recorded in terms of displacement (µm) by a linear variable differential transformer (Make: Syscon) and the frictional force is measured by a load cell (Make: IPA) from which the values of COF are obtained.

Table 2. Process parameters and their levels.

<table>
<thead>
<tr>
<th>Design factors</th>
<th>Unit</th>
<th>Levels</th>
</tr>
</thead>
<tbody>
<tr>
<td>Load (L)</td>
<td>N</td>
<td>10 20* 30</td>
</tr>
<tr>
<td>Speed (S)</td>
<td>rpm</td>
<td>60 80* 100</td>
</tr>
<tr>
<td>Time (T)</td>
<td>min</td>
<td>5 10* 15</td>
</tr>
</tbody>
</table>

*initial condition

To perform the experiments, the combinations given in Taguchi’s L27 OA (26 degrees of freedom) are utilized since the degrees of freedom for the present experimental study is seen to be 18.

3.3 Coating characterization

Coating characterization is very important to ensure the deposit uniformity of the coatings and identify the morphological and microstructural changes occurring due to heat treatment, which in turn influences the tribological behavior of the deposits to a large extent. Composition study in terms of weight percentages of nickel, phosphorus and tungsten is done using energy dispersive X-ray (EDX) analyzer (FEI, FEG Quanta 250) for as-deposited and heat treated coatings. The phase transformation analysis is done using an X-ray diffraction (XRD) analyzer (Rigaku, Miniflex) using Cu Kα radiation at a scan rate of 1°/min in as-deposited and heat treated condition to identify the compounds before and after heat treatment. Surface morphology of as-deposited as well as heat treated electroless Ni-P-W coating is done using scanning electron microscope (SEM) (JEOL, JSM 6360). SEM of a cross-cut section of heat treated Ni-P-W coating is also done to determine the coating thickness and compared with the highest value of wear depth that a specimen undergoes during the tribological tests. Finally, a worn out Ni-P-W coated surface is observed under SEM to determine the mechanism responsible for the wear of the deposits.

4. RESULTS AND DISCUSSION

4.1 Grey fuzzy reasoning

The values of wear depth and COF as obtained from the tribological tests are shown in Table 3. These values are further used to obtain the multi performance index representing both COF and wear depth which is the GFG in this study. The data obtained for wear depth and COF is first normalized based on lower–the–better criteria (Eqn. 2) since both of them are undesirable and needs to be minimized. From the normalized values, the GRCs for wear depth and COF are calculated using Eqn. 4.

The GRCs for wear depth and COF are fuzzified using triangular membership functions. Five fuzzy subsets, namely very small (VS), small (S), medium (M), large (L) and very large (VL) represent the GRCs for wear depth and COF (Fig. 1(a)). The output space is divided into nine fuzzy subsets i.e. tiny (T), very small (VS), small (S), small medium (SM), medium (M), medium large (M), large (L), very large (VL) and high (H) (Fig. 1(b)).

To activate the fuzzy inference system, 25 rules are written which is obtained directly based on the fact that higher the value of GRC, better is the system response. The GFG for each of the 27 experiments is obtained by the centroid defuzzification method and they are enlisted in Table 3. Again, a higher value of GFG indicates a better multi performance index. The fuzzy reasoning analysis is done using Matlab 7 (Fuzzy logic toolbox).
Table 3. Experimental data and calculation of GRC and GFG.

<table>
<thead>
<tr>
<th>Exp. No.</th>
<th>Experimental Data</th>
<th>GRC</th>
<th>GFG</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Wear (µm)</td>
<td>COF</td>
<td>Wear</td>
</tr>
<tr>
<td>1</td>
<td>5.031</td>
<td>0.1659</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>7.887</td>
<td>0.2221</td>
<td>0.732</td>
</tr>
<tr>
<td>3</td>
<td>10.196</td>
<td>0.2673</td>
<td>0.601</td>
</tr>
<tr>
<td>4</td>
<td>7.849</td>
<td>0.0804</td>
<td>0.734</td>
</tr>
<tr>
<td>5</td>
<td>9.486</td>
<td>0.0894</td>
<td>0.636</td>
</tr>
<tr>
<td>6</td>
<td>11.203</td>
<td>0.1147</td>
<td>0.558</td>
</tr>
<tr>
<td>7</td>
<td>12.649</td>
<td>0.0329</td>
<td>0.461</td>
</tr>
<tr>
<td>8</td>
<td>14.14</td>
<td>0.0355</td>
<td>0.601</td>
</tr>
<tr>
<td>9</td>
<td>15.407</td>
<td>0.0494</td>
<td>0.429</td>
</tr>
<tr>
<td>10</td>
<td>10.283</td>
<td>0.1353</td>
<td>0.597</td>
</tr>
<tr>
<td>11</td>
<td>11.134</td>
<td>0.1598</td>
<td>0.561</td>
</tr>
<tr>
<td>12</td>
<td>12.239</td>
<td>0.1653</td>
<td>0.519</td>
</tr>
<tr>
<td>13</td>
<td>9.55</td>
<td>0.1127</td>
<td>0.633</td>
</tr>
<tr>
<td>14</td>
<td>11.509</td>
<td>0.1236</td>
<td>0.546</td>
</tr>
<tr>
<td>15</td>
<td>12.866</td>
<td>0.1366</td>
<td>0.498</td>
</tr>
<tr>
<td>16</td>
<td>13.481</td>
<td>0.1179</td>
<td>0.48</td>
</tr>
<tr>
<td>17</td>
<td>16.816</td>
<td>0.1532</td>
<td>0.398</td>
</tr>
<tr>
<td>18</td>
<td>18.946</td>
<td>0.1655</td>
<td>0.359</td>
</tr>
<tr>
<td>19</td>
<td>10.979</td>
<td>0.2889</td>
<td>0.567</td>
</tr>
<tr>
<td>20</td>
<td>13.1</td>
<td>0.3045</td>
<td>0.491</td>
</tr>
<tr>
<td>21</td>
<td>14.088</td>
<td>0.3172</td>
<td>0.462</td>
</tr>
<tr>
<td>22</td>
<td>9.825</td>
<td>0.1373</td>
<td>0.619</td>
</tr>
<tr>
<td>23</td>
<td>13.718</td>
<td>0.1497</td>
<td>0.473</td>
</tr>
<tr>
<td>24</td>
<td>16.21</td>
<td>0.1559</td>
<td>0.411</td>
</tr>
<tr>
<td>25</td>
<td>17.318</td>
<td>0.1668</td>
<td>0.388</td>
</tr>
<tr>
<td>26</td>
<td>19.423</td>
<td>0.173</td>
<td>0.351</td>
</tr>
<tr>
<td>27</td>
<td>20.602</td>
<td>0.1755</td>
<td>0.333</td>
</tr>
</tbody>
</table>

Since the experimental design considered is orthogonal, it is very easy to obtain the mean value of GFG for a particular parameter at any given level. The mean GFG for the applied normal load, sliding speed and sliding time at their designated levels is laid down in Table 4 which is also called the response table. The mean GFG is also calculated. The table also contains ranks based on delta values. A delta value for each of the parameters is obtained by subtracting the lowest value from the highest one in each column. This value indicates the contribution of a process parameter in controlling the response parameters. It is reflected from the results that load has the highest contribution followed by time and speed. The corresponding main effects plot is shown in Fig. 2. From the response table (Table 4) and main effects plot for GFG (Fig. 2), the optimum combination of parameters is seen to be L1S2T1 i.e. lowest value of applied normal load, mid value of sliding speed and lowest value of sliding time.

![Fig. 1. Fuzzy subsets for (a) GRCs of wear depth and COF and (b) GFG.](image)

![Fig. 2. Main effects plot for GFG.](image)
4.2 Confirmation test

A confirmation test is carried out to validate the result obtained for the optimization of wear depth and COF. The values of GFG obtained at the initial setting of the tribometer-testing parameters i.e. L2S2T2 (mid value of the process parameters) and the one obtained at the optimum setting of the parameters (L1S2T1) are compared with a predicted value of GFG. The GFG at the optimal setting of the process parameters can be predicted using the following formula:

\[
\hat{\eta} = \eta_m + \sum_{i=1}^{a} (\bar{\eta}_i - \eta_m)
\]

where \(\hat{\eta}\) is the predicted value of GFG, \(\eta_m\) is the mean value of GFG for all the 27 experiments, \(\bar{\eta}_i\) is the mean GFG at the optimal level for each of the process parameters, and \(o\) is the number of main design parameters that significantly affect the wear and friction of electroless Ni-P-W coating.

Table 5. Result of validation test.

<table>
<thead>
<tr>
<th>Source</th>
<th>Initial Parameter</th>
<th>Optimal Parameter</th>
<th>Predicted</th>
<th>Experimental</th>
</tr>
</thead>
<tbody>
<tr>
<td>Level</td>
<td>L2S2T2</td>
<td>L1S2T1</td>
<td>L1S2T1</td>
<td></td>
</tr>
<tr>
<td>Wear (µm)</td>
<td>11.509</td>
<td>7.849</td>
<td></td>
<td></td>
</tr>
<tr>
<td>COF</td>
<td>0.124</td>
<td>0.080</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Grade</td>
<td>0.588</td>
<td>0.76</td>
<td>0.739</td>
<td></td>
</tr>
<tr>
<td>Improvement of GFG = 0.151 (25.68 %)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

The validation run is carried out to determine if there is any improvement in GFG at the optimal condition compared to the initial setting of the design variables. Results for the confirmation test are shown in Table 5. It is seen that the predicted and experimental GFG are in close agreement. There has been an improvement of 25.68 % in the GFG at the optimum condition compared to the initial test levels. The wear depth and COF are seen to have reduced by a considerable amount. Hence, the optimal parameter settings can be justified as L1S2T1 (applied normal load of 10 N, sliding speed 80 rpm and sliding duration 5 min).

4.3 Analysis of variance (ANOVA)

ANOVA, an important statistical technique helpful in revealing the influence of process parameters and their interactions in controlling the design variables, is carried out on the GFG. It separates the total variability of the responses into contributions of each of the factors and error [5,8]. The significance of a factor is determined by the F – ratio or the variance ratio. The F – ratio is the ratio of regression mean square to the mean square error. At a particular confidence level, if the obtained F – ratio for a factor is higher than the tabulated value, then it is considered to be significant at the same confidence level. The results of ANOVA performed on the GFG are tabulated in Table 6. It can be seen that at a confidence level of 97.5 % the applied normal load has the highest contribution in controlling the friction and wear of electroless Ni-P-W coating followed by sliding time and sliding speed. The interaction of load and speed is also seen to influence the response variables to some extent. The percentage contribution of each of the factors along with their interactions to the process is also enlisted in Table 6. Applied normal load is again seen to have the highest contribution of 57.67 % followed by sliding duration (19.25 %). The interaction of load and speed is also seen to be statistically significant.

A check of the deviation of data from normality is also done and the plot for the standardized residuals is shown in Fig. 3. It can be seen that the residuals fall on a straight-line indicating that the errors are normally distributed. This validates the normality assumption and the independence of data and the adequacy of the ANOVA model is also tested. Thus, the proposed model can be considered to be adequate. In the present study, ANOVA results and residuals plot for GFG are obtained using Minitab software [44].

Table 6. Results of ANOVA.

<table>
<thead>
<tr>
<th>Source</th>
<th>DF</th>
<th>SS</th>
<th>MS</th>
<th>F - Ratio</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>L</td>
<td>2</td>
<td>0.1741</td>
<td>0.0870</td>
<td>75.15*</td>
<td>57.67</td>
</tr>
<tr>
<td>S</td>
<td>2</td>
<td>0.0205</td>
<td>0.0103</td>
<td>8.87*</td>
<td>6.81</td>
</tr>
<tr>
<td>T</td>
<td>2</td>
<td>0.0581</td>
<td>0.0291</td>
<td>25.08*</td>
<td>19.25</td>
</tr>
<tr>
<td>L*S</td>
<td>4</td>
<td>0.0294</td>
<td>0.0074</td>
<td>6.35*</td>
<td>9.74</td>
</tr>
<tr>
<td>L*T</td>
<td>4</td>
<td>0.0074</td>
<td>0.0018</td>
<td>1.59</td>
<td>2.44</td>
</tr>
<tr>
<td>S*T</td>
<td>4</td>
<td>0.0031</td>
<td>0.0008</td>
<td>0.67</td>
<td>1.03</td>
</tr>
<tr>
<td>Error</td>
<td>8</td>
<td>0.0093</td>
<td>0.0012</td>
<td>3.07</td>
<td></td>
</tr>
<tr>
<td>Total</td>
<td>26</td>
<td>0.3019</td>
<td></td>
<td>100</td>
<td></td>
</tr>
</tbody>
</table>

*Significant at 97.5 % confidence level
4.4 Composition, phase transformation, surface morphology and microhardness

From EDX analysis, it is observed that as deposited Ni-P-W coated specimen is composed of 82.71 % Ni, 13.56 % P and 3.73 % W by weight. The weight percentage of phosphorus in the coatings indicates that the deposits fall in the high phosphorus category and the deposits are expected to be amorphous.

SEM micrographs of some coated specimens are taken in random in as-deposited and heat treated condition to analyze the effect of heat treatment on the surface morphology. Similar qualitative change is observed in all the specimens. SEM micrographs of an as-deposited and a heat treated sample are shown in Fig. 5(a) and Fig. 5(b) respectively. Nodulated structures can be observed on the coated surface in the SEM image of as-deposited coating (Fig. 5(a)). No visible surface damage or porosity can be observed and appears to be dense. The coating seems to become more compact on heat treatment (Fig. 5(b)). On heat treating the deposits at 400 °C, crystallization occurs, leading to the precipitation of Ni₃P out of the supersaturated solid solution of Ni as matrix containing P and W. Thus, volume contraction and a compaction of the EN matrix are observed. The globular surface morphology of the deposits is responsible for the inherent self lubricating nature of EN coatings and low COF [2,3].

XRD results reveal that as-deposited Ni-P-W coating is amorphous/nanocrystalline in nature (Fig. 4). The incorporation of tungsten improves the crystallinity of the coatings [33]. Hence a broad peak superimposed by a sharp peak of Ni is observed in the XRD plot in Fig. 4. It becomes crystalline on heat treatment due to the precipitation of nickel and its phosphides (Fig. 4). On heat treating the deposits in a furnace at 400 °C for 1 hr, the intensity of the Ni peak increases and short peaks of stable Ni₃P is visible in Fig. 4. This is the indication of the crystalline and microcrystalline nature of the coatings on heat treatment. The crystallization temperature for binary Ni-P alloy system is 335 °C. Due to the incorporation of W in the alloy, the phase transformation temperature is raised with the precipitation of Ni₃P and crystallization of Ni [17]. The phase transformation behavior observed in the present work agrees well with the theories proposed by other researchers [17,33]. A high value of hardness and consequently high wear resistance is observed in heat treated Ni-P-W coated specimens [16-22] due to the precipitation of phosphides and solid solution strengthening by Ni (W) on heat treatment.
A cross-cut section of Ni-P-W coating has been shown in Fig. 5(c). It is evident that the coating thickness is around 30 µm and the thickness is uniform. Since the bath composition is kept constant, the deposit thickness of all the coated specimens can be assumed to have nearly the same value. The highest value of wear encountered by a pin is seen to be 20.602 µm, which is less than the deposit thickness. Hence, it can be inferred that wear of the coatings only has taken place during the tribological tests and not the substrate material.

Microhardness of the as-deposited coatings is seen to be 678HV0.5 while it increases to 994HV0.5 on heat treatment. Microhardness of EN coatings has a direct relationship with wear resistance. The coatings expectedly exhibit enhanced microhardness on heat treatment due to crystallization and phosphides precipitation. Though, heat treatment to a very high temperature leads to deterioration of the tribological behavior of the coatings due to grain coarsening and softening of the deposits [16].

4.5 Analysis of wear mechanism

The wear mechanism is also investigated using SEM (Fig. 6). In the surface morphology of a worn surface, micro – cutting and ploughing effects are clearly visible. No pits or prows and detachment of the coating from the substrate are observed. Parallel grooves along the direction of sliding with high degree of plasticity can be seen which is indicative of ductile failure of the coatings and wear debris getting displaced to the sides of the wear track [8,12]. From the results, it can be assimilated that abrasive wear mechanism is predominant.

4.6 Effect of process parameters on response variables

The interaction effects of process parameters i.e. applied normal load, sliding speed and sliding time on the wear depth and COF are investigated by generating three dimensional surface and
contour plots in Matlab 7. In each of the plots, one parameter is kept constant at its mid level value and the variation is plotted against the other two parameters. The plots for wear depth and COF are depicted in Fig. 7 and Fig. 8 respectively.

**Effect of process parameters on wear depth:** The effects of load, speed and time on wear depth are shown in Fig. 7. The plots reveal that the wear depth increases with the increase in load and speed (Fig. 7(a)). The specific wear rate of two mating surfaces depends on the actual contact area of the asperities. With the increase in applied normal load, the asperities on the coated surface and the counterface material advance more towards each other increasing the actual area of contact, thereby increasing the volume of material sheared. Hence, an increase in wear depth is observed. The wear depth is also seen to increase with the load time interaction (Fig. 7(b)). For a constant track diameter, with the increase in sliding speed or the sliding time, the sliding distance increases. This results in an increased wear of the coatings. The increase in wear depth with speed-time interaction (Fig. 7(c)) can also be attributed to the aforesaid reasons. The results obtained for the variation of wear depth with the lower ranges of tribo – testing parameters considered in the present study is in agreement with the ones obtained in previous studies [16] and is also in accordance with the relation given by Archard for the wear volume encountered by a material subjected to abrasive wear condition. Curvatures in the plot for wear depth vs. load, speed (Fig. 7(a)) and wear depth vs. speed, time (Fig. 7(c)) are seen indicating a low value of wear encountered by the coated specimen at mid – value of sliding speed. It can also be deduced from the Fig. 7(a), Fig. 7(b) and Fig. 7(c) that a lower wear of the deposits occurs at lower values of applied normal load and sliding time. Thus, the optimum combination of parameters for enhanced wear behavior i.e. L1S2T1 is justified. Low wear is suffered by the coatings on account of the high hardness obtained on heat treatment due to transformation of phase from amorphous to crystalline Ni and its phosphides (Fig. 4).

**Effect of process parameters on COF:** The COF is seen to be increasing with load and speed as well as load and time in Fig. 8(a) and Fig. 8(b) respectively. But with the increase in speed and time, the COF is seen to decrease in Fig. 8(c). This can be attributed to the fact that EN coatings are self lubricating in nature due to their nodulated structure which is also visible in the SEM image of the as – plated and heat treated coatings (Fig. 5(a) and Fig. 5(b)).

Fig. 7. Surface plots showing the variation of wear vs. (a) load, speed, (b) load, time and (c) speed, time.
5. CONCLUSION

The tribological behavior of electroless Ni-P-W coating under dry condition is investigated in the present work by following Taguchi’s L27 OA. A hybrid grey-fuzzy logic multi criterion decision making method is adopted to minimize the wear depth and COF simultaneously. It is seen that the combination of the lowest level of load (10 N), mid – level of speed (80 rpm) and the lowest level of time (5 min) is the optimum combination of tribological testing parameters for the same. ANOVA results indicate that the normal load has the highest contribution in controlling the friction and wear behavior of Ni-P-W coating followed by sliding time and speed. The interaction of load and speed also contributes significantly to the friction and wear behavior of the deposits. SEM results indicate that a Ni-P-W coated surface has nodulated structures while heat treatment (400 °C for 1 hour) leads to a compaction of the EN matrix. Heat treatment also results in change of phase of the coating from amorphous/nanocrystalline to crystalline due to the precipitation of nickel and its phosphides. Abrasive wear of the coatings takes place which is confirmed from the SEM image of a worn surface. Surface plots indicate that the wear depth increases with the increase in values of all the interaction terms i.e. load and speed, load and time as well as speed and time. The COF increases with the variation of load–speed and load–time interaction while it has a decreasing trend for the speed–time interaction. A significant amount of reduction in wear and friction is obtained from the optimization study. The present investigation can prove to be
beneficial for selection of a potentially hard surface coating like Ni-P-W in applications requiring the reduction of friction and wear of mating surfaces working under non – lubricated environment.

Acknowledgement

The authors gratefully acknowledge the financial support of University Grants Commission (UGC), Govt. of India through Major Research Project vide Ref. No. F No. 41-984/2012 (SR) dated 25.07.2012.

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