

Reliability Investigation and Hazard Rate Diagnosis of Shaliwahana MSW 12MW Green Power Plant Components

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Abstract: Investigation of reliability is by default a must needed process to access the status of the machine. The machine condition of performance and ability to perform its intended tasks are understood by the investigation of the reliability. Basically reliability is the ability of the machine to perform its intended task with zero or minimum number of failures. The more is the reliability of the machine the better is the performance. Hence this research is very much needed and useful one for the understanding of the machine. Here the term machine describes the boiler and turbine in the power plant. Although there are many numbers of components available in the plant only boiler and turbine are considered here in this research for the analysis. The power plant considered here in this research is a municipal solid waste power plant MSW named Shaliwahana MSW 12MW green power plant. While performing the analysis the boiler and turbine performance and failure data for the previous 4 years is considered that is from the 2015 to 2019. The data pertaining to these components is refined and trend analysis is done to know the machine trend to find the position of the machine in bath tub curve. Then the hazard rate diagnosis is done by two parameter weibull analysis is made to ascertain the nature of machine performance with respect to reliability. From the results of the weibull analysis the functioning of the machine is formulated and further an arena is made towards its performance and maintenance formulation.

Keywords: Hazard rate diagnosis, Reliability investigation, Weibull analysis.

1. Introduction

Electricity is a major requirement to the mankind for every need; one cannot imagine the present world without electricity. In this scenario the performance of the electric power plant is very much needed to make the productivity as maximum as possible. In the other terms making the machines to work with zero or minimum number of failures. In this context the present research presented in this paper focuses on the behavior of machines such as boiler and turbine used in the plant. If the past and present performance of the machine is analyzed means the future performance can be controlled with ease. With this as motto of the present research is made to analyze the boiler and turbine of the Shaliwahana 12 MSW green power plant since 2015 to 2019. In each year around two months of time is spent on the annual maintenance hence the net effective working days of the power plant becomes 300days in a year. Within the 300 working days of operation in each year, how much time the boiler and turbine suffered with failures and their repairing times along with their networking time is recorded, collected and refined [1]-[3]. The data obtained from the plant log books and systems along with the manpower are collected, refined, analyzed and summarized in such a way that the time of operation is formulated as time between failures (TBF) and the time consumed for the machines repair is formulated as time to repair (TTR). Further the data formulated as TBF and TTR are converted in to ordered time between failures (OTBF) and ordered time to repair (OTTR) as well as cumulative time between failures (CTBF) and cumulative time to repair (CTTR) [4], [5]. And also the refined data is used to calculate the mean time between failures (MTBF) and mean time to repair (MTTR).

2. Literature survey

There is a lot of need and necessity behind the investigation of reliability of the power plant. By investigating the ability and capability of the machines employed in the power plant a better maintenance policy can be formulated. And also the present status of the machine and its position in its life cycle can be found [6]-[8]. Be far this is very important consideration for the decision making about the replacement analysis of the power plant components. In general, one may experience a doubt like why to make all this analysis as the product or machine manufacturer used to detail the life span of the machine. But in reality the machines performance alters from the numbers given while manufacturing to the numbers obtained while functioning. It is because of many uncertainties, out of these uncertainties some are of system generated and some are of manmade. Hence in order to understand the machine's performance in terms its ability to discharge the duties well one has to undergo the reliability investigation as well as hazard rate diagnosis [9], [10].



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3. Methodology



Fig. 1. Methodology of reliability investigation and hazard rate diagnosis of Shaliwahana MSW 12MW green power plant components

4. Analysis

The analysis of the work starts with the refined data at first, then the trend analysis is carried out using six tests then the majority of them are considered as actual trend. Out of the six tests two are of graphical type and the remaining four are of analytical type. Graphical tests are Eye ball test, Cumulative plot test and analysis tests are Karl Pearson's correlation test in this again two types first (i) vs(i-1) and second (i) vs (i-2), Serial correlation test in this first (i) vs(i-1) and second (i) vs (i-2). While performing the analysis in each year boiler and turbine are considered as two separate machines and they are given below in table 1.

Table 1

List of machines considered for the analysis							
S. no.	Machine	Machine	Machine	Machine			
	number	name	number	name			
1	1	2015 Boiler	6	2015 Turbine			
2	2	2016 Boiler	7	2016 Turbine			
3	3	2017 Boiler	8	2017 Turbine			
4	4	2018 Boiler	9	2018 Turbine			
5	5	2019 Boiler	10	2019 Turbine			

The results of the eye ball tests and cumulative plot tests based on the TBF and TTR are illustrated below. In the graphs the x ordinate represents cumulative number of failures (CNF) and y axis represents cumulative time between failures (CTBF) for the reliability trend analysis and in the same way for the availability calculation x axis is same as CNF and Y axis changed as cumulative time to repair (CTTR).



Fig. 2. Cumulative plot test result for the machine 2015 boiler







Fig. 4. Eye ball test result for the machine 2015 Turbine





Fig. 5. Eye ball test result for the machine 2015 Boiler



Fig. 6. Serial correlation test (i vs i-1) result for the machine 2015 Turbine



Fig. 7. Serial correlation test (i vs i-1) result for the machine 2015 Boiler

In the above graphs the figure 2 and 3 represents the trend test results using the cumulative plot test for the machines 2015 boiler and turbine respectively and in the figure 4 and figure 5 they belongs to the trend tests results (for reliability) of the eye ball test. In the same way using the TTR for the availability calculations eye ball tests are drawn for the trend analysis of all the 10 machines. And similarly using the CTTR for the availability calculations cumulative plot tests are drawn for the trend analysis of all the 10 machines.



Fig. 8. Serial correlation test (i vs i-2) result for the machine 2016 Boiler



Fig. 9. Serial correlation test (i vs i-2) result for the machine 2016 Turbine

The above figures 6,7,8 and 9 shows the serial correlation test of i^{th} vs (i-1)th and i^{th} vs (i-2)th test results for 10 machines each for TBF and TTR. 4 graphs are shown above similarly the graphs can be plotted for all the machines.

And after these graphical tests the analytical test is carried out to know the trend of 10 machines. As mentioned above in the analytical test used here is Karl Pearson's coefficient of correlation test first (i) vs(i-1) and second (i) vs (i-2). The formula to calculate the trend is given below.

Coefficient of Correlation {r}=

sum x*y/sqrt{(sum x**2)*(sum y**2)} = 0.9976



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The above table 2 shows Karl Pearson's coefficient of correlation test (i) vs(i-1) of 2016 Boiler based on reliability (TBF) and likewise for all the machines computations are calculated.

5. Hazard rate diagnosis

To make the hazard rate diagnosis the 2 parameter weibull analysis is done. In this the two parameters are α and β . α is known as scale parameter of weibull distribution and β is known as shape parameter of weibull distribution. The analysis is done using the following governing equation.

$$F_{(t)} = 1 - \exp(-(t/\alpha)^{\beta})$$

It can be further modified into Ln ln $[1/{1-F_{(t)}}] = \beta \ln t - \beta \ln \alpha$

In the above equation $F_{(t)}$ can be calculated using the

Table 3	
d tests results based on	reliabil

Trend tests results based on reliability							
Item	TBF based	Eye	Karl	Karl	Serial	Serial	Result
	Cumulative	Ball Test	Pearson (i-1) Pearson (i-2)		Correlation	Correlation	
	Plot Test				(i) vs (i-1)	(i-1) vs (i-2)	
2015 T	Week - ve Trend	Week + ve Trend	- ve Trend	+ ve Trend	- ve Trend	- ve Trend	- ve Trend
2016 T	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend
2017 T	No Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	+ ve Trend	+ ve Trend	+ ve Trend
2018 T	Week + ve Trend	Week - ve Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	+ ve Trend
2019 T	Week - ve Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	+ ve Trend	+ ve Trend	+ ve Trend
2015 B	Week + ve Trend	- ve Trend	- ve Trend	Week - ve Trend	+ ve Trend	- ve Trend	- ve Trend
2016 B	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend
2017 B	Week - ve Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	+ ve Trend	+ ve Trend	+ ve Trend
2018 B	Week - ve Trend	Week + ve Trend	Week + ve Trend	Week + ve Trend	+ ve Trend	+ ve Trend	+ ve Trend
2019 B	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend

Table 4 Trend tests results based on availability

Tiend tests festilis based on availability							
Item	TTR based	Eye Ball Test	Karl Pearson (i-1)	Karl Pearson (i-2)	Serial Correlation	Serial Correlation	Result
	Cumulative				(i) vs (i-1)	(i-1) vs (i-2)	
	Plot Test						
2015 T	Week + ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend
2016 T	No Trend	Week - ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2017 T	Week - ve Trend	+ ve Trend	Week + ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2018 T	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2019 T	Week - ve Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2015 B	Week + ve Trend	Week - ve Trend	Week - ve Trend	Week + ve Trend	- ve Trend	- ve Trend	- ve Trend
2016 B	No Trend	Week + ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2017 B	No Trend	No Trend	Week - ve Trend	Week + ve Trend	- ve Trend	- ve Trend	- ve Trend
2018 B	Week + ve Trend	Week - ve Trend	Week - ve Trend	Week - ve Trend	- ve Trend	- ve Trend	- ve Trend
2017 B	No Trend	No Trend	Week - ve Trend	Week + ve Trend	- ve Trend	- ve Trend	- ve Trend

Table 5

Summary of results						
S. no.	no. Trend analysis			Hazard rate diagnosis using 2 parameter weibull analysis		
	Parameter	Machine	Trend	α	β	Nature of machine behavior
1	TBF	2015 Boiler	-ve	18.7	0.8	Decreasing failure rate
2	TBF	2016 Boiler	-ve	19.6	0.7	Decreasing failure rate
3	TBF	2017 Boiler	+ve	18.8	1.4	Increasing failure rate
4	TBF	2018 Boiler	+ve	21.2	1.3	Increasing failure rate
5	TBF	2019 Boiler	+ve	19.8	1.5	Increasing failure rate
6	TBF	2015 Turbine	-ve	22.7	0.9	Decreasing failure rate
7	TBF	2016 Turbine	-ve	18.9	0.8	Decreasing failure rate
8	TBF	2017 Turbine	+ve	24.7	1.6	Increasing failure rate
9	TBF	2018 Turbine	+ve	31.39	1.8	Increasing failure rate
10	TBF	2019 Turbine	-ve	30.8	0.8	Decreasing failure rate
11	TTR	2015 Boiler	-ve	28.7	0.7	Decreasing failure rate
12	TTR	2016 Boiler	-ve	29.8	0.7	Decreasing failure rate
13	TTR	2017 Boiler	-ve	26.7	0.8	Decreasing failure rate
14	TTR	2018 Boiler	-ve	27.9	0.9	Decreasing failure rate
15	TTR	2019 Boiler	-ve	30.8	0.9	Decreasing failure rate
16	TTR	2015 Turbine	-ve	31.7	0.8	Decreasing failure rate
17	TTR	2016 Turbine	-ve	30.3	0.7	Decreasing failure rate
18	TTR	2017 Turbine	-ve	29.7	0.8	Decreasing failure rate
19	TTR	2018 Turbine	-ve	31.6	0.7	Decreasing failure rate
20	TTR	2019 Turbine	-ve	29.7	0.7	Decreasing failure rate



relationship $F_{(t)} = \frac{j-0.3}{n+0.4}$ and α , β values are computed from the graph.

6. Results

The results of trend tests are as follows shown in table 3 and 4.

From the figure 10 the shape parameter β value is 1.8, scale parameter α value is 31.39, intercept value is -5.963. These values indicate that the machine 2018 turbine has the concave distribution and its property is it has the increasing failure rate. This implies that the machine is in old age from the bath tub curve.



Fig. 10. Weibull plot for 2018 Turbine (TBF)

7. Conclusion

It is concluded from the above analysis the trend based on reliability and availability of all the machines are calculated and listed in table 5. And also the hazard rate diagnosis based on reliability and availability of all the machines are calculated and listed in table 5.

Based on the above research the machine exact status is found in its life cycle it gives the arena to understand about the machine and from that a decision can be made towards the betterment of the machine.

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