



Analysis for Induction Motor Fault Detection with Machine Learning

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ABSTRACT

Induction motors are one of the most critical electrical hardware and are broadly utilized in ventures in a wide scope of utilizations. This paper presents an machine learning or AI model for the fault detection and arrangement of acceptance engine issues by utilizing three stage voltages and flows as data sources. The point of this work is to ensure essential electrical parts and to forestall unusual occasion movement through early discovery and conclusion. This work presents a quick forward counterfeit neural organization model to recognize a portion of the generally happening electrical shortcomings like overvoltage, under voltage, single staging, uneven voltage, over-burden, ground issue. A different model free observing framework wherein the actual engine behaves like a sensor is introduced and the main checked signs are the info given to the engine. Cutoff points for current and voltage esteems are set for the defective and solid conditions, which is finished by a classifier. Continuous information from a 0.33 HP acceptance engine is utilized to prepare and test the neural organization. The model so created investigations the voltage and current qualities given at a specific moment and groups the information into no issue or the particular issue. The model is then interfaced with a genuine engine to precisely distinguish and arrange the fault so further fundamental move can be made.

Key words: Induction motor, machine learning or AI, fault analysis, ANN, projecting maintenance.

1.INTRODUCTION

Induction motors are the most widely utilized electrical motors as a result of their straightforward development, roughness and minimal expense. Over 90% of the enterprises use enlistment motors, for the most part as electrical drives since they can be intended for a wide scope of force appraisals. Disregarding their adaptability and vigor, they are exposed to numerous disastrous disappointments. Recognizing these deficiencies at a beginning phase

and taking care of them is vital as in any case they

Lead to enormous creation and monetary misfortunes.

Pre-shortcoming location and Separation of the sound parts additionally forestalls shortcoming movement and disappointment of other more crucial parts. Ventures utilize an enormous number of motors and subsequently, their manual support is dreary and untrustworthy. Thusly, many endeavors have been made towards programmed support. Prior, restrictive observing of electrical machines was utilized and the electromechanical transfers were put to use to accomplish this. Yet, these transfers are delayed in activity and furthermore lead to tremendous force misfortunes because of mechanical parts included [6]. Subsequently, they can't be utilized in basic applications requiring little reaction times. Electromechanical transfers were subsequently supplanted by strong state transfers as they devour extremely less force and are similarly quick. With the appearance of microchips, endeavors were made for restrictive observing of machines by utilizing pre-composed projects downloaded onto the chip chips.

The previously mentioned procedures, but couldn't ensure greatest wellbeing and dependability as they can't address calamitous disappointments. PC upheaval with the beginning of AI drew the consideration of researchers and they started to consider courses through which these strategies could be utilized to screen and defend machines. AI models replace human to brilliantly screen and keep up with the predetermined framework assignments [4]. Counterfeit neural organizations are extremely convenient in such manner as they can deal with tremendous measure of information, have little reaction time and can successfully deal with non-linearity (which more often than not is the innate trait of electromechanical frameworks) [10].

The point of this work is to forestall issue movement and ensure essential parts of the force framework by early identification of electrical deficiencies of three stage acceptance motors utilizing counterfeit neural organization. We have tended to seven classes of electrical shortcomings of acceptance motors; over-burden, ground issue, locked rotor, single staging, over voltage, under voltage and lopsided stockpile voltage.

Segment II gives brief data about the took on neural organization approach. Consequences of the work are examined in segment III. Genuine framework reconciliation is clarified in segment IV. Segment V gives a far reaching perspective on comparative AI based methods while area VI offers future extension. The discoveries are again examined in segment VII.

2. ARTIFICIAL NEURAL NETWORK APPROACH.

The ANN goes under AI draws near.

Any AI issue is settled through the accompanying advances.

1. Defining targets: The initial move towards addressing any AI issue is to characterize the objectives. The issue may be to track down a mathematical yield esteem (straight relapse), to isolate into classes (arrangement) or grouping. The sort of AI approach that should be utilized is chosen by concentrating on the issue and removing the goals out of it.

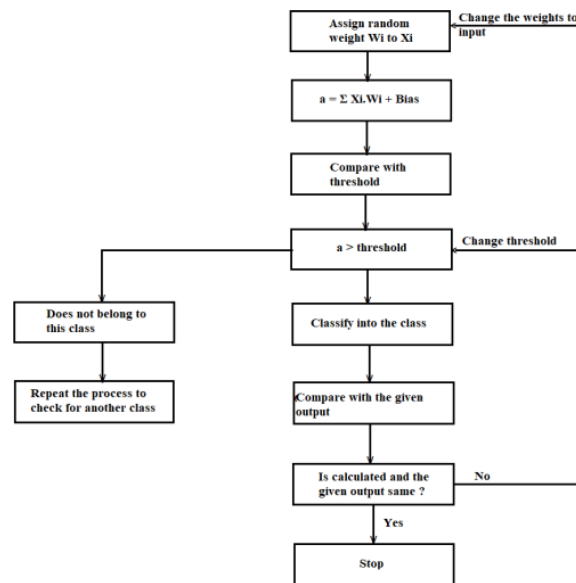


Fig 1: Block diagram of ANN algorithm

2. Data procurement: The information needed to prepare the neural organization is acquired progressively. The sort and measure of information required relies upon the application.

3. Separating train and test information: The gathered information is then isolated into train and test information to prepare the neural organization model and afterward test its working and compute its exactness [12].

4. Analysing the discoveries: The acquired outcomes are then examined to check for any mistakes and furthermore to recognize scope for development, assuming any.

5. Interfacing with genuine framework: Once the legitimate working of the prepared model is affirmed, it is interfaced with a genuine framework and is then utilized progressively applications.

The Artificial Neural Network (ANN) is intended to mirror human mind to have an independent perspective and make a move without being expressly modified. Actually like a human mind, the ANN model comprises of fake neurons or hubs as the essential structure blocks. The hubs are organized in layers: input layer, yield layer and at least one secret layers. The secret layer is utilized to expand the precision of grouping. The working of the calculation can be clarified from the beneath steps.

1) The calculation doles out loads or needs to every one of the information variable.

2) It then, at that point, duplicates the load with the information variable, x_i and adds inclination to it. Exactly the same thing is done to every one of the hub and the summation of these qualities is sent to the following layer. The learning continues as information travels through each layer.

3) The information toward the finish of yield hub is then contrasted and a set edge esteem and if the determined worth surpasses the edge, the info vector is ordered to the expected class or, more than likely is disregarded to check for different classes. The square outline of the working of a feed forward back circle neural organization calculation is as displayed in fig.1.

4) The calculated class is contrasted and the real yield and any blunder is amended by fluctuating the quantity of layers and the quantity of neurons.

5) The cycle is proceeded till a precise fit is gotten and the model is consequently, prepared. The condition of the calculation is,

$$\sum W_i X_i + \text{Bias} \geq \text{Threshold}$$

In this work, three stage voltages and flows of a three stage 1/3 HP, 208 V acceptance engine is utilized as the information. Information is gathered continuously for every one of the classes and is utilized to prepare and test the neural organization. MATLAB is the product utilized whose neural organization tool kits train and tests the information.

III. RESULTS AND DISCUSSION

In this work, the fake neural organization model is made and prepared by utilizing three stage voltages and flows as the information. The model is prepared for every one of the seven electrical fault utilizing roughly 800 worth arrangements of three stage voltages and flows. An example of the pre-owned data set is as displayed in table 1.

Table 1: Sample database

CLASS	V1	V2	V3	I1	I2	I3
1	2.661025	2.624276	2.701274	0.490768	0.478549	0.493368
1	2.660319	2.624661	2.700700	0.491114	0.478722	0.492584
2	2.647625	2.598815	2.671626	0.006194	0.643518	0.640217
2	2.650816	2.601661	2.673722	0.006123	0.641548	0.638553
3	0.919570	2.621412	2.626511	0.172113	0.772419	0.662758
3	0.919852	2.621627	2.625342	0.172072	0.772106	0.662653
4	1.874796	1.855089	1.874878	0.286777	0.287052	0.281013
4	1.452803	1.449902	1.441935	0.245231	0.249739	0.234152
5	2.865128	2.871906	2.855436	0.482896	0.499206	0.496894
5	2.868791	2.875877	2.860353	0.483453	0.499363	0.496879
6	2.657179	2.613409	2.687374	1.671357	1.650515	1.668712
6	2.661374	2.613395	2.688907	1.416786	1.397752	1.411372
7	2.637658	2.600486	2.673771	0.803147	0.782514	0.797477
7	2.650468	2.608143	2.682578	0.857336	0.837601	0.847664

Feed forward back loop algorithm is used and the trainedmodel is as shown in fig. 2

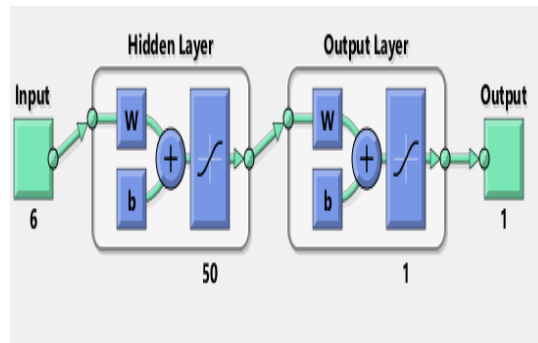


Fig 2: Trained ANN model

The weight (need) allotted to each information can be seen from fig 2 and furthermore the numerical capacity $\sum XiWi + Bias$ working can be envisioned. After every summation, a sigmoid capacity is created which sends the current worth to the following layer. Furthermore, the relapse plot showing the precision of preparing is as displayed in fig 3. From fig 3, it tends to be seen that the prepared model fits precisely with the ideal. Consequently, the preparation of the organization is acceptable. The prepared neural organization was then tried with test data. The ANN model classified each of the testdata into their respective classes.

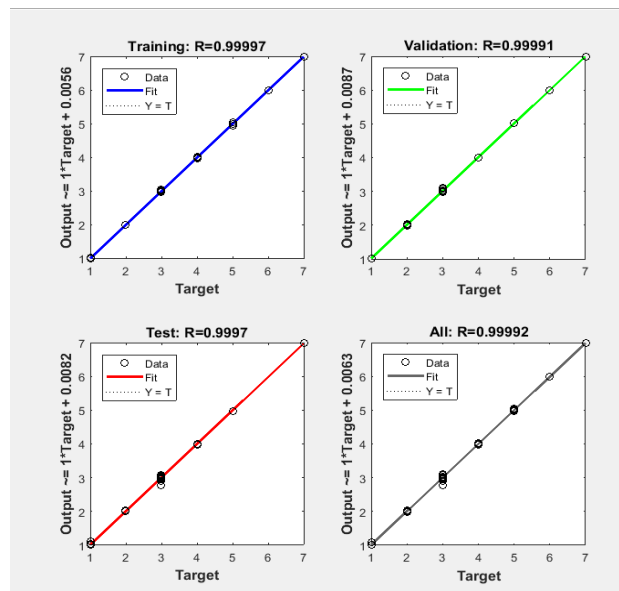


Fig 3: Regression plot of trained model

Table 2, gives the frequency of classification of each fault. The proposed ANN model classified all the test data accurately.

Table 2: Test results

Fault	Frequency of classification
No fault (1)	11
Overload (2)	12
Ground fault (3)	15
Locked rotor (4)	17
Unbalanced voltage (5)	3
Single phasing, undervoltage (6)	3
Overvoltage (7)	5
Total	67

IV. INTERFACING WITH REAL MOTOR SYSTEM

To interface the created ANN model with the engine framework, the information should be gathered ceaselessly continuously from current and voltage sensors. The got values are communicated to Wi-Fi switch by utilizing a solid Wi-Fi association. This information is then moved and gathered on a cloud stage (any of the numerous accessible) by utilizing an Ethernet link between the switch and capacity framework. From the cloud, the information is consistently communicated onto the MATLAB work area. At the point when a shortcoming happens, the prepared ANN model identifies the issue promptly and advises the control framework to make a move.

V. SIMILAR MACHINE LEARNING BASED TECHNIQUES.

Of late, numerous different strategies like Support vector machines (SVM), fluffy rationale, MCSA are being considered for applications including prescient upkeep of electrical gear. Nonetheless, they include parcel of science, are boundary based and henceforth, are blunder inclined when utilized for enormous and complex frameworks. High numerical reliance of these procedures additionally makes them not really proficient for electromechanical frameworks because of their failure to deal with non-linearity [13]. ANN then again is non-parametric, doesn't need complex arithmetic and can be acknowledged utilizing straightforward apparatuses and subsequently, is an extremely intense system for issue location and upkeep [9].

VI. FUTURE SCOPE

Predictive support of electrical gear, including enlistment motors utilizing MACHINE LEARNING approaches is as yet in proposition stage in many agricultural nations including India. ANN because of its capacity to gain from past encounters can successfully address calamitous disappointments and is wonderful to deal with non-linearity in information. The proposed ANN model is sufficiently interfaced with genuine frameworks and is demonstrated to be dependable. Subsequently it can fill in as the most effective and conservative prescient upkeep device in future. The proposed model can likewise be reached out for D.C and simultaneous motors.

VII. CONCLUSION

This work portrays the capability of AI apparatus (ANN) in recognizing the electrical fault of a three stage acceptance engine. The information gathered continuously from a 1/3 HP, 208 V three stage enlistment engine was utilized to prepare and test the neural organization. The created neural organization groups all the test information into their separate classes with practically 100% precision.

The precision of the prepared model is reflected by the outcomes got utilizing the test information and furthermore from the relapse plot displayed in figure 3. In the event that there are blunders in grouping, the exactness can be improved by expanding the quantity of stowed away layers and by picking ideal number of neurons. The proposed model gives exact outcomes and thus, forestalls occasion movement during the event of deficiencies and ensures fundamental electrical hardware.

REFERENCES

- [1] Hammo Rama, "Fault Identification in Three-Phase Induction Motors Using Support Vector Machines" (2014). Master of Technology Management Plan II Graduate Projects. Paper I.
- [2] M. A. Wahab, "Fake neural organization based expectation strategy for transformer oil breakdown voltage", *Electric Power System Research*, Vol. 71, No. 1, pp. 73-84, 2004.
- [3] C. Li, R. V. Sanchez, G. Zurita, and M. Cerrada, "Multimodal profound help vector order with homologous components and its application to gearbox shortcoming conclusion," *Neuro-registering*, vol. 168, pp. 119-127, 2015.
- [4] Y. Lei, F. Jia, X. Zhou, and J. Lin, "A Deep Learning-based Method for Machinery Health Monitoring with Big Data", *Journal of Mechanical Engineering*, vol. 51, pp. 49-56, Nov. 2015.
- [5] C. Li, R. V. Sanchez, and G. Zurita, "Shortcoming Diagnosis for Rotating Machinery Using Vibration Measurement Deep Statistical Feature Learning," *Sensors*, vol. 16, pp. 895, June 2016.
- [6] Faiz, J., Ghorbanian, V., Ebrahimi, B. "An overview on condition of inspection and issue fortitude in line-start and inverter-took care of broken bar receiving motors". *IEEE Int. Conf. on Power Electronics, drive and Energy system (PEDES)*, 2012, pp. 15

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- [7] Georgakopoulos, I., Mitronikas, E., Safacas, A. "Recognition of acceptance engine deficiencies in inverter drives utilizing inverter input current examination", *IEEE Trans. Ind. Electron.*, 2011, 58, (9), pp. 4365–4373
- [8] Karabadjji, N.E.I., Seridi, H., Khelf, I., et al.: "Further developed choice tree development dependent on characteristic choice and information inspecting for issue finding in pivoting machines", *Eng. Appl. Artif. Intell.*, 2014, 35, pp. 71–83.
- [9] Ghate, V., Dudul, S. "Shortcoming finding of three stage acceptance engine utilizing neural organization methods", *Second Int. Conf. on Emerging Trends in Engineering and Technology (ICETET)*, December 2009, pp. 922–928.
- [10] Seera, M., Lim, C., Ishak, D., et al. "absence recognition and finding of induction motors utilizing engine recent mark theory test and a partially and half FMM-CART model". *IEEE Trans. Neural Netw. Learning Syst.*, 2012, 23, (1), pp. 97–108.
- [11] Baccarini, L., Avelar, V., Silva, V., et al. "Astute scaffold plan for stator windings fault analysis: appropriate for upkeep work", *J. Softw. Eng. Appl.*, 2013, 6, (10), pp. 526–532.
- [12] W. Lu, X. Wang, C. Yang, and T. Zhang, "A Novel Feature Extraction Method utilizing Deep Neural Network for Rolling," 29th Chinese Control and conclusion consultation (CCDC), Qingdao China, pp. 2427-2431, May 2015.
- [13] E. Byvatov, U. Fechner, J. Sadowski, and G. Schneider, "connection of carry Vector Machine and Artificial Neural institute Systems for Drug/Nondrug arrangement," *Journal of element in sequence and Computer Sciences*, vol. 43, pp. 1882–1889, Sept. 2003.
- [14] J. Schmidhuber, "Profound learning in neural organizations: An outline," *Neural Networks*, vol. 61, pp. 85-117, 2015..
- [15] C. Li, R. V. Sanchez, and G. Zurita, "Shortcoming Diagnosis for Rotating Machinery Using Vibration Measurement Deep Statistical Feature Learning," *Sensors*, vol. 16, pp. 895, June 2016.