

OPEN ACCESS

Volume: 11

Special Issue: 1

Month: July

Year: 2023

E-ISSN: 2582-0397

P-ISSN: 2321-788X

Impact Factor: 3.025

Received: 08.05.2023

Accepted: 13.06.2023

Published: 01.07.2023

Citation:

Gudada, Priyanka V., and KR Rohith. "A Detailed Analysis of Data Mining Procedures for Problem Detection In Energy Rcd Systems." *Shanlax International Journal of Arts, Science and Humanities*, vol. 11, no. S1, 2023, pp. 177–81.

DOI:

<https://doi.org/10.34293/sijash.v11iS1-July.6334>

A Detailed Analysis of Data Mining Procedures for Problem Detection In Energy Rcd Systems

Priyanka V. Gudada

*Department of Master of Computer Applications
Raja Rajeswari College of Engineering, Bengaluru*

Rohith KR

*Department of Master of Computer Applications
Raja Rajeswari College of Engineering, Bengaluru*

Abstract

Critically important early recognition of failures within electrical electrical appliances (PESs) to guarantee reliability is an obstacle that has garnered a great deal attention recently. After examining diverse literature on defect detection in PESs, this study introduces data mining-based techniques include deep learning computations, learning from data, and artificial neural networks. Then, in PESs, the defect detection process is described by adding signal measuring sensors and how to extract the characteristic from them. Finally, the efficacy of various data mining algorithms in detecting PESs flaws is evaluated based on research. The assessment findings reveal that deep learning-based techniques with being able to extract features from measured signals are markedly more effective than alternative techniques and serve as an appropriate device for the prospective applications in the power electronics sector are discussed

Keywords: Mininig, PESs, Neural Network

Introduction

Recent Days, electrical energy is a powerful force in the scientific, economic, and welfare spheres of human existence. In recent years, the progress of electrical energy applications Considering the amount of electrical energy users has caused distributed generation to significantly outperform traditional power systems. Renewable energy sources and However, storage for energy systems received widespread attention deployed to minimise fossil fuel usage and tackle environmental concerns. The main point is that It is essential to have power electronic systems. for the creation, storage, and Power usage in the economic and everyday life cycle. PESs show an vital character in integrating RESs, energy management, power grid dependability, and other relevant infrastructures and systems. PESs provide for simple and low-cost energy/power conversion.

Converters for electricity were utilised more and more frequently in electric motor drive networks, industrial robotics, high-voltage straight-current gearbox, green vehicles, and high-power hydrolysis in recent years.

Prediction and description models could be used to transfer out DM tasks. In general, prediction models are built by learning known data classes, whereas description models are built from the results of a dataset. In this sense, predictive DM methods are the greatest basic alternative for doing apparatus and system diagnostics since their various operating conditions may be learnt and predicted by a forecast

Literature Survey

Recent Days, electrical energy is a powerful force in the scientific, economic, and welfare spheres of human existence. In the last few years, technological progress has applications and the quantity of Energy users has caused distributed generation to significantly outperform traditional power systems. Renewable energy sources and Yet storing energy devices received widespread attention deployed to minimise fossil fuel usage and tackle environmental concerns. The main point is that It is essential to have power electronic systems. for the creation, storage, Regarding physical usage in the economic and everyday life cycle. PESs show an vital character in integrating RESs, energy management, power grid dependability, and other relevant infrastructures and systems. PESs provide for simple and low-cost energy/power conversion.

Converters for electricity become utilised ever more frequently on electric motor drive networks, industrial robotics, high-voltage straight-current gearbox, green vehicles, and high-power hydrolysis in recent years.

Related Studies

Several publications in the current state of regression and classifier research This was displayed in literature. methods that may be applied in a a range of applications. Djeflal et al. outline a method for removing instances that have little impact on SVM educational results using screening and modification phases, with the objective of reducing working out time without sacrificing accuracy. This technique might be use to handle and reduce large datasets prior to using any other procedure. Zhao et al. handle the difficult problem of marker sound in categorization. They offer a Markov chain sampling system that reliably learns good categories and precisely identifies samples with incorrect labels in this context.

Existing Approach

The main issue is that power electronic systems (PESs) are required for the generation, storage, and use of electrical energy in the economic and everyday life cycle. PESs play an imperative role in integrating RESs, energy management, power grid dependability, and other relevant infrastructures and systems. PESs provide for simple and low-cost energy/power conversion. Despite all of its advantages, among those with the greatest important disadvantages of PESs is their high sensitivity to natural catastrophes, frequent switching in hostile environments, and so on, which leads in power outages or system shutdown and, the outcome, increased operating costs.

Among the maximum essential criteria in the necessities of PESs applications is long-term sustainability without power interruption. High temperatures, overvoltage and overcurrent, wear-out of electrical components, radiation, vibration and mechanical damages, thermal damages, hardware design or control defects, and electromagnetic stresses are the most common causes of critical failures in PESs. According to several research, primary side semiconductors (low voltage, high current) and resonance components employed in PESs can be the principal contributors of damage owing to a variety of circumstances.

Proposed System

Fortunately the study only comprises a few defect diagnostic approaches based on Artificial Neural Network (ANN) algorithms. The authors of investigate capacitor condition monitoring strategies in power electronic converters, emphasising parameter identification methods based on ANNs. Furthermore, used a range of ANN-based approaches dubbed the dynamic Bayesian network and object-oriented Bayesian network for industrial applications such as transient and intermittent defect detection in complicated electronic systems. Another useful study introduces the hybrid applications rule-based algorithm and back propagation neural network (BPNN) for defect identification in a diesel engine. The data in the research has been conducted utilising wavelet threshold denoising and ensemble empirical mode decomposition.

Rapid failure identification in the use of motors with asynchronous actuators that are accomplished using data-driven systems known as Bayesian networks. In this work, wavelet threshold denoising and minimal entropy deconvolution techniques are used to pre-process and denoise the input signals in instruction to enhance the fault identification process. A overview of machine learning approaches used to control the dependability of energy systems is offered in this paper.

Current DM research has concentrated on building classification systems capable of managing datasets with various properties, such as proportional imbalance and vast volumes of facts. This capability is particularly important in the final since, on the one hand, data availability is increasing but, on the other hand, their presentation might be jeopardised if only restricted datasets are analysed. In reality, the sum of data accessible through the working out of a neural network (NN) is a critical factor in its success. Taylor et al., for example, explored three strategies: simple predicted rulesets learned using biological approaches, neural networks developed by a mixture of evolving exploration and return propagation, and neural networks educated employing straightforward backpropagation to The findings show that generated NNs perform better than NNs developed by backpropagation.

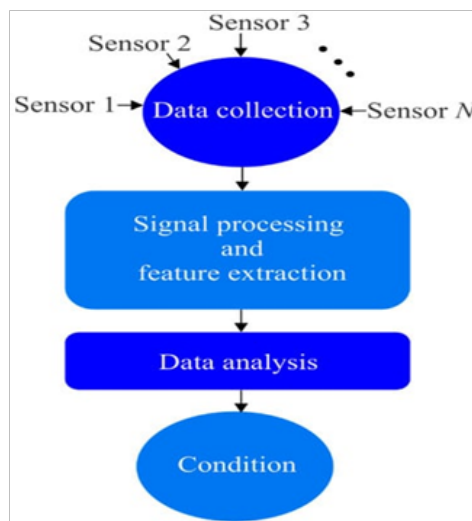


Figure 1 Proposed Architecture

Implementation

The autoencoder (AE) is a deep learning technology that is frequently used to decrease data dimension and extract data features with low reconstruction error. As shown, an AE is made up of an encoder and a decoder network, The hidden layer or layers, an input layer, then a conclusion

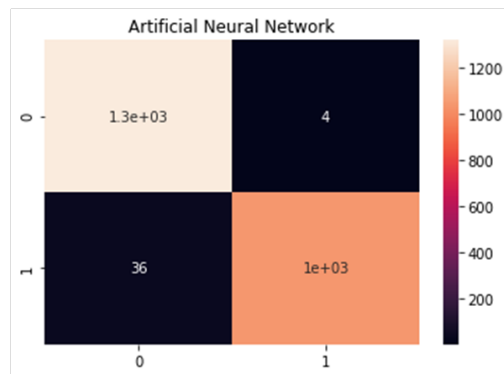
component. The structure and function of AE are such that after dimensional reduction, it may maintain the most data-related characteristics and the most relevant information based on nonlinear layers. During the AE training phase, with proper retrieval of the input data, hidden cells in the hidden layers keep the most evidence about the original input data.

PES exchangers have nonlinear behaviour, and their behaviour varies dramatically in the presence of anomalies. Any problem or anomaly that happens on the PES converters has an influence on the system's behaviour, and the extraction of these effects from the system's characteristic curves begins the process of anomaly identification indicated in the literature, intelligent learning-based approaches have been employed as a range of defect detection tools in most research.

However, in most situations, these strategies have encountered questions when industry with composite systems and high-dimensional data that contain noise and anomalous behaviour in the signals received from the system

Result

Data Selection						
	Output (S)	Ia	Ib	...	Vc	Unnamed: 7 Unnamed: 8
0	-170.472196	9.219613	...	0.605431	NaN	NaN
1	-122.235754	6.168667	...	0.526202	NaN	NaN
2	-90.161474	3.813632	...	0.464251	NaN	NaN
3	-79.904916	2.398803	...	0.445963	NaN	NaN
4	-63.885255	0.590667	...	0.411050	NaN	NaN
5	-55.954681	-1.001882	...	0.397281	NaN	NaN
6	-45.248446	-2.586980	...	0.371743	NaN	NaN
7	-47.845428	-3.428094	...	0.373727	NaN	NaN
8	-43.294259	-4.511300	...	0.357840	NaN	NaN
9	-43.474722	-5.388233	...	0.355587	NaN	NaN
10	-37.006454	-6.567132	...	0.337111	NaN	NaN
11	-40.491588	-7.334343	...	0.339412	NaN	NaN
12	-38.489593	-8.385392	...	0.327893	NaN	NaN
13	-40.907121	-9.331471	...	0.328015	NaN	NaN
14	-35.638927	-10.639162	...	0.311706	NaN	NaN



Conclusions

Power electronic systems (PESs) are regarded as critical components of power/energy systems, and the stability of these systems is largely dependent on the health of PESs. Anomalies in PESs are caused by a range of causes, which might jeopardise system security and dependability. Such concerns necessitate the urgent requirement for early defect identification in PES. Different forms of PES faults and diagnostic procedures were investigated in this work. Since the commencement of the defect detection procedure in PESs, several pieces of literature have been evaluated. Following the displaying the numerous forms of defects associated with PESs, a fundamental categorization of fault detection methods in PESs based on data mining techniques and signal measuring sensors was offered, and each approach was introduced with a thorough explanation.

Considering how feature extraction is thought to be critical to the success of deep learning approaches in diagnostic applications, a quick overview of the feature extraction procedure from measured signals was provided. Finally, deep learning techniques were explained and offered being among the data mining applications for usage in various sectors of the power electronics industry in a broad review.

References

1. Bhattacharyya, R., & Basu, S. (2018). India Inc looks to deal with rising stress in employees. Retrieved from 'The Economic Times'
2. OSMI Mental Health in Tech Survey Dataset, 2017 from Kaggle
3. Van den Broeck, J., Cunningham, S. A., Eeckels, R., & Herbst, K. (2005). Data cleaning: detecting, diagnosing, and editing data abnormalities. *PLoS medicine*, 2(10), e267.
4. Shwetha, S, Sahil, A, Anant Kumar J, (2017) Predictive analysis using classification techniques in healthcare domain, *International Journal of Linguistics & Computing Research*, ISSN: 2456-8848, Vol. I, Issue. I, June-2017
5. Tomar, D., & Agarwal, S. (2013). A survey on Data Mining approaches for Healthcare. *International Journal of Bio-Science and Bio-Technology*, 5(5), 241-266.
6. Pedregosa, F., Varoquaux, G., Gramfort, A., Michel, V., Thirion, B., Grisel, O., & Vanderplas, J. (2011). Scikit-learn: Machine learning in Python. *Journal of machine learning research*, 12(Oct), 2825-2830.
7. Gender and Stress. (n.d.). Retrieved from APA press release 2010[1] Bhattacharyya, R., & Basu, S. (2018). India Inc looks to resolve rising stress in employees. Retrieved from 'The Economic Times'
8. Y. Wu, Y. Wang, Y. Jiang, and Q. Sun, "Multiple parametric faults diagnosis for power electronic circuits based on hybrid bond graph and genetic algorithm," *Measurement: Journal of the International Measurement Confederation*, vol. 92, pp. 365–381, Oct. 2016, doi: 10.1016/j.measurement.2016.06.018.
9. H. Soliman, H. Wang, and F. Blaabjerg, "A Review of the Condition Monitoring of Capacitors in Power Electronic Converters," *IEEE Transactions on Industry Applications*, vol. 52, no. 6, pp. 4976–4989, Nov. 2016, doi: 10.1109/TIA.2016.2591906.
10. M. Pecht and R. Jaai, "A projections and health administration roadmap for information- and electronics-rich systems," *Microelectronics Reliability*, vol. 50, no. 3, pp. 317–323, Mar. 2010, doi: 10.1016/j.microrel.2010.01.006.
11. H. S. H. Chung, H. Wang, F. Blaabjerg, and M. Pecht, *Reliability of Power Electronic Converter Systems*. Institution of Engineering and Technology, 2015.
12. M. Riera-Guasp, J. A. Antonino-Daviu, and G.-A. Capolino, "Advances in Electrical Machine, Power Electronic, and Drive Condition Monitoring and Fault Detection: State of the Art," *IEEE Transactions on Industrial Electronics*, vol. 62, no. 3, pp. 1746–1759, Mar. 2015, doi: 10.1109/TIE.2014.2375853.
13. L. Duchesne, E. Karangelos, and L. Wehenkel, "Recent "Machine Learning Developments for Energy System Reliability Management," *Transactions of the IEEE*, vol. 108, no. 9, pp. 1656–1676, Sep. 2020, doi: 10.1109/JPROC.2020.2988715.
14. S. Zhao, F. Blaabjerg, and H. Wang, "An Overview of Artificial Intelligence Applications for Power Electronics," *IEEE Transactions on Power*