Transformer fault diagnosis using Dissolved Gas Analysis technology and Bayesian networks

A. Lakehal, Z. Ghemari, and S. Saad

Abstract— Bayesian model is developed for transformer faults diagnosis using dissolved gas in oil analysis. DGA (Dissolved Gas Analysis) is the traditional and conventional transformer fault diagnosis method, which mainly depends on the experience of operators and of the percentages of dissolved gases. In addition, the only measurement of the gases percentage is not sufficient to evaluate the equipment health. There are several cases where the proportions of dissolved gases remain trapped in the transformer. Regarding this uncertainty and in order to make decisions in a certain environment, the model developed in this study represents a powerful tool for decision making. In addition, one traditional method of DGA does not enable the diagnosis of all faults, for example the Rogers Ratio Method diagnose five faults only, but using the proposed Bayesian network (BN) it is possible to diagnose all faults from the same model. To illustrate the advantages of Bayesian methods in transformer fault diagnosis, a study of power station main transformer is conducted and the results are analyzed and discussed.

I. INTRODUCTION

The transformer is the most important equipment in a power plant. The main concern for maintenance managers is to ensure optimal availability of this equipment. In the field of power distribution, the transformer is an omnipresent element, its failure influences directly he continuity of service. Thus in order to optimize the availability of this equipment rigorous monitoring of its behavior and an effective supervision must be carried out.

Dissolved gas analysis (DGA) is the most common technique used to monitor and diagnose transformer health. Each gas has a percentage, and each percentage should not exceed a limited value. Each anomaly gives a gas or a mixture of gas with different proportions, the proportions percentage depends on the anomaly nature [1]. The analysis technique of dissolved gas is based on traditional (conventional) methods of faults diagnosis such as, Key Gas Method, Dornenburg Ratio Method, Rogers Ratio Method, Method nomograph, IEC Ratio Method, Duval Triangle Method, and CIGRE Method [2]. Recently, artificial intelligence methods have been introduced to improve the

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Salah Saad is with Laboratoire Systèmes Electromécaniques (LSELM), University of Badji-Mokhtar Annaba, 23000, Algeria (saadsalah2006@yahoo.fr). results accuracy and remove the inherent uncertainty in the diagnosis and data reliability (Fuzzy Logic (FL), Artificial Neural Networks (ANN), Support Vector Machine (SVM), Evolutionary Algorithms and Expert System) [3].

The literature shows that in recent years, many reported works were carried out using the combination of two or more methods of artificial intelligence to better diagnose the causes of transformer faults. Combining ANN with expert system for transformer fault diagnosis [4], SVM, ANN and clonal selection algorithms optimization [5], and a model based on the SVM method and genetic algorithm [6].

This paper focuses on the transformer fault diagnosis using Bayesian network (BN). The BNs are artificial intelligence tools which can be employed for faults prediction and diagnosis. Using a BN, the probability of oil breakdown voltage drop as a function of the water content can be quantified [7]. BNs are based primarily on experimental data by using DGA to define the network parameters. They also depend on the experience feedback of maintenance staff operators to mount equipment failures various scenarios, and consequently to define the network structure.

II. BAYESIAN NETWORKS

Probabilistic graphical models such as BNs [8] have been widely used to solve various problems (for example, diagnosis, anomalies prediction and risk analysis) [7,9]. These models are characterized by their ability to treat uncertain information and represent the interdependencies between different variables of a given problem. The advantage of probabilistic graphical models is interesting graphical representation of models, easy to understand and analyze. In addition, the probabilistic failure analysis evaluates the probability of fault of a complex system that its weak points can be identified.

In Bayesian methods, *a priori* information, the likelihood and the *a posteriori* information are represented by probability distributions. Thus, *a priori* probability represents the probability distribution of knowledge or belief concerning a subject or a variable before the parameter it represents is observed. The likelihood is a parameter function of a statistical model, reflecting the possibility of observing a variable when these parameters have a value. *A posteriori* probability is the conditional probability on collected data by a combination of *a priori* probability and likelihood through Bayes' theorem [10].

In a BN, an edge between two variables implies a direct dependence between these two variables: one is the parent and the other is the child. It must provide the behavior of the child variable regarding the behavior of its Parents (if there



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Introduction

- Transformer fault diagnosis
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The transformer is an omnipresent equipment

Its failure influences the continuity of supply

Vital and not redundant



Optimize the availability

Rigorous monitoring of its behaviour

Predictive maintenance



Dissolved gas analysis (DGA)

≻Each gas has a percentage

Each percentage should not exceed a limited value



- ≻Key Gas Method
- ➢Dornenburg Ratio Method
- ≻Rogers Ratio Method
- ➤Method nomograph
- **≻**IEC Ratio Method
- >Duval Triangle Method

➤CIGRE Method



- ➤ Each anomaly gives a gas
- ≻A mixture of gas with different proportions
- ≻The proportions percentage depends on the anomaly nature

Artificial intelligence methods

Fuzzy Logic (FL)
Artificial Neural Networks (ANN)
Support Vector Machine (SVM)
Evolutionary Algorithms
Expert System

Bayesian networks tools

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Abnormal electrical or thermal stresses cause insulation oil to break down and to release small quantities of gases

DGA technology allows early fault detection and prevents other failures occurrence.

Various gases concentrations

Type of fault : Thermal or electricFault severity



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Traditional methods of DGA provide more details on fault nature by the use of ratios.



The existence of a gas is not always the result of a fault. In practice, there are several cases where gases are found but no fault is observed on the transformer; that is the case when gases remain trapped in the transformer.

DGA interpretation schemes are based on empirical assumptions and practical knowledge gathered by experts worldwide.

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For what reason we will use the Bayesian network?

Reliability of data :

•Non-compliance with sampling instructions

•There are several cases where the proportions of dissolved gases remain trapped in the transformer.

Effectiveness of the method :

•One traditional method of DGA does not enable the diagnosis of all faults *Rogers ratio method diagnoses only four faults*

•When several gas ratios occur, the problem becomes a problem of decision making

If interpretation schemes are not applied cautiously, they may incorrectly identify faults because they only indicate possible faults.

In some cases, DGA interpretation schemes may differ in terms of identified faults, which is clearly unacceptable for a reliable fault diagnosis system.

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Definition:

Bayesian network is defined by a directed acyclic graph (DAG). Where C (Vi) is the set of parents (or causes) of Vi in DAG.



The relationship between the nodes of the network is given by the following formula

$$p(V_1, V_2, \dots, V_n) = \prod_{i=1}^n p(V_i / C(V_i))$$

A Bayesian network for what?

- Knowledge modeling in an uncertain domain
- A simple graphical representation
- •A large variety of uses

 A Bayesian network is a tool allowing the visualization of variables and their dependencies (or independences)

•One model \rightarrow multiple uses :

oPrediction (variables influenced by those which are known)

oDiagnosis ("guess" the variables that influence those which are known)



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•the random variables are modeled as nodes

The edges between variables can explain a causal phenomenon (causal networks)

•Indicating an edge between two variables implies a direct dependence between these two variables: one is the parent and the other is the child



•To provide the behaviour of the child variable in view of the behaviour of his (or her if there are several) parents, each node has a conditional probability table.

•A conditional probability table associated with a node allows to quantify the effect of the parent nodes on this node.

•For root nodes (without parents), the probability table is not conditional and it fixed a priori probabilities.

 The Bayesian inference is to update the probability distribution to account for the new observations.

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Defining the structure of the BN is to identify the different types of faults associated with different key gases .

The second step is to find firstly the causal links between faults and key gases, and secondly between key gases and ratios, because ratios gases exist only if key gases exist.

The third step is to define the probabilities of occurrence of each fault based on the results of oil analysis and experience feedback.

$$p(Ratio_1) = \prod_{i=1}^{2} p(Ratio_i / Keygaz_i)$$



The most common gas ratio method is the Rogers ratio method

The Rogers method analyzes four gas ratios: CH4/H2, C2H6/CH4, C2H4/C2H6 and C2H2/C2H4.

The four detectable conditions of an oil-insulated transformer are normal ageing, partial discharge with or without tracking, and electrical and thermal faults of varying severity.

Fault	Key gas or Ratio gas	Code
Overheating of cellulose insulation	CO, CO2	F1
Electric discharge (corona effect, low partial discharge)	H2	F2
Electric fault (arc, spark, partial discharge with or without tracking)	C2H2, CH4, C2H4/C2H6, C2H6/CH4	F3
Thermal fault (low temperature overheating (150 to 300 °C), middle temperature overheating (300 to 700 °C) and high temperature overheating (700 °C).	C2H4, C2H6, C2H2/C2H4	F4
Normal ageing	CH4/H2	F5



Ratio gas	Value	Fault type
C2H2/C2H4	0.008	low or middle temperature overheating
CH4/H2	1.6	
C2H4/C2H6	1	



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Conclusion

- The probabilities given by the Bayesian network show that the model presented in the studied case diagnose five faults including healthy state.
- It supports various electrical and thermal severities.
- Compared to Rogers Ratio Method which diagnoses only four faults in contrast, the developed model can diagnose all faults.
- It is also possible to use all existing gas ratios in the same model.
- Each method of DGA uses a number of ratios but the Bayesian model developed in this study supports all ratios of gas and key gases which is a great advantage.



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Future work

Our future work will be focused on a new contribution concerning transformer faults diagnosis by multi information fusion based on operating information and Dissolved Gas Analysis Transformer fault diagnosis using Dissolved Gas Analysis technology and Bayesian networks



Thank you for your attention



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