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## **Abstract**

Roller bearing data has already been used in many publications, to evaluate and proof condition monitoring concepts (Marwala, 2012), (Nelvamondo, Marwala, & Mahola, 2006), (Li, Chow, Tipsuwan, & Hung, 2000). Most of these papers used both normal and outlier training data to evaluate the efficiency of supervised classifiers such as neural networks or support vector machines. Goal of this thesis is the design and evaluation of an exemplary semi-supervised condition monitoring approach, using the roller bearing data from (Case Western University) as benchmark.

## My contribution involves

- Research on *condition monitoring* and a concise presentation of a general condition monitoring approach (1)
- Research into *feature extraction* methods and selection of suitable methods with respect to the benchmark data set (2,4.2)
- Implementation of a Higuchi fractal dimension (HFD) feature extraction method (4.2.2)
- Research on *one-class-classification* and selection of suitable classifiers in relation with the benchmark data set and the available features (3,4)
- Design and implementation of a new one-class-classifier based on random forest
- Implementation and evaluation of a complete *semi-supervised condition monitoring* approach, including *feature extraction*, *feature selection*, classification and tests (4).

The first part of this thesis introduces several techniques, starting with *feature extraction* methods in chapter 2. Chapter 3 provides a general approach for the construction of *one-class-classifiers* and continues with an introduction of various *semi-supervised* classifiers which can be found in literature. The final part of this chapter discusses several aspects of the *random forest* classifier (3.7), of which a *semi-supervised* version was developed as part of this thesis (3.7.6), (4.3).

The techniques introduced in the first part, are evaluated in the second part of this thesis (4), which starts with a description of the *roller bearing* data set in 4.1. The *feature extraction* results are briefly discussed in 4.2 and 4.3 provides a proof of concept of the *semi-supervised random forest* classifier introduced in 3.7.6. Two experiments where designed to evaluate the complete *outlier detection* scenario, of which the first one (4.4.1) compares classifier performances, while the second analyzes the applicability of *random forest* for feature reduction purposes.