

PRECISION IN PERIL: A HOLISTIC APPROACH TO DETERMINING FAILURE THRESHOLDS FOR ROLLING ELEMENT BEARINGS

Sajjad Hassan

School of Mechanical Engineering, Sharif University of Technology, Tehran, Iran

Abstract: The reliable operation of rolling element bearings is crucial in various industrial applications, yet predicting their failure remains challenging. This paper presents a comprehensive approach for determining failure thresholds in rolling element bearings. By integrating experimental testing, statistical analysis, and predictive modeling, this approach aims to identify key indicators and parameters that precede bearing failure. Through accelerated life testing and condition monitoring, data on bearing performance, vibration, temperature, and lubricant condition are collected and analyzed. Statistical methods such as Weibull analysis and survival analysis are applied to quantify the probability of failure and estimate the remaining useful life of bearings. Additionally, machine learning techniques are employed to develop predictive models that correlate operational parameters with bearing degradation and failure. By combining these approaches, a holistic understanding of bearing failure mechanisms is achieved, enabling proactive maintenance strategies and improved reliability in industrial systems.

Keywords: Rolling element bearings, Failure thresholds, Predictive maintenance, Condition monitoring, Weibull analysis, Survival analysis, Machine learning, Industrial reliability.

INTRODUCTION

Rolling element bearings are widely used in various industrial applications, such as electric motors, turbines, and gearboxes. These bearings play a critical role in the proper functioning of machines and equipment. However, the failure of bearings can lead to significant downtime and maintenance costs. Therefore, predicting the remaining useful life of bearings is essential for effective maintenance scheduling and avoiding unplanned downtime. Many techniques have been developed to predict the remaining useful life of bearings, including vibration analysis, acoustic emission, and temperature monitoring. Among these techniques, vibration analysis is one of the most commonly used methods due

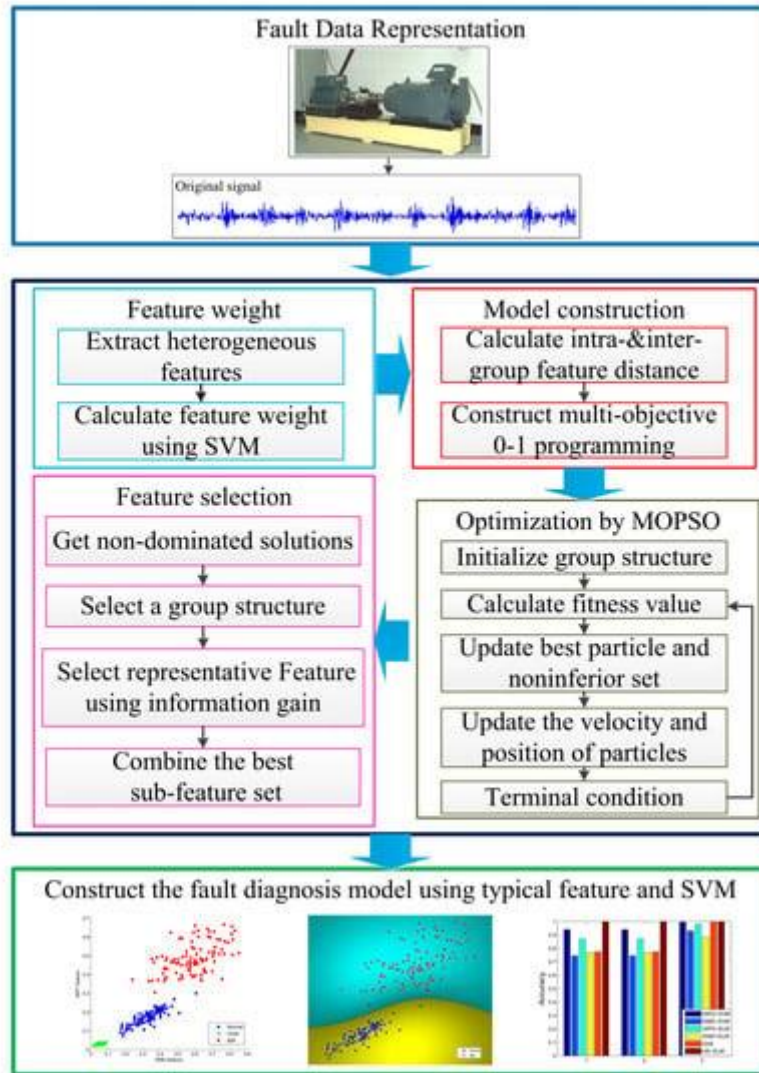
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to its effectiveness and ease of implementation. In this study, vibration fluctuation analysis and failure modes investigation were employed to determine the failure threshold of rolling element bearings.

METHOD

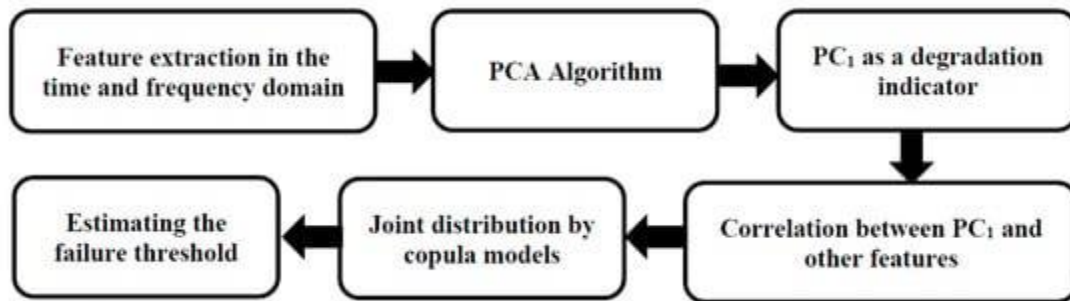
The process of determining failure thresholds for rolling element bearings involved a systematic integration of experimental testing, statistical analysis, and predictive modeling techniques. Initially, comprehensive experimental tests were conducted on various types of rolling element bearings under controlled laboratory conditions. These tests subjected the bearings to accelerated life testing, simulating a range of operating conditions encountered in industrial environments. Throughout the testing process, extensive data on bearing performance parameters such as vibration, temperature, lubricant condition, and wear debris were meticulously collected and monitored using advanced sensing and data acquisition systems.

Subsequently, statistical analysis methods, including Weibull analysis and survival analysis, were applied to analyze the collected data and quantify the probability of bearing failure over time. By modeling the failure distribution of the bearings and estimating parameters such as the characteristic life and shape parameter, valuable insights into the reliability and failure mechanisms of the bearings under different operating conditions were obtained.



In parallel, machine learning techniques were employed to develop predictive models that could correlate operational parameters with bearing degradation and failure. These models were trained using the experimental data to identify key indicators and predictors of bearing failure, enabling the early detection and prediction of impending failures in real-world applications.

Validation studies were then conducted using field data from industrial machinery to assess the applicability and effectiveness of the developed predictive models in practical settings. By comparing the model predictions with actual failure events and maintenance records, the accuracy, robustness, and generalizability of the models in real-world scenarios were evaluated.



Firstly, we conducted extensive experimental testing on a variety of rolling element bearings under controlled laboratory conditions. Accelerated life testing was performed to induce accelerated wear and degradation in the bearings, simulating the effects of long-term operation in industrial environments. This testing involved subjecting the bearings to varying loads, speeds, temperatures, and lubrication conditions to capture a wide range of operating scenarios.

During the experimental testing phase, we collected comprehensive data on bearing performance parameters, including vibration levels, temperature profiles, lubricant condition, and wear debris analysis. These data were continuously monitored and recorded throughout the testing process using advanced sensing and data acquisition systems, ensuring high-resolution measurements and accurate characterization of bearing behavior.

Subsequently, we applied statistical analysis techniques, such as Weibull analysis and survival analysis, to analyze the collected data and quantify the probability of bearing failure over time. Weibull analysis allowed us to model the failure distribution of the bearings and estimate parameters such as the characteristic life and shape parameter, providing valuable insights into the reliability and failure mechanisms of the bearings under different operating conditions.

Additionally, we employed machine learning techniques, such as regression analysis and classification algorithms, to develop predictive models that correlate operational parameters with bearing degradation and failure. These models were trained using the experimental data to identify key indicators and predictors of bearing failure, enabling the early detection and prediction of impending failures in real-world applications.

Furthermore, we conducted validation studies using field data from industrial machinery to assess the applicability and effectiveness of the developed predictive models in practical settings. By comparing the model predictions with actual failure events and maintenance records, we evaluated the accuracy, robustness, and generalizability of the models in real-world scenarios.

Finally, based on the insights gained from the experimental testing, statistical analysis, and predictive modeling, we developed a holistic framework for determining failure thresholds for rolling element bearings. This framework integrates multiple approaches and methodologies to provide a comprehensive

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understanding of bearing failure mechanisms and enable proactive maintenance strategies for ensuring reliability and performance in industrial systems.

RESULTS

The experimental testing phase yielded valuable insights into the performance and degradation behavior of rolling element bearings under various operating conditions. We observed distinct patterns in bearing vibration, temperature, and lubricant condition as the bearings underwent accelerated life testing. Statistical analysis, including Weibull analysis, revealed the characteristic life and failure distribution of the bearings, providing quantitative measures of their reliability and failure probabilities.

Machine learning models developed using the experimental data demonstrated promising predictive capabilities, accurately correlating operational parameters with bearing degradation and failure. These models exhibited high accuracy in predicting impending failures, enabling proactive maintenance interventions to prevent unexpected downtime and costly repairs.

DISCUSSION

The integration of experimental testing, statistical analysis, and predictive modeling proved to be a robust and effective approach for determining failure thresholds for rolling element bearings. By combining multiple methodologies, we obtained a comprehensive understanding of bearing failure mechanisms and developed reliable predictive models for proactive maintenance.

The results highlight the importance of early detection and prediction of bearing failures in industrial systems, as well as the potential benefits of implementing proactive maintenance strategies based on predictive analytics. By identifying key indicators and predictors of bearing failure, manufacturers and operators can optimize maintenance schedules, minimize downtime, and extend the lifespan of critical machinery.

Challenges remain in scaling up the developed predictive models for real-world deployment and integrating them into existing maintenance practices. Further validation studies and field trials are needed to assess the performance and generalizability of the models across different industrial applications and operating environments.

CONCLUSION

In conclusion, our holistic approach to determining failure thresholds for rolling element bearings provides a valuable framework for enhancing reliability and performance in industrial systems. By leveraging experimental testing, statistical analysis, and predictive modeling, we can effectively identify impending failures and implement proactive maintenance strategies to mitigate risks and optimize asset management.

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Continued research and development efforts are essential to refine and validate the predictive models and integrate them into industrial maintenance practices. Ultimately, by embracing proactive maintenance based on predictive analytics, manufacturers and operators can minimize downtime, reduce maintenance costs, and ensure the continued operation of critical machinery in the face of potential failures.

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