

Severity Estimation of Faulty Bearings Based on Strain Signals From Physical Models and FBG Measurements

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ABSTRACT

Condition based maintenance (CBM) is the preferred approach in rotating machinery and aim to replace the commonly used approach of maintenance based on service time. To achieve an effective CBM, different types of sensors should be placed in the system for condition monitoring to detect the location of the fault and its severity. In this research, a Fiber Bragg Grating (FBG) has been used for condition monitoring on spalls in deep groove ball bearings. The motivation for using these sensors is the ability to get a high-noise signal (SNR) ratio. The usage of FBG sensors is relatively new for health monitoring systems of rotating machinery. Therefore, there is not enough understanding of the strain signature measured by the FBG. To examine the phenomena in the strain signals, a physics-based model of the strain signature has been developed. In this model, two complementary models were integrated, a finite element (FE) model and a dynamic model. The strain model describes the interaction between the rolling elements (REs) and the bearing housing and simulates the strain behavior measured on the bearing housing. The simulation results are validated with strain signals measured by the FBG sensor at different stages of an endurance test. The model allows simulation of a wide range of spall lengths and describes the behavior of the strain signals for different levels of misalignment. The insights from the model enabled the development of an automatic algorithm that assess the severity of the

defect and to track spall length during bearing operation, based on strain signals.

1. INTRODUCTION

Bearings are important components in rotating machinery to enable smooth operation. Failure in one of the bearings can lead to severe damage and even total failure. The most common failure mechanism in bearings is spall formation (Jalalahmadi *et al.*, 2009), where metallic particle flakes are released from the surface of the bearing raceways and/or the rolling elements (REs) (Raje, Slack and Sadeghi, 2009; Rosado *et al.*, 2010; Gazizulin, Klein and Bortman, 2017). Interactions with a defective component within a bearing lead to undesirable vibrations which increase as the defect evolves. Fault detection and diagnosis, based on vibration signals (such as, accelerometers, load cell, acoustic emission etc.), have been widely studied in the literature (Qing *et al.*, 1991; Zhen *et al.*, 2008; Eftekharnjad *et al.*, 2010; Randall and Rôme Antoni, 2010; Randall, 2011; Cui *et al.*, 2015; El-Thalji and Jantunen, 2015; French and Hannon, 2015; W.S. Siew, W.A. Smith, Z. Peng, 2015; Wang, Sawalhi and Becker, 2016; Tarawneh *et al.*, 2019; Zhang *et al.*, 2021, 2022). This study examines strain signals measured with Fiber Bragg Grating (FBG). FBG are small, flexible and gives a high signal to noise ratio (SNR).

Fault diagnosis of bearings via FBG sensors has not been thoroughly studied in the literature (Khmelnitsky *et al.*, 2015; Wei *et al.*, 2016; Alian *et al.*, 2019). Numerous researchers have developed physics-based models to investigate the RE-edge impact reaction (Arakere *et al.*, 2009; Branch *et al.*, 2009, 2013;

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Gazizulin *et al.*, 2020). Branch *et al.* (Branch *et al.*, 2013) presents a FE analysis of RE-spall impact to determine the relative contributions of four bearings' properties on spall propagation: (i) ball material density, (ii) subsurface residual stress, (iii) gradient in yield strength with depth (case hardening), and (iv) raceway surface hardness/yield strength that are thought to affect spall propagation. None of these studies investigated the strain states in a remote area from the spall. A fundamental study to examine the physical phenomenon in the strain signal, during the interaction of the RE with the damaged part of the bearing, is presented. In this study, a strain model has been developed and compared to the experimental strain signal measured by FBG sensors. This article deals with small spalls, such that the RE meets the spall edge before it collides with the floor of the spall.

The paper is organized as follows. Section 2 presents the theoretical background. Section 3 presents the experimental setup. Section 4 presents the strain model. Section 5 presents a comparison between simulations and experimental results and Section 6 presents the algorithm for spall length estimation.

2. BACKGROUND

Strain signal has lack of interpretation in the literature. For this purpose, two complementary models were integrated into a strain model: a FE model and a dynamic model. A strain-contact force relation was developed using the FE model. This relation is multiplied by the contact force from the dynamic model to simulate the real strain behavior of the bearing house. The simulated strain and the strain signals measured by the FBG in the endurance test were compared to validate the model and to explain different phenomena detected in the strain signals. The flowchart methodology presented in Figure 1. Based on the strain model insight, a spall size estimation algorithm was developed.

2.1. FBG Signal

A FBG sensor is made by the inscription of periodic variations in the index of refraction along a short section of the core of a single-mode optical fiber (Hill and Meltz, 1997). The strain signals measured by the FBG sensor, given by

$$\varepsilon = \frac{\lambda_b - \lambda_0}{C_\varepsilon \lambda_0} \quad (1)$$

Where λ_b is the light reflection in a very narrowband of wavelengths around the Bragg wavelength of peak reflection, λ_0 and C_ε is a material constant.

The usage of FBG sensors is motivated by the ability to install them close to the tested bearing, which gives a high signal to noise ratio (SNR).

2.2. Data Analysis

The strain signal in the time domain was synchronized with the rotational speed signal, by angular resampling, and was transformed to the cycle domain. A de-phase algorithm (Klein *et al.*, 2011) was applied to the strain in the cycle domain to remove the synchronous elements, e.g. shaft, that were masking the effect of the bearing in the signal. First, the synchronous average (SA) is calculated by averaging the resampled signal over a cycle of rotation. The de-phased signal is obtained by removing the SA from the angular resampled signal, leaving only the asynchronous elements. The data analysis process of the strain signal is presented in Figure 2.

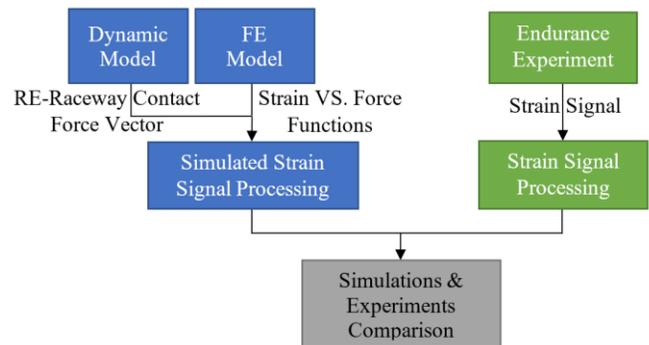


Figure 1: Research flowchart structure

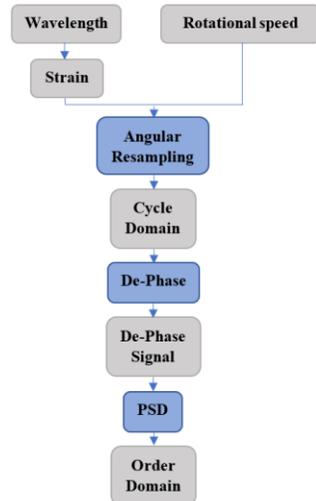


Figure 2: Strain signal analysis process

3. ENDURANCE TEST

The experimental setup for the endurance test consists of a test rig and a measurement unit. The test rig includes a shaft, two support bearings, the tested bearing (6206 ETN9) and a pneumatic piston for applying different loads. All components are listed in Table 1 and marked in Figure 3. All the bearings were lubricated. A FBG sensor mounted along the bearing house of the tested bearing, see Figure 3. The sensor is connected to a Smart Fiber™ “SmartScan Aero Mini” interrogator, sampling data at 10[kS/s]. To accelerate the defect initiation process, a small bore was seeded into the outer raceway of the bearing using an electrical discharge machining (EDM). The initial defect was in the center of the loading zone. The operational conditions during the test were constant with a load of 2.2 kN and the rotation speed was 35 Hz.

Figure 4 presents the strain de-phased signal at three different stages of the experiment: green signal-at the beginning of the experiment, the blue signal-at the middle of the experiment and the red signal-at the end of the experiment. All three of the signals, have a form of a periodic signal wave with BPFO frequency. An additional pulse was observed at the advanced stages of the experiment. In addition, the pulse widths and amplitude levels are higher for advanced stages of the experiment with wider spalls.

Table 1: Experimental system components

Component number	Component	Amount
1,2	Supporting bearing SYF 30 TF	2
3	Coupling	1
4	Shaft 30 mm diameter	1
5	Bearing house + tested bearing 6206 ETN9	1
6	Gear 20 teeth 2.5 module	1
7	Pneumatic piston	1
8	Three-phase motor	1

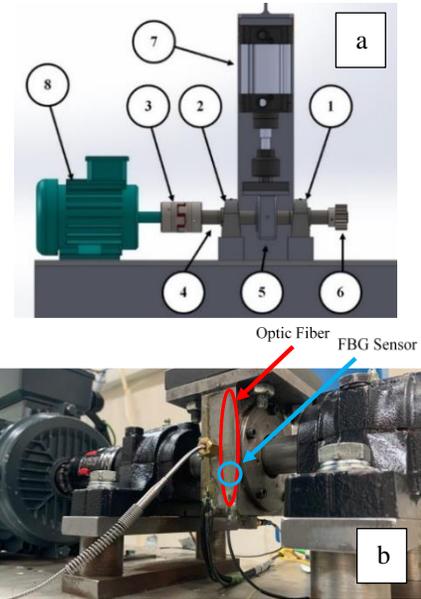


Figure 3: Endurance test rig: (a) Illustration of the experimental system. The component details are listed in Table 1. (b) Location of the FBG sensor.

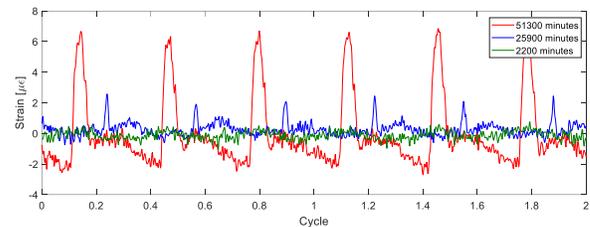


Figure 4: Strain de-phased signal at different stages during the endurance experiment.

4. STRAIN SIGNAL MODEL

To analyze the behavior of the measured FBG strain signals for healthy and faulty bearings, the strain model integrates between a dynamic model and a FE model as illustrated in the research flowchart diagram presented in Figure 1.

4.1. Non-Linear Dynamic Model

A validated non-linear dynamic model was used to calculate the contact force between the RE and the outer raceway generated during the operation of both a healthy and a faulty bearing. The non-linear dynamic model developed by Kogan et al. (Kogan *et al.*, 2015; Kogan, Bortman and Klein, 2017) is based on the classic kinematics and the dynamic equations. The boundary conditions for the dynamic model were defined according to the endurance experimental setup. Two different spall lengths were examined: 2mm and 4.8 mm. An example of the contact force of a faulty

bearings with a spall length of 2mm versus the rotation angle, θ , (as illustrated in Figure 6a), of one RE is presented in Figure 6b. In Figure 6d, the green dot, and the red dot in the $\theta=270^\circ$ region, represents the entry and exit of the RE into and from the spall, respectively. The high frequency appears in the signal is a result of the RE-spall interaction at the trailing edge. The zero force in this region represent the RE disconnection from the raceways (destressing), i.e., free flight of the RE.

4.2. FE Model

A FE model was built to study the relation between (i) the contact force between the REs and the outer raceway, and (ii) the strain developed within the bearing house. The strain response location was chosen to be the same as the FBG sensor located in the test rig. A quasi-static, two-dimensional, plane strain model was developed in ABAQUS software with the ABAQUS/standard solver. The assembly included a single RE and a bearing house (Figure 5a). The FE model geometry and dimensions presented in Figure 5a and in Table 2. The Young’s modulus was taken as $E = 208 \text{ MPa}$ and the Poisson ratio was $\nu = 0.3$. An eight-node biquadratic element was used (CPE8R). A fine mesh was used in the vicinity of the contact zone and became coarser as the distance from the contact zone increased. The narrow edge was fixed at the top surface (see Figure 5a). The RE was pressed, with constant force, against the bearing house. Then, the RE was rotated 360 degrees counterclockwise with angle increments of 0.005 radians ($\Delta\theta = 0.005 \text{ rad}$) until it returned to the starting point. The simulated strain from the FE model for constant force is shown in Figure 5b. Since the contact force between the RE and the outer raceway is changing during the rotation of the RE, i.e., inside, and outside of the loading zone, the integration of the dynamic model is essential. A linear correlation between the FE strain and the constant force acting on the RE, was sought regarding the fact that, the measurement of the strain was at a far distance from the contact zone and the material is linear elastic. It is noteworthy to mention that the linear function was calculated for every angle increment.

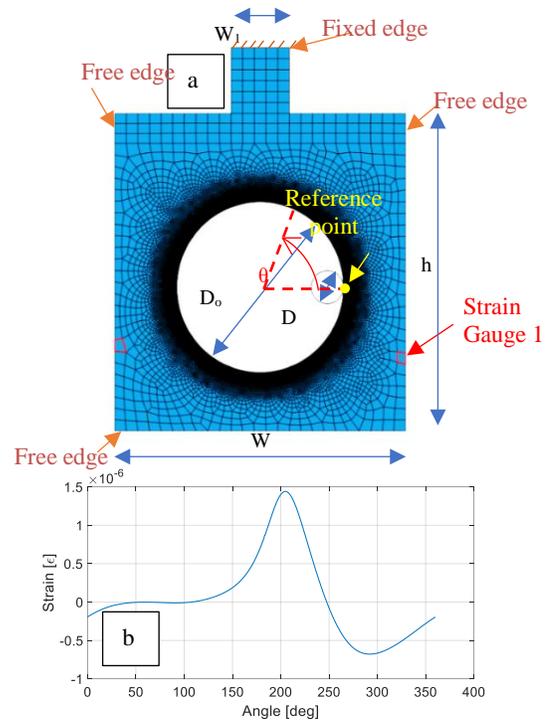


Figure 5: (a) FE model – dimensions. (b) Simulated strain results obtained from the FE model after applying a constant force of $F=100 \text{ [N]}$ on the RE

Table 2: Model dimensions

Parameter	Comment	Value	Units
h	Bearing house height	106	mm
w	Bearing house width	100	mm
d	Bearing house depth	32	mm
D	RE diameter	11	mm
D_o	Outer race diameter	57	mm
w_1	Piston axis diameter	20	mm
N	RE number	8	-

4.3. Simulated Strain Signal

The simulated strain signal has obtained by multiplying the contact force from the dynamic model (Section 4.1) and the slope of the linear functions from the FE model (Section 4.2) for every angle increment. An example of the simulated strain of faulty bearing with 2mm spall length is shown in Figure 6c and e. The combined effect of (i) the behavior of the contact force distribution (Figure 6b), and (ii) the behavior of the simulated strain from the FE model (Figure 5b) is observed. The

calculated strain in the faulty state (Figure 6e) has the same shape as the contact force signal in the RE-spall interaction (Figure 6d), which means that the disconnection of the RE from the raceways can also be observed in the strain signals.

5. RESULTS

The model was validated by comparing the simulation results, Figure 7a, to the experiments, Figure 7b. The green curve in both graphs represent a healthy bearing and the red curve represent a faulty bearing with 4.8mm spall length. There is good agreement between the

results of the simulation and the experiments, both in behavior and in scale. The simulated signal is biased, because the model simulates the total strain including the loading effect while the FBG sensor measures only the strain fluctuations. As illustrated in Figure 8, when a single RE reach to the spall leading edge it may impact different locations with the trailing edge. In this Figure, two examples of different impact locations are shown. For each one of the impact a different pulse width will be measured. Although the pulse width represents very well the hovering of the RE, it can give a good estimation of the spall length.

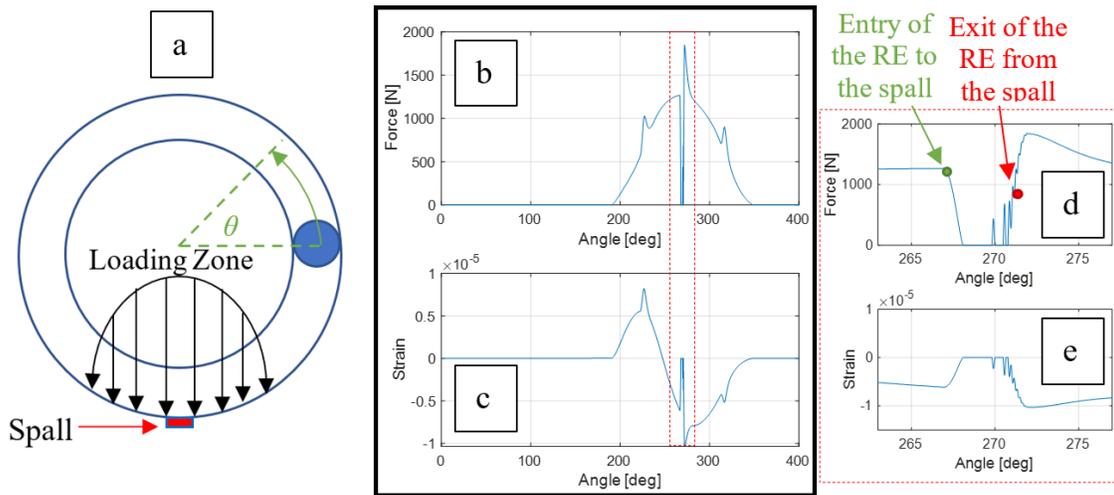


Figure 6: (a) Schematic description of the bearing from the endurance test: the loading zone is depicted by the black curve and the location of the spall can be seen in red rectangle. The angle of rotation, θ , relative to the reference point (b) contact force of faulty bearing with 2mm spall length, and (c) simulated strain signal in strain gauge 1 (Figure 5a) for faulty bearing (2mm spall), (d) and (e) close-up of the RE-spall interaction of the contact force and the simulated strain signal respectively

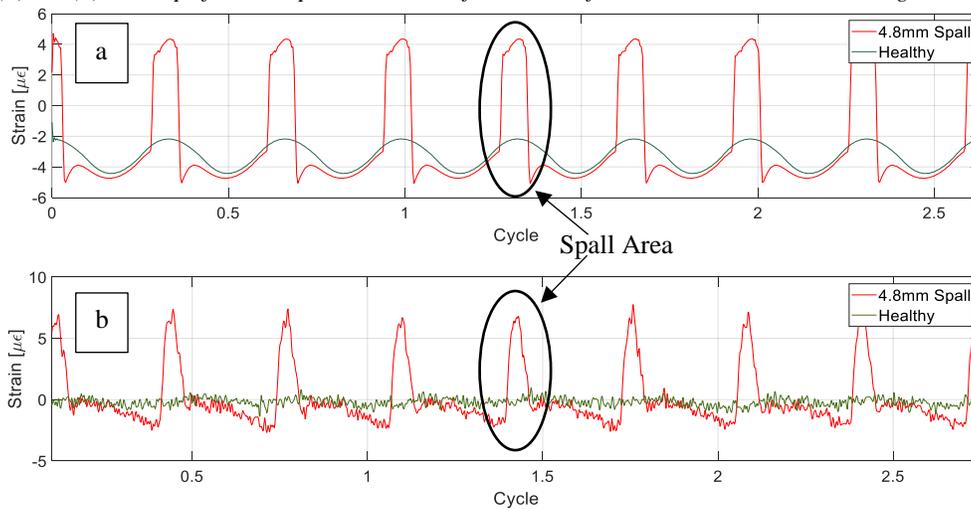


Figure 7: Comparison between: (a) simulation and (b) experimental endurance test (healthy - green and 4.8mm spall - red)

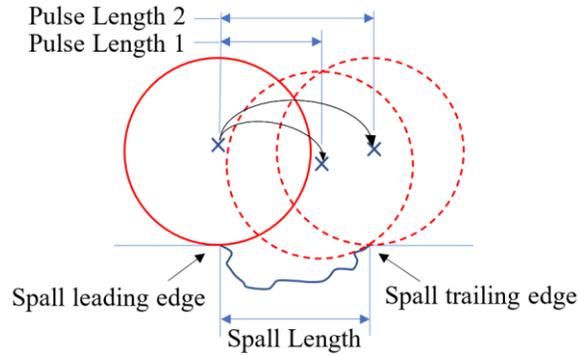


Figure 8: Illustration of a RE impact in two different possible locations at the spall edge in the experiments

6. SPALL LENGTH ESTIMATION ALGORITHM DESCRIPTION

The spall length estimation algorithm is divided into two main stages and presented in Figure 9: (i) selection of the RE-spall interactions and (ii) evaluation of the pulse width that represents the RE flight, based on the Defect Size Ratio (DSR) equation (Alian *et al.*, 2019):

$$DSR = \frac{\Delta x}{X} = \frac{\Delta C}{C} \quad (2)$$

Where Δx is the spall length, X is the distance between two RE, ΔC is the pulse width in the cycle domain and C is the distance between two pulses in the cycle domain.

The selection of the RE-spall interactions is performed on the squared de-phased strain signal, x^2 , in order to emphasize the pulses in the strain signal. Since the spall is located on the outer raceway of the bearing, it is possible to detect BPFO interactions in one cycle. A dynamic threshold, which is a percentage of the highest value in the signal, is used to determine which peaks represent the correct interactions. After the RE-spall interaction has been detected, the algorithm searches the entrance and the exit points of the RE into the spall by defining the first negative minimum adjacent to the peak of the pulse. The evaluated pulse width, that it is assumed to be the spall length, is obtained by the average of all the estimated pulse widths.

The pulses started to be observed in the signals after 24,000 minutes, therefore, the algorithm was applied to all the strain de-phased signals after 24,000 minutes. Figure 10 presents the algorithm output which is a trend line of the estimated spall length. The trend line is approximately monotonic as expected. The minimal spall length that could be estimated by the algorithm was 1.6 mm. After the bearing was disassembled, a 5.1 mm spall was measured, while the estimated fault

length was 5.4 mm. The approximation error is close to 6%.

This work focuses on outer ring defects. For further research, a general algorithm (both the outer and inner rings defects and rolling elements defects) will enable to develop an industrial application.

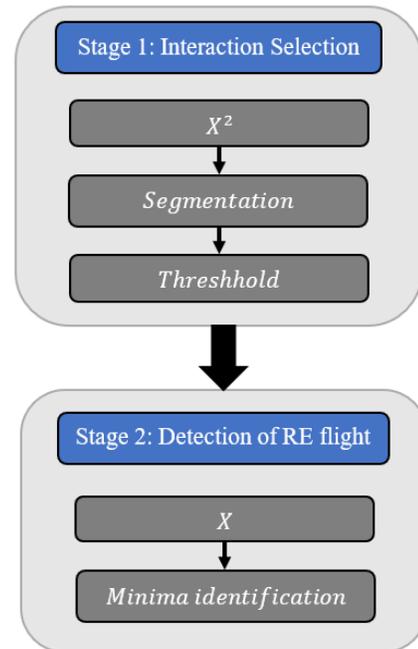


Figure 9: Algorithm stages for spall length estimation in the bearing outer race for the strain signal

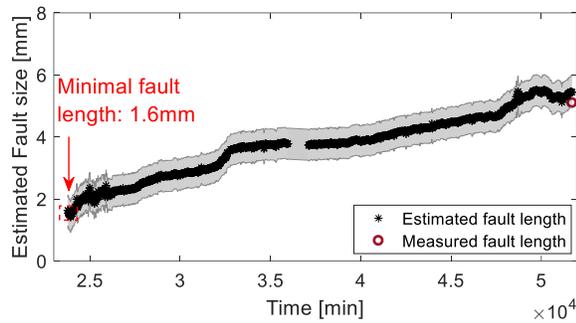


Figure 10: Trend line of the fault length estimation during the endurance experiment

7. SUMMARY AND CONCLUSIONS

This study examines the strain signals of healthy and faulty bearings, for diagnostic capabilities via FBG sensors. The theoretical analysis is based on two complementary models that were integrated into a strain model. The validation of the model has been done by comparing the simulation results with the endurance test results. Based on the model insights, it was possible to determine that the additional pulses correspond to the free flight of the ball over the spalled area. In addition, the pulse width is directly related to the arc length of the free flight of the ball. Based on the strain model, a reliable algorithm for spall length estimation has been developed.

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Prof. Jacob Bortman joined the academic faculty of Ben-Gurion University of the Negev in September 2010 as a full Professor. Prof. Bortman spent thirty years in the Israel Air Force (IAF), retiring with rank of Brigadier General. His areas of research in the Dept. of Mechanical Engineering include: Health usage monitoring systems (HUMS); Conditioned based maintenance (CBM); Usage and fatigue damage survey; Finite Element Method; and Composite materials.