

Bergische Universität Wuppertal

Fachbereich Mathematik und Naturwissenschaften

Institute of Mathematical Modelling, Analysis and Computational Mathematics (IMACM)

Preprint BUW-IMACM 13/20

S. Schmitz, H. Gottschalk, G. Rollmann and R. Krause

Probabilistic Analysis of LCF Crack Initiation Life of a Turbine Blade under Thermomechanical Loading

October 2, 2013

http://www.math.uni-wuppertal.de

Probabilistic Analysis of LCF Crack Initiation Life of a Turbine Blade under Thermomechanical Loading

S. Schmitz^{1, 2}, G. Rollmann¹, H. Gottschalk³, R. Krause²
¹ Siemens AG, Mellinghoferstraße 55, 45473 Mülheim an der Ruhr, Germany
² Universita della Svizerra Italiana, Institute of Computational Science,
Via Giuseppe Buffi 13, 6904 Lugano, Switzerland
³ Bergische Universität Wuppertal,
Department of Mathematics and Computer Science, Gaußstraße 20, 42097 Wuppertal, Germany

October 2, 2013

Abstract

An accurate assessment for fatigue damage as a function of activation and deactivation cycles is vital for the design of many engineering parts. In this paper we extend the probabilistic and local approach to this problem proposed in [1], [2] and [3] to the case of non-constant temperature fields and thermomechanical loading. The method has been implemented as a finite element postprocessor and applied to an example case of a gas-turbine blade which is made of a conventionally cast nickel base superalloy.

Keywords: Fatigue Crack Initiation; Probabilistic Fatigue; FE Analysis; Hazard Function;

1 Introduction

The necessity for a flexible service of a lot of engineering parts such as gas turbines leads to the importance of fatigue analysis, where probabilistic models can be very valuable. In this work, we present a probabilistic model for low-cycle fatigue (LCF) which can be derived from the Poisson point process or from a spatial hazard approach, confer [1] and [2], respectively. The model can be applied to polycrystalline metal which is sufficiently finegrained so that isotropic material behavior can be assumed and continuum mechanics can be employed. Here, failure of a component is defined to be given by the initiation of the first LCF crack. The probabilistic model yields the probability of failure (PoF) as a function of the number of load cycles. We have extended the model proposed in [1], [2] and [3] by a temperature model of the LCF parameters as well as the percentile bootstrap method in order to be able to consider uncertainties due to LCF test data. In contrast to the deterministic safe-life approach [4], this extended probabilistic LCF model takes inhomogeneous temperature and strain fields, size effects and uncertainties due to specific calibration data into account.

We apply the probabilistic model to a gas-turbine blade which is subjected to thermomechanical loading during the operating state. In this case, the corresponding LCF failure mechanism is surface driven. Having computed the total PoF we also consider and visualize the hazard density on the blade's surface.

2 A Probabilistic Model for LCF

In the following, we first briefly revisit the spatial hazard approach for surface driven LCF presented in [2] and [3] and consider LCF as a failure-time process. If N denotes the random variable which represents the cycle of first crack initiation and P the underlying probability measure, the hazard rate is defined by

$$h(n) = \lim_{\Delta n \to 0} \frac{P(n < N \le n + \Delta n | N > n)}{\Delta n} = \frac{f_N(n)}{1 - F_N(n)},\tag{1}$$

confer [5]. Here, we model N as a continuous random variable in agreement with the literature, confer [4] and [6]. $F_N(n) = P(N \le n)$ is the cumulative distribution function and $f_N(n) = dF_N(n)/dn$ the corresponding density function. The hazard rate h is also called instantaneous failure rate as for a small step Δn the expression $h(n) \cdot \Delta n$ is an approximation for the propensity of failure in the next time step Δn , given no failure to time n.

Considering strain controlled LCF failure mechanism on a component which is made of polycrystalline metal and represented by a domain Ω , we assume according to [2] and [3] that the surface zone which is affected from the crack initiation process of a single LCF crack is small with respect to the surface of the component. Thus, we suppose that in any subregion Aof the component's surface $\partial\Omega$, the corresponding hazard rate h_A is a local functional of the displacement field **u** and the temperature field T in that particular region with

$$h_A(n) = \int_A \rho(n; \nabla \mathbf{u}, T) \, dA. \tag{2}$$

Here, $\nabla \mathbf{u}$ is the Jacobian matrix of \mathbf{u} . We call the integrand ρ hazard density function which is the core of this spatial hazard approach. For inhomogeneous strain fields $\varepsilon_a = \varepsilon_a(\nabla \mathbf{u}, T)$ we obtain $h(n) = \int_{\partial \Omega} \rho(n; \varepsilon_a, T) \, dA$.

Taking $F_N(n) = 1 - \exp\left(-\int_0^n h(s) \, ds\right)$ into account the probability of LCF crack initiation on the surface $\partial\Omega$ until cycle *n* is given by

$$F_N(n) = 1 - \exp\left(-\int_0^n \int_{\partial\Omega} \rho(s;\varepsilon_a,T) \, dAds\right). \tag{3}$$

In Section 3 the hazard density function ρ will be employed to identify the critical and possibly overengineered regions of the component. Considering the statistical evaluations in [1] and following [2] and [8] we assume that the number N of cycles to crack initiation are Weibull distributed. Therefore, we choose the Weibull hazard ansatz

$$\rho(n;\varepsilon_a,T) = \frac{m}{N_{\det}(\varepsilon_a,T)} \left(\frac{n}{N_{\det}(\varepsilon_a,T)}\right)^{m-1},\tag{4}$$

where *m* is the Weibull shape and $N_{det}(\varepsilon_a, T)$ determines the corresponding Weibull scale parameter. If the Coffin-Manson-Basquin (CMB) equation – confer [4] and [6] – is taken for the strain-life relationship the scale field $N_{det}(\mathbf{x}) = N_{det}(\varepsilon_a(\mathbf{x}), T(\mathbf{x}))$ is given by the solution of

$$\varepsilon_a(\mathbf{x}) = \frac{\sigma'_f(T(\mathbf{x}))}{E(T(\mathbf{x}))} (2N_{\text{det}}(\mathbf{x}))^{b(T(\mathbf{x}))} + \varepsilon'_f(T(\mathbf{x})) (2N_{\text{det}}(\mathbf{x}))^{c(T(\mathbf{x}))}$$
(5)

on every point $\mathbf{x} \in \partial \Omega$, where the surface $\partial \Omega$ is subjected to an equivalent strain field $\varepsilon_a(\mathbf{x})$ and a temperature field $T(\mathbf{x})$. Having chosen an appropriate temperature model¹ for the CMB parameters $\sigma'_f, b, \varepsilon'_f, c$ and for Young's modulus E, the probabilistic model for LCF is given by the Weibull distribution

$$F_N(n) = 1 - \exp\left[-\left(\frac{n}{\eta}\right)^m\right] \quad \text{for scale} \quad \eta = \left(\int_{\partial\Omega} \frac{1}{N_{\text{det}}^m} dA\right)^{-1/m} \tag{6}$$

and for some shape parameter $m \ge 1$, which yields the probability for LCF crack initiation to cycle n.

The model can be calibrated by means of usual maximum likelihood methods, confer [5] and [9], for example. The parameter m determines the scatter of the distribution where small values for $m \ge 1$ correspond to a large scatter.

The CMB parameters of the model are not the same as obtained from fitting standard specimen data. Due to the size effect, see [6], the original CMB approach leads to different values of the parameters for specimens under the same temperature and strain conditions but with different gauge

¹In this work we use a proprietary temperature model by Siemens AG.

areas. The new and more physical interpretation of these parameters – in context of the probabilistic model according to [1] and [2] – takes the size effect and inhomogeneous strain and temperature fields into account, so that the CMB parameters can be calibrated with LCF-test results of specimens with arbitrary geometry² and under arbitrary strain and temperature fields. Thus, these newly interpreted parameters can be assigned to every such geometry.

3 LCF Crack Initiation Life of a Turbine Blade

In this section we consider a turbine blade which is made of a polycrystalline cast nickel-based superalloy such as RENE 80. The following probabilistic analysis of its LCF crack initiation life is based on an FEA model for the operating state and on the Weibull distribution (6).

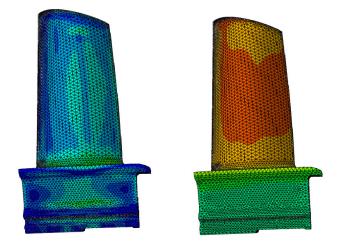


Figure 1: FEA results of Abaqus 6.9-2 for the von Mises stress field (left) and temperature field (right) of the turbine blade.

The temperature and strain field of the turbine blade in the operating state is computed by means of a thermoelastic FEA-model within Abaqus 6.9-2 and of the Neuber shakedown method³ which considers plasticity. The model includes approximately 190,000 tetrahedral, affine Lagrange elements of Serendipity class with 10 nodes. Figure 1 shows the von Mises stress and temperature field in the operating state. In the shutdown state the von Mises stress is everywhere zero and the temperature field is set equal to that one of the operating state which is an conservative approximation and avoids

 $^{^{2}}$ The specimen must consist of sufficiently many grains so that continuum mechanics can be applied. Moreover, information on when the first LCF crack initiation occurred has to be provided which can be practically difficult, however.

³Confer [4] and [7].

the treatment of thermo-mechanical fatigue (TMF). The transition from the shutdown state to the operating state and then back to the shutdown state is considered as one load cycle. It is further assumed that the shutdown and operating state stay the same during the cycles.

The probabilistic LCF model has been calibrated with strain controlled LCF test results⁴ for standard specimens which were subjected to different temperatures and strain amplitudes. The maximum likelihood method has been used for the calibration which constitutes a statistical estimation of the model parameters, confer [5]. This estimation in conjunction with computing the surface integral (6) results in a value for the Weibull scale parameter η which we call the maximum likelihood value for η .

As the estimation depends on the LCF test results there are uncertainties for the values of the model parameters. This affects the total PoF with respect to LCF crack initiation. We employ the fully parametric bootstrap sampling procedure in conjunction with the percentile method – confer [5] – to consider theses additional uncertainties. We used 2,000 bootstrap samples which were obtained from the maximum likelihood estimation. These samples are different parameter realizations of the probabilistic model for LCF and describe the distributions of the model parameters. Using these parameter realizations in conjunction with the FEA postprocessing results in 2,000 values for the Weibull shape and scale parameter. Then, the law of total probability yields the total PoF with respect to LCF crack initiation. Note that computationally most expensive parts of the FEA postprocessing are identifying the blade's surface and computing the surface integral in (6) for the bootstrap samples.

Figure 2 shows cumulative Weibull distributions corresponding to 72 of the 2,000 bootstrap samples. Each black curve is one parameter realization of the probabilistic model and yields different values for the PoF depending on multiples N^* of the maximum likelihood value of η . The law of total probability results in the total PoF (red curve). For $N^* = 0.0796$ the total PoF is 3.032%, for example. From a design perspective one decides which PoF is acceptable and then chooses the corresponding number of allowable shutdown and operating cycles. Figure 3 shows the hazard density on the turbine's surface, where the red regions are areas with higher risk for crack initiation. Blue regions in some cases may indicate overengineered areas of the turbine blade. To obtain a final judgment on overengineered areas a more complete design perspective is needed which includes performance and efficiency criteria, for example.

⁴The results were provided by Siemens AG.

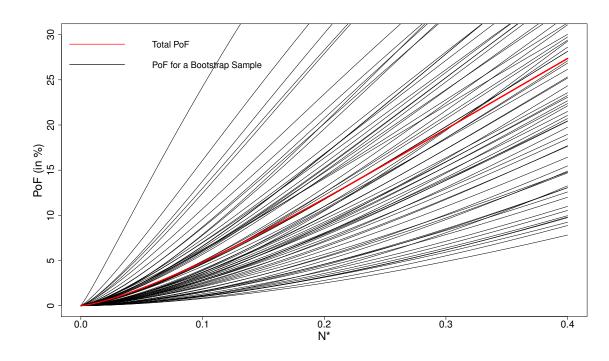


Figure 2: Total PoF (red) due to LCF crack initiation and single PoF curves (cumulative Weibull distributions, black) corresponding to 72 of the 2,000 bootstrap samples. N^* are multiples of the maximum likelihood value for the Weibull scale η .

4 Conclusion

In this work, we computed the hazard density and the total PoF of a turbine blade under cyclic loading due to LCF crack initiation on the surface. The computations are based on the probabilistic model for LCF as proposed in [1], [2] and [3]. In order to consider thermomechanical loading and uncertainties due to LCF test data we extended the model by a temperature model and included the percentile bootstrap method, respectively.

In future, we plan to consider information on local strain gradients. LCF cracks initiating at the surface grow into the component, where a different local strain field may result to a different speed of crack growth, confer [4]. Moreover, we plan to extend the probabilistic model to consider HCF, TMF and non-stationary FEA. Finally, note that the model can be used to optimize the total PoF with respect to the shape Ω , i.e. to find a design Ω under certain constraints such that surface integrals of the form of (6) are minimized. This is also called optimal reliability, confer [2].

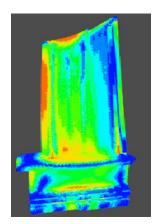


Figure 3: Hazard density of the analyzed turbine blade: Red areas show critical regions and blue ones may indicate overengineered regions of the blade.

Acknowledgment

This work has been supported by the German federal ministry of economic affairs BMWi via an AG Turbo grant. We wish to thank the gasturbine technology department of the Siemens AG and the institute of energy and climate research (IEK-2) of the FZ Jülich for stimulating discussions and many helpful suggestions.

References

- Schmitz, S.; Seibel, T.; Beck, T.; Rollmann, G.; Krause R.; Gottschalk, H.: A Probabilistic Model for LCF, Computational Materials Science, 79 (2013), 584-590, arXiv: in preparation
- [2] Gottschalk, H.; Schmitz, S.: Optimal Reliability in Design for Fatigue Life, Part I – Existence of Optimal Shapes, arXiv:1210.4954, (2012)
- [3] Schmitz, S.; Gottschalk, H.: Rollmann, G.; Krause R.: Risk Estimation for LCF Crack, Proc. ASME Turbo Expo 2013, GT2013-94899, arXiv:1302.2909, (2013)
- [4] Bäker, M.; Harders, H.; Rösler, J.: Mechanisches Verhalten der Werkstoffe, Third edition, Vieweg+Teubner, Wiesbaden, 2008
- [5] Escobar, L. A.; Meeker, W. Q.: Statistical Methods for Reliability Data, Wiley-Interscience Publication, New York, 1998
- [6] Radaj, D.; Vormwald, M.: Ermüdungsfestigkeit, Third edition, Springer, Berlin Heidelberg, 2007

- [7] Neuber, H.: Theory of Stress Concentration for Shear-Strained Prismatical Bodies with Arbitrary Nonlinear Stress-Strain Law, J. Appl. Mech., 26 (1961), 544
- [8] Fedelich, B.: A stochastic theory for the problem of multiple surface crack coalescence, International Journal of Fracture, 91 (1998), 2345.
- [9] Georgii, H.-O.: Stochastics Introduction to Probability and Statistics, de Gruyter, Berlin, 2008