



Probabilistic Analysis of Accelerated Life Tests for Thermal Barrier Coatings

1. Dresdner-Probabilistik-Workshop
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Research Center Karlsruhe,
Institute for Materials Research II

Research Center Jülich,
Institute for Materials and Processes of Energy Systems 2

Siemens AG, Energy
Gas Turbine Engineering

Overview

Motivation

- for Accelerated Life Testing
- for Probabilistic Analysis
- for Bayesian Approach

Application to Thermal Barrier Coatings (TBCs)

- Cyclic Oxidations Test

Bayesian Statistics

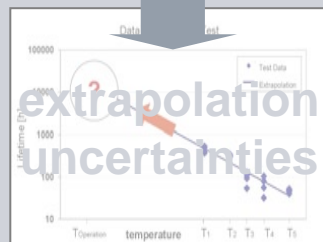
- Basic Ideas
- Application to Cyclic Oxidation Test Data

Uncertainties in Lifetime Prediction

Motivation

reliable life
time
predictions

accelerated
life testing



probabilistic
methods

Increasing turbine inlet temperatures require prime
reliant TBCs for turbine blades and vanes.

- Reliable Thermal Barrier Coatings
- Reliable Lifetime Tests

Testing under real gas turbine conditions is

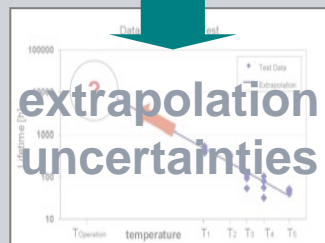
- difficult to realize
- too time consuming

→ **Accelerated Life Testing**

Motivation

reliable life time predictions

accelerated life testing



probabilistic methods

Accelerated Life Testing:

tests under overload conditions

- e.g. non stop operation,
- lower/higher temperature,
- higher stress levels,
- more aggressive environment, ...

extrapolation from overload conditions to operating conditions

→ **Extrapolation Uncertainties**

Motivation

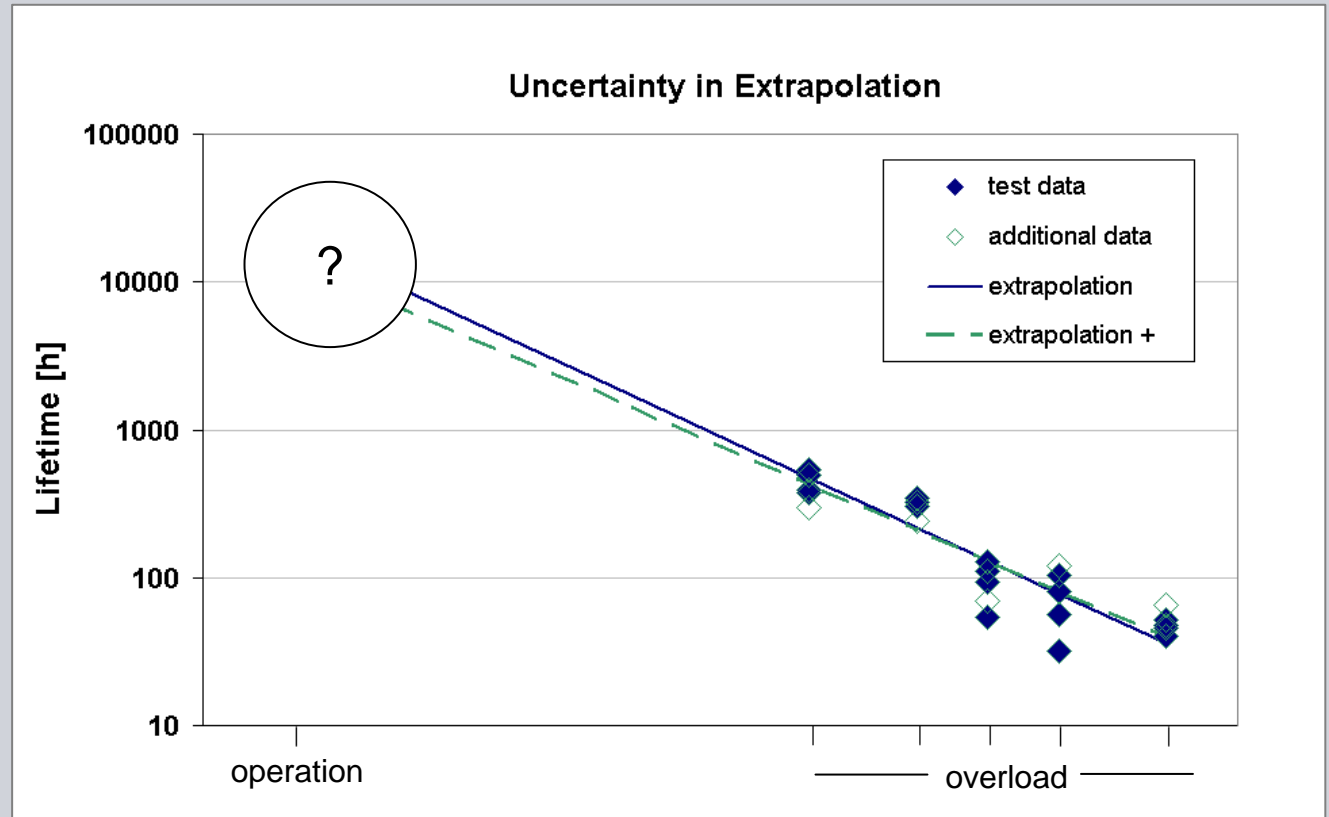
reliable life time predictions

accelerated life testing

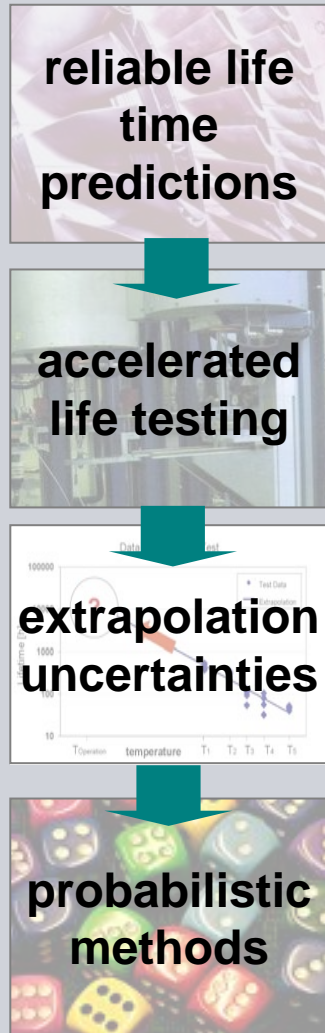
extrapolation uncertainties

probabilistic methods

typical test data contain scatter and uncertainties:



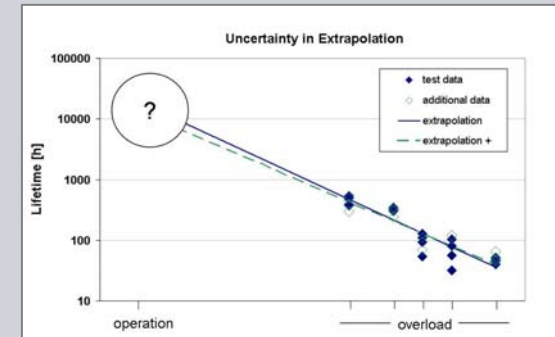
Motivation



Extrapolation Uncertainties

- natural scatter in material
- scatter in test data
- uncertainties in acceleration model

→ high uncertainties in extrapolation



Lifetime prediction should

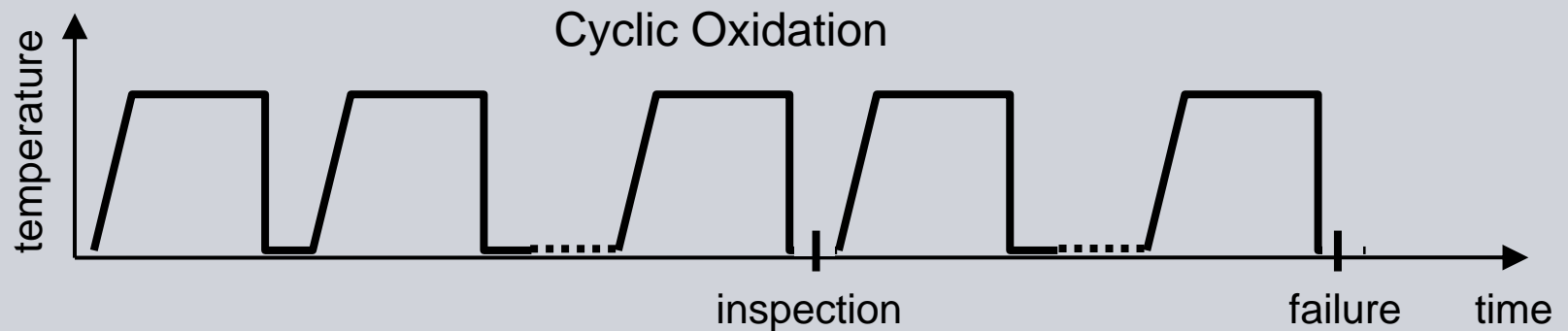
- be precise
- be reliable
- consider uncertainties

→ **Probabilistic Methods**

Accelerated Life Test for Thermal Barrier Coatings

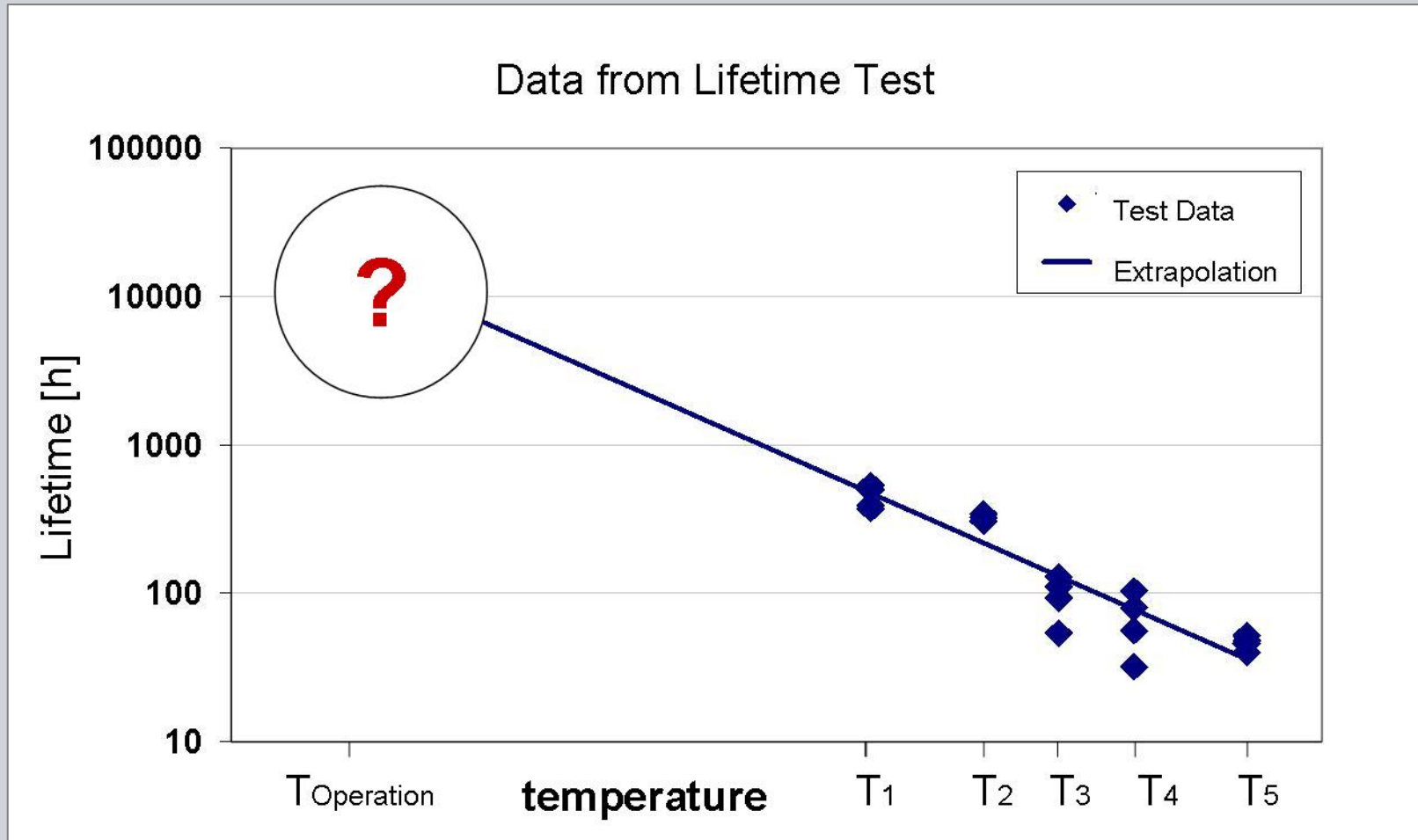


Failure detected by visual inspection:
 Bond Coat Oxidation
 Crack Initiation
 Spallation



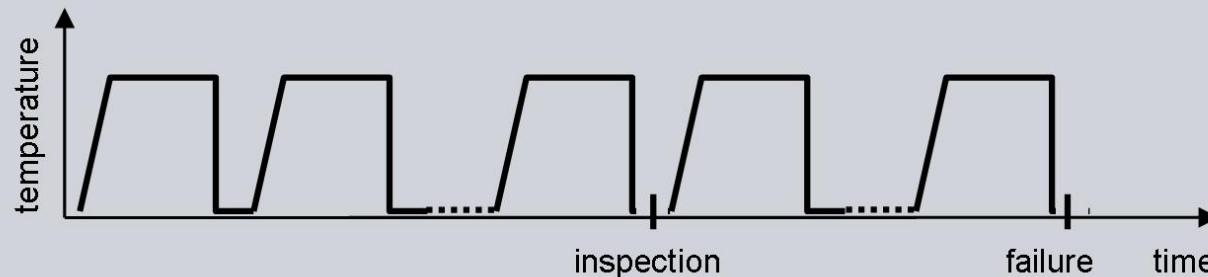
Test Results

Extrapolation from Overload to Operating Conditions



Requirements for a Probabilistic Approach

- low number of specimen
- censored data (inspection intervals)



nice to have:

- easy to include new data points
- considering single data points in analysis
- no further assumptions, except probability distribution type

Application of Bayesian-Statistics

lifetime prediction under operating conditions

a-posteriori-probability
this value is derived from data analysis

lifetime data from cyclic oxidation test at several temperature levels

Likelihood:
information from data analysis

$$P(\lambda | D, H) = P(\lambda | H) \cdot \frac{P(D|\lambda, H)}{P(D|H)}$$

model quality

evidence:
normalization

λ estimated parameter
D information from data
H additional information
(hypothesis)

a-priori-probability:
value from prior knowledge

no previous knowledge: $P(\lambda|H)=1$

Application on Cyclic Oxidation Test

given from test data: temperature level T_i , $i=1, \dots, 5$
 lifetime t_i
 failure rate λ_i

Likelihood Function for Exponential Distribution

$$\begin{aligned} P(D | \lambda_i, H) &= P(x_1, \dots, x_n | \lambda_i, H) \\ &= \prod_{k=1}^n P(x_k | \lambda_i, H) = \prod_{k=1}^n \lambda_i \exp(-x_k) \end{aligned}$$

Application on Cyclic Oxidation Test

Arrhenius-Approach for Acceleration Function

$$\lambda_i = \lambda_0 \cdot \exp\left(-\frac{B}{R}\left(\frac{1}{T_i} - \frac{1}{T_0}\right)\right)$$

given:

operating temperature T_0

temperature level T_i

failure rate λ_0

failure rates λ_i

universal gas constant R

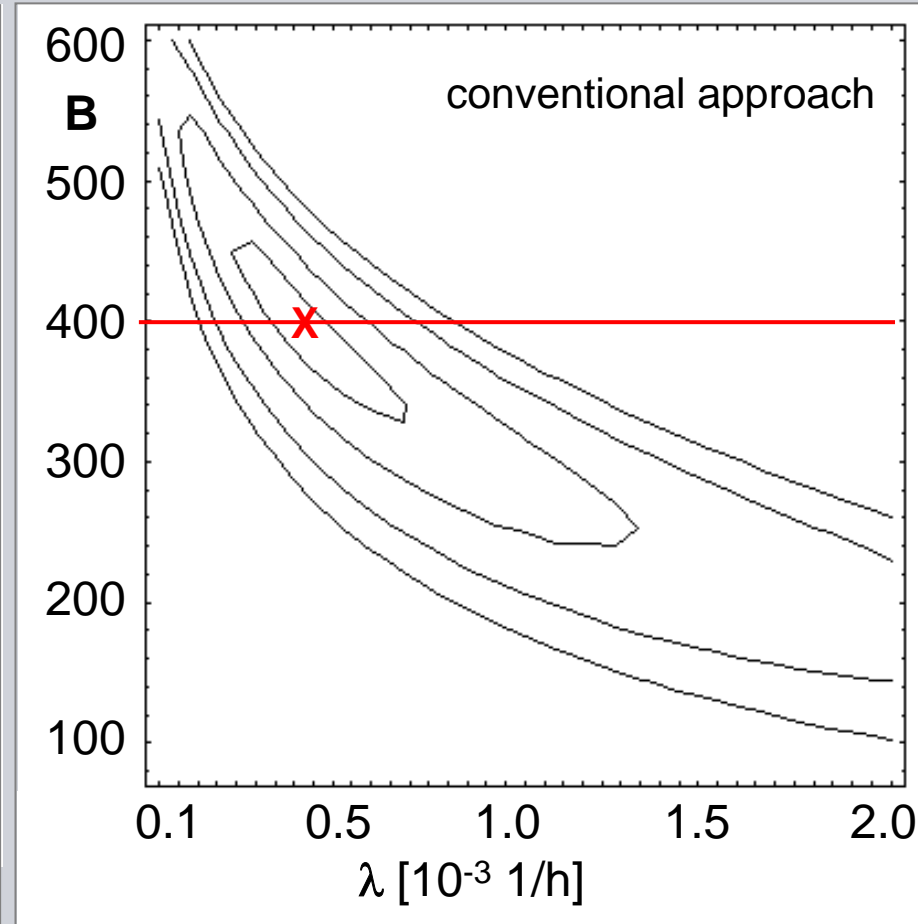
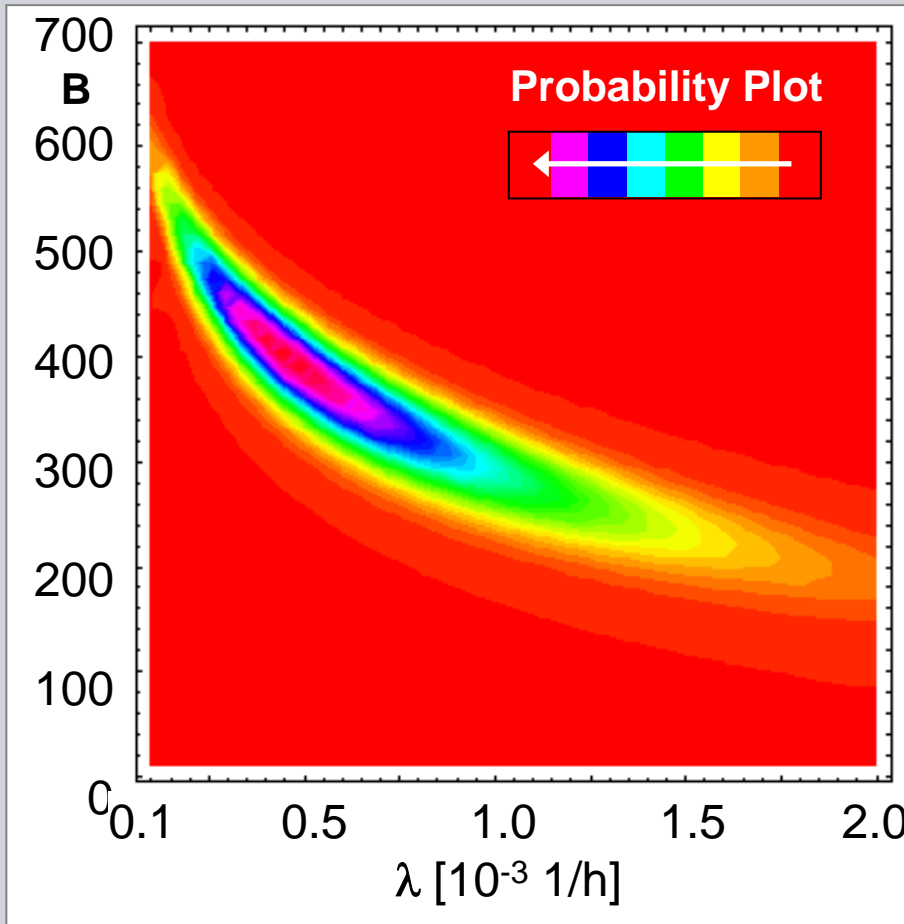
Likelihood Funktion for Lifetime Prediction

$$P(D | \lambda_0, B, H) = \lambda_0^n \cdot \exp\left(-\lambda_0 \cdot \sum_{i=1}^n \exp\left(-\frac{B}{R}\left(\frac{1}{T_i} - \frac{1}{T_0}\right)\right) \cdot t_i - \sum_{i=1}^n \frac{B}{R}\left(\frac{1}{T_i} - \frac{1}{T_0}\right)\right)$$

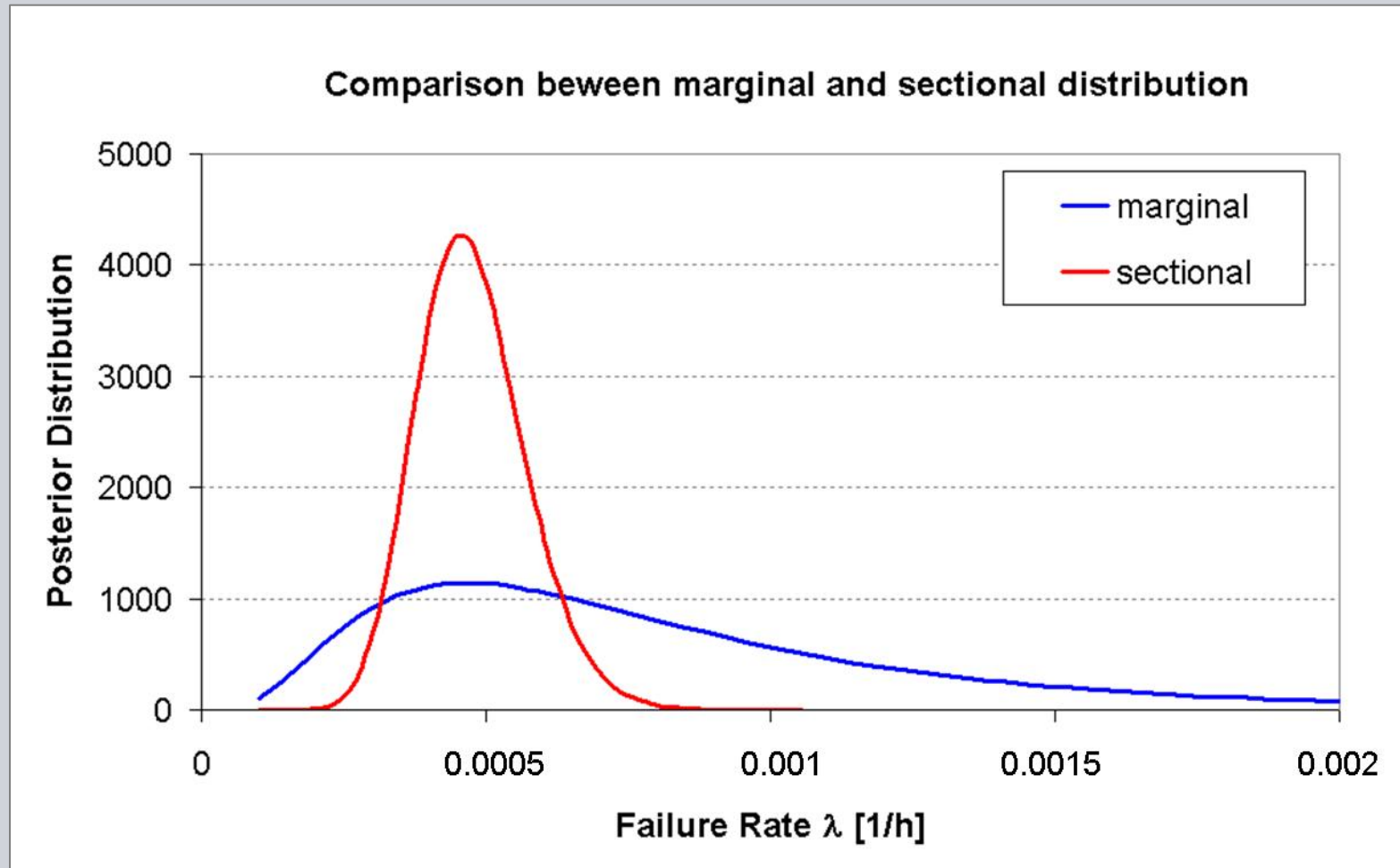
wanted: failure rate λ_0

unknown: parameter B

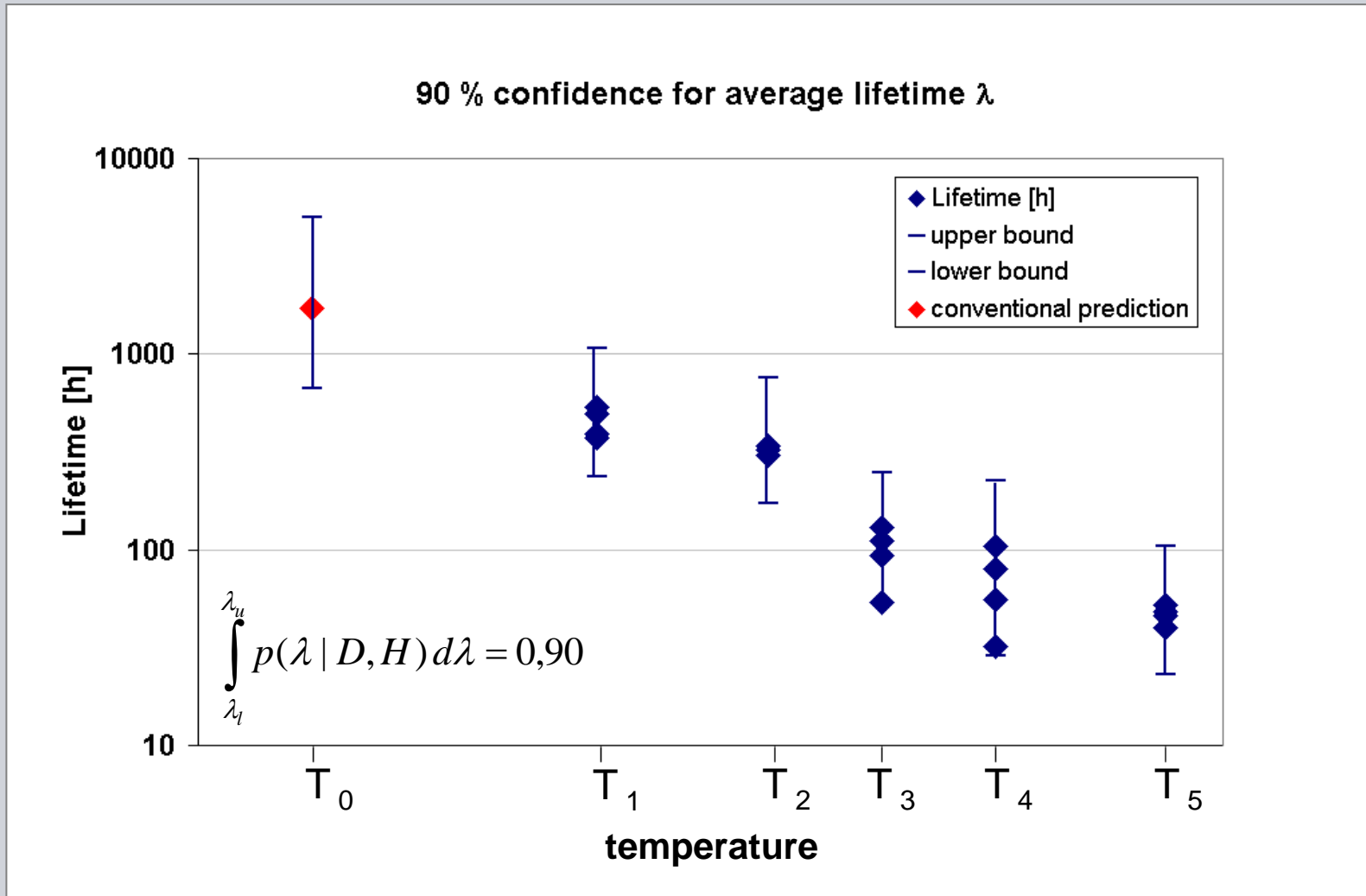
Prediction of Failure Rate λ



Comparison of Failure Rate Distribution



Uncertainties in Lifetime Prediction



Conclusion

- A probabilistic evaluation of Accelerated Life Testing was performed based on a Bayesian approach.
- The Bayesian set up has a lot of advantages compared to the „conventional“ probabilistic approach.
- Marginalization allows to eliminate nuisance parameters, while still taking into account their uncertainty.
- Results show high uncertainties in test data.
- High uncertainties in lifetime prediction were quantified.

Thank you
for your attention!