Chapter 6. Reliability Data Sources

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Learning Objectives

The main learning objectives associated with these slides are to:

- Introduce and discuss different data types
- Give examples of different data sources
- Present an approach for esimating reliability data, when limited experience is available for the prevailing operating environment

The slides include topics from Chapter 6 in **Reliability of Safety-Critical Systems: Theory and Applications**. DOI:10.1002/9781118776353.

Outline of Presentation

Introduction







5 New Technology

Types of Data

Many different types of data (or information) may be relevant in the analysis of system reliability:

- Technical data/information: Data and information that is needed to identify and understand how elements, channels and subystems are operating
- **Operational data:** Data and information about mode of operation, environemental exposure, operating conditions and so on
- Reliability data: Failure rates or mean time to failure (MTTF), or data that support the estimation of these. The following sub-categories of reliability data are often used:
- Test and maintenance data: Associated data of relevance for the analysis related to testing and maintenenance, such as proof test intervals, mean test or inspection time, mean repair time, diagnostic test intervals, mean restoration time after a DD failure, proof test coverage, and so on.

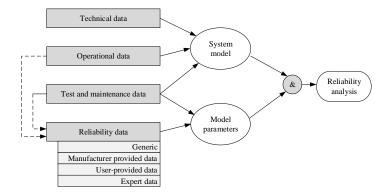
Types of Reliability Data

Reliability data may be classified into:

- Generic data: Data collected by an organization and published in handbooks. The collected data may be for specific component types (not brands), and may be a combination of operating experience, manufacturer data *applicable for a specific industry sector or specific usagen conditions* (e.g., offshore oil and gas industry).
- Manufacturer (brand) data: Data provided for a particular component brand, based on manufacturers internal statistics on reported failures, in-house testing, or failure rate estimation techniques.
- User-provided (experience) data: Data collected by a specific user, at a specific site or plant or for a selection of sites/plants.
- Expert judgment: Data constructed on the basis of experts opinions and experience. May be an option when systematic data collection has not been carried out or when new technology is introduced to a system.

Data Types

Application of Data



Data Sources

Types of Reliability Data Sources

Maintenance system:

A site's or plant's maintenance system, and is not accessible unless approved by the site/plant owner

Accident and incident databases:

Published by organizations and authorities), often with basis in mandatory reporting of serious incidents (e.g., events with a major accident potential) or accidents.

Component reliability databases:

Data handbooks and data bases constructed on the basis of generic data.

A high number of generic data sources are available:

Standards	General	Offshore/Process industry
IEC 61709 IEC TR 62380 ISO 13849-1	MIL-HDBK-217F NPRD-2011 (RIAC) FIDES Telecordia SR332 Siemens SN29500 MechRel Handbook (NSWC-11)	OREDA PDS data handbook Exida

More information about each data source is provided in the textbook.

It is important to document where the failure rates origin from, and what assumptions that have been made.

■ Data dossier: A data sheet that presents and justifies the choice of data for each element included in the reliability model.

An example of a data dossier is shown in the textbook. Also data handbooks may provide similar layout of their data dossiers.

Data Dossier

Data Dossier in PDS Data Handbook

5.1.10 Smoke Detector

 $\lambda_{crit} = 3.7$ $\lambda_D = 1.3 \text{ per } 10^6 \text{ hrs}$

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$\lambda_{00} = 0.8 \text{ per } 10^6 \text{ hrs}$ $\lambda_{5TU} = 1.4 \text{ per } 10^6 \text{ hrs}$ I_{12} Assumed $c_0 = 40\%$	failure rate (per 10 ⁶ hrs)		Data source/con	nment		
	$\lambda_{crit} = 3.7$	$\lambda_{DU} = 0.8 \text{ per } 10^6 \text{ hrs}$ $\lambda_{STU} = 1.4 \text{ per } 10^6 \text{ hrs}$	[12]	,		

Recommended values for calculation in 2004- and

New technology

New technology is by DNV-GL RP considered as:

- (Totally) new technology (unproven design principles)
- Proven technology in new environment
- Proven technology operated in a new way

How can we identify applicable reliability data in this case?

New technology and SIS

E/E/PE technology is developing fast, an generic data provided at the component level (e.g., a logic solver) becomes outdated almost before it is published.

Is it possible to apply previous reliability data at all?

Brissaud et al (2010) have proposed an approach for this purpose. See the more exact reference in the textbook.

A Suggested Approach

- 1. Identify λ_B , using generic reliability databases, observed failures, or expert judgments for the reference (proven) technology. There is no specific rules about confidence limits, but it may be considered if a more conservative value should be selected rather than a maximum likelihood/mean value.
- 2. Identify factors that are assumed to be highly influential for this failure rate. Try to keep the number of factors as low as possible, and combine factors that are highly dependent on each other. For the *k* remaining factors, do as follows:
 - Denote each influencing factor y₁, y₂ ...y_k
 - Define the *nominal level* (e.g. industry average) for these factors, and denote these by: y_{0,1}, y_{0,2}...y_{0,k}.

Example of nominal values for a selection of influencing factors:

Example

No	Influencing factor	Nominal value	No	Influencing factor	Nominal value
1	Temperature	-5° - +20° C	4	Technology maturity	High
2	Environment	Outside, offshore	5	Diagnostic coverage	75%
3	Frequency of use	< 1 per year	6	Testing/ inspection frequency	1 year/ 1 year

3. Weight the influencing factors, using input from several experts and physical and engineering knowledge. Make sure that $\sum_{i=1}^{k} \omega_i = 1$.

Example of values assigned as nominal for a selection of influencing factors, with possible weights added:

Example							
No	Influencing factor	Nominal value	Weight	No	Influencing factor	Nominal value	Weight
1	Temperature	-5° - +20°	15%	4	Technology maturity	High	15%
2	Environment	Outside, off- shore	20%	5	Diagnostic coverage	75%	15%
3	Frequency of use	< 1 per year	20%	6	Testing/ inspection frequency	1 year/ 1 year	15%

4. Identify the new values of the influencing factors and denote these by *y*_{c,1}, *y*_{c,2}..*y*_{c,k}

Example								
No	Influencing factor	New value	Weight	No	Influencing factor	New value	Weight	
1	Temperature	0 - +4 ^{<i>o</i>}	15%	4	Technology maturity	Medium (due to design changes)	15%	
2	Environment	Subsea, 400 meter water depth	20%	5	Diagnostic coverage	90%	15%	
3	Frequency of use	< 1 per year	20%	6	Testing/ inspection frequency	1 year/ 5 years	15%	

New Technology

A Suggested Approach (cont.)

- 5. Determine the effect of the new values $\sigma_{c,i}$ for each influencing factor y_i , $i = 1, 2 \dots k$, using the following rules:
 - $\sigma_{c,i} = 1$, when $y_{c,i} \approx y_{0,i}$
 - $\sigma_{c,i} < 1$, when $y_{c,i}$ is more *benign* than $y_{0,i}$
 - $\sigma_{c,i} > 1$, when $y_{c,i}$ is more *hostile* than $y_{0,i}$
- 6. Calculate the failure rate λ_P for the new technology using equation (1) on next slide

The approach does not specify how much > 1 or how much < 1.

Equation:

$$\lambda_{\rm P} = \lambda_{\rm B} \cdot \sum_{i=1}^{k} \omega_{\rm i} \cdot \sigma_{\rm c,i} \tag{1}$$

where k is the number of influencing factors, ω_i is the weight of the influencing factor, and $\sigma_{c,i}$ is the value assigned to influencing factor *i*, with $i = 1 \cdots k$.

A Suggested Approach: Case study

In this example, we assume for the 6 influencing factors that:

- $\sigma_{c,i} = 0.5$, when $y_{c,i}$ is more *benign* than $y_{0,i}$
- $\sigma_{c,i} = 2.0$, when $y_{c,i}$ is more *hotile* than $y_{0,i}$ (except for influencing factor *i*=2, where a value σ of 10 is used)

We also assume that $\lambda_{\rm B} = 2.6 \cdot 10^{-6}$ per hour.

$$\lambda_{\rm P} = \lambda_{\rm B} \cdot \sum_{i=1}^{k} \omega_{\rm i} \cdot \sigma_{\rm c,i}$$

= 2.6 \cdot 10^{-6} \cdot [0.15 \cdot 1 + 0.20 \cdot 10.0 + 0.20 \cdot 1
+0.15 \cdot 2.0 + 0.15 \cdot 0.5 + 0.15 \cdot 2.0] per hour = 1.02 \cdot 10^{-5} per hour

A Suggested Approach: Some Considerations

There are several questions that may arise, and that may influence the uncertainty associated with the new failure rate:

- Have all relevant influencing factors been captured?
- Are the weights reasonable? Will also the weights change with the new technology?
- Are the values of $\sigma_{c,i}$ reasonable?

A reasonable question is how to capture and express uncertainty about the failure rate in this context. This is not a part of the method as it is now.