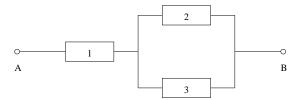
## Chapter 3

# Reliability and Survival

## 3.1 Systems of components

We will study *systems* that in a simple case might look like in Figure 3.1 below:

Figure 3.1. A simple system.



In the system in Figure 3.1, the *component* with component number i may be *healthy*, for  $i = 1, 2, \ldots$ , otherwise it is *unhealthy* or *dead*. The system is healthy if there is a path from the point A to the point B which only passes healthy components. Otherwise the system is unhealthy.

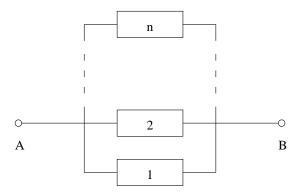
**Example 3.1.** The *series coupling* in Figure 3.2 below is healthy if and only if all its components are healthy.

Figure 3.2. Series coupling.



**Example 3.2.** The *parallel coupling* in Figure 3.3 below is healthy if and only if at least one of its components is healthy.

Figure 3.3. parallel coupling.

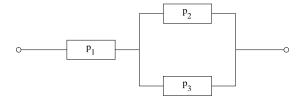


A system can be built by means of a finite number of series couplings and parallel couplings. See Figures 3.4 and 3.5 below for an example of how this works in a practical application.

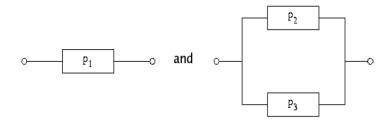
We will study systems with component numbers i = 1, 2, ... which are healthy with certain health probabilities  $p_1, p_2, ...$ . Unless otherwise is stated, the components of a system are assumed to be independent of each other.

**Example 3.3.** The system in the Figure 3.4 below is a series coupling of two systems with health probabilities  $p_1$  and  $1 - (1 - p_2)(1 - p_3)$ , respectively. The health probability for the whole system is  $p_1[1 - (1-p_2)(1-p_3)]$ .

Figure 3.4. A simple system with health probabilities.



**Figure 3.5.** Composition of a simple system as a series coupling of a single component with a parallel coupling.



The components with component numbers  $i = 1, 2, \ldots$  of a system have certain life lengths

 $T_1, T_2, \ldots$  The life lengths are modeled as random variables that are mutually independent, unless otherwise is stated.

The relation between the health probability and the life length  $T_i$  of the component with component number i is as follows:

$$p_i = p_i(t)$$
  
=  $\mathbf{P}$ {component with component number  $i$  is healthy at time  $t$ }  
=  $\mathbf{P}$ { $T_i > t$ }  
=  $1 - F_{T_i}(t)$ 

Here  $F_{T_i}(t) = \mathbf{P}\{T_i \leq t\}, i = 1, ..., n$ , are the distribution functions of the life lengths  $T_1, T_2, ...$ These distribution functions in turn are assumed to be continuous, unless otherwise is stated.

**Definition 3.1.** The survival function  $R_T$  of a system with life length T is given by

$$R_T(t) = \mathbf{P}\{the \ system \ is \ healthy \ at \ time \ t\} = \mathbf{P}\{T > t\} = 1 - F_T(t)$$

for t > 0, where  $F_T(t)$  is the distribution function of the life length.

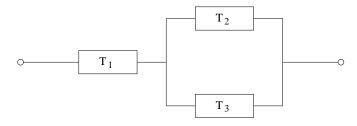
**Example 3.4.** The life length of the system in the Figure 3.6 below is given by  $T = \min\{T_1, \max[T_2, T_3]\}$ , see also Example 3.3. Hence the survival function for the system is given by

$$R_T(t) = \mathbf{P}\{T > t\}$$

$$= \mathbf{P}\{T_1 > t\} [1 - (1 - \mathbf{P}\{T_2 > t\}) (1 - \mathbf{P}\{T_3 > t\})]$$

$$= R_{T_1}(t) [1 - (1 - R_{T_2}(t)) (1 - R_{T_3}(t))].$$

**Figure 3.6.** A simple system with life lengths.



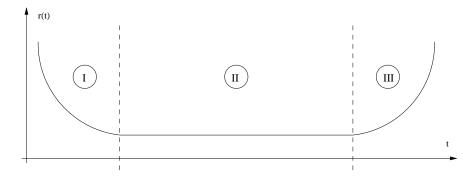
**Definition 3.2.** The death intensity  $r_T$  of a system with life length T is given by

$$r_T(t) = -\frac{d}{dt} \ln \left( R_T(t) \right) \quad for \ t > 0.$$

**Definition 3.3.** A system with life length T has increasing failure rate, IFR, if the death intensity is increasing  $(r'(t) \ge 0 \text{ for } t > 0)$ . A system with life length T has decreasing failure rate, DFR, if the death intensity is decreasing  $(r'(t) \le 0 \text{ for } t > 0)$ .

**Example 3.5.** It is quite common that life lengths have failure rates that are neither IFR or DFR, but instead follow a so called *bath tub curve*, BTC, the principal appearance of which is displayed in Figure 3.7 below. In that figure the region I corresponds to an early phase with a comparatively high probability of unhealth (as for example, small children). The region II corresponds to a component that has survived these early hazards and has settled at a lesser death intensity (as for example, grown up people). Finally, the region III corresponds to an aged component, where the death intensity increases with accumulated age (as for example, aged people).

Figure 3.7. Example of a life length with a death intensity that displays a BTC shape.



## 3.2 More on systems

The following theorem explains that the death intensity really is the (infinitesimal) intensity at which unhealth occurs:

**Theorem 3.1.** For a life length T with death intensity  $r_T$ , we have

$$\mathbf{P}\{T \le t + h|T > t\} = r_T(t)h + o(h) \quad as \ h \downarrow 0.$$

*Proof.* Writing  $F_T$  and  $f_T$  for the distribution function and probability density function of the life length T, respectively, we have

$$\mathbf{P}\{T \le t + h | T > t\} = \frac{F_T(t+h) - F_T(t)}{1 - F_T(t)} = \frac{f_T(t)h + o(h)}{R_T(t)}$$
$$= -\frac{d}{dt} \ln(R_T(t))h + o(h) = r_T(t)h + o(h). \quad \Box$$

**Theorem 3.2.** A function  $r:(0,\infty)\to [0,\infty)$  is a death intensity if and only if

$$\int_0^\infty r(t)dt = \infty.$$

In that case the corresponding survival function is given by

$$R(t) = \exp\left\{-\int_0^t r(s)ds\right\}.$$

*Proof.* If r is a death intensity of a system with survival function  $R_T$ , then a differentiation of the function

$$R(t) = \exp\left\{-\int_0^t r(s)ds\right\}$$

gives

$$-\frac{d}{dt}\ln(R(t)) = -\frac{-r(t)\exp\left\{-\int_0^t r(s)ds\right\}}{\exp\left\{-\int_0^t r(s)ds\right\}} = r(t).$$

As the function  $-\ln(R(t))$  has the same derivative as the function  $-\ln(R_T(t))$ , namely the death intensity r(t), it follows that the functions  $-\ln(R(t))$  and  $-\ln(R_T(t))$  can differ only by a additive constant, so that the functions R(t) and  $R_T(t)$  differ only by a multiplicative constant. Since  $R(0) = 1 = \mathbf{P}\{T > 0\} = R_T(0)$ , we conclude that the functions R(t) and  $R_T(t)$  are equal. Finally,

$$\lim_{t \to \infty} \exp\left\{-\int_0^t r(s)ds\right\} = \lim_{t \to \infty} R_T(t) = \lim_{t \to \infty} \mathbf{P}\{T > t\} = 0,$$

we must have  $\int_0^\infty r(s)ds = \infty$ . Conversely, if  $\int_0^\infty r(s)ds = \infty$  and we define the function

$$R(t) = \exp\left\{-\int_0^t r(s) \, ds\right\},\,$$

then R(t) in decreasing with R(0) = 1 and  $R(\infty) = 0$ , so that F(t) = 1 - R(t) is increasing with F(0) = 0 and  $F(\infty) = 1$ . This makes F a probability distribution function, so that R is a survival function. 

**Theorem 3.3.** For a life length T we have the following formula for expectations

$$\mathbf{E}\{T\} = \int_0^\infty R_T(t) \, dt$$

*Proof.* By integration by parts, we get

$$\mathbf{E}\{T\} = \int_0^\infty t \left(-\frac{d}{dt}R_T(t)\right)dt = -\left[tR_T(t)\right]_0^\infty + \int_0^\infty R_T(t)dt = \int_0^\infty R_T(t)dt. \quad \Box$$

A life length T with a constant death intensity  $r_T(t) = \lambda$  has the *lack of memory* property (cf. Theorem 3.1). By Theorem 3.2, a life length T lacks memory if and only if T is exponentially  $\exp(\lambda)$  distributed with parameter  $\lambda$ , i.e.,  $R_T(t) = e^{-\lambda t}$ .

**Example 3.6.** The second simplest form of death intensity, after a constant one, is a polynomial death intensity  $r_T(t) = ba^b t^{b-1}$ . In this case, Theorem 3.2 gives that the survival function is  $R_T(t) = \exp\{-(at)^b\}$ , that is, a Weibull distribution with parameters a and b, Weibull $(a,b)^6$ . And so we have the expectation

$$\mathbf{E}\{T\} = \int_0^\infty R_T(t)dt = \int_0^\infty \exp\left\{-a^b t^b\right\} dt$$

by Theorem 3.3.

If the life lengths  $T_1, \ldots, T_n$  are exponentially  $\exp(\lambda)$  distributed, then their sum  $T \equiv T_1 + \ldots + T_n$  is gamma $(n, \lambda)$  distributed with parameters n and  $\lambda$ . Hence the corresponding probability density function is given by

$$f_T(t) = \frac{\lambda^n t^{n-1}}{(n-1)!} e^{-\lambda t}$$
 for  $t > 0$ ,

giving the survival function

$$R_T(t) = \sum_{k=0}^{n-1} \frac{\lambda^k t^k}{k!} e^{-\lambda t} \quad \text{for } t > 0.$$

To achieve a high health probability of a system, the system may be equipped with more components than are actually needed for its health, if all the components are healthy. In other words, the system is not a pure series coupling, but a series coupling of subsystems, some of which are parallel couplings, to achieve higher health probability.

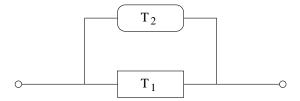
A component that is not required for the health of a system, when all other components of the system are healthy, is called a *redundant component*, RC.

A warm redundant components, WRC, is incorporated with the system already from the start of the system, while a cold redundant component, CRC, is incorporated with the system first at the time at which it is required for the health of the system.

<sup>&</sup>lt;sup>6</sup>It should be noted that there is a lot of variation in the parametrization of the Weibull distribution, so that, e.g., what is denoted Weibull(a, b) by us could be denoted Weibull(b, 1/a) in a software package.

**Example 3.7.** Figure 3.8 below depicts a system where a first component with life length  $T_1$  is supported by a second redundant component with life length  $T_2$ .

**Figure 3.8.** A system where a first component is supported by a second redundant component.



For the life length T of the system we have  $T = \max\{T_1, T_2\}$  when the redundant component is warm, so that

$$R_T(t) = 1 - (1 - R_{T_1}(t))(1 - R_{T_2}(t)).$$

If the redundant component is cold, we get  $T = T_1 + T_2$  instead, so that,

$$R_T(t) = 1 - \int_0^t (1 - R_{T_1}(t - x)) R_{T_2}(x) r_{T_2}(x) dx.$$

A quantity of great interest for a system, is the probability that component with component number  $i = 1, 2, \ldots$  causes the death (unhealth) of the system. That probability, in turn, coincides with the probability that the life length of component with component number  $i = 1, 2, \ldots$  is equal to the life length of the whole system!

Primarily, component that have high probabilities to cause the death (unhealth) of the system, are those who should be supported by (warm or cold) redundant components.

**Example 3.8.** For the system in Figure 3.6, we have

$$\mathbf{P}\{\text{component with number 1 causes death}\}$$

$$= \mathbf{P}\left\{T_1 \leq \max[T_2, T_3]\right\}$$

$$= \int_0^\infty \mathbf{P}\{\max(T_2, T_3) \geq t\} f_{T_1}(t) dt$$

$$= \int_0^\infty (1 - F_{T_2}(t) F_{T_3}(t)) f_{T_1}(t) dt$$

$$= \int_0^\infty \left(1 - (1 - R_{T_2}(t))(1 - R_{T_3}(t))\right) R_{T_1}(t) r_{T_1}(t) dt.$$

(This probability must be 2/3 when the life lengths  $T_1, T_2$  and  $T_3$  are identically distributed.)

### 3.3 Computer assignment

In order to get a pass on this lab, 5 out of 11 points are needed.

Be careful to include in your report not only a description of the code that is needed, but also your understanding of the problem and an explanation to your "theoretical strategy". In addition to the general instructions for report writing that can be found on the homepage, here are some extra advice on things you should remember to include in the "Theory and implementation" sections:

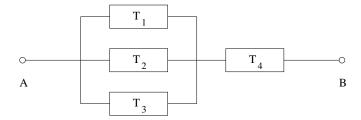
- In Assignment 1.1, show and explain how you derived the survival function for the system.
- In Assignment 1.2, show and explain how you derived the expression that can be used to calculate the probability that it is component number four that causes the death of the system.
- In Assignment 2.1, show and describe how you derived the two different survival functions for the new systems, with emphasis on the survival function for the system with a cold redundant component (since the derivation in the warm case is similar to the one you did in Assignment 1.1).
- In Assignment 2.2 and 3, remember to describe your strategy to solve the task and show the functions (e.g. survival function) you will use.

Remember also to comment on your results in the "Results and discussion" section (for example, are the results as expected)?

#### 3.3.1 Assignment 1 (4 p)

In the system in Figure 3.9 below, the first three components have life lengths  $T_1, T_2, T_3$  that are Weibull $(1, \frac{1}{2})$  distributed, while the fourth component have a life length  $T_4$  that is  $\exp(\frac{1}{2})$  distributed.

Figure 3.9. A system with four components.



#### Assignment 1.1

Find the expected life length  $\mathbf{E}\{T\}$  of the system. Also, plot the death rate  $r_T(t)$ ,  $t \in (0, 10)$ , for the system: Is the system IFR, DFR or neither?

#### Assignment 1.2

Find the probability that it is component with component number 4 that causes the death of the system.

#### 3.3.2 Assignment 2 (3 p)

#### Assignment 2.1

Redo the calculation of expected life length for the system, first with the component with component number 4 doubled with a warm redundant  $\exp(\frac{1}{2})$  distributed component, and then with the component with component number 4 doubled with a cold redundant  $\exp(\frac{1}{2})$  distributed component. Also, plot the death rates  $r_T(t)$  (including the one from task 1).

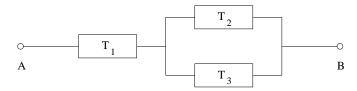
#### Assignment 2.2

For which values of the parameter  $\rho < \frac{1}{2}$  does a change of the component with component number 4 to a  $\exp(\rho)$  distributed component, have the same effect on the expected life length  $\mathbf{E}\{T\}$  of the system, as have the incorporation of a warm and cold (respectively) redundant  $\exp(\frac{1}{2})$  distributed component?

#### 3.3.3 Assignment 3 (3 p)

In the system in Figure 3.10 below, the first component has a life length  $T_1$  that is Weibull( $\mu, \frac{1}{3}$ ) distributed, while the second and third components have life length  $T_2$  and  $T_3$  that are Weibull( $\lambda, \frac{1}{3}$ ) distributed.

Figure 3.10. Yet another system with three components!



The cost of a Weibull $(\gamma, \frac{1}{3})$  distributed component is  $1/5 + 1/\gamma$  monetary units. Display graphically the values of the parameters  $\lambda$  and  $\mu$ , that maximizes the expected life length  $\mathbf{E}\{T\}$  of the system, at the total costs  $1, 2, \ldots, 10$  monetary units, respectively, of the system. Also plot the expected life length  $\mathbf{E}\{T\}$  as a function of the costs of  $1, 2, \ldots, 10$  monetary units, for the optimal values of the parameters  $\lambda$  and  $\mu$ .

#### 3.3.4 Software - some hints

The lab is to be done in R.

• Useful functions:

integrate, numDeriv::grad, Vectorize, uniroot, alabama::constrOptim.nl, where the notation numDeriv::grad means the function grad in the package numDeriv. The function integrate returns not only the numerical value of an integral, but an R object containing the value as well as other information.

- When using (for example) maximization/minimization procedures, make sure to apply suitable constraints.
- For the Weibull and Gamma distribution, read carefully on page 6 in the lecture notes how the distributions are defined (read the footnote also!). For example, note that we use the rate parameter in the Gamma distribution. Then, if you use the inbuilt functions, check how R defines the distributions.
- Some functions that take another function as an argument require that the argument function is vectorized. That means that when the function is applied to a vector, it should return a vector consisting of the result from applying the function to each element in the input vector. If it is inconvenient to change a non-vectorized function into a vectorized one, the Vectorize function can be used. An automatically vectorized function may be slower than a manually vectorized function, though.
- You may encounter the need to numerically calculate a double integral. To calculate, for example, the integral of x + t over a triangle in x-t-space you can do:

```
g <- function(t) {
  f <- function(x) x+t
  integrate(f, 0, t)$value
}
g <- Vectorize(g)
integrate(g, 0, 1)</pre>
```

But a little bit of thinking with pen and paper may save you the need to calculate a double integral!

- When doing numerical calculations, remember to check how exact the results are! Do not give answers with non-significant digits, but do not give too inexact answers either!
- A reference for reading more about the death intensity function is chapter 7.1.1 in the book Probability and Risk Analysis by Igor Rychlik and Jesper Ryden. It is available as an e-book at the Chalmers library.