

MCTR 702_1

Master Advanced Mechatronics

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2021

Mechatronics common framework Lecture 1







Contents

Lecture 1

RELIABILTIY

- Introduction to Dependability
- Reliability Functions & Estimators
- Predictive Reliability
- Reliability Modelling







Reliability Functions & Estimators

Reliability functions

- The failure function F(t)
- The reliability function R(t)
- Complementarity between F(t) and R(t)
- The failure rate $\lambda(t)$
- The density function f(t)

Reliability function estimators

- Estimation of F(t)
- Estimation of R (t)
- Estimation of $\lambda(t)$
- Estimation of f (t)
- Estimation of the mean E(t)





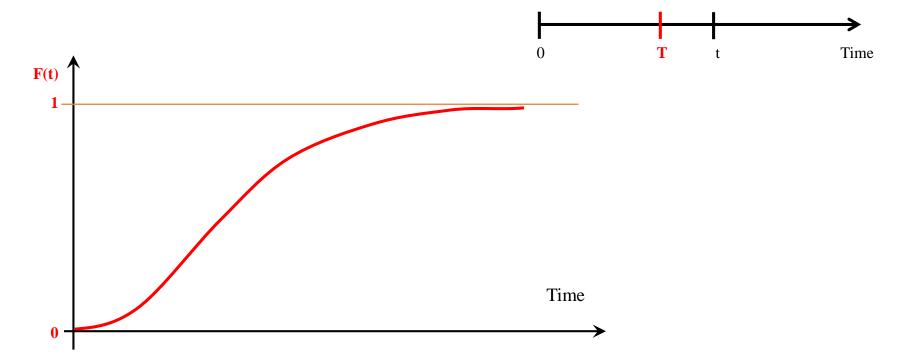


The failure function: F(t)

It is measured by the probability that an entity E fails over the time interval [0, t]: F(t) = Pr(E failing on [0, t])

That is, if we assume that an entity is failing at a date T:

$$F(t) = Pr(T \le t)$$







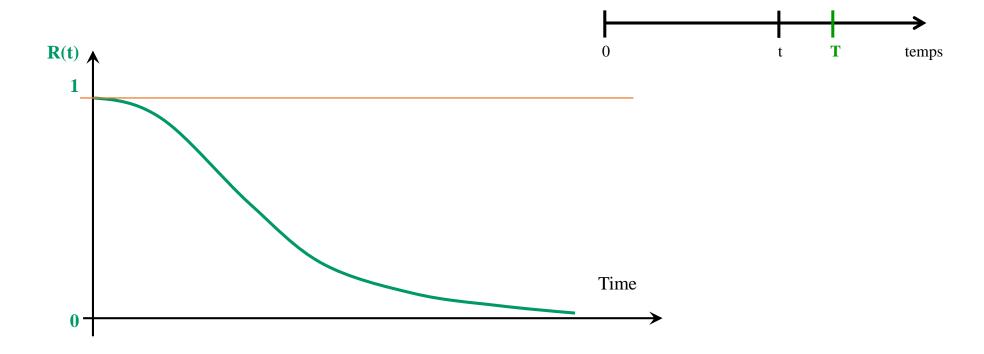


The reliability function: R(t)

It is measured by the probability that an entity E is non-faulty over the time interval [0, t]: R(t) = Pr(E not failing on [0, t])

That is, if we assume that an entity is failing at a date T:

$$R(t) = Pr(T>t)$$





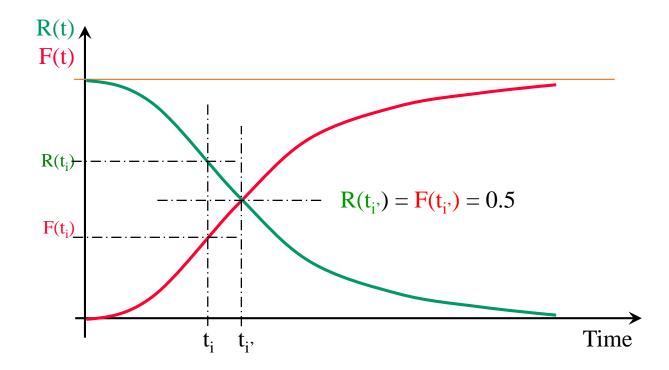




Complementarity between F(t) and R(t)

For a given system, $\forall t = ti$: we have

$$R(ti) + F(ti) = 1$$









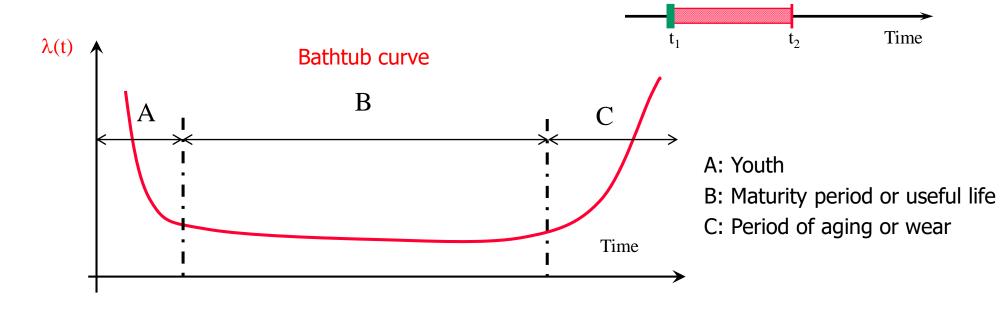
The failure rate: $\lambda(t)$

It expresses the **speed of arrival of failures** at time t, or The **evolution of the conditional probability of failure** during the life of the material (failure/product and unit of time)

A: Event "product runs at t1"

B: Event "product fails between t1 and t2"

 $\lambda(t) = \Pr(B/A) / (t2 - t1)$



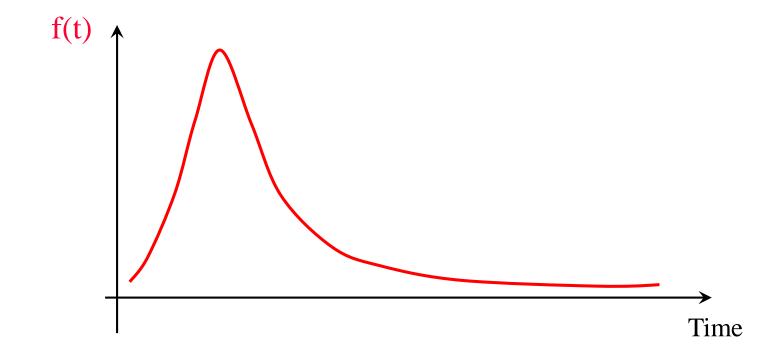






The density function: f(t)

It represents the histogram of the relative failures as a function of time or the probability frequency of the relative failures relative to the unit of time.

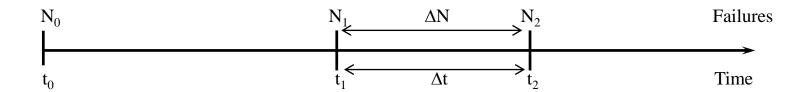








Estimation of the reliability functions



N: total number of products put into operation at time t0 (sample size)

N0: number of failures at t0 (generally equal to 0)

N1: number of failures at t1 N2: number of failures at t2

 ΔN : number of failures on the interval [t1, t2]

 Δt : time interval [t1, t2]

Reliability function estimators depend on the value of N (number of products put into operation)

N	1 <n≤20< th=""><th>20<n th="" ≤50<=""><th>N>50</th></n></th></n≤20<>	20 <n th="" ≤50<=""><th>N>50</th></n>	N>50
Estimator	Median Ranks	Average Ranks	Cumulative Frequencies







Failure estimators: F(t)

Point estimator at time t

Median ranks if 1<N≤20

$$F(t) = \frac{N_t - 0.3}{N + 0.4}$$

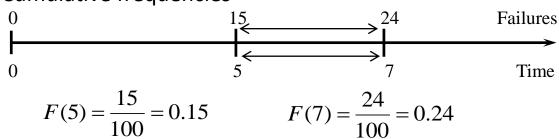
Average ranks if 20<N≤50

$$F(t) = \frac{N_t}{N+1}$$

Cumulative frequencies if N>50

$$F(t) = \frac{N_t}{N}$$

Example (N = 100) \rightarrow Cumulative frequencies





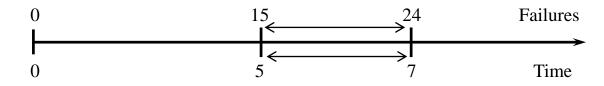


Reliability estimators: R(t)

Point estimator at time t

Whatever the value of N, first compute the estimator of F and then apply the complementarity relation R(t) = 1 - F(t)

Example (N = 100)



$$R(5) = 1 - 0.15 = 0.85$$

$$R(5) = \frac{100 - 15}{100} = 0.85$$

$$R(7) = 1 - 0.24 = 0.76$$

$$R(7) = \frac{100 - 24}{100} = 0.76$$







Density estimators: f(t)

Interval estimator (failure/product and unit of time)

Median ranks if 1<N≤20

$$f_{[t1;t2[} = \frac{\Delta N}{(N+0.4) \times \Delta t}$$

Average ranks if 20<N≤50

$$f_{[t1;t2[} = \frac{\Delta N}{(N+1) \times \Delta t}$$

Cumulative frequencies if N>50

$$f_{[t1;t2[} = \frac{\Delta N}{N \times \Delta t}$$

Example (N = 100) 0 15 9 24 Failures 5 2 7 Time $f_{[5;7[} = \frac{9}{100 \times 2} = 0.045 failures/product & t.u.$







Failure rate estimators: $\lambda(t)$

Estimator on interval (failure/product and unit of time) (f/R)

Median ranks if 1<N≤20

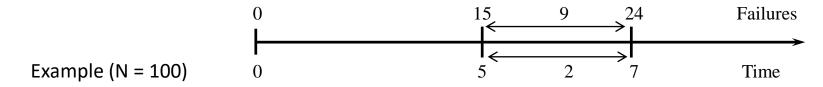
$$\lambda_{[t1;t2[} = \frac{\Delta N}{(N+0.7-N_1)\times \Delta t}$$

Average ranks if 20<N≤50

$$\lambda_{[t1;t2[} = \frac{\Delta N}{(N+1-N_1)\times \Delta t}$$

Cumulative frequencies if N>50

$$\lambda_{[t1;t2[} = \frac{\Delta N}{(N - N_1) \times \Delta t}$$



$$\lambda_{[5;7[} = \frac{9}{85 \times 2} = 0.053 \text{ failures/product \& t.u.}$$





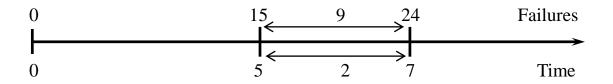


Average estimators: E(t)

Interval Estimator

Whatever the value of N, first calculate the estimator of λ and then apply the inverse relation $E(t) = 1/\lambda(t)$

Example (N = 100)



$$E(t)_{[5;7[} = \frac{1}{0.053} = \frac{85 \times 2}{9} = 18.89 \text{ u.t.}$$









Predictive Reliability

Definition

 Predictive reliability (predicted or theoretical) is calculated on the basis of a mathematical model defined from the functional decomposition of the system into subsets, components, etc. and the estimated or predicted reliability of its components

Analysis and calculation procedure

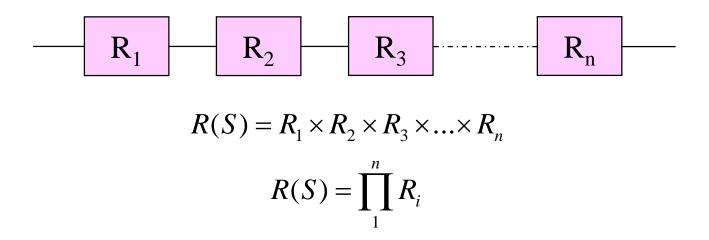
- Decompose the system into components (parts, subsystems ...) and establish the functional links between the components
- Identify component reliability models or collect reliability at a given time
- Search for a model of the system and calculate its reliability







Reliability of the serial system (S)



 R1, R2, R3, ... Rn are the elementary reliabilities of the system components at a given time

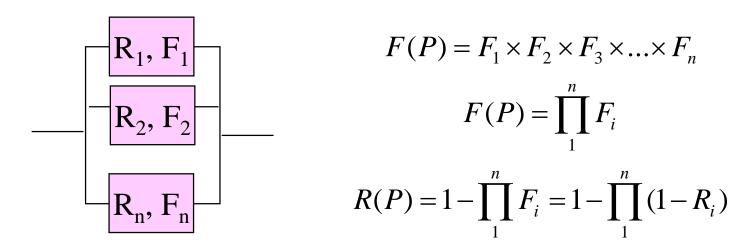






Reliability of the parallel system (P)

1 only component on "n" must operate (system 1/n by default)



 R1, R2,... Rn and F1, F2,... Fn are respectively the elementary reliabilities and failures of the components of the system at a given time

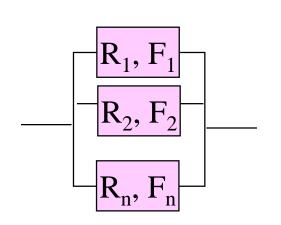






Reliability of the parallel system (k/n)

"k" components on "n" must operate simultaneously (system k/n notation required)



$$B(p,k,n) = C_n^k p^k (1-p)^{n-k}$$
 binomial

$$with: C_n^k = \frac{n!}{k!(n-k)!}$$

$$R(P) = B(p, k, n) + B(p, k+1, n) + B(p, k+2, n) + ... + B(p, n, n)$$

$$R(P) = \sum_{i=k}^{n} C_{n}^{i} p^{i} (1-p)^{n-i}$$

 $n \rightarrow total number of components$

 $p \rightarrow probability of success (reliability of a component)$

 $k \rightarrow number\ of\ components\ to\ operate\ simultaneously$

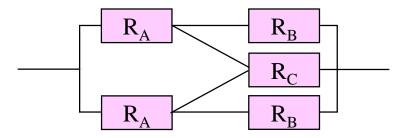
R1, R2, ... Rn and F1, F2, ... Fn are respectively the elementary reliabilities and failures of the components of the system at a given time (binomial law applicable only if the R are identical, otherwise, application of the truth table)



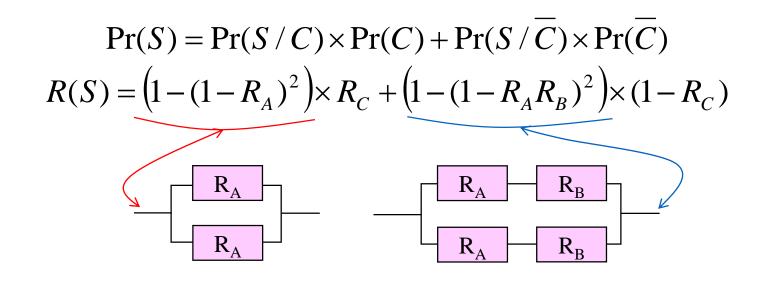




Reliability of a complex system



Baye's theorem









Application to exponential model

R(t) =
$$e^{-\lambda t}$$
 and F(t) = $1-e^{-\lambda t}$ with: $\lambda = 1/E(t)$ (failure rate)

The serial system

$$R(t) = e^{(-t\sum_{i=1}^{n}\lambda_i)} = e^{(-t\lambda_S)}$$

$$R(t) = 1 - \prod_{i=1}^{n} (1 - e^{(-\lambda_i t)})$$

$$R(t) = e^{-\lambda_1 t} + e^{-\lambda_2 t} - e^{-(\lambda_1 + \lambda_2)t}$$

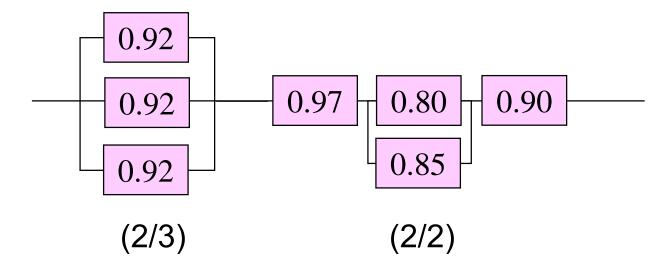






Example

The given values correspond to the reliability of the components at a given time



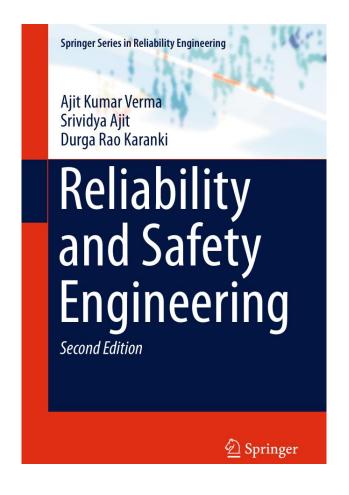
Calculate the predictive reliability of this system

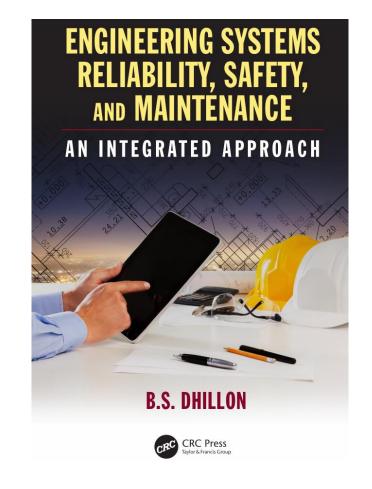






Relevant books











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