

**MCTR 701** 

#### **Master Advanced Mechatronics**

Lecturers: Luc Marechal, Christine Barthod, Georges Habchi







**Mechatronics** common framework **Lecture 3** 

2021







#### **Contents**

#### Lecture 1

#### **RELIABILTIY**

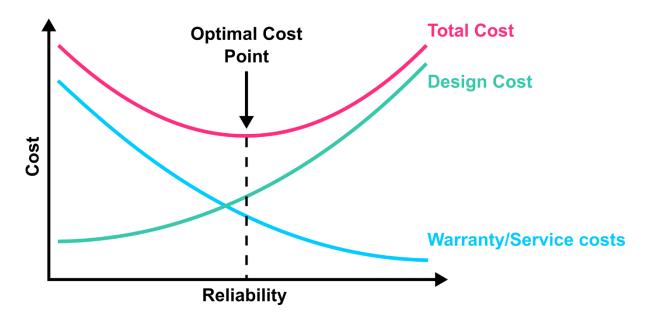
- Introduction to Dependability
- 1. Reliability Functions & Estimators
- 2. Predictive Reliability
- 3. Reliability Modelling







#### **Costs vs Reliability**



- Effects of Over-reliability in Development
  - Product is too expensive for target market
  - Product is later getting to market
  - Company is behind technology leaders due to slow program development cycles

- Effects of Under-reliability in Development
  - High field Return Rate
  - High Warranty Cost
  - Loss of product sales once low reliability is known in market
  - Loss of market share in all product lines due to poor brand perception.





## 1. RELIABILITY FUNCTIONS & ESTIMATORS







### **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)

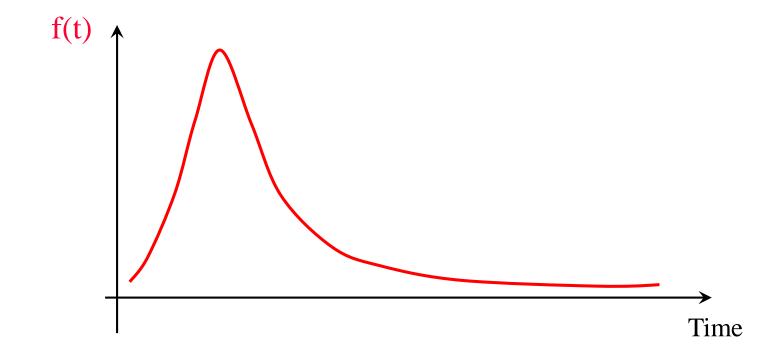






### The density function: f(t)

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.







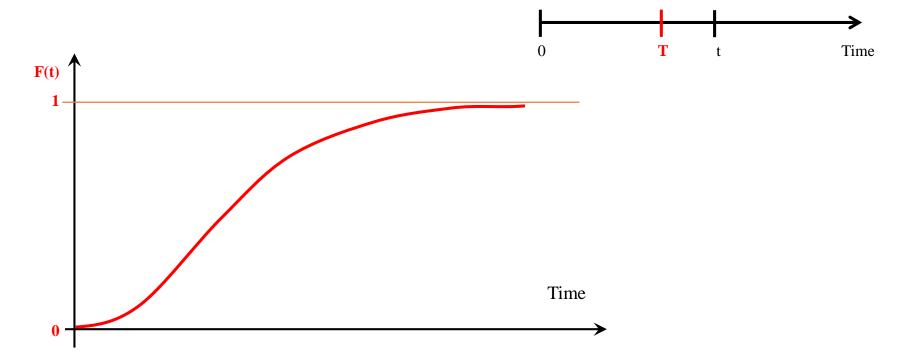


### The failure function: F(t)

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:

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







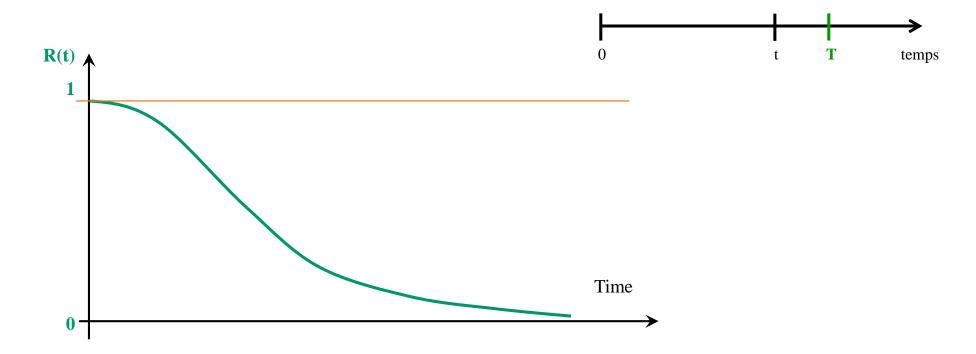


## The reliability function: R(t)

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:

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





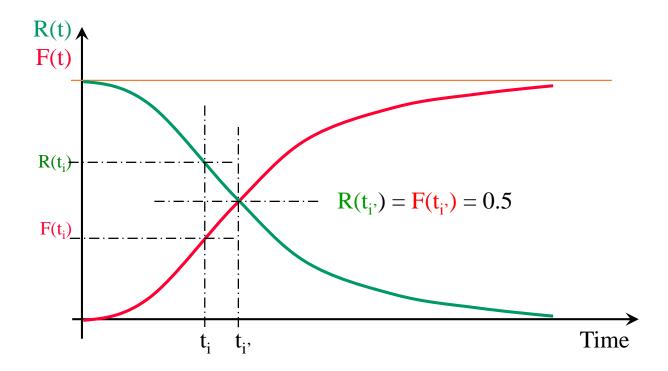




## Complementarity between F(t) and R(t)

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

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

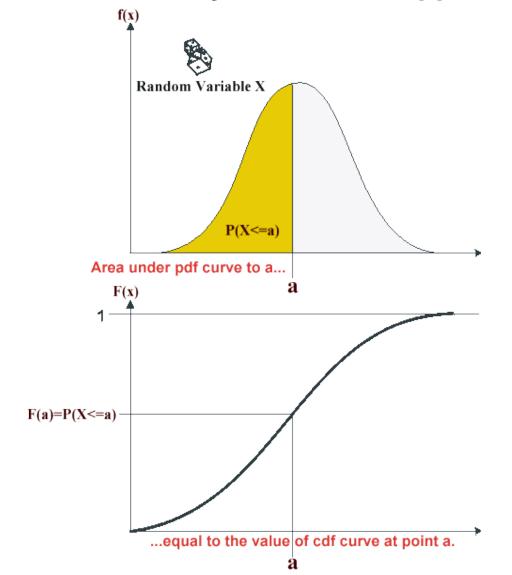


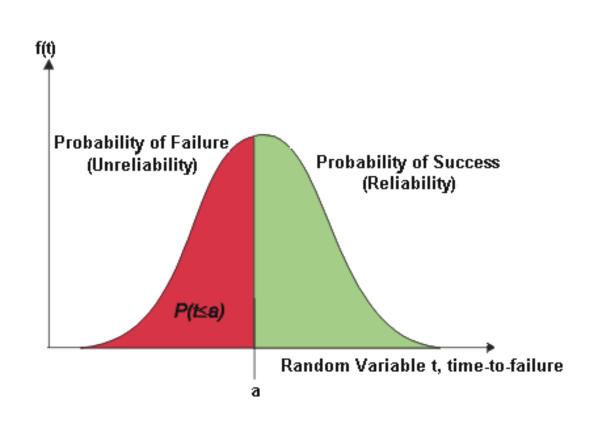






### The density function f(t) and Failure function F(t)









### **Reliability Function**

Several distributions defined by parameters are commonly used for reliability models, including:

- Exponential
- Weibull
- Gamma
- normal (Gaussian)
- log-normal
- log-logistic.

The choice of parametric distribution for a particular application can be made using graphical methods or using formal tests of fit.

The exponential law is well adapted when the failure rate ( $\lambda$ ) is constant, (i.e. independent of time).

The **normal law** represents behaviours where the lifetime of the population is homogeneous, the probability of failure is **centred and symmetrical**. It can be used to model maturity behaviour (without breakdowns) and then rapid wear and tear.

The Weibull's law models each of the three phases of a material's life. It generalises the two previous laws but is more difficult to use and interpret.







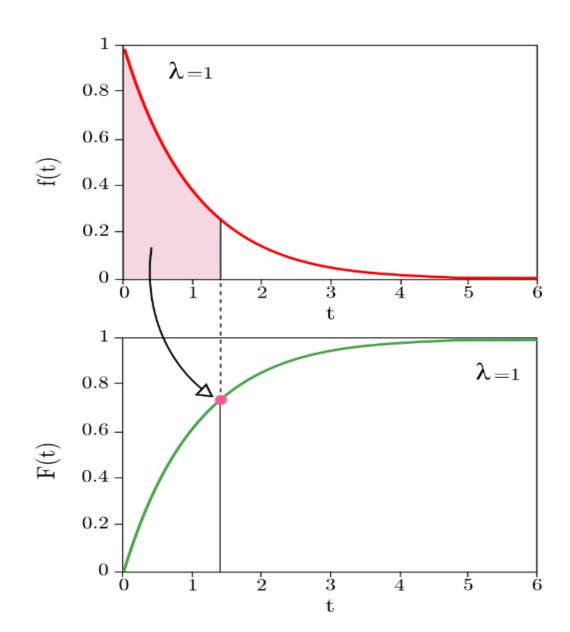
### **Exponential distribution**

F(t) in this particular case with the exponential distribution, is the Failure function of a product when the failure rate is constant. The failure density is linked to the failure rate by the following relation:

$$f(t) = \lambda e^{(-\lambda t)}$$

$$f(t) = \lambda(t). [1 - F(t)]$$

$$\mathbf{F}(\mathbf{t}) = 1 - \mathbf{e}^{(-\lambda.\mathbf{t})}$$

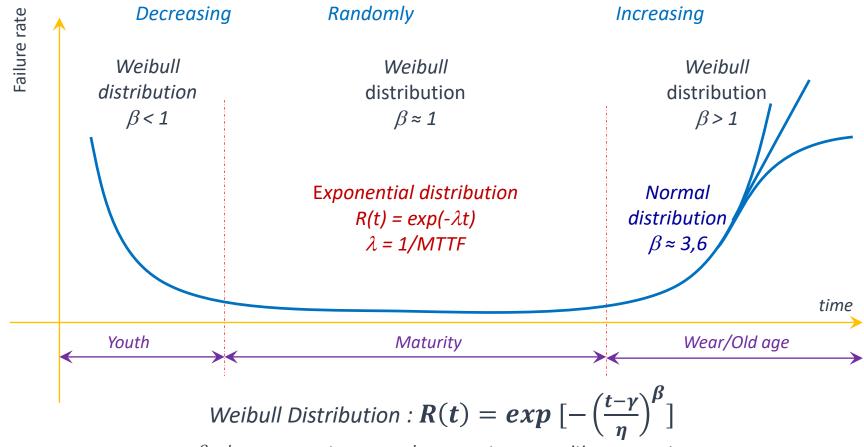








#### Weibull's Law



 $\beta$ : shape parameter;  $\eta$ : scale parameter;  $\gamma$ : position parameter







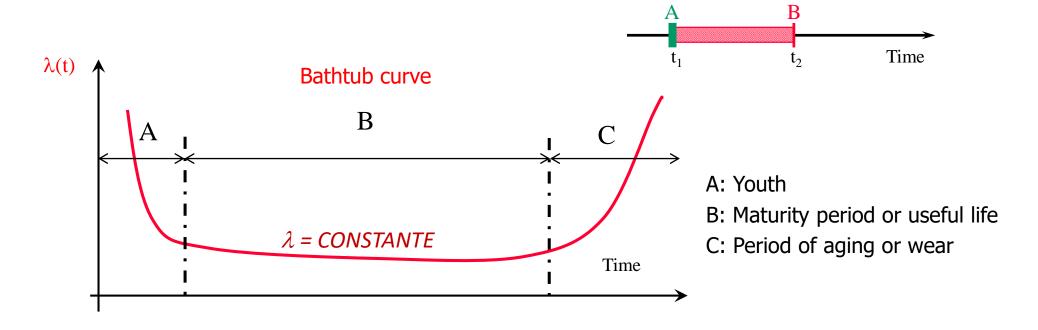
### The failure rate: $\lambda(t)$

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 syst. (failure/product and unit of time)

A: Event "product runs at t1"

B: Event "product fails between t1 and t2"

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







## **Example**

The failure time (T) of an electronic circuit board follows exponentially distribution with failure rate  $\lambda = 10^{-4} / h$ .

- a) What is the probability that it will fail before 1000 h?
- b) What is the probability that it will survive at least 10,000 h
- c) What is the probability that it will fail between 1000 and 10,000 h
  - a) For exponential distribution  $F(T) = 1 e^{-\lambda t}$  F(T = 1000) = 0.09516
  - b) For exponential distribution  $R(T) = e^{-\lambda t}$  R(T = 10,000) = 0.3678
  - **C)** F(10,000) F(1000) = [1 R(10,000)] F(1000) = [1 0.3678] 0.09516 = 0.537





#### **Estimators**

It is obviously not realistic to calculate failure rates by building many units and running them for many hours, under expected operating conditions.

Lecture 1

**Mechatronics common framework** 

This is especially true for well-designed and properly built supplies, with extremely low failure rates, where the number of supplies and hours required to get valid results would be in the thousands.

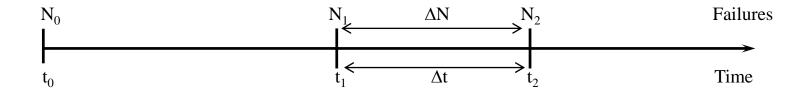
Instead based on representative samples of a population statistical analysis techniques can be used to estimate failure rates.







#### **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

 $\Delta N$ : number of failures on the interval [t1, t2]

 $\Delta t$ : 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&gt;50</th></n></th></n≤20<>	20 <n th="" ≤50<=""><th>N&gt;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

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

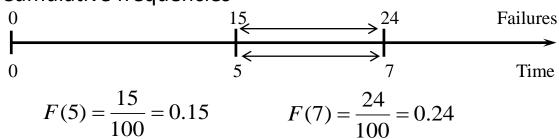
**Average ranks** if 20<N≤50

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

**Cumulative frequencies** if N>50

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

Example (N = 100)  $\rightarrow$  Cumulative frequencies





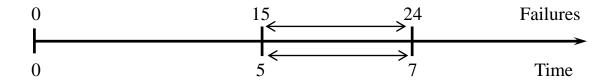


### Reliability estimators: R(t)

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







## Failure rate estimators: $\lambda(t)$

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

Median ranks if 1<N≤20

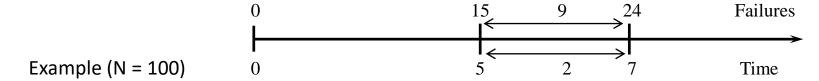
$$\lambda_{[t1;t2[} = \frac{\Delta N}{(N+0.7-N_1)\times \Delta t}$$

Average ranks if 20<N≤50

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

Cumulative frequencies if N>50

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



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





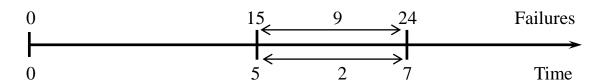


### **Average estimators: E(t)**

#### **Interval** Estimator

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

Example (N = 100)



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















#### **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

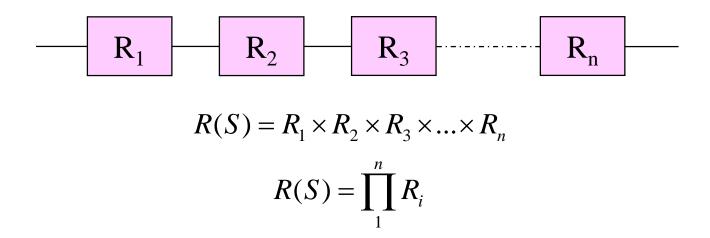
- Decompose the system into components (parts, subsystems ...) and establish the functional links between the components
- Identify component reliability models or collect reliability at a given time
- Search for a model of the system and calculate its reliability







### Reliability of the serial system (S)



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

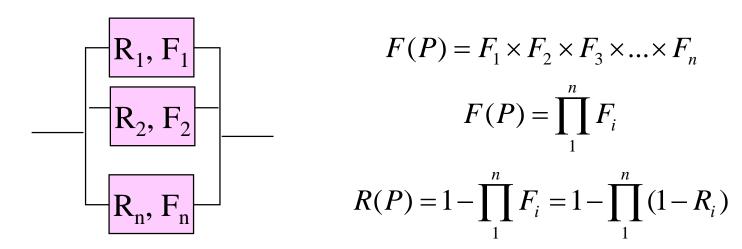






### Reliability of the parallel system (P)

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

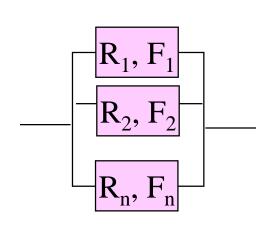






### Reliability of the parallel system (k/n)

"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

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

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

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

 $n \rightarrow total \ number \ of \ components$ 

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

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

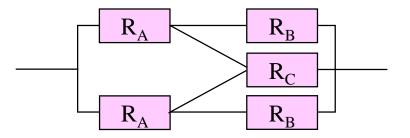
R1, R2, ... Rn and F1, F2, ... Fn are respectively the elementary reliabilities and failures of the components of the system at a given time (binomial law applicable only if the R are identical, otherwise, application of the truth table)



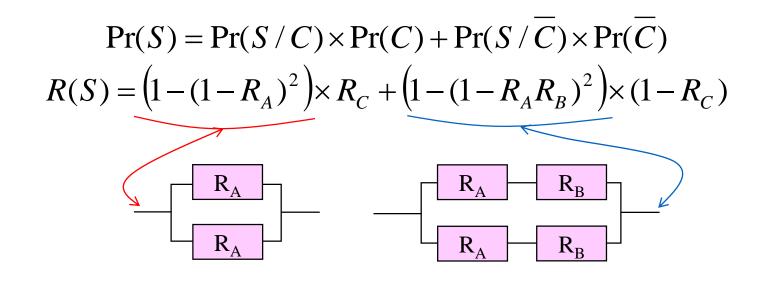




#### Reliability of a complex system



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

The parallel system (n components)

$$R(t) = e^{\left(-t\sum_{i=1}^{n} \lambda_{i}\right)} = e^{\left(-t\lambda_{S}\right)}$$

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

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

# Mechatronics common framework Lecture 1

**Example 1** A personal computer consists of four basic sub systems: motherboard (MB), hard disk (HD), power supply (PS) and processor (CPU). The reliabilities of four subsystems are 0.98, 0.95, 0.91 and 0.99 respectively. What is the system reliability for a mission of 1000 h?

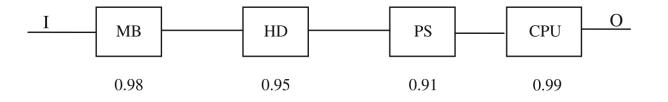
Solution: As all the sub-systems need to be functioning for the overall system success, the RBD is series *configuration* 

The reliability of system is

$$R_{sys} = R_{MB} \times R_{HD} \times R_{PS} \times R_{CPU}$$

$$R_{sys} = 0.98 \times 0.95 \times 0.91 \times 0.99$$

$$R_{sys} = 0.8387$$





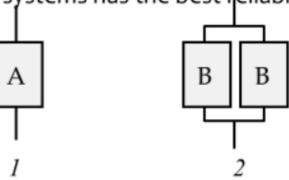


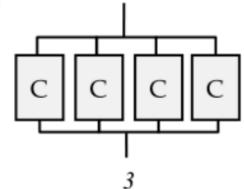




- We consider three components A, B and C of respective prices 150, 75 and 45 €. Be at a given moment, F<sub>A</sub>=0.1, F<sub>B</sub>=0.3 and F<sub>C</sub>=0.4 the respective failures of the three components.
  - Which of the following three systems has the highest reliability?

II. Which of the three systems has the best reliability/price ratio?





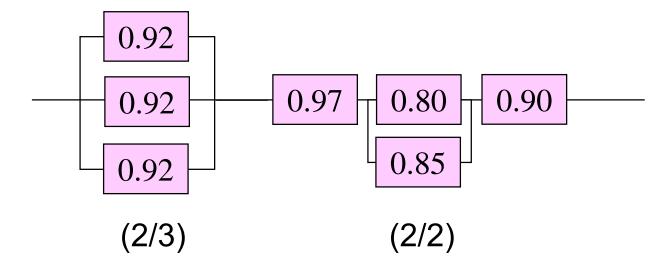






#### **Example**

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



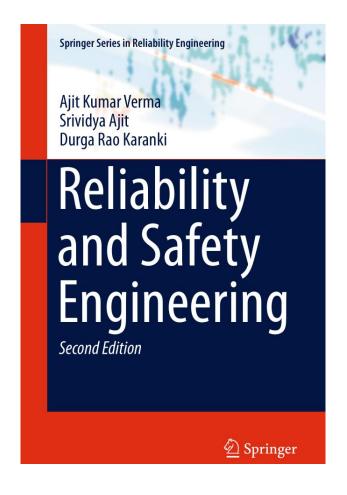
Calculate the predictive reliability of this system

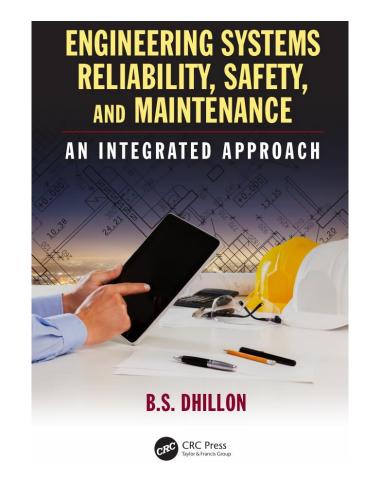






#### Relevant books











#### **Contact Information**

#### Université Savoie Mont Blanc

Polytech' Annecy Chambery Chemin de Bellevue 74940 Annecy France

https://www.polytech.univ-savoie.fr





#### Lecturer

Dr Luc Marechal (luc.marechal@univ-smb.fr)
SYMME Lab (Systems and Materials for Mechatronics)



#### Acknowledgement

Pr Georges Habchi
Pr Christine Barthod
SYMME Lab (Systems and Materials for Mechatronics)
for the original writing of this lecture