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AMMSI

An imperfect repair model based on reduction of  
virtual age and uniform distributed repair degrees

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- 5 Estimation of the Fisher Information
- 6 Simultaneous confidence region based on the likelihood ratio
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## Repair strategies

- Perfect repair (as good as new) : (PR)
- Minimal repair (as bad as old) : (MR)
- Imperfect repair : (IR)

- ▶ Nakagawa (1979-1980) - Optimum Preventive Maintenance policies.

The imperfect PM is modeled in this way: after PM a unit is returned to the "as good as new" state (perfect PM) with probability  $p$  and returned to the "as bad as old" state (minimal PM) with probability  $(1 - p)$ .

- ▶ Brown and Proschan (1983) - Repair is perfect with probability  $p$  and minimal with probability  $(1 - p)$ .
- ▶ Kahle(1991) - Simultaneous confidence region (case of perfect repair and case of minimal repair).
- ▶ Bathe and Franz (1996) - Modelling of repairable systems with various degrees of repair.

- ▶ Last and Szekli (1998) - Stochastic comparison of repairable systems by coupling.
- ▶ Kijima (1988,1989) - Virtual age models or Generalized Renewal Process (GRP).
- ▶ Baxter, Kijima, and Tortorella (1996) - Generalization of Kijima's models.
- ▶ Gasmi, Love and Kahle (2003) - Modelling and estimation of repair effects of complex repairable systems.
- ▶ Doyen and Gaudoin (2004) - ARA and ARI models based on an Arithmetic Reduction of virtual Age or failure Intensity.

- **Aim:** Establish general statistical models for repairable systems
- **Concept:** virtual age  
The first who discovered this process are *Malik* (1979), *Kijima, Morimura and Suzuki* (1988), *Kijima* (1989) and *Stadje & Zuckerman* (1991).
- **Advantages**
  - flexibility for modeling repairable systems.
  - better representation of statistical models of the real situation.

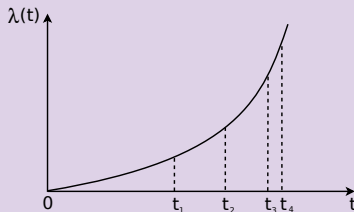
## Basic models

- 1 Minimal maintenance model
- 2 Perfect maintenance model

### (1) Minimal maintenance model

- the maintenance effect is to put the system in operation in the same state just before failure.
- the system is said **(ABAO)**.
- the failure intensity  $\lambda(t)$  does not depend on the past of the process.

### (1) Failure intensity in the case MR





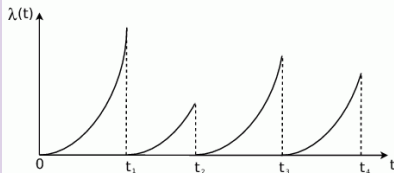
## Basic models

- 1 Minimal maintenance model
- 2 Perfect maintenance model

### (2) Perfect maintenance model

- the maintenance effect is to put the system into operation in the same state as new.
- the system is said **(AGAN)**.
- operating times are independent and identically distributed.
- the failure intensity  $\lambda(t)$  depends on the past of the process.

### (2) Failure intensity in the case PR



# Imperfect maintenance model

## Imperfect maintenance model

- In practice, the situation is between the two extreme cases:
  - ◇ minimal maintenance (ABAO)
  - ◇ perfect maintenance (AGAN)
- Industrial systems are difficult to refurbish after maintenance
- In the industrial field the maintenance has an effect more than minimal

## Remark

*We can note that most of the models concerning the modeling of repairable systems identify the minimal repair and the imperfect repair actions. Naturally, this popular assumption is a very unreal one.*

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## Description of the general model

$$\lambda_1(t) = \lambda(t) \quad \text{for } t \in [0, t_1).$$

▷ At  $t_1$  failure  $\rightarrow$  random degree of repair  $z_1 \in [0, 1]$

$$\text{Virtual age in } t_1 \quad \rightarrow v_1 = z_1 t_1.$$

$$\lambda_2(t) = \lambda(t - t_1 + v_1) \quad \text{with } t \in [t_1, t_2).$$

▷ At  $t_2$  failure  $\rightarrow$  random degree of repair  $z_2 \in [0, 1]$

$$\text{Virtual age in } t_2 \quad \rightarrow v_2 = z_2(v_1 + t_2 - t_1).$$

⋮

## Description of the general model

$$\lambda_{k+1}(t) = \lambda(t - t_k + v_k) \quad \text{for } t_k \leq t < t_{k+1}$$

$$v_k = z_k(v_{k-1} + t_k - t_{k-1}), \quad v_0 := 0; \quad t_0 := 0;$$

## Definition

$v(t) := t - t_k + v_k$ ,  $t_k \leq t < t_{k+1}$ ,  $k \geq 1$ , is called **the virtual age Process**.

# Virtual Age Process

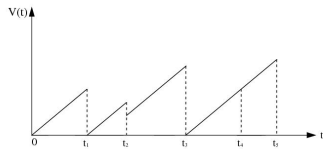


Figure 1: Virtual Age Process

# Failure Intensity

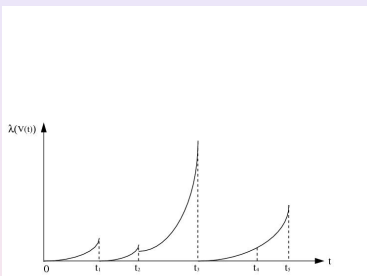


Figure 2: Failure Intensity

## Special cases Model

- Perfect maintenance model
- Minimal maintenance model
- Kijima's models (1989)
- Brown and Proschan model (1983)
- Stadje and Zuckerman model (1991)
- Reliability model with alternating repairs (Gasmi (2011))



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# Assumptions of the model

- ① Repairs affect the failure intensity at any instant via a virtual age process from type Kijima 2.
- ② After failure, one of the three following cases is possible:
  - ◇ perfect repair
  - ◇ minimal repair
  - ◇ imperfect repair with uniform distributed degree of repair
- ③ All repair times are small and can be neglected.
- ④ The baseline failure intensity of the system is from Weibull type:

$$\lambda(x, \theta) = \frac{\beta}{\alpha} \left( \frac{x}{\alpha} \right)^{\beta-1}, \alpha > 0, \beta > 0,$$

with  $\theta = (\alpha, \beta)$ ,  $\alpha$  scale parameter,  $\beta$  form parameter.



# The imperfect repair model based on uniform distributed repair degrees

- $(t_k)_{k=1,2,\dots}$  failure times
- $(z_k)_{k=1,2,\dots}$  repair degrees
- $N(t) = \sum_{k=1}^{\infty} \mathbf{1}(t_k \leq t)$  the number of failures until  $t$  for the failure repair process, with  $\mathbf{1}(A)$  is the indicator function of  $A$

- ◇ If the degree of repairs  $z_k \in (0, 1)$ ,  $\forall k = 1, \dots, N(t)$  we obtain imperfect repairs. (model 1)
- ◇ If the degree of repairs  $z_k = 0$ ,  $\forall k = 1, \dots, N(t)$  we obtain only perfect repairs. (model 2)
- ◇ If the degree of repairs  $z_k = 1$ ,  $\forall k = 1, \dots, N(t)$  we obtain only minimal repairs. (model 3)

# The imperfect repair model based on uniform distributed repair degrees

## Description

We consider a marked point process  $\Phi = ((t_k, z_k))$ .  $\Phi$  is described by:

- 1 the counting process  $\{N(t), t \geq 0\}$  and the corresponding intensities  $\lambda(v(t), \theta)$  with  $V(t) := t - t_k + v_k, \quad t_k \leq t < t_{k+1}, \quad k = 1, 2, \dots$  is the virtual age process.
- 2 the marks  $z_k$  are repair degrees at  $t_k$ .

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## The LL function

The log-likelihood function for observation of point processes is of the form (Liptser and Shirayev (1978))

$$\ln L(t; \theta) = \sum_{k=1}^{N(t)} \ln \lambda(v(t_k-)) - \int_0^t \lambda(v(s)) ds,$$

where  $v(t_k-) = t_k - t_{k-1} + v_{k-1}$ .

## The LL function

We obtain:

$$\begin{aligned}\ln L(t; \theta) &= (\beta - 1) \sum_{i=1}^{N(t)} \ln(v_{k-1} + t_k - t_{k-1}) + \\ &N(t)(\ln \beta - \beta \ln \alpha) - \frac{1}{\alpha^\beta} S_1.\end{aligned}$$

Where

$$S_1 = \sum_{k=1}^{N(t)} \left\{ (v_{k-1} + t_k - t_{k-1})^\beta - (v_k)^\beta \right\} + (t - t_{N(t)} + v_{N(t)})^\beta.$$

The MLE of the parameters  $\alpha$  and  $\beta$  are obtained by solving the nonlinear system:

$$\frac{\partial \ln L(t; \theta)}{\partial \alpha} = 0$$

and

$$\frac{\partial \ln L(t; \theta)}{\partial \beta} = 0.$$



- The estimator of the scale parameter  $\alpha$  is explicitly determined:

$$\hat{\alpha} = \left( \frac{S_1 | \hat{\beta}}{N(t)} \right)^{1/\hat{\beta}}$$

- $\hat{\alpha}$  involves the usual parameter estimation in terms of Weibull intensities.
- This estimator depends on the virtual age of the system and the number of failures  $N(t)$ .

- The estimator of the shape parameter  $\beta$  can be found by numerical solve of the following equation:

$$\frac{1}{\hat{\beta}} + \frac{1}{N(t)} \left\{ \sum_{k=1}^{N(t)} \ln(v_{k-1} + t_k - t_{k-1}) \right\} - \frac{S_2|_{\hat{\beta}}}{S_1|_{\hat{\beta}}} = 0.$$

Where

$$S_2 = \sum_{k=1}^{N(t)} \left\{ (v_{k-1} + t_k - t_{k-1})^\beta \ln(v_{k-1} + t_k - t_{k-1}) - v_k^\beta \ln v_k \right\} \\ + (t - t_{N(t)} + v_{N(t)})^\beta \ln(t - t_{N(t)} + v_{N(t)}).$$

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# Estimation of the Fisher Information

- Because the MLE of the vector  $\theta = (\alpha, \beta)$  is not obtained in closed form, it is not possible to derive the exact distribution of the MLE.
- An approximation of the Fisher information matrix  $I(\theta)$  is given.
- In this case  $n$  independent failure repair processes are observed.
- Let  $\theta = (\theta_1, \theta_2)$  with  $\theta_1 = \alpha$  and  $\theta_2 = \beta$ .
- The elements of the 2x2 matrix  $I(\theta)$ ,  $I_{r,s}(\theta)$ ,  $r, s = 1, 2$ , can be

$$\widehat{I_{r,s}(\theta)} = -\frac{1}{n} \sum_{l=1}^n \frac{\partial^2 \ln L_l(t; \theta)}{\partial \theta_r \partial \theta_s}.$$

# Estimation of the Fisher Information

- Let  $l \in \{1, 2, \dots, n\}$ .

## Notations:

- $L_l(t; \theta)$  – the likelihood function of the  $l$ -th failure repair process.
- $N^l(t)$  – the number of failures until  $t$  for the  $l$ -th failure repair process.
- $t_{l,1}, \dots, t_{l,N^l(t)}$  – failure times of the  $l$ -th failure repair process.
- $x_{l,1}, \dots, x_{l,N^l(t)}$  – operating times of the  $l$ -th failure repair process.
- $v_{l,1}, \dots, v_{l,N^l(t)}$  – virtual age of the  $l$ -th failure repair process.

## Definition

*Let  $i = 1, 2, \dots, N^l(t)$ ,  $x_{l,i} = t_{l,i} - t_{l,i-1}$  is the operating time between two successive failures of the  $l$ -th failure repair process.*

# Estimation of the Fisher Information

- The observed information matrix  $I$  for this model is given by:

$$I = \begin{pmatrix} -\frac{1}{n} \sum_{l=1}^n \frac{\partial^2 \ln L_l(x, \theta)}{\partial \alpha^2} & -\frac{1}{n} \sum_{l=1}^n \frac{\partial^2 \ln L_l(x, \theta)}{\partial \alpha \partial \beta} \\ -\frac{1}{n} \sum_{l=1}^n \frac{\partial^2 \ln L_l(x, \theta)}{\partial \beta \partial \alpha} & -\frac{1}{n} \sum_{l=1}^n \frac{\partial^2 \ln L_l(x, \theta)}{\partial \beta^2} \end{pmatrix}$$

- The variance-covariance matrix  $V$  is the inversion of the observed information matrix  $I$

$$V = \begin{pmatrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{pmatrix} = I^{-1}$$

# Estimation of the Fisher Information

- It follows then that the asymptotic distribution of the MLE  $(\hat{\alpha}, \hat{\beta})$  (Miller (1981)):

$$\begin{pmatrix} \hat{\alpha} \\ \hat{\beta} \end{pmatrix} \sim N \left( \begin{pmatrix} \alpha \\ \beta \end{pmatrix}, \begin{pmatrix} V_{11} & V_{12} \\ V_{21} & V_{22} \end{pmatrix} \right)$$

- If we replace the parameters  $\alpha, \beta$  by the corresponding MLE's, we get then an estimate of the variance-covariance matrix  $V$ , denoted by  $\hat{V}$  and defined as follows:

$$\hat{V} = \begin{pmatrix} \widehat{l_{11}} & \widehat{l_{12}} \\ \widehat{l_{21}} & \widehat{l_{22}} \end{pmatrix}^{-1}$$

with  $\widehat{l_{ij}} = l_{ij}$  are obtained if we replace  $(\alpha, \beta)$  by  $(\hat{\alpha}, \hat{\beta})$ .

# Estimation of the Fisher Information

- We obtain then approximate  $100(1 - \nu)\%$  confidence intervals for the parameters  $\alpha, \beta$  respectively as:

$$\hat{\alpha} \pm z_{\frac{\nu}{2}} \sqrt{\widehat{V}_{11}}, \quad \hat{\beta} \pm z_{\frac{\nu}{2}} \sqrt{\widehat{V}_{22}},$$

where  $z_{\frac{\nu}{2}}$  is the upper  $\frac{\nu}{2}$ -th percentile of the standard normal distribution.



## Theorem

$$\widehat{I_{1,1}(\theta)} = \frac{\hat{\beta}^2}{n\hat{\alpha}^2} \sum_{l=1}^n N^l(t),$$

$$\widehat{I_{1,2}(\theta)} = -\frac{1}{n\hat{\alpha}} \sum_{l=1}^n \left\{ N^l(t)(1 - \hat{\beta} \ln \hat{\alpha}) + \hat{\beta} \sum_{i=1}^{N^l(t)} \ln(x_{l,i} + v_{l,i-1}) \right\},$$

$$\widehat{I_{2,2}(\theta)} = \frac{1}{n} \sum_{l=1}^n \left\{ N^l(t) \left( \frac{1}{\hat{\beta}} - \ln \hat{\alpha} \right)^2 - 2 \ln \hat{\alpha} \sum_{i=1}^{N^l(t)} \ln(x_{l,i} + v_{l,i-1}) + \frac{1}{\hat{\alpha}\hat{\beta}} W_{3,l}(t, \hat{\beta}) \right\}.$$

Where

$$W_{3,l}(t, \hat{\beta}) = \sum_{i=1}^{N'(t)} \left\{ (x_{l,i} + v_{l,i-1})^{\hat{\beta}} \ln^2(x_{l,i} + v_{l,i-1}) - (v_{l,i})^{\hat{\beta}} \ln^2(v_{l,i}) \right\} \\ + (t - t_{l,N'(t)} + v_{l,N'(t)})^{\hat{\beta}} \ln^2(t - t_{l,N'(t)} + v_{l,N'(t)}).$$

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## LR

- Method: Likelihood ratio (LR)
- Under regularity conditions (Barndorff-Nielsen and Blaesild, (1986)) the log-likelihood ratio (LLR)  
 $q = 2 \left\{ \ln L(x, \hat{\theta}) - \ln L(x, \theta) \right\}$  converges in distribution to a central  $\chi^2$  distribution with 2 degrees of freedom.
- The simultaneous confidence region is defined by the inequality  $q \leq \chi_{1-\mu, 2}^2$ .
- $\chi_{1-\mu, 2}^2 = -2 \ln \mu$  is the  $(1 - \mu)$ - quantile of the  $\chi^2$ - distribution with 2 degrees of freedom.

# Simultaneous confidence region based on the likelihood ratio

- We study the case of  $n$  independent failure repair processes.
- The border of the simultaneous confidence region using the likelihood ratio is given as follows:

$$2\{\ln L_r(t; \hat{\theta}) - \ln L_r(t; \theta)\} = -2 \ln \mu.$$

- Notations:

- $N^r(t)$  - the number of failures until  $t$  for the  $r$ -th failure repair process.
- $t_{r,1}, \dots, t_{r,N^r(t)}$  - failure times of the  $r$ -th failure repair process.
- $v_{r,1}, \dots, v_{r,N^r(t)}$  - virtual ages of failures until  $t$  for the  $r$ -th failure repair process.

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# Simultaneous confidence region based on the likelihood ratio

- We obtain then the following simultaneous confidence region for  $n$  independent failure repair processes:

$$\frac{1}{\alpha^\beta} \sum_{r=1}^n S_{r,1} + (\hat{\beta} - \beta) \sum_{r=1}^n \sum_{k=1}^{N^r(t)} \ln(v_{r,k-1} + t_{r,k} - t_{r,k-1}) \\ + (\ln \hat{\beta} - \ln \beta - 1 - \hat{\beta} \ln \hat{\alpha} + \beta \ln \alpha) \sum_{r=1}^n N^r(t) = -\ln \mu.$$

where

$$S_{r,1} = \sum_{k=1}^{N^r(t)} \left\{ (v_{r,k-1} + t_{r,k} - t_{r,k-1})^\beta - (v_{r,k})^\beta \right\} + (t - t_{N^r(t)} + v_{N^r(t)})^\beta.$$

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## Example 1

- The simulation study to generate data of the IR process and following the Kijima type 2 repair model with uniform distributed degrees of repair.
- Let  $\alpha = 1.2$  and  $\beta = 3$ .
- We obtain  $N(t) = 10$  failures until the time  $t = 10$ .

$t_k$	1.3515	1.8118	2.4047	3.3637	3.4218
$z_k$	0.4159	0.4900	0.3486	0.6980	0.1740
$t_k$	4.1991	4.6912	5.6913	7.9375	8.0719
$z_k$	0.2247	0.8906	0.1236	0.1357	0.4553

Table 1: Failure times and repair degrees for  $\alpha = 1.2$  and  $\beta = 3$

- A sample 1 was observed until  $t = 100$ . Let  $\alpha = 1.2$ ,  $\beta = 3$  and  $s = 10$ , where  $s$  is the number of simulations.

$\hat{\alpha}$	$\hat{\beta}$	LL
1.1907	2.9978	-25.0579
1.2014	2.9880	-24.9123
1.2337	3.0314	-26.5588
1.2216	3.1723	-26.7767
1.2537	3.0977	-28.6813
1.2158	2.9855	-26.9009
1.1789	2.9456	-26.6481
1.1666	2.9134	-28.7947
1.1855	3.0615	-25.9890
1.2104	2.9219	-26.1210

Table 2: Estimations of  $\alpha$ ,  $\beta$  and LL from data of sample 1

- The mean squared errors (MSE) of  $\hat{\alpha}$  and  $\hat{\beta}$  are given in Table 3.

s	50	100	500	1000
MSE( $\hat{\alpha}$ )	0.0022	0.0018	0.0017	0.0015
MSE( $\hat{\beta}$ )	0.0993	0.0542	0.0514	0.0504

Table 3: MSE of  $\hat{\alpha}$  and  $\hat{\beta}$

- We could remark that if the number of simulations  $s$  increases, then the mean squared errors of  $\hat{\alpha}$  and  $\hat{\beta}$  decrease.

- A sample 2 was observed until  $t = 100$ . Let  $\alpha = 1.2$ ,  $\beta = 3$  and  $s = 200$ , where  $s$  is the number of simulations.
- The bias and variance of the estimators are estimated by their empirical version on 200 replicates.
- The estimations are given in Table 4.

	$\hat{\alpha}$	$\hat{\beta}$
Estimation	1.1903	2.9896
Empirical mean	1.2001	3.0370
Empirical variance	0.0016	0.0541

Table 4: Estimations results considering an average of 200 simulations

- A sample 1 was observed until  $t = 100$ . Let  $\alpha = 1.2$ ,  $\beta = 3$ .
- The parameter estimates are  $\hat{\alpha} = 1.1907$  and  $\hat{\beta} = 2.9978$  and the  $LL = -25.0579$ .
- Figure 3 illustrates the LL function with respect to  $\alpha$  and  $\beta$ .

# Simulation study

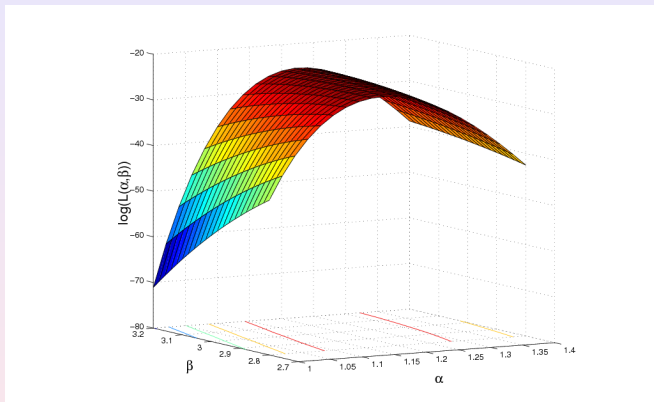


Figure 3: Graph of the LL function for one sample with respect to  $\alpha$  and  $\beta$



- Figure 4 illustrates simultaneous confidence region of the parameter estimations  $\hat{\alpha} = 1.2030$  and  $\hat{\beta} = 3.0375$  by given  $\mu = 0.05$ .
- for  $n = 50$  (curve in dash).
- for  $n = 100$  (curve in line).

# Simulation study

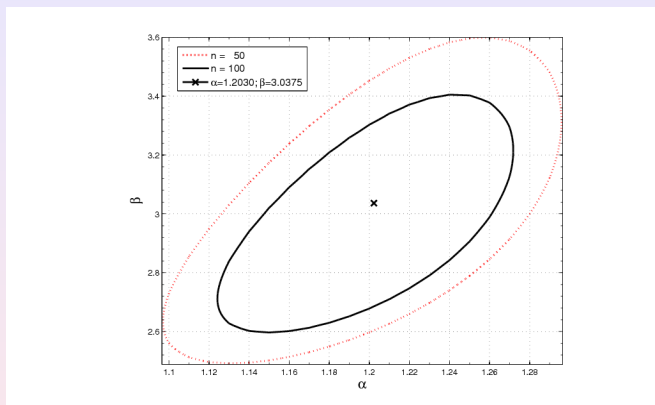


Figure 4: Simultaneous confidence region for the estimations of  $\alpha = 1.2$  and  $\beta = 3$

- The simultaneous confidence region based on the likelihood ratio for  $n = 100$  is smaller than that for  $n = 50$ .
- In the case of  $n = 100$ , the parameter  $\alpha$  varies between 1.124 and 1.270 and the parameter  $\beta$  varies between 2.6 and 3.4.
- In the case of  $n = 50$ , the parameter  $\alpha$  varies between 1.096 and 1.296 and the parameter  $\beta$  varies between 2.5 and 3.6.

## Example 2

- Figure 5 illustrates simultaneous confidence region of the parameter estimations  $\hat{\alpha} = 1.4955$  and  $\hat{\beta} = 3.4975$  by given  $\mu = 0.05$ .
- for  $n = 50$  (curve in dash).
- for  $n = 100$  (curve in line).

## Example 2

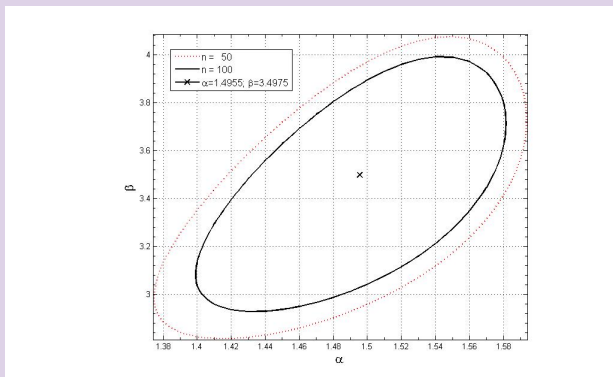


Figure 5: Simultaneous confidence region for the estimations of  $\alpha = 1.5$  and  $\beta = 3.5$

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# Illustrative Example

- We provide a data analysis to investigate how this model works in practice.
- We illustrate the modeling and estimation procedure.
- A well known data on airplane air-conditioning failures on a fleet of Boeing aircraft (Plane 7914) given in Hollander and Wolfe are considered.
- For the data from Plane 7914, the number of failures is  $N(t)=24$ .
- The rest time after the last failure is assumed to be equal to zero.

# Illustrative Example

- Our objective is to compare the model with uniform distributed degree of repair, denoted by **model 1** and models using a fixed degree of repair.
- Models using a fixed degree of repair:
  - ① The **model 2** (perfect repairs); RP.
  - ② The **model 3** (minimal repairs); NHPP.
  - ③ The **model 4** (average repairs); (degrees of repair  $z_k = 0.5$ ).



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# Illustrative Example

- Table 5 gives the inter-failure times for Plane 7914 and the uniform distributed degrees of repair used in model 1.

$i$	1	2	3	4	5	6	7	8
$x_i$	50	44	102	72	22	39	3	15
$z_k$	0.83	0.63	0.54	0.65	0.73	0.09	0.87	0.01
$i$	9	10	11	12	13	14	15	16
$x_i$	197	188	79	88	46	5	5	36
$z_k$	0.29	0.18	0.93	0.07	0.58	0.64	0.65	0.86
$i$	17	18	19	20	21	22	23	24
$x_i$	22	139	210	97	30	23	13	14
$z_k$	0.05	0.81	0.53	0.69	0.21	0.54	0.70	0.96

**Table 5:** Inter-failure times from Plane 7914 and degrees of repair (model 1)

# Illustrative Example

- For comparison purpose, we use the mean square of the difference between the empirical cdf and the fitted cdf, say MSD.

$$MSD = \frac{1}{N(t)} \sum_{k=1}^{N(t)} \left( \hat{F}_k - F_{E,k} \right)^2,$$

- $\hat{F}_k$  ... the empirical cdf computed at the cumulative failure times  $t_k$ .
- $F_{E,k}$  ... the estimated cdf computed at the cumulative failure times  $t_k$ .

# Illustrative Example

- The ML estimates of the parameters  $\alpha$  and  $\beta$ , the MSD and the LL values are given in Table 6.

	$\hat{\alpha}$	$\hat{\beta}$	MSD	LL
Model 1	56.8875	0.9224	0.0010	-123.8080
Model 2	65.4085	1.0339	0.0014	-123.9939
Model 3	82.9261	1.0880	0.0057	-123.9455
Model 4	55.8537	0.9138	0.0011	-123.7917

Table 6: Estimations of  $\alpha$  and  $\beta$ , MSD and LL from data of Plane 7914

- For all introduced models, the empirical, the estimated cdf and the 95% lower and upper confidence bounds for the cdf of the data from Plane 7914 are shown in Figure 6.

# Simulation study

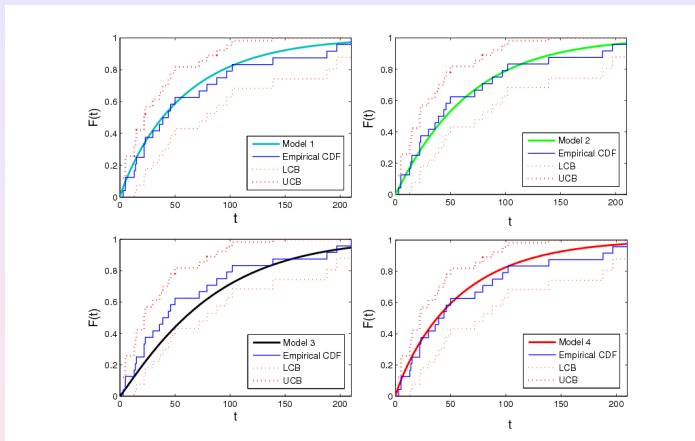


Figure 6: Cdf and empirical cdf of data from Plane 7914 for models 1-4

Based on Table 6 and Figure 6 we can conclude that:

- 1 Model 4 fits the data better than Model 2 and Model 3.
- 2 Model 1 fits the data better than Model 4.
- 3 Model 3 gives the worst fit of the data.
- 4 Model 1 concord the data better than all other models.

Based on Table 6 and Figure 6 we can conclude that:

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## Conclusions

- 1 Parameter estimation of an imperfect repair model is proposed.
- 2 Simultaneous confidence region based on the likelihood ratio for the parameters of the Weibull intensity is developed in the case of uniform distributed degrees of repair.
- 3 The obtained results are applied on sets of simulated data.
- 4 Interesting results on real data are obtained.

## Future Work








- 1 Study other models.
- 2 Change the distribution of repair degrees.
  - ◇ Beta distribution with equal shape parameters.
  - ◇ Beta distribution by fixing one of its parameters.
  - ◇ Beta distribution.
- 3 Include time dependent repair effectiveness.
- 4 Study the Kijima's type 1 virtual age imperfect repair model and compare results.
- 5 Compare models with AIC criterion.

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Thank you for your attention