

# Stochastic Maintenance Optimization of Complex Industrial Systems under Uncertainty of Repair Time and Resources

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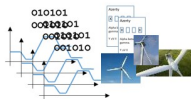
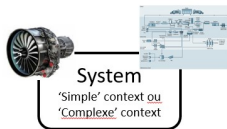
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# INTRODUCTION

# Context : MPO Project

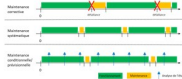
Technical and economic criteria  
for maintenance objectives



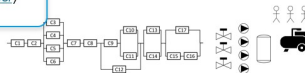
Collect and evaluate complex and  
heterogenous data for the maintenance



Methods and tools for diagnosis,  
prognosis  
(model-based and data-based)



The MPO project (Predictive Maintenance and Optimization)  
IRT SystemX (<https://www.irt-systemx.fr/en/projets/mpo/>)  
Over a period of 4 years (started in September 2018)



Optimizing maintenance scheduling (correctives and preventives) according to technical and economic criteria and execution results (analytic or simulation)

Consider different relevant maintenance according to results of monitoring, diagnosis and prognosis, and the technical and economic criteria

# Maintenance Planning

A maintenance strategy determines the:

- 1 Components requiring preventive maintenance (PM).
- 2 Optimal frequency of PM actions.
- 3 Maintenance interval.
- 4 Failure risks.
- 5 Maintenance cost.

# Assumptions

01

Multi-component systems

02

Corrective Maintenance (ABAO)  
Preventive Maintenance (AGAN)

03

Resources +Repairmen

04

One Degradation Mode

⚠ However this may not always be the reality !

# Maintenance Constraints



# OPTIMAL INDIVIDUAL MAINTENANCE DATES

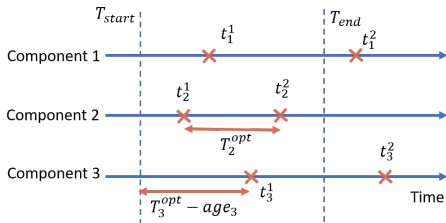


# Optimal Individual Maintenance Dates

## Individual maintenance dates

- We determine the individual maintenance date  $T_i^{opt}$
- Scheduling maintenance tasks

$$t_i^1 = T_i^{opt} - age_i + T_{start} \quad (1)$$



# PROBLEM FORMULATION

# Problem identification

We consider:

- A system with  $n$  components

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- A Planning horizon  $[0, T]$

We define the following sets:

- $\mathcal{J}$  : Planning Horizon Set  $\mathcal{J} = 1, 2, \dots, T$

# Problem identification

We consider:

- A system with  $n$  components
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- A Planning horizon  $[0, T]$

We define the following sets:

- $\mathcal{J}$  : Planning Horizon Set  $\mathcal{J} = 1, 2, \dots, T$
- $\mathcal{I}$  : Maintenance components Set  $\mathcal{I} = 1, 2, \dots, n$

# Parameters

- $C_i^{prev}$  : component-specific PM action cost
- $C_i^{correc}$  : component-specific CM action cost
- $\alpha_i, \beta_i$ : Weibull Law shape and scale parameters for component  $i$
- $d_i$  : maintenance action duration of component  $i$
- $r_i$  : the number of resources required for the maintenance action of component  $i$
- $R$  : The total available resources in the system
- $R_j$  : The total available resources at time  $j$



# Variables

## 1 Binary Variables:

$$m_{ij} = \begin{cases} 1, & \text{if object } i \text{ is maintained at instant } j \\ 0, & \text{otherwise} \end{cases}$$

## 2 Continuous Variables:

- $x_i$  : maintenance date of component  $i$

## Objective Function

- $cost(x_i)$  is the cost of maintenance of component  $i$  (at maintenance time  $x_i$ ):

$$cost(x_i) = \frac{C_i^{prev} + C_i^{correc} \left(\frac{x_i}{\alpha_i}\right)^{\beta_i}}{x_i} \quad x_i \neq 0 \quad (2)$$

- Minimize the total cost of maintenance (a non linear function):

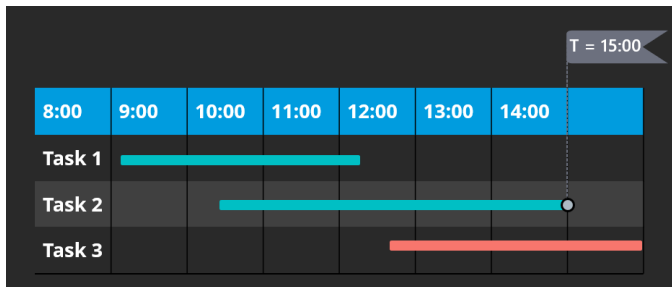
$$F(x) = \sum_{i=1}^n cost(x_i). \quad (3)$$

# Constraints

## Maintenance duration

- If a maintenance action for component  $i$  starts at  $j$ , it must be executed within the planning horizon.

$$jm_{ij} + (d_i - 1) \leq T \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T} \quad (4)$$



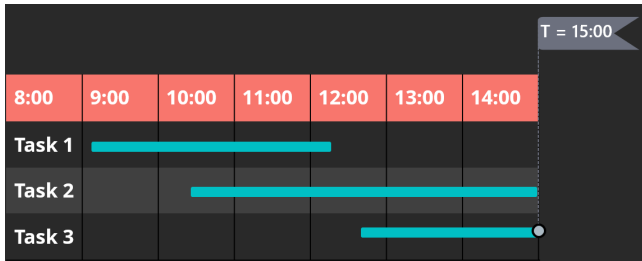
# Constraints

## Maintenance frequency

- Each component must be maintained once.

$$\sum_{j=1}^T m_{ij} \geq 1 \quad \forall i \in \mathcal{I} \quad (5)$$

$$m_{ij'} \leq 1 - m_{ij} \quad \forall j' > j \in \mathcal{T} \quad \forall i \in \mathcal{I} \quad (6)$$

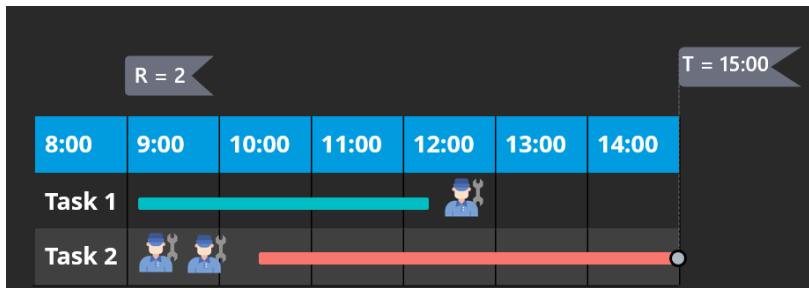


# Constraints

## Resource constraints

- A component can be maintained if and only if there are available resources.

$$\sum_{i=1}^n m_{ij} r_i \leq R_j, \forall j \in \mathcal{T} \quad (7)$$

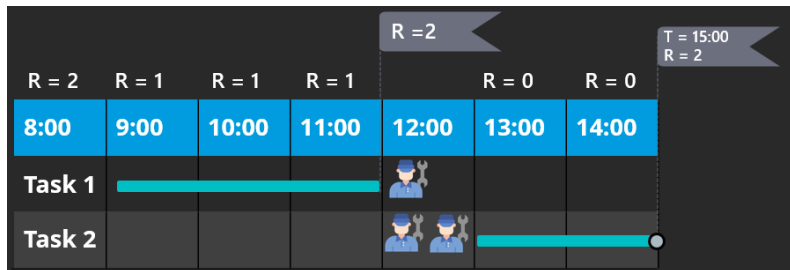


# Constraints

## Resources release constraints:

- Used resources are restored.
- PM actions can be executed simultaneously.

$$m_{ij}r_i \leq R - \sum_{\substack{k=0 \\ k \neq i}}^n \sum_{\substack{j' < j \\ j' + d_k > j}}^T m_{kj'}r_k \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T} \quad (8)$$



# MNLP Model

$$\min F(x) = \sum_{i=1}^n \text{cost}(x_i)$$

$$\text{s.t.} \quad jm_{ij} + (d_i - 1) \leq T \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T}$$

$$\sum_{j=1}^T m_{ij} \geq 1 \quad \forall i \in \mathcal{I}$$

$$\sum_{i=1}^n m_{ij} r_i \leq R_j \quad \forall j \in \mathcal{T}$$

$$m_{ij} r_i \leq R - \sum_{\substack{k=0 \\ k \neq i}}^n \sum_{\substack{j' < j \\ j' + d_k > j}}^T m_{kj'} r_k \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T}$$

$$m_{ij'} \leq 1 - m_{ij} \quad \forall i \in \mathcal{I} \quad \forall j' > j \in \mathcal{T}$$

$$x_i \leq m_{ij} T + \epsilon \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T} \quad \epsilon > 0$$

$$x_i \geq 0 \quad \forall i \in \mathcal{I}$$

$$m_{ij} \in \{0, 1\} \quad \forall i \in \mathcal{I} \quad \forall j \in \mathcal{T}$$

# NUMERICAL SIMULATIONS



# Simulation settings

## Remark

- Pyomo optimization language
- The Mixed-Integer Nonlinear Decomposition Toolbox (MindtPy) solver



## Example 1

We consider a system with 5 components using the following parameters:

- $T = 20, R = 18$



Component 1



Component 2



Component 3



Component 4

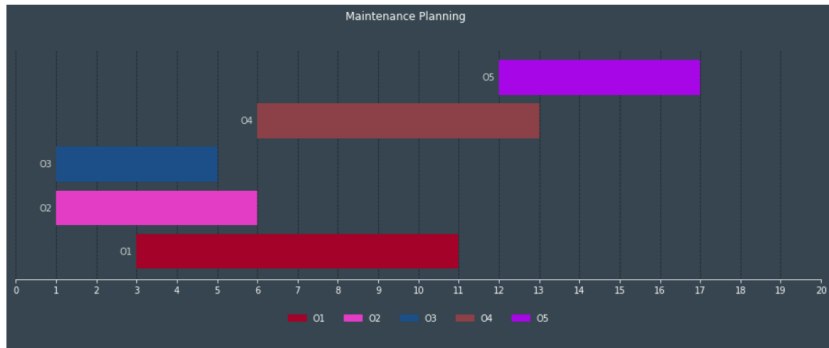


Component 5



# Example 1

We obtain the following planification



## Example 2

We consider a system with 10 components using the following parameters:

- $T = 25$ ,  $R = 17$



Component 1



Component 2



Component 3



Component 4



Component 5



Component 6



Component 7



Component 8



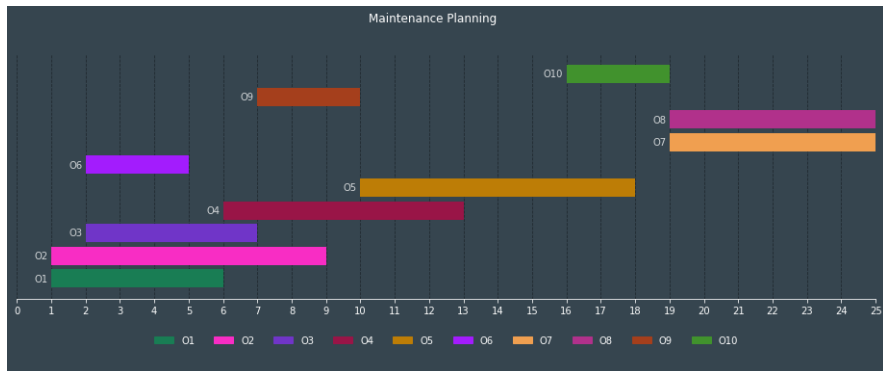
Component 9



Component 10

## Example 2

We obtain the following result :



# Performance Assessment

- Randomly generated systems
- Cost parameters
- Considered metrics:
  - ① Execution time
  - ② Maintenance Cost
  - ③ Convergence Rates
  - ④ Scalability

# Performance Assessment

Table: Simulation parameters

<b>Parameters</b>	$x_i$	$t_i$	$C_i^{Prev}$	$C_i^{correc}$	$r_i$	$d_i$	$R_j$
<b>Values</b>	[10, 40]	[1.25, 2.75]	[40, 150]	[140, 350]	[1, 10]	[2, 8]	[0, 20]

# Performance Assessment

- Algorithms performance for complex systems with  $n=5$  components

		Planning Horizon						
		$T = 25$	$T = 35$	$T = 45$	$T = 55$	$T = 65$	$T = 75$	$T = 100$
MNLP	$C(x)$	<b>1126.24</b>	653.28	626.50	<b>487.51</b>	494.34	814.06	652.95
	$\tau_{system}$	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>
	Time (s)	0.628	1.178	1.085	1.829	<b>2.289</b>	2.292	3.177

- Algorithms performance for complex systems with  $n=10$  components

		Planning Horizon						
		$T = 25$	$T = 35$	$T = 45$	$T = 55$	$T = 65$	$T = 75$	$T = 100$
MNLP	$C(x)$	<b>1288.19</b>	900.50	<b>355.99</b>	412.182	1272.09	480.73	937.107
	$\tau_{system}$	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>	<b>100%</b>
	Time (s)	2.091	4.771	4.624	9.012	5.872	<b>9.027</b>	11.027



Stochastic Maintenance  
Optimization of  
Complex Industrial  
Systems under  
Uncertainty of Repair  
Time and Resources

# Stochastic Maintenance Optimization

## Problem formulation

- Let  $O$  be the set of system's components.
- For each component  $o_i \in O$  we associate a duration random variable  $d_i$  as Gaussian probabilistic distribution :  $\forall o_i \in O, d_i \sim \mathcal{N}(\mu(d_i), \sigma^2(d_i))$
- At each time step  $j$ , we assume  $R_j$  as a random variable following a Gaussian distribution :  
 $\forall j \in T, R_j \sim \mathcal{N}(\mu(R_j), \sigma^2(R_j))$

# Stochastic Maintenance Optimization

## Problem formulation

- Uncertainty of the repair time

$$\mathbb{P}(jm_{ij} + (d_i - 1) \leq T) \geq \alpha_i, \forall i \in \mathcal{I}, \forall j \in \mathcal{T} \quad (9)$$

- Uncertainty of resources availability

$$\mathbb{P}\left(\sum_{i=1}^n m_{ij}r_i \leq R_j\right) \geq \beta_i, \forall j \in \mathcal{T} \quad (10)$$

$$\mathbb{P}\left(m_{ij}r_i \leq R_j - \sum_{k=1, k \neq i}^n \sum_{\substack{j' < j \\ j' + d_k > j}}^T m_{kj'}r_k\right) \geq \gamma_i \forall j \in \mathcal{T} \quad (11)$$

# Stochastic Maintenance Optimization

## Approximations of Chance Constraints

- Deterministic equivalent formulation of the stochastic repair time constraints :

$$jm_{ij} \leq T + 1 - \mu_i(d) + \phi^{-1}(\alpha_i)\sigma_i(d) \quad (12)$$

- Deterministic equivalent formulation of the stochastic resources limitation constraints

$$\sum_{i=1}^n m_{i,j} \times r_i \leq \mu(R_j) + \phi^{-1}(1 - \beta) \sigma(R_j) \quad (13)$$

$$m_{ij}r_i - \sum_{k=1, k \neq i}^n \sum_{\substack{j' < j \\ j' + d_k > j}}^T m_{kj'}r_k \leq \mu(R_j) + \phi^{-1}(1 - \gamma) \sigma(R_j)$$

# Deterministic Equivalent Model

$$(DEM) : \left\{ \begin{array}{l}
 \min_x \quad C(x) = \sum_{i=1}^n \text{cost}(x_i) \\
 s.t. : \\
 jm_{ij} \leq T + 1 - \mu_i(d) + \phi^{-1}(\alpha_i)\sigma_i(d), \forall i \in \mathcal{I} \\
 \sum_{i=1}^n m_{i,j} \times r_i \leq \mu(R_j) + \phi^{-1}(1 - \beta)\sigma(R_j); \\
 \forall i \in \mathcal{I} \\
 m_{ij}r_i - \sum_{k=1, k \neq i}^n \sum_{\substack{j' < j \\ j'+d_k > j}}^T m_{kj'}r_k \leq \mu(R_j) + \\
 + \phi^{-1}(1 - \gamma)\sigma(R_j) \\
 \sum_{x_i=1}^T m_{ix_i} \leq \theta\sigma_i + \mu_i, \forall i \in \mathcal{I} \\
 x_i \leq m_{i,j} \times T, \forall i \in \mathcal{I} \\
 \sum_{j=1}^T m_{ij} \geq 1 \quad \forall i \in \mathcal{I} \\
 m_{ij'} \leq 1 - m_{ij}, \forall i \in \mathcal{I} \quad \forall j' > j \in \mathcal{T} \\
 m_{ij} = \{0, 1\}^n, x_i > 0, \forall i \in \mathcal{I}
 \end{array} \right. \quad (14)$$

$$jm_{ij} \leq T + 1 - \mu_i(d) + \phi^{-1}(\alpha_i)\sigma_i(d), \forall i \in I \quad (15)$$

For  $\sigma(d_i) = 1$ , we discuss the following three cases:

- $\alpha = 0$ : Then we have  $\phi^{-1}(\alpha) = -\infty$ . (15) is indicating that no maintenance is possible with confidence levels very low or close to 0.
- $\alpha = 0.5$ : Then we have  $\phi^{-1}(\alpha) = 0$ . (15) is converging totally to the provided original deterministic valid inequality.
- $\alpha = 1$ : Then we have  $\phi^{-1}(\alpha) = +\infty$ . (15) is indicating that maintenance operations are totally relaxed and can be done according to the availability of the repair man.

$$\sum_{i=1}^n m_{i,j} \times r_i \leq \mu(R_j) + \phi^{-1}(1 - \beta) \sigma(R_j); \forall j \in \mathcal{T} \quad (16)$$

we propose to assess three cases as follows:

- $\beta = 0$ : Then we have  $\phi^{-1}(1 - \beta) = +\infty$ . (16) is indicating that for low values of confidence levels (close to zero), the resources available become theoretically large (infinity).
- $\beta = 0.5$ : Then we have  $\phi^{-1}(1 - \beta) = 0$ . (16) is converging to the original and deterministic valid inequality.
- $\beta = 1$ : Then we have  $\phi^{-1}(1 - \beta) = -\infty$ . (16) is indicating that maintenance could not occur and this due to lack of resources (repair man).

# CONCLUSION & FUTURE WORK



# Conclusion & Perspectives

## Conclusion

- Considered Resource and Duration Constraints
- A new MNL P model.
- Stochastic formulation to deal with uncertainty

## Future Work

- We will extend the result to address maintenance grouping problems
- We will consider routing constraints for maintenance optimisation in multi-sites factories

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The background is an abstract composition of thick, expressive brushstrokes. The color palette is dominated by vibrant greens and blues, with accents of orange and yellow. The strokes are layered and textured, creating a sense of movement and depth. The overall effect is reminiscent of a modern, expressive painting or digital artwork.

Thank you for your attention  
Questions ?