

# Scheduling elective surgeries under uncertainty: a multi-objective stochastic approach

YAMC 2022

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**Ambrogio Maria Bernardelli**, Lorenzo Bonasera, Eleonora Vercesi

Advisor: Davide Duma

2022-09-22



1. Introduction
2. Mathematical Models
3. Methodology
4. Computational Analysis
5. Conclusions

# Introduction

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# Problem Statement

$p_1$

$p_2$

$p_3$

$p_4$

$p_5$

$p_6$

# Problem Statement

$p_1$

$p_2$

$p_3$

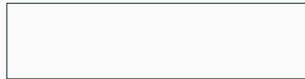
$p_4$

$p_5$

$p_6$



OR 1



OR 2

# Problem Statement

$p_1$

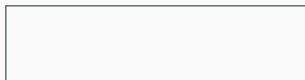
$p_2$

$p_3$

$p_4$

$p_5$

$p_6$



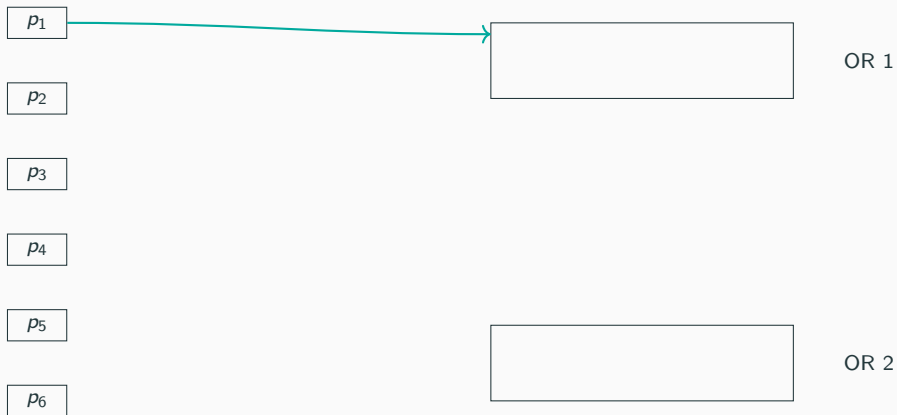
OR 1



OR 2

Advance Scheduling:

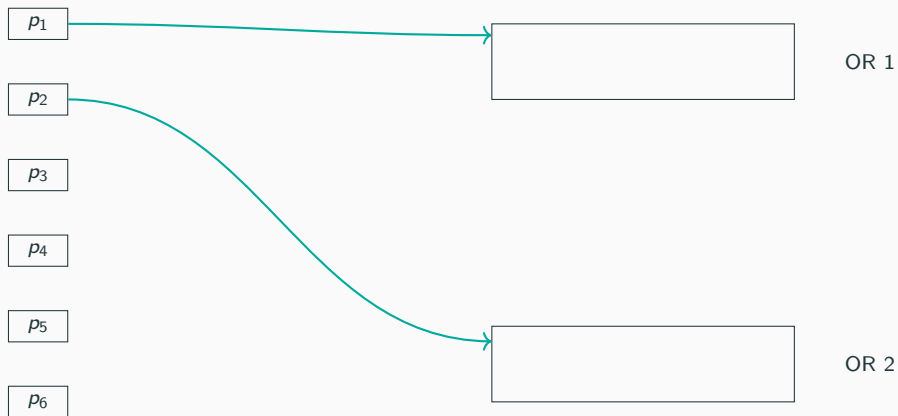
# Problem Statement



Advance Scheduling:

Assignment Procedure (AP).

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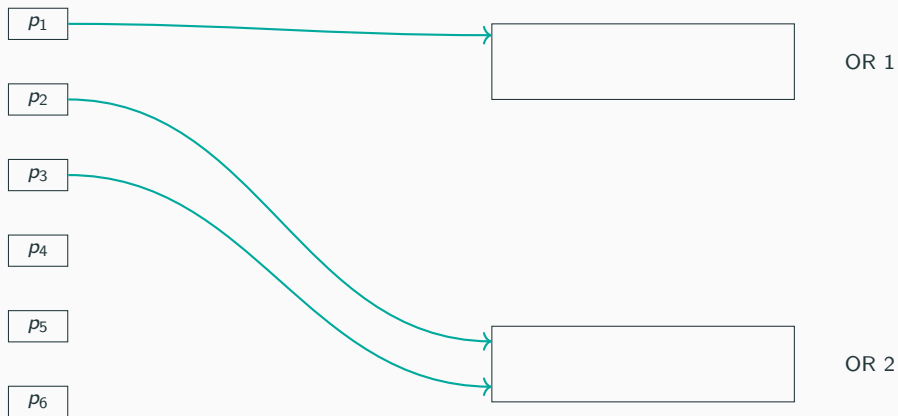


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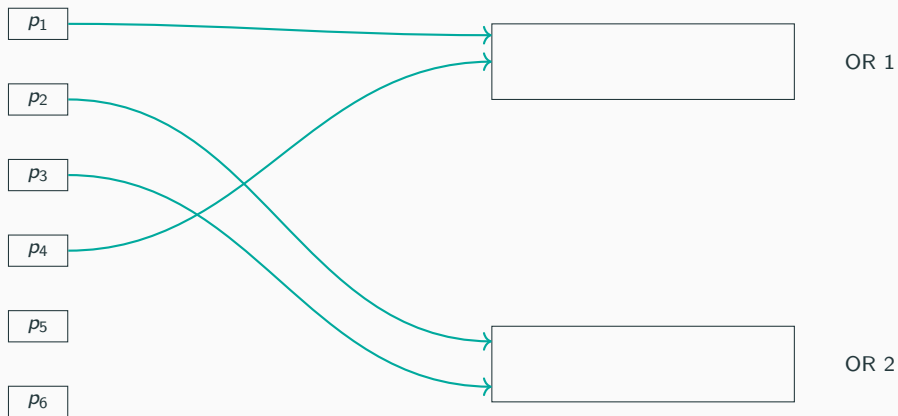
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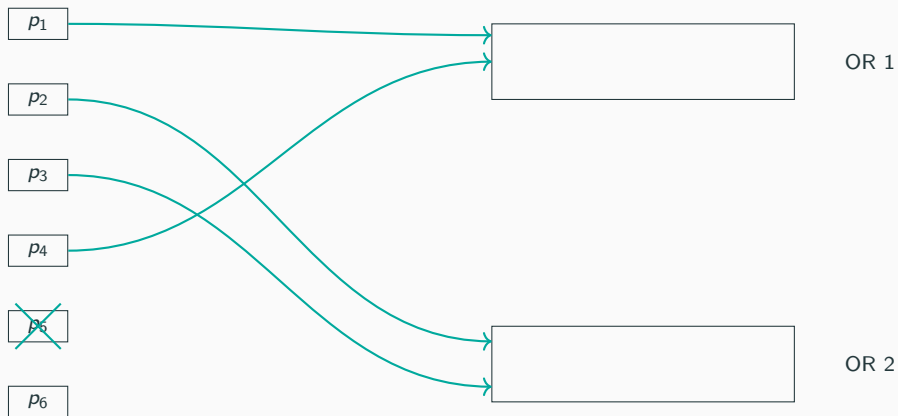
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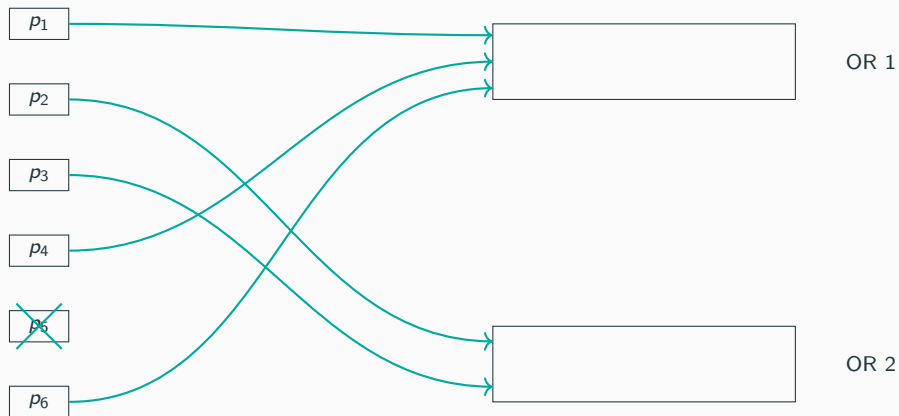
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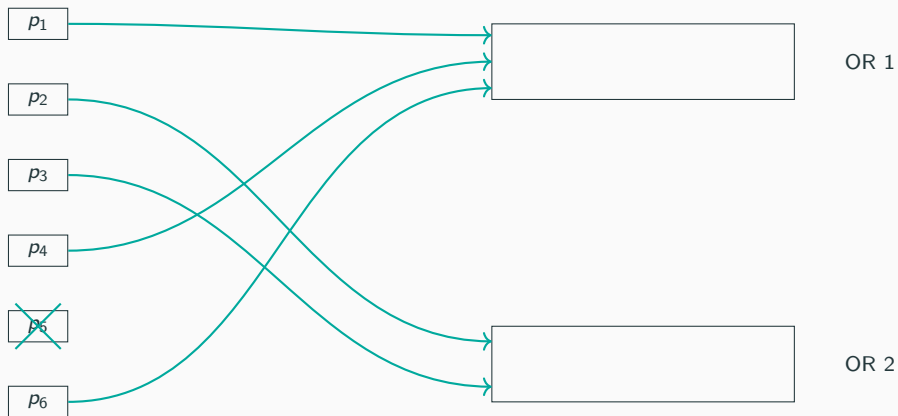
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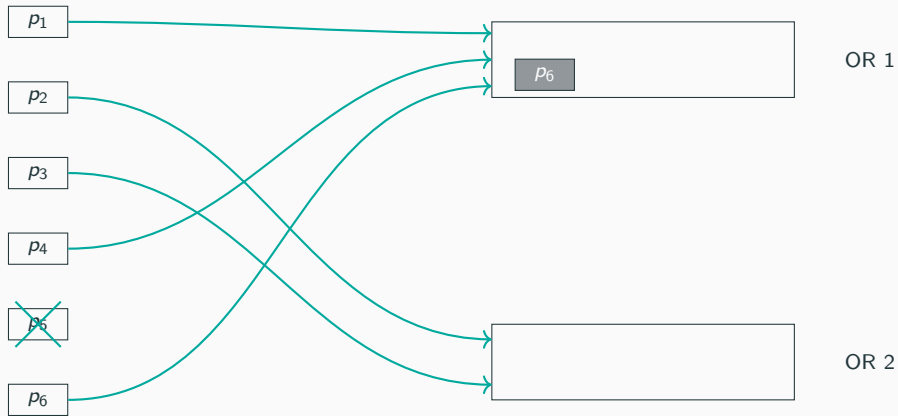


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Allocation Scheduling:

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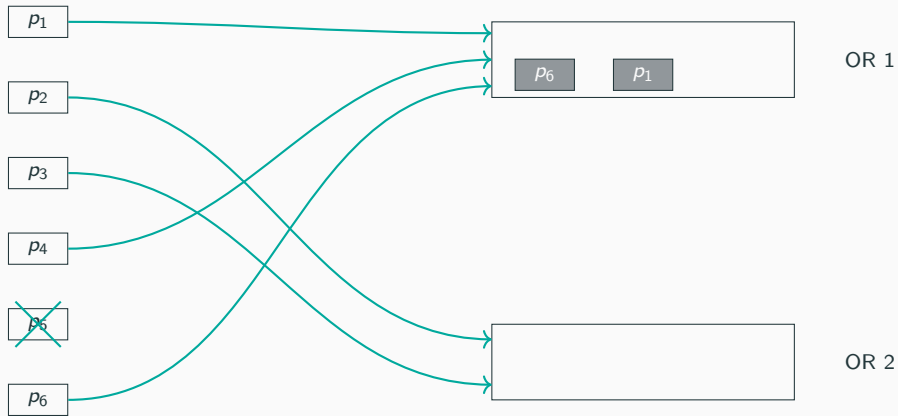
Advance Scheduling:

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Allocation Scheduling:

Sequencing Procedure (SP),

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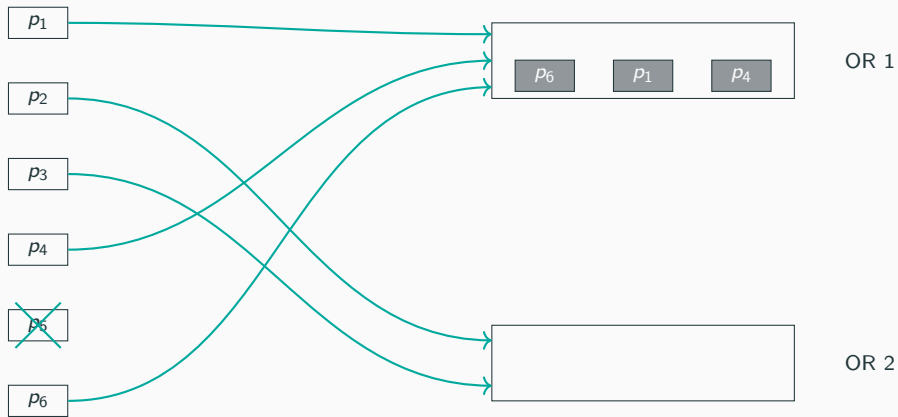
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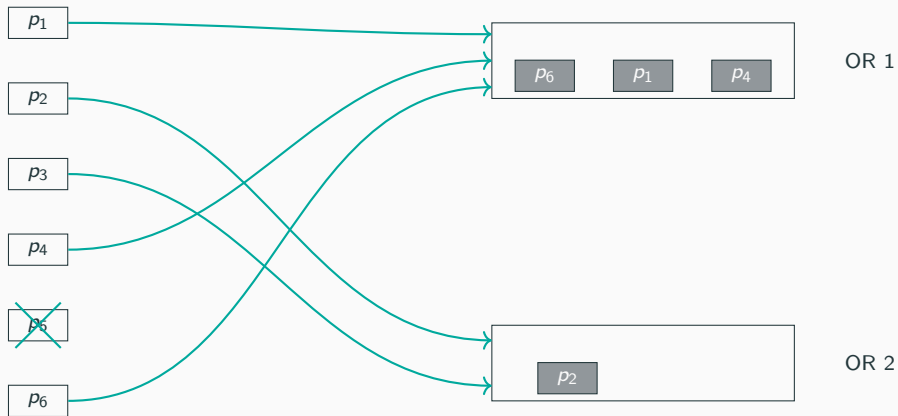
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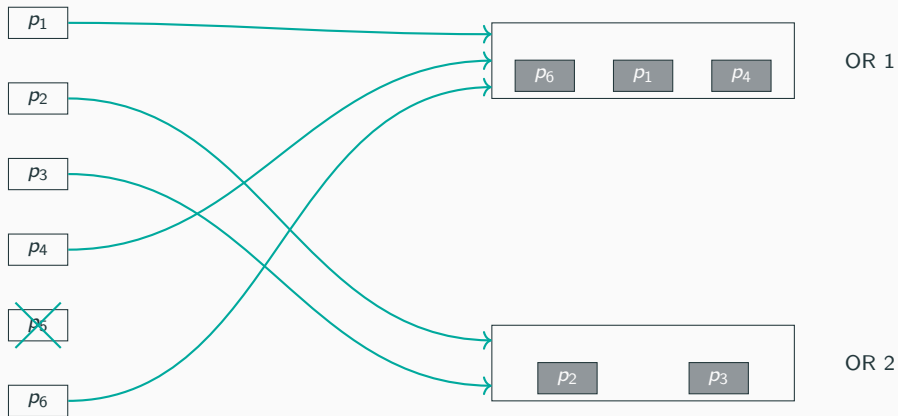
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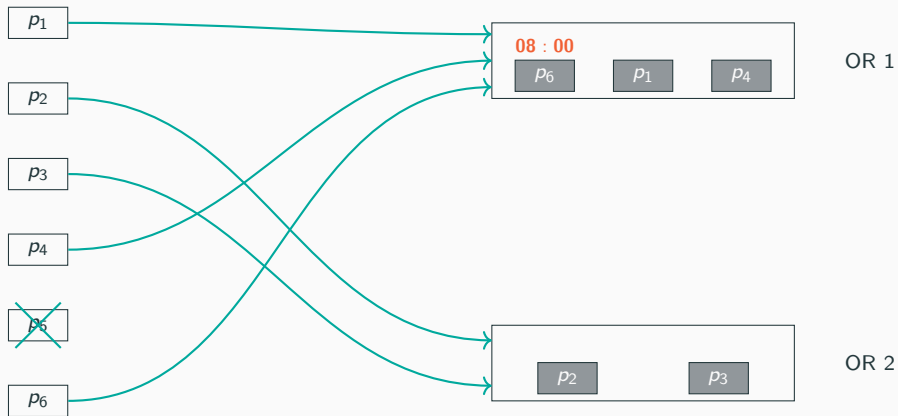
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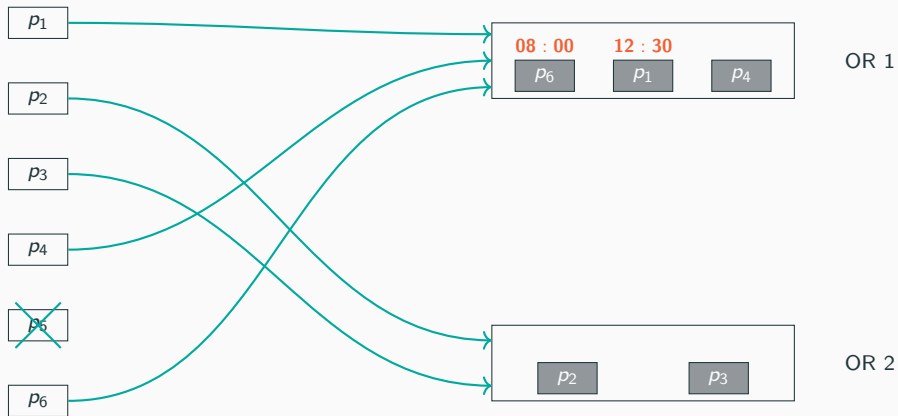
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Allocation Scheduling:

Sequencing Procedure (SP),

Timing Procedure (TP).

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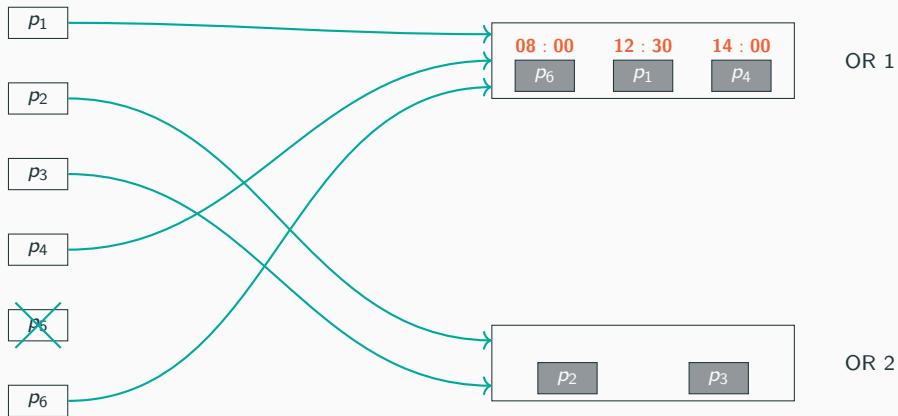
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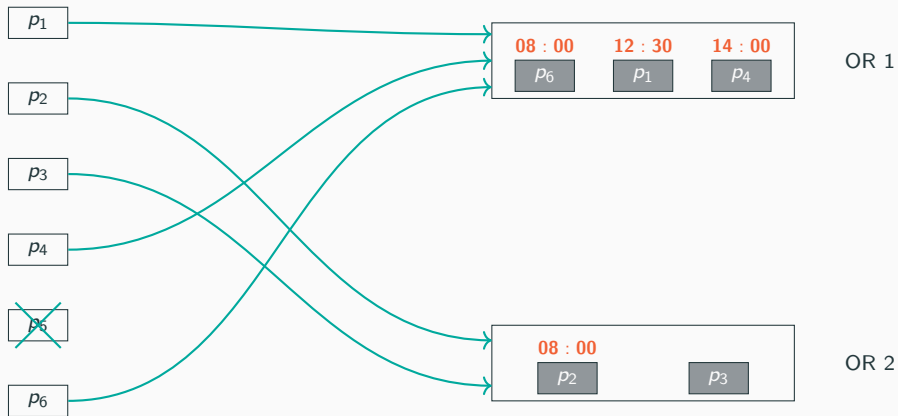
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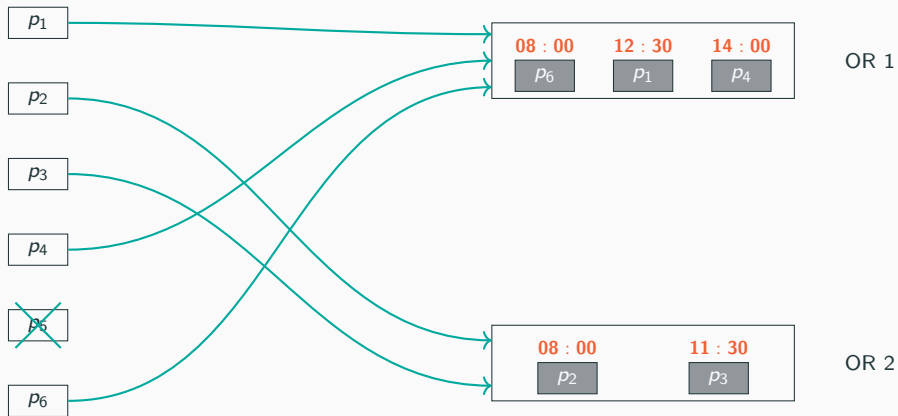
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# Objectives

Cancellations

Indirect  
Waiting  
Time

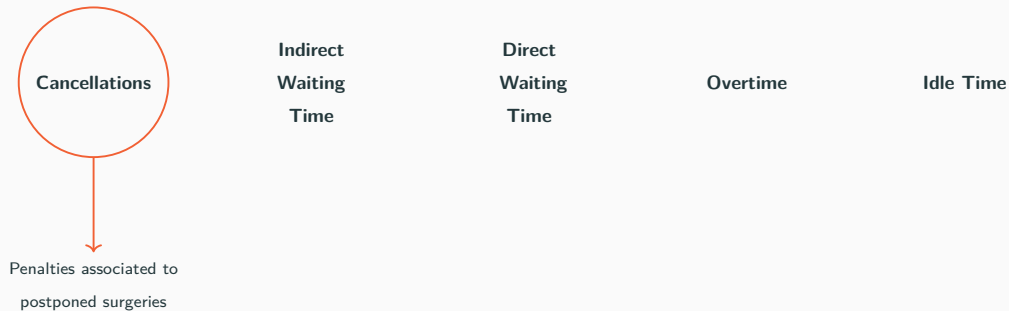
Direct  
Waiting  
Time

Overtime

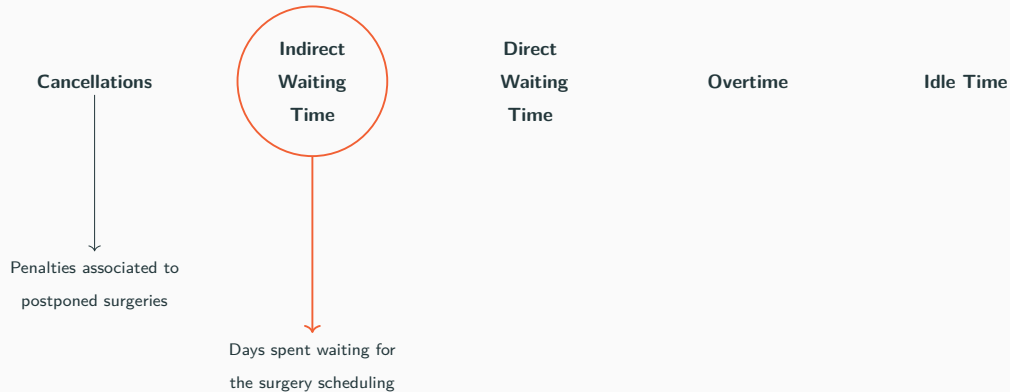
Idle Time



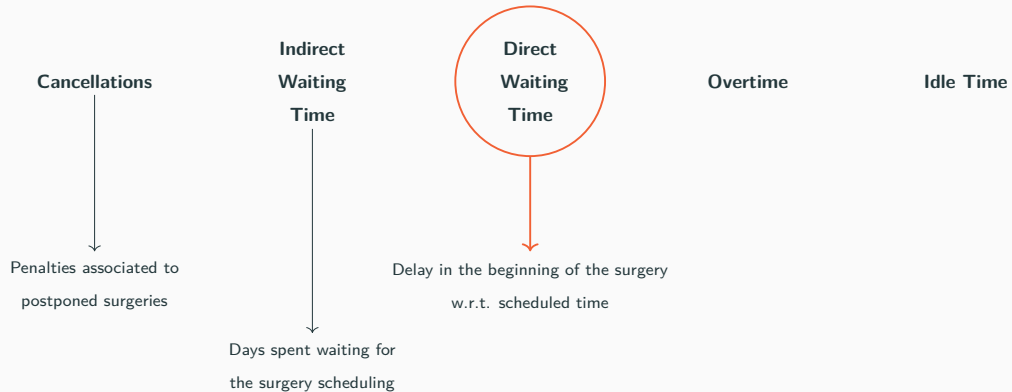
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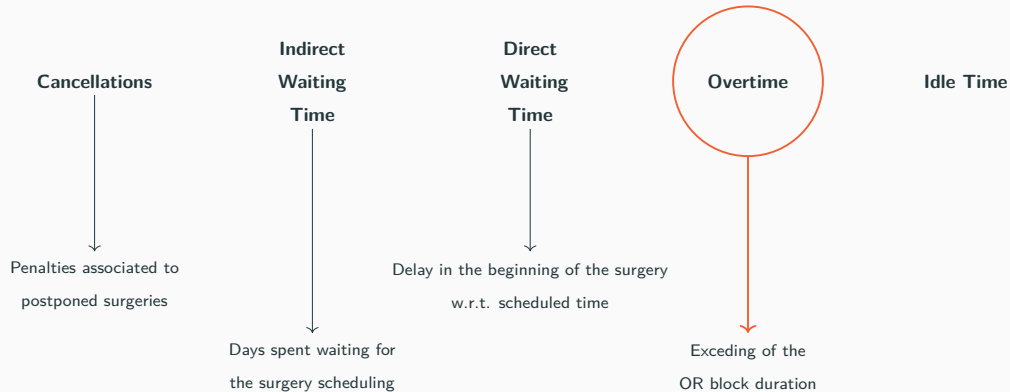
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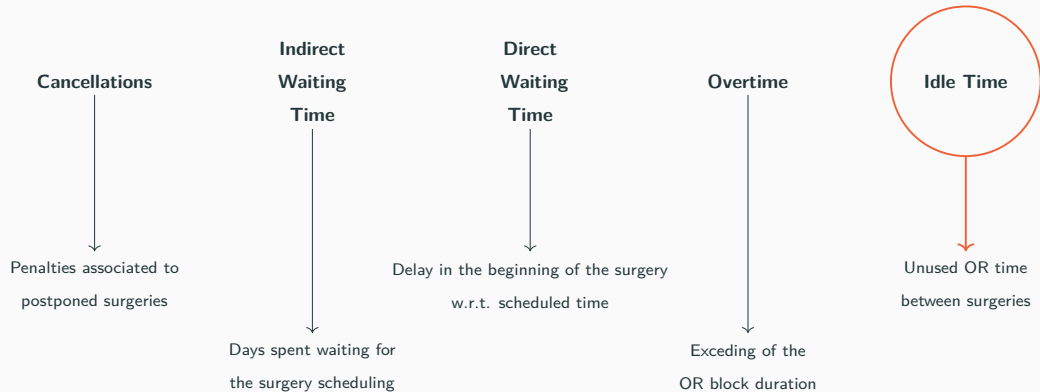
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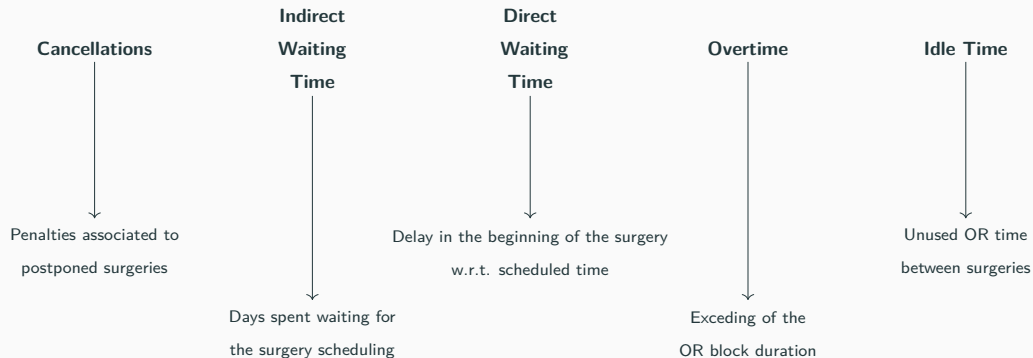
# Objectives



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Cardoen et al. (2009). Optimizing a multiple objective surgical case sequencing problem. *Int. J. Prod. Econ* 119(2) pp. 354-366.

Duma & Aringhieri (2019). The management of non-elective patients: shared vs. dedicated policies. *Omega* 83 pp 199-212.

# Uncertainty



**Surgery Duration:** the Real Operating Time (ROT) differs from the Estimated Operating Time (EOT).



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**No-shows:** some patients do not showing up the day of surgery.

## Prior Papers

Few prior studies deal with at least two of the three defined **procedures** (AP, SP, and TP) under **uncertainty**.

Paper	AP	SP	TP	Other decisions	Uncertainty	Objectives	Methodologies
Testi et al. (2007)	✓	✓	✗	MSS	surgery duration	overtime, OR utilization, throughput, bed utilization	DES, ILP, heuristics
Batun et al. (2011)	✓	✓	✗	ORs to be opened, physician-patient assignment	surgery duration	overtime, idle time, financial costs	SMIP
Landa et al. (2016)	✓	✓	✗	overtime allocation	surgery duration	OR utilization, cancellations	SMIP, metaheuristics
Aringhieri et al. (2016)	✓	✓	✗	real-time management	surgery duration	overtime, OR utilization, throughput, cancellations, indirect waiting time, % patient within due date	DES, online algorithms
Duma et al. (2019)	✓	✓	✗	OR policy, real-time management	surgery duration, non-elective patients	overtime, OR utilization, throughput, cancellations, indirect waiting time, % patient within due date	DES, online algorithms
Wang et al. (2022)	✓	✓	✗	partitioning	surgery duration, non-elective patients	overtime, idle time, OR utilization, throughput, cancellations, indirect waiting time, % patient within due date	DES

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<b>This work</b>	✓	✓	✓	-	surgery duration, non-elective patients, <b>no-shows</b>	overtime, idle time, cancellations, <b>direct and indirect waiting time</b>	SMIP, metaheuristics

# Mathematical Models

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# Assumptions



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Strum et al. (2003). Estimating times of surgeries with two components procedures comparison of the lognormal and normal models. Anesthesiology 98(1) pp. 232-240.

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Denton et al. (2007). Optimizing of surgery sequencing and scheduling decisions under uncertainty. Health Care Manag. Sci. 10(1) pp. 13-24.

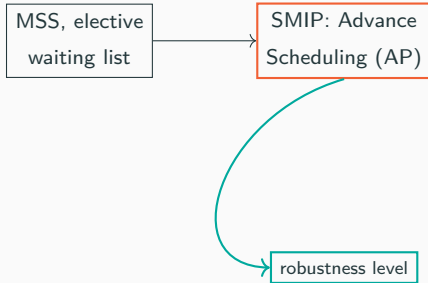
# Our Approach

MSS, elective  
waiting list

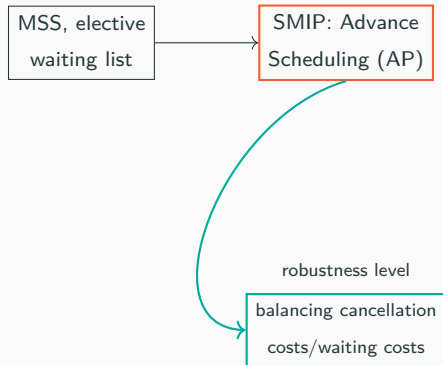
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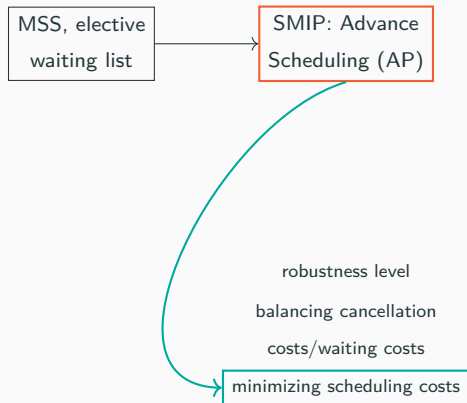
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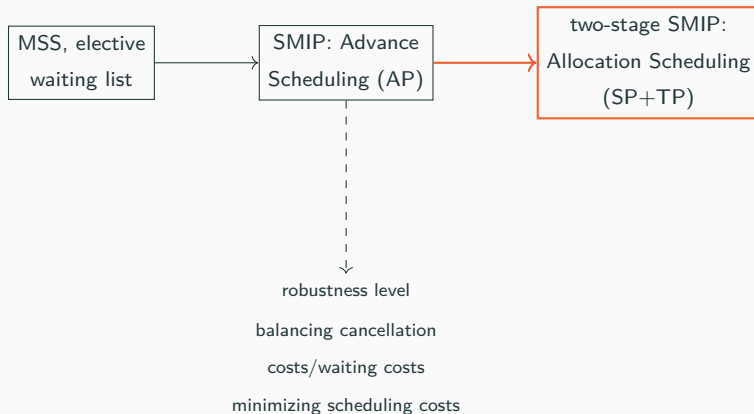
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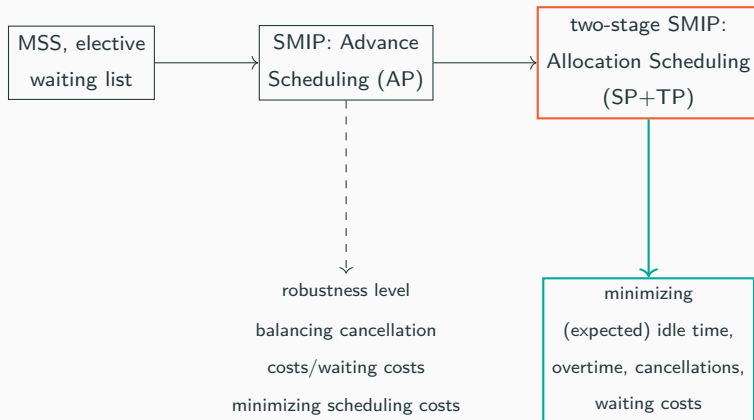


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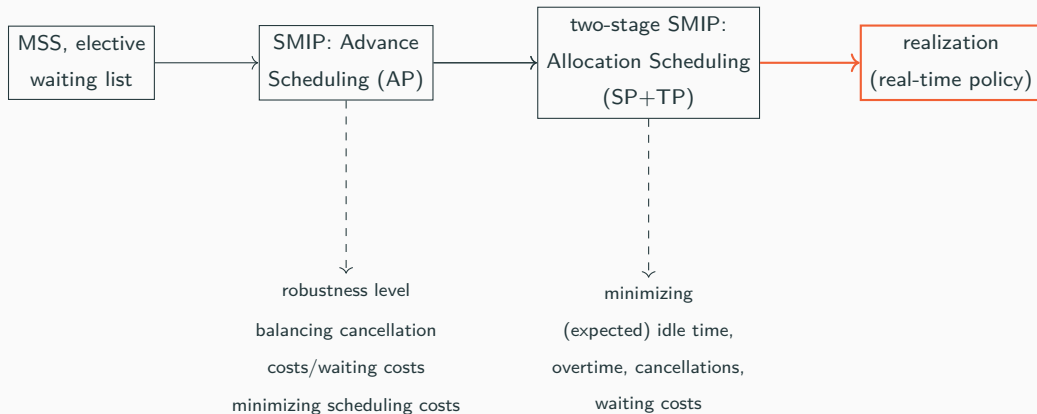




# Our Approach



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$$\mathcal{A}_s(\alpha) : \quad \min_{\mathbf{x}} \sum_{i \in W_s} c_i^{sched} \left( 1 - \sum_{(j,k) \in B_s} x_{ijk} \right) \quad (1a)$$

## Advance Scheduling - First Model

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$$\mathbb{P}_{\xi} \left[ \delta_{jk}(\omega) + \sum_{i \in W_s} \rho_i(\omega) x_{ijk} \leq L + H \right] \geq 1 - \alpha, \quad \forall (j,k) \in B_s, \quad (1d)$$

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$$\sum_{i \in W_s} q_i x_{ijk} \leq 1, \quad \forall (j, k) \in B_s, \quad (1e)$$

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$$x_{ijk} \in \{0, 1\}, \quad \forall i \in W_s, \forall (j,k) \in B_s. \quad (1g)$$



## Hierarchy and balance constants

$$C_1 = \frac{c_{min}^{sched}}{1 + \sum_{i \in W_s} c_i^{sched}}, \quad C_2 = \frac{c_{min}^{canc}}{c_{min}^{wait}}.$$

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## The Model

$$\mathcal{B}_s(\alpha, \beta, \nu) : \min_{\mathbf{x}, \Gamma^{canc}, \Gamma^{wait}} \sum_{i \in W_s} c_i^{sched} \left( 1 - \sum_{(j,k) \in B_s} x_{ijk} \right) + C_1 \left( \beta \Gamma^{canc} + \nu C_2 \Gamma^{wait} \right) \quad (2a)$$

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# Advance Scheduling - Second Model

## Hierarchy and balance constants

$$C_1 = \frac{c_{min}^{sched}}{1 + \sum_{i \in W_s} c_i^{sched}}, \quad C_2 = \frac{c_{min}^{canc}}{c_{min}^{wait}}.$$

## The Model

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$$\text{s.t.} \quad (1b) - (1g),$$

$$\Gamma^{canc} = \max_{(j,k) \in B_s} \left\{ \sum_{i \in W_s} c_i^{canc} x_{ijk} \right\}, \quad (2b)$$

$$\Gamma^{wait} = \max_{(j,k) \in B_s} \left\{ \sum_{i \in W_s} c_i^{wait} x_{ijk} \right\}. \quad (2c)$$

$$C_{jk}^l : \min_{\mathbf{o}, \mathbf{t}} \mathbb{E}_{\xi} [Q(\mathbf{o}, \mathbf{t}; \xi(\omega))] \quad (3a)$$

# Allocation Scheduling Model - First Stage

$$C_{jk}^l : \quad \min_{\mathbf{o}, \mathbf{t}} \mathbb{E}_{\xi} [Q(\mathbf{o}, \mathbf{t}; \xi(\omega))] \quad (3a)$$

$$\text{s.t. } t_i \leq (L - \mu_i) \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (3b)$$

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$$t_i + \mu_i \leq t_{i'} + (1 - o_{i'i})M_{i'i}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (3c)$$

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$$t_i + \mu_i \leq t_{i'} + (1 - o_{i'i})M_{i'i}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (3c)$$

$$\sum_{i \in I_{jk} \setminus \{i'\}} o_{i'i} \leq 1 - q_{i'}, \quad \forall i' \in I_{jk}, \quad (3d)$$

$$\sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i} \leq 1 - u_i, \quad \forall i \in I_{jk}, \quad (3e)$$



# Allocation Scheduling Model - First Stage

$$C_{jk}^l : \quad \min_{\mathbf{o}, \mathbf{t}} \mathbb{E}_{\xi} [Q(\mathbf{o}, \mathbf{t}; \xi(\omega))] \quad (3a)$$

$$\text{s.t. } t_i \leq (L - \mu_i) \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (3b)$$

$$t_i + \mu_i \leq t_{i'} + (1 - o_{i'i})M_{i'i}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (3c)$$

$$\sum_{i \in I_{jk} \setminus \{i'\}} o_{i'i} \leq 1 - q_{i'}, \quad \forall i' \in I_{jk}, \quad (3d)$$

$$\sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i} \leq 1 - u_i, \quad \forall i \in I_{jk}, \quad (3e)$$

$$\sum_{i \in I_{jk}} \sum_{i' \in I_{jk} \setminus \{i\}} = |I_{jk}| - 1, \quad (3f)$$

# Allocation Scheduling Model - First Stage

$$C_{jk}^l : \quad \min_{\mathbf{o}, \mathbf{t}} \mathbb{E}_{\xi} [Q(\mathbf{o}, \mathbf{t}; \xi(\omega))] \quad (3a)$$

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$$t_i + \mu_i \leq t_{i'} + (1 - o_{i'i})M_{i'i}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (3c)$$

$$\sum_{i \in I_{jk} \setminus \{i'\}} o_{i'i} \leq 1 - q_{i'}, \quad \forall i' \in I_{jk}, \quad (3d)$$

$$\sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i} \leq 1 - u_i, \quad \forall i \in I_{jk}, \quad (3e)$$

$$\sum_{i \in I_{jk}} \sum_{i' \in I_{jk} \setminus \{i\}} = |I_{jk}| - 1, \quad (3f)$$

$$o_{ii'} \in \{0, 1\}, t_i \geq 0, \quad \forall i, i' \in I_{jk}, i \neq i'. \quad (3g)$$

## Allocation Scheduling Model - Second Stage

$$C_{jk}^{II}(\omega) : \min_{\mathbf{o}, \mathbf{t}} c^h h_{jk} + c^g g_{jk} + \sum_{i \in I_{jk}} c_i^{canc} (1 - y_i) + \sum_{i \in I_{jk}} c_i^{wait} a_i \quad (4a)$$

## Allocation Scheduling Model - Second Stage

$$C_{jk}^{II}(\omega) : \quad \min_{\mathbf{o}, \mathbf{t}} \quad c^h h_{jk} + c^g g_{jk} + \sum_{i \in I_{jk}} c_i^{canc} (1 - y_i) + \sum_{i \in I_{jk}} c_i^{wait} a_i \quad (4a)$$

$$\text{s.t. } o_{ii'} = 1 \implies q_{i'} = \max \{c_i, t_{i'}\} \wedge \hat{q}_{i'} = \max \{\hat{c}_i, t_{i'}\}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4b)$$

$$q_i, \hat{q}_i \leq M \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (4c)$$

## Allocation Scheduling Model - Second Stage

$$C_{jk}^{II}(\omega) : \min_{\mathbf{o}, \mathbf{t}} c^h h_{jk} + c^g g_{jk} + \sum_{i \in I_{jk}} c_i^{canc} (1 - y_i) + \sum_{i \in I_{jk}} c_i^{wait} a_i \quad (4a)$$

$$\text{s.t. } o_{ii'} = 1 \implies q_{i'} = \max \{c_i, t_{i'}\} \wedge \hat{q}_{i'} = \max \{\hat{c}_i, t_{i'}\}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4b)$$

$$q_i, \hat{q}_i \leq M \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (4c)$$

$$c_i = q_i + \rho_i(\omega) \theta_i(\omega) y_i + z_i + \delta_{jk}(\omega) e_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4d)$$

$$\hat{c}_i = \hat{q}_i + \rho_i(\omega) \theta_i(\omega) y_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4e)$$

# Allocation Scheduling Model - Second Stage

$$C_{jk}^{II}(\omega) : \min_{\mathbf{o}, \mathbf{t}} c^h h_{jk} + c^g g_{jk} + \sum_{i \in I_{jk}} c_i^{canc} (1 - y_i) + \sum_{i \in I_{jk}} c_i^{wait} a_i \quad (4a)$$

$$\text{s.t. } o_{ii'} = 1 \implies q_{i'} = \max \{c_i, t_{i'}\} \wedge \hat{q}_{i'} = \max \{\hat{c}_i, t_{i'}\}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4b)$$

$$q_i, \hat{q}_i \leq M \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (4c)$$

$$c_i = q_i + \rho_i(\omega) \theta_i(\omega) y_i + z_i + \delta_{jk}(\omega) e_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4d)$$

$$\hat{c}_i = \hat{q}_i + \rho_i(\omega) \theta_i(\omega) y_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4e)$$

$$C \geq \theta_i(\omega) (q_i + \rho_i(\omega) y_i) + z_i + \delta_{jk}(\omega) e_i - (1 - y_i) M, \quad \forall i \in I_{jk}, \quad (4f)$$

$$C \geq \tau_{jk}(\omega) + \delta_{jk}(\omega), \quad \forall i \in I_{jk}, \quad (4g)$$

# Allocation Scheduling Model - Second Stage

$$C_{jk}^{II}(\omega) : \min_{\mathbf{o}, \mathbf{t}} c^h h_{jk} + c^g g_{jk} + \sum_{i \in I_{jk}} c_i^{canc} (1 - y_i) + \sum_{i \in I_{jk}} c_i^{wait} a_i \quad (4a)$$

$$\text{s.t. } o_{ii'} = 1 \implies q_{i'} = \max \{c_i, t_{i'}\} \wedge \hat{q}_{i'} = \max \{\hat{c}_i, t_{i'}\}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4b)$$

$$q_i, \hat{q}_i \leq M \sum_{i' \in I_{jk} \setminus \{i\}} o_{i'i}, \quad \forall i \in I_{jk}, \quad (4c)$$

$$c_i = q_i + \rho_i(\omega) \theta_i(\omega) y_i + z_i + \delta_{jk}(\omega) e_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4d)$$

$$\hat{c}_i = \hat{q}_i + \rho_i(\omega) \theta_i(\omega) y_i, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4e)$$

$$C \geq \theta_i(\omega) (q_i + \rho_i(\omega) y_i) + z_i + \delta_{jk}(\omega) e_i - (1 - y_i) M, \quad \forall i \in I_{jk}, \quad (4f)$$

$$C \geq \tau_{jk}(\omega) + \delta_{jk}(\omega), \quad \forall i \in I_{jk}, \quad (4g)$$

$$z_i \leq M e_i, \quad \forall i \in I_{jk}, \quad (4h)$$

$$\sum_{i \in I_{jk}} e_i = 1, \quad \forall i \in I_{jk}, \quad (4i)$$

$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$



$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$

$$\begin{cases} e_i = 1, \\ \tau_{jk}(\omega) > q_i + \rho_i(\omega)\theta_i(\omega)y_i \end{cases} \implies z_i = \tau_{jk}(\omega) - (q_i + \rho_i(\omega)\theta_i(\omega)y_i), \quad \forall i \in I_{jk}, \quad (4k)$$

$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$

$$\begin{cases} e_i = 1, \\ \tau_{jk}(\omega) > q_i + \rho_i(\omega)\theta_i(\omega)y_i \end{cases} \implies z_i = \tau_{jk}(\omega) - (q_i + \rho_i(\omega)\theta_i(\omega)y_i), \quad \forall i \in I_{jk}, \quad (4k)$$

$$\theta_i(\omega)(q_i + \mu_i) \leq L + H \iff y_i = 1, \quad \forall i \in I_{jk}, \quad (4l)$$

## Allocation Scheduling Model - Second Stage

$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$

$$\begin{cases} e_i = 1, \\ \tau_{jk}(\omega) > q_i + \rho_i(\omega)\theta_i(\omega)y_i \end{cases} \implies z_i = \tau_{jk}(\omega) - (q_i + \rho_i(\omega)\theta_i(\omega)y_i), \quad \forall i \in I_{jk}, \quad (4k)$$

$$\theta_i(\omega)(q_i + \mu_i) \leq L + H \iff y_i = 1, \quad \forall i \in I_{jk}, \quad (4l)$$

$$y_i \geq 1 - \theta_i(\omega), \quad \forall i \in I_{jk}, \quad (4m)$$

# Allocation Scheduling Model - Second Stage

$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$

$$\begin{cases} e_i = 1, \\ \tau_{jk}(\omega) > q_i + \rho_i(\omega)\theta_i(\omega)y_i \end{cases} \implies z_i = \tau_{jk}(\omega) - (q_i + \rho_i(\omega)\theta_i(\omega)y_i), \quad \forall i \in I_{jk}, \quad (4k)$$

$$\theta_i(\omega)(q_i + \mu_i) \leq L + H \iff y_i = 1, \quad \forall i \in I_{jk}, \quad (4l)$$

$$y_i \geq 1 - \theta_i(\omega), \quad \forall i \in I_{jk}, \quad (4m)$$

$$a_i \geq q_i - t_i - M(1 - y_i\theta_i(\omega)), \quad \forall i \in I_{jk}, \quad (4n)$$

$$h_{jk} \geq C - L, \quad \forall i \in I_{jk}, \quad (4o)$$

$$g_{jk} \geq \max \{L, C\} - \sum_{i \in I_{jk}} \rho_i(\omega)\theta_i(\omega)y_i - \delta_{jk}(\omega), \quad \forall i \in I_{jk}, \quad (4p)$$

# Allocation Scheduling Model - Second Stage

$$e_i = 1 \wedge o_{ii'} = 1 \iff \hat{q}_i \leq \tau_{jk}(\omega) < \hat{q}_{i'}, \quad \forall i, i' \in I_{jk}, i \neq i', \quad (4j)$$

$$\begin{cases} e_i = 1, \\ \tau_{jk}(\omega) > q_i + \rho_i(\omega)\theta_i(\omega)y_i \end{cases} \implies z_i = \tau_{jk}(\omega) - (q_i + \rho_i(\omega)\theta_i(\omega)y_i), \quad \forall i \in I_{jk}, \quad (4k)$$

$$\theta_i(\omega)(q_i + \mu_i) \leq L + H \iff y_i = 1, \quad \forall i \in I_{jk}, \quad (4l)$$

$$y_i \geq 1 - \theta_i(\omega), \quad \forall i \in I_{jk}, \quad (4m)$$

$$a_i \geq q_i - t_i - M(1 - y_i\theta_i(\omega)), \quad \forall i \in I_{jk}, \quad (4n)$$

$$h_{jk} \geq C - L, \quad \forall i \in I_{jk}, \quad (4o)$$

$$g_{jk} \geq \max \{L, C\} - \sum_{i \in I_{jk}} \rho_i(\omega)\theta_i(\omega)y_i - \delta_{jk}(\omega), \quad \forall i \in I_{jk}, \quad (4p)$$

$$h_{jk}, g_{jk}, q_i, \hat{q}_i, c_i, \hat{c}_i, C, z_i, a_i \geq 0, \quad y_i, e_i \in \{0, 1\}, \quad \forall i \in I_{jk}. \quad (4q)$$

# Methodology

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## SAA



## SAA





## SAA



SAA



SAA<sub>N</sub>



SAA



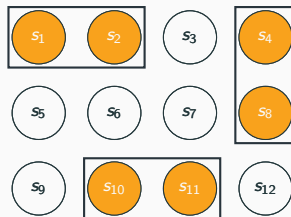
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SAA



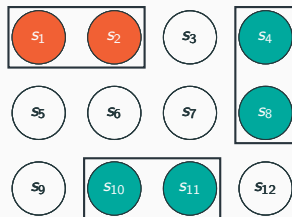
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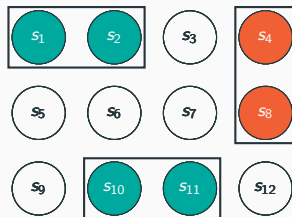
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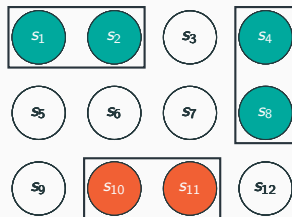
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SAA



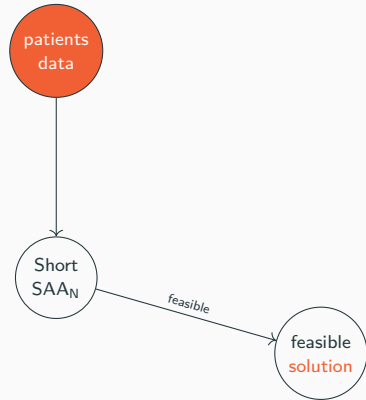
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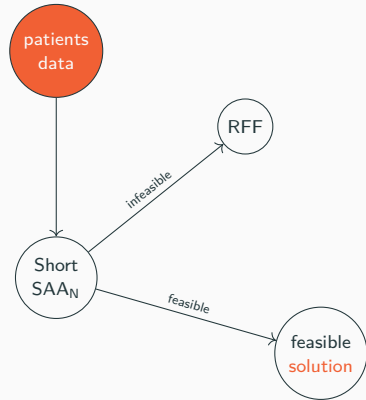


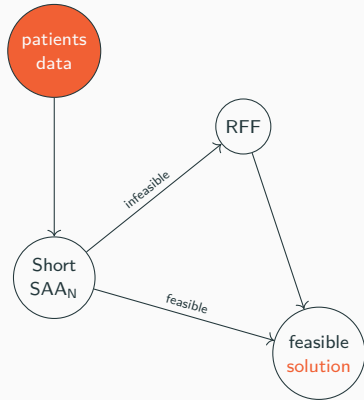


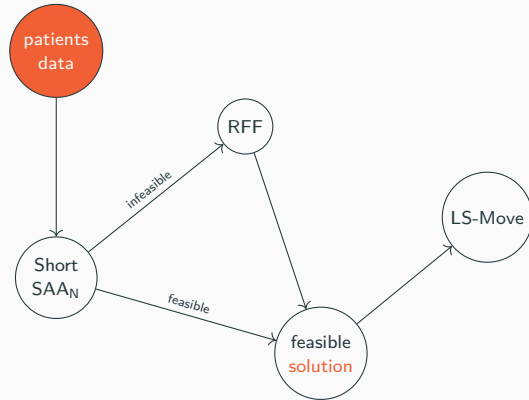




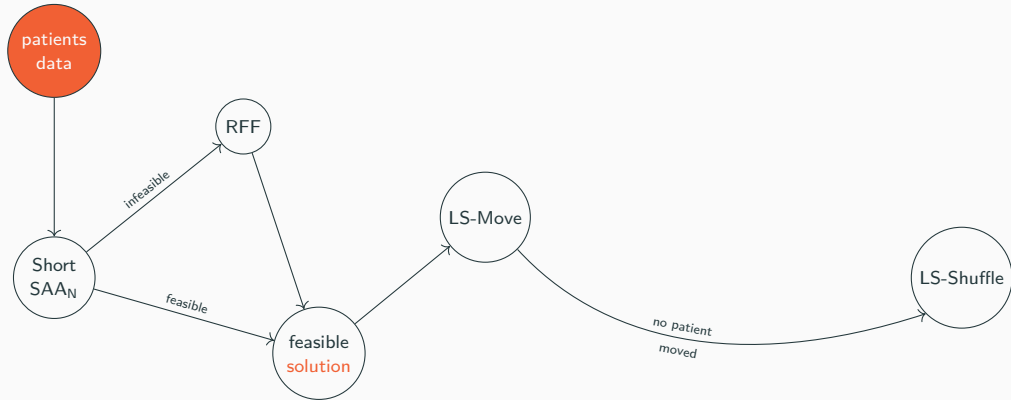




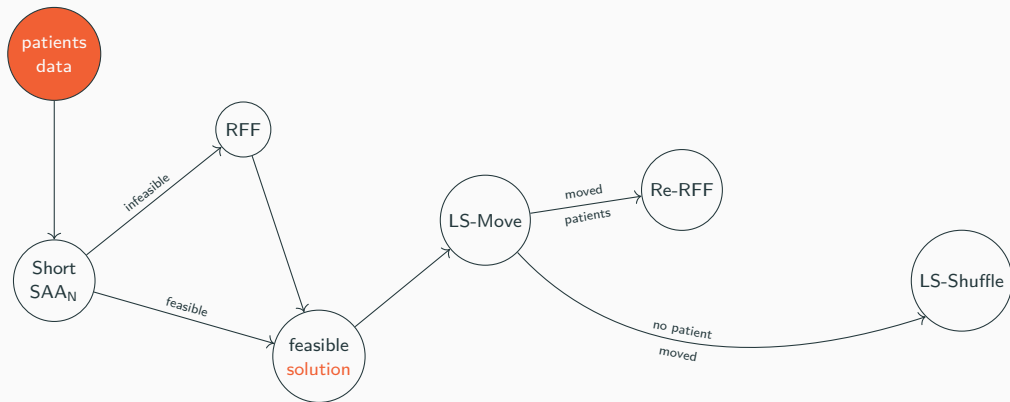




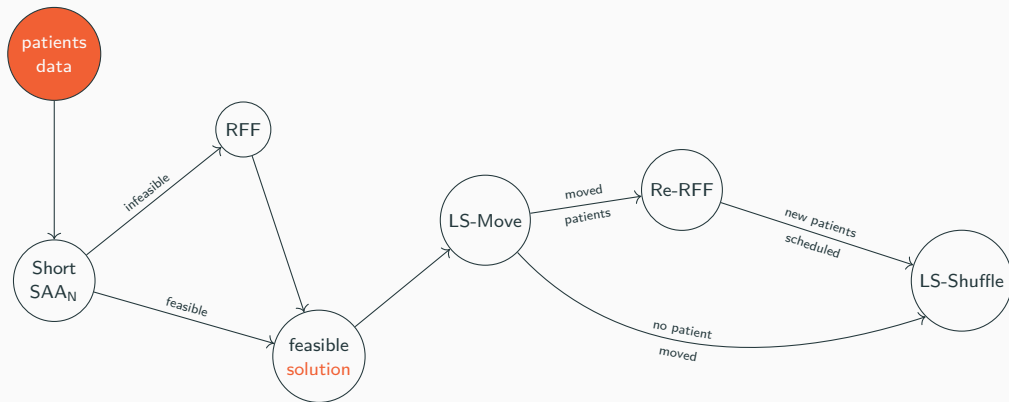
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# Advance Scheduling - SCI

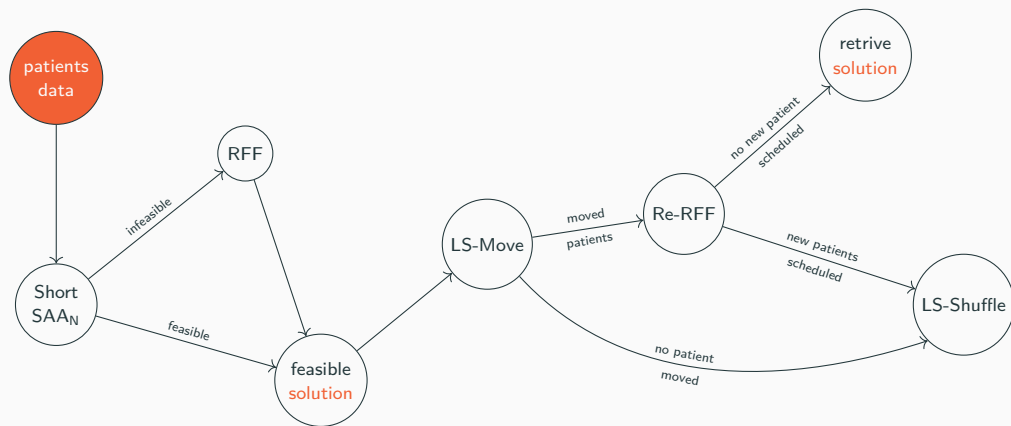


# Advance Scheduling - SCI

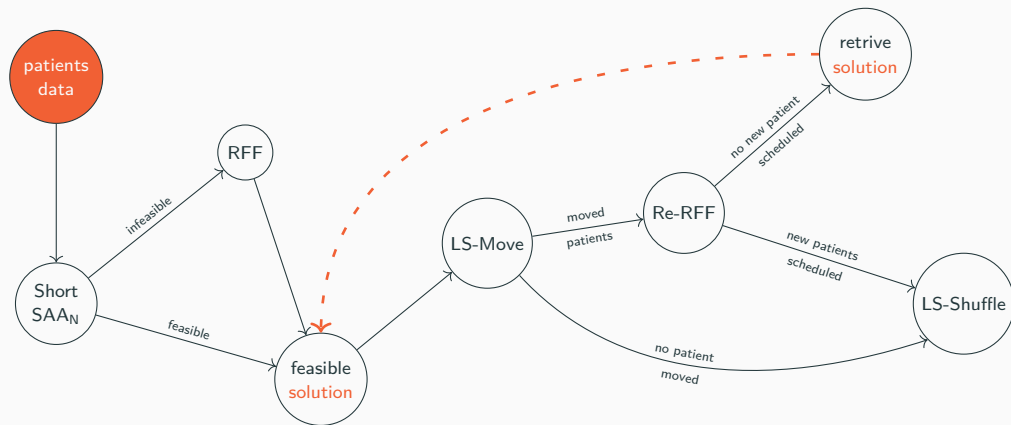




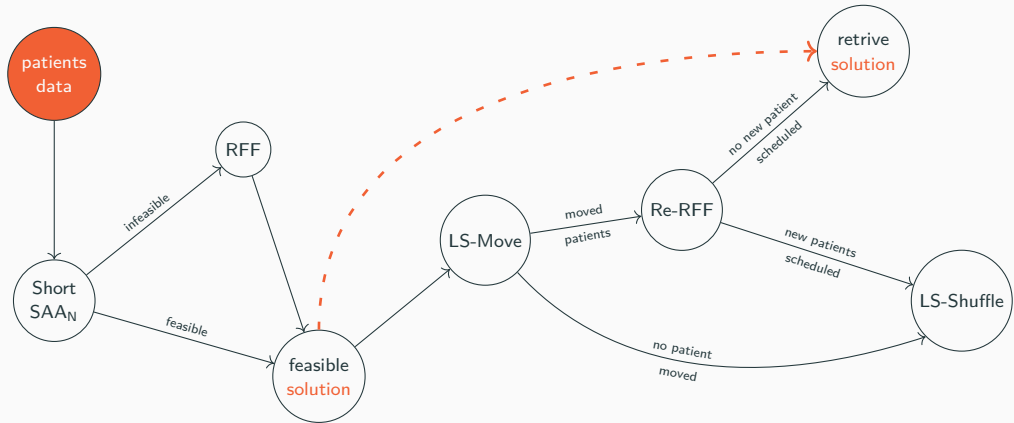
# Advance Scheduling - SCI



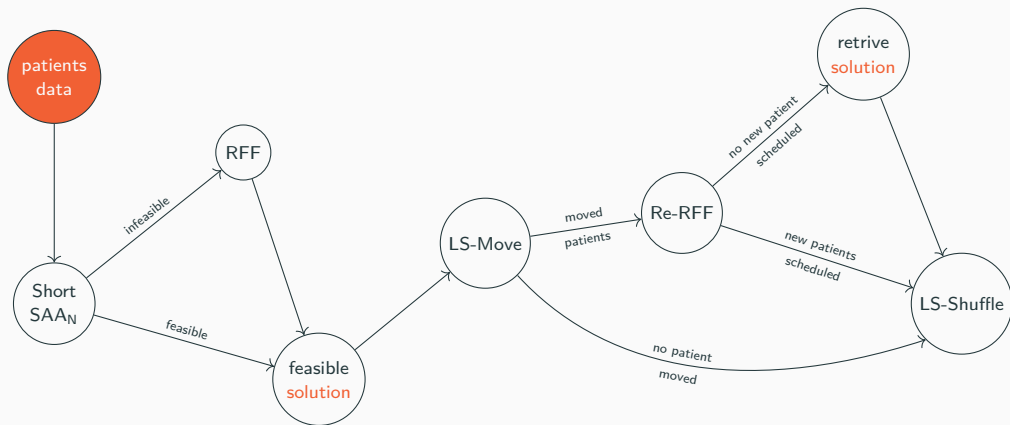
# Advance Scheduling - SCI



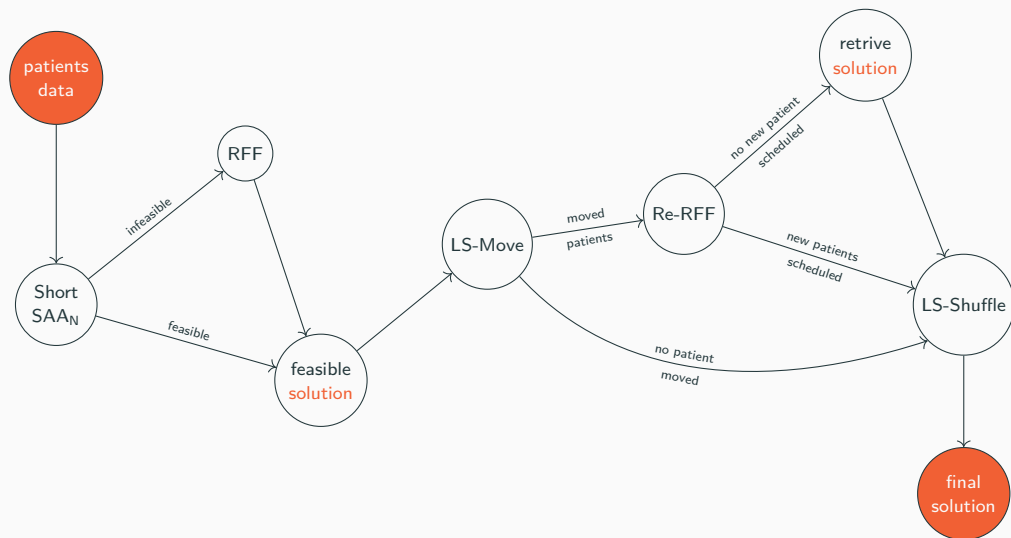
# Advance Scheduling - SCI



# Advance Scheduling - SCI



# Advance Scheduling - SCI





EOTs

# Allocation Scheduling - BRKGA



EOTs



chromosome

# Allocation Scheduling - BRKGA

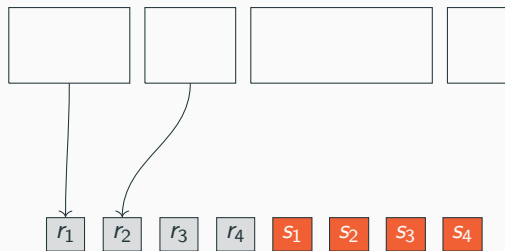


EOTs

chromosome



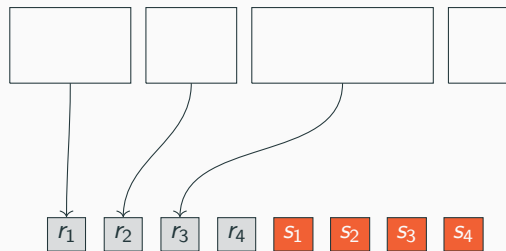
# Allocation Scheduling - BRKGA



EOTs

chromosome

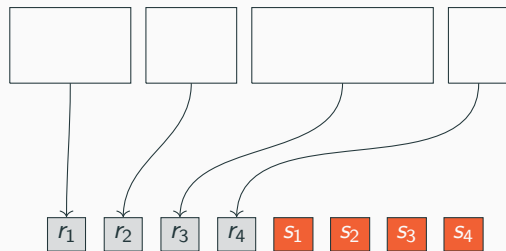
# Allocation Scheduling - BRKGA



EOTs

chromosome

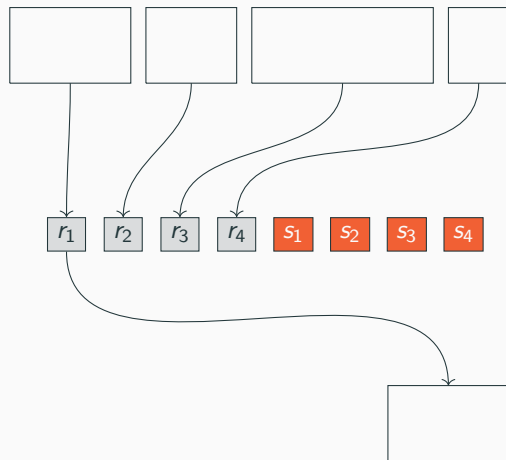
# Allocation Scheduling - BRKGA



EOTs

chromosome

# Allocation Scheduling - BRKGA

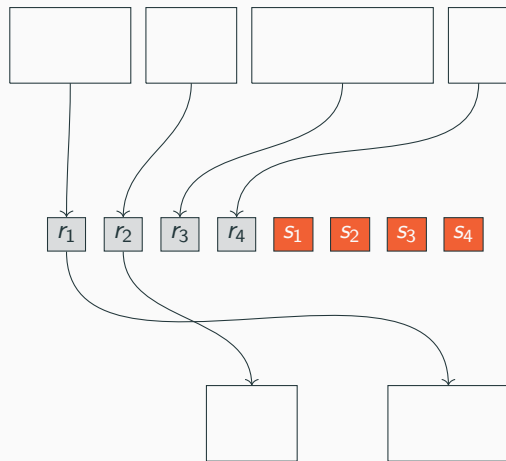


EOTs

chromosome

sequencing and  
starting times

# Allocation Scheduling - BRKGA

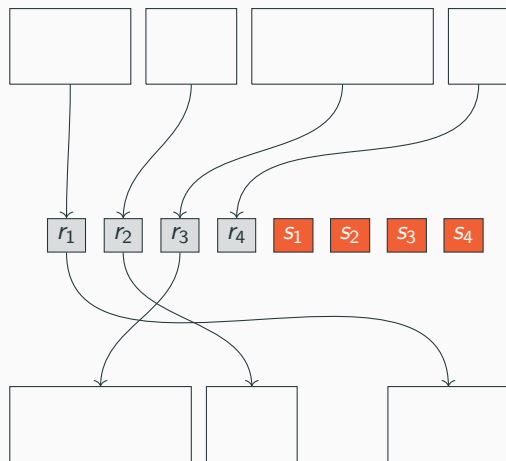


EOTs

chromosome

sequencing and  
starting times

# Allocation Scheduling - BRKGA

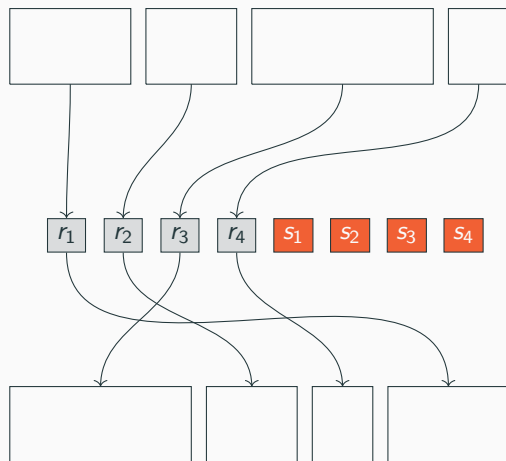


EOTs

chromosome

sequencing and  
starting times

# Allocation Scheduling - BRKGA

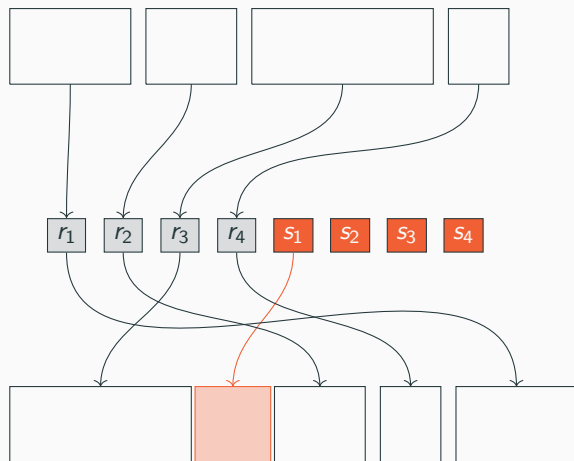


EOTs

chromosome

sequencing and  
starting times

# Allocation Scheduling - BRKGA



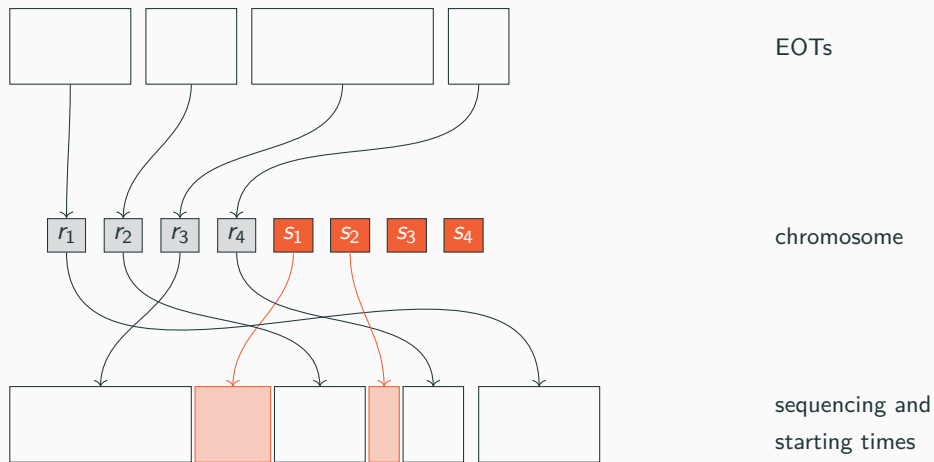
EOTs

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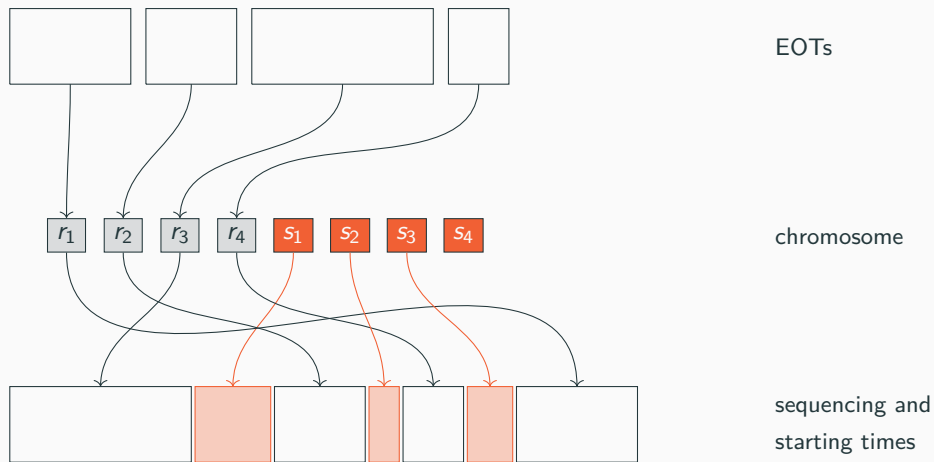
sequencing and  
starting times



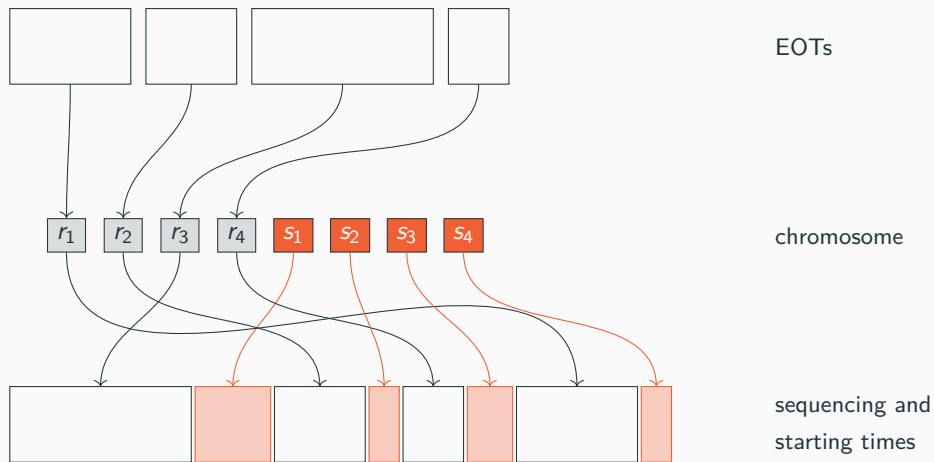
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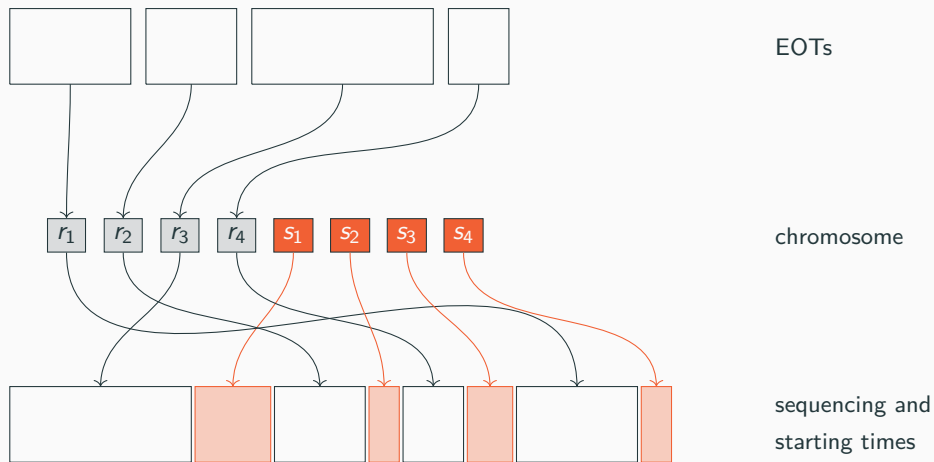
# Allocation Scheduling - BRKGA



# Allocation Scheduling - BRKGA



# Allocation Scheduling - BRKGA



# Computational Analysis

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# Data - Sykehuset Asker og Bærum HF

OR room	Monday	Tuesday	Wednesday	Thursday	Friday
1	GASTRO	GASTRO	GASTRO		
2			GASTRO	GASTRO	GASTRO
3	CARD		CARD		CARD
4	ORTH	ORTH		ORTH	ORTH
5		ORTH	MED		
6	GYN	GYN	GYN	GYN	
7		GYN	GYN	GYN	GYN
8	URO	URO		URO	URO
9	CARD		URO		CARD
10	URO		ORTH		

# Data - Sykehuset Asker og Bærum HF

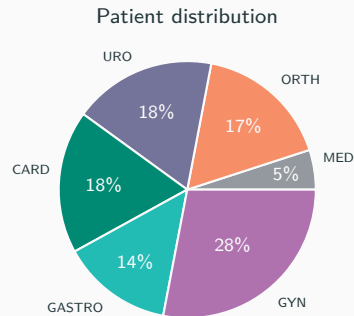
OR room	Monday	Tuesday	Wednesday	Thursday	Friday
1	GASTRO	GASTRO	GASTRO		
2			GASTRO	GASTRO	GASTRO
3	CARD		CARD		CARD
4	ORTH	ORTH		ORTH	ORTH
5		ORTH	MED		
6	GYN	GYN	GYN	GYN	
7		GYN	GYN	GYN	GYN
8	URO	URO		URO	URO
9	CARD		URO		CARD
10	URO		ORTH		

Surgery type	Mean	STDEV
CARD	99	53
GASTRO	132	76
GYN	78	52
MED	75	72
ORTH	142	58
URO	72	38

# Data - Sykehuset Asker og Bærum HF

OR room	Monday	Tuesday	Wednesday	Thursday	Friday
1	GASTRO	GASTRO	GASTRO		
2			GASTRO	GASTRO	GASTRO
3	CARD		CARD		CARD
4	ORTH	ORTH		ORTH	ORTH
5		ORTH	MED		
6	GYN	GYN	GYN	GYN	
7		GYN	GYN	GYN	GYN
8	URO	URO		URO	URO
9	CARD		URO		CARD
10	URO		ORTH		

Surgery type	Mean	STDEV
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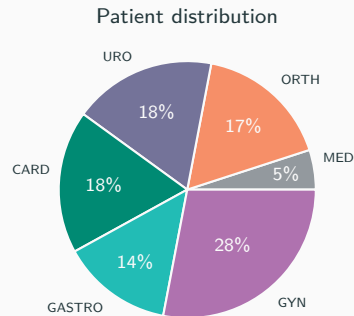




# Data - Sykehuset Asker og Bærum HF

OR room	Monday	Tuesday	Wednesday	Thursday	Friday
1	GASTRO	GASTRO	GASTRO		
2			GASTRO	GASTRO	GASTRO
3	CARD		CARD		CARD
4	ORTH	ORTH		ORTH	ORTH
5		ORTH	MED		
6	GYN	GYN	GYN	GYN	
7		GYN	GYN	GYN	GYN
8	URO	URO		URO	URO
9	CARD		URO		CARD
10	URO		ORTH		

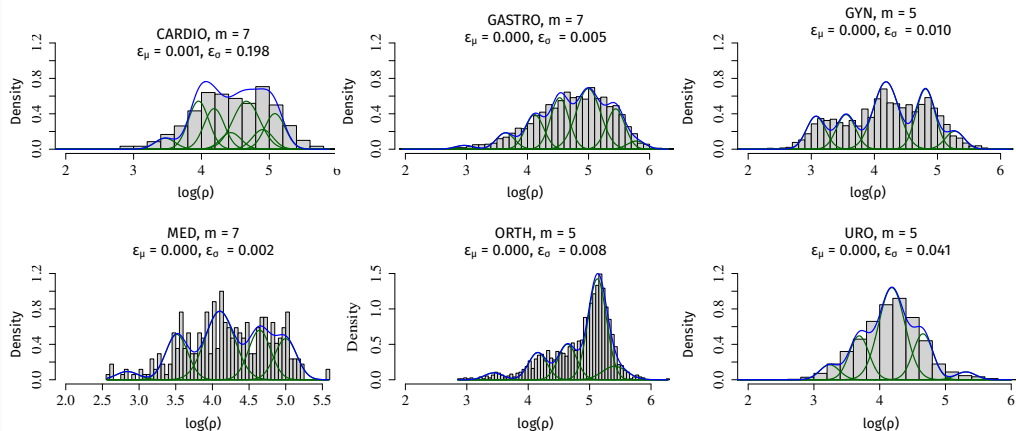
Surgery type	Mean	STDEV
CARD	99	53
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ORTH	142	58
URO	72	38



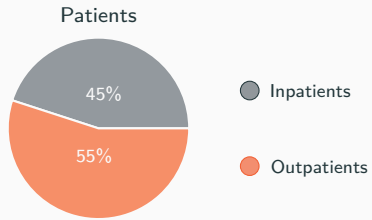
2 emergency surgeries per day.     ROTs  $\sim$  Lognormal(93, 60) min.

Mannino et al. (2010). SINTEF ICT: MSS adjusts surgery data. URL: <https://www.sintef.no/Projectweb/Health-care-optimization/Testbed/>  
Karmel S. Shehadeh and Luis F. Zuluaga (2022). "14th AIMMS-MOPTA Optimization Modeling Competition. Surgery Scheduling in Flexible Operating Rooms Under Uncertainty", Modeling and Optimization: Theory and Application (MOPTA)

# Instance Generation

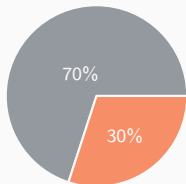
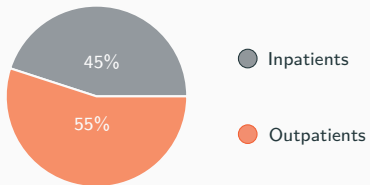


# Instance Generation

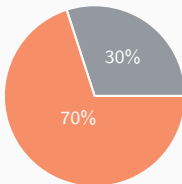


# Instance Generation

Patients



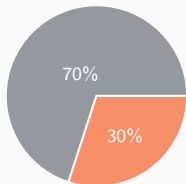
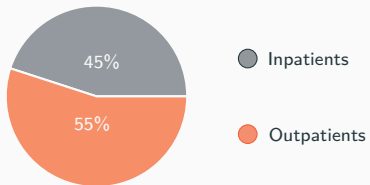
Patients with  
higher CoV



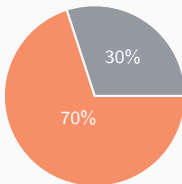
Patients with  
lower CoV

# Instance Generation

Patients



Patients with  
higher CoV



Patients with  
lower CoV

$$\text{CoV}_i = \sigma_i / \mu_i \in \{0.1518, 0.202\}$$

# Instance Generation

● Outpatients    ● Inpatients

No-show rate



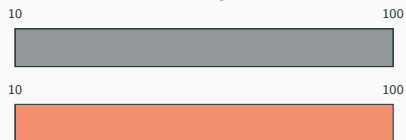
# Instance Generation

● Outpatients    ● Inpatients

No-show rate



Scheduling costs



# Instance Generation

● Outpatients    ● Inpatients

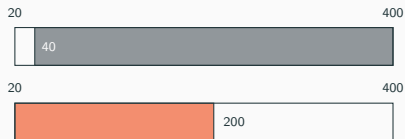
No-show rate



Scheduling costs



Cancellation costs





# Instance Generation

● Outpatients    ● Inpatients

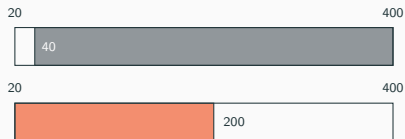
No-show rate



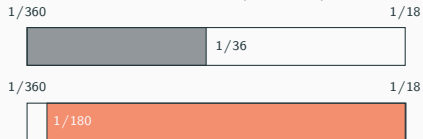
Scheduling costs



Cancellation costs



Waiting costs (per min)



# Instance Generation



Outpatients



Inpatients

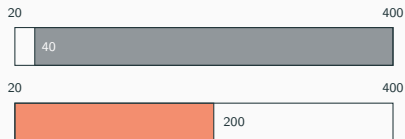
No-show rate



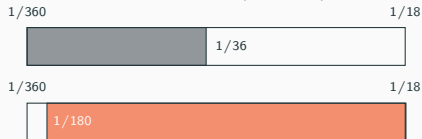
Scheduling costs



Cancellation costs



Waiting costs (per min)



Idle time cost =  $1/6$  per minute

Overtime cost =  $1/9$  per minute

# SAA vs SCI

Instances				SSA <sub>N</sub> (600s)			SSA <sub>N</sub> (60s)			SCI			
Spec.	W	SMIP	$\alpha$	o.f. value	r. time (sec)	robust. ratio	o.f. value	r. time (sec)	robust. ratio	short SAA <sub>N</sub> o.f. value	o.f. value	r. time (sec)	robust. ratio
ORTH	70	$\mathcal{A}_{ORTH}$	0.1	0.0	33.0	0.903	0.000	22.4	0.905	0.000	0.000	0.3	0.984
			0.3	0.000	33.1	0.909	0.000	22.5	0.929	0.000	0.000	0.3	0.980
			0.1	5.408	43.7	0.917	5.408	35.5	0.913	5.408	5.408	2.1	0.914
		$\mathcal{B}_{ORTH}$	0.3	5.408	40.7	0.904	5.408	TL	0.913	5.408	5.408	1.8	0.895
			0.1	-	46.5	-	-	31.3	-	0.000	0.000	1.6	0.956
			0.3	0.000	45.3	0.852	0.000	31.0	0.891	0.000	0.000	1.3	0.957
	100	$\mathcal{A}_{ORTH}$	0.1	-	196.00	-	-	TL	-	RFF	29.035	39.2	0.998
			0.3	9.389	158.5	0.822	9.389	TL	0.833	9.389	9.389	3.4	0.878
		$\mathcal{B}_{ORTH}$	0.1	-	76.8	-	-	51.9	-	RFF	310.000	41.8	0.984
			0.3	163.000	75.2	0.839	163.000	50.6	0.831	163.000	163.000	1.1	0.855
		$\mathcal{B}_{ORTH}$	0.1	-	282.2	-	-	TL	-	RFF	314.004	58.1	0.987
			0.3	166.021	245.1	0.85	166.021	TL	0.861	166.021	166.021	5.3	0.853
	150	$\mathcal{A}_{ORTH}$	0.1	-	107.1	-	-	TL	-	RFF	795.000	46.4	0.988
			0.3	579.000	92.3	0.837	579.000	TL	0.869	579.000	579.000	0.9	0.853
		$\mathcal{B}_{ORTH}$	0.1	-	205.2	-	-	TL	-	RFF	801.736	56.7	0.981
			0.3	585.463	127.4	0.807	585.463	TL	0.822	585.463	585.463	3.9	0.873
	200	$\mathcal{A}_{ORTH}$	0.1	-	157.9	-	-	TL	-	RFF	1316.000	39.0	0.983
			0.3	1151.000	146.3	0.816	1151.000	TL	0.793	1151.000	1151.000	2.1	0.829
		$\mathcal{B}_{ORTH}$	0.1	-	324.7	-	-	TL	-	RFF	1321.326	57.8	0.988
			0.3	1155.714	280.6	0.830	1155.714	TL	0.822	1155.714	1155.714	5.6	0.861
	300	$\mathcal{A}_{ORTH}$	0.1	-	395.3	-	-	TL	-	RFF	2743.000	TL	0.974
			0.3	2541.000	TL	0.812	2541.000	TL	0.827	2541.000	2541.000	2.6	0.851
		$\mathcal{B}_{ORTH}$	0.1	-	TL	-	-	TL	-	RFF	2748.111	TL	0.959
			0.3	2543.992	TL	0.832	2544.031	TL	0.811	2544.452	2544.167	TL	0.827

For instances with higher level of robustness the SAA approach is not able to find a feasible solution when the number of patients increases. SCI's solutions are very close to that of SAA, but it provides always a feasible solution due to RFF.

# SAA vs BRGKA

Spec.	# Pat.	a.f.	SAA(10 min)			SAA <sub>N</sub> (10 min)			BRGKA	
			o.f.	time	#feas./tot	o.f.	time	#feas./tot	o.f. (10min)	o.f. (1min)
ORTH	2-5	yes	115.30	8.27	24/24	99.83	0.49	24/24	<b>97.41</b>	<b>97.41</b>
		no	-	-	0/20	<b>155.17</b>	4.76	20/20	159.26	161.78
	6+	no	-	-	0/4	172.06	10	4/4	<b>145.23</b>	147.83
URO		yes	35.97	7.51	4/4	34.26	1.05	4/4	<b>33.28</b>	<b>33.28</b>
	2-5	no	-	-	0/9	<b>40.28</b>	6.71	9/9	67.41	70.49
	6+	no	-	-	0/35	170.19	10	34/35	<b>143.33</b>	150.13
GYN		yes	92.70	3.10	3/3	91.21	0.21	3/3	<b>64.43</b>	<b>64.43</b>
	2-5	no	-	-	0/15	102.26	8.97	15/15	<b>89.18</b>	<b>89.18</b>
	6+	no	-	-	0/46	167.37	10	42/46	<b>149.15</b>	152.22
MED		yes	-	-	0/2	78.25	10	2/2	<b>76.71</b>	<b>76.71</b>
	2-5	no	-	-	0/6	222.78	10	5/6	<b>176.60</b>	176.75
	6+	no	-	-	0/6	-	-	-	-	-
CARDIO		yes	46.63	9.89	4/4	36.22	1.76	4/4	<b>31.39</b>	<b>31.39</b>
	2-5	no	-	-	0/20	<b>157.08</b>	6.53	20/20	191.53	199.38
	6+	no	-	-	0/16	254.73	10	16/16	<b>223.97</b>	234.57
GASTRO		yes	139.60	6.84	13/13	<b>108.74</b>	1.87	13/13	109.40	111.01
	2-5	no	-	-	0/22	112.41	6.62	22/22	<b>106.70</b>	117.37
	6+	no	-	-	0/13	223.15	10	13/13	<b>182.55</b>	<b>182.55</b>

BRGKA always finds a better solution as soon as the **dimension of the problem** becomes challenging (6+ patients).

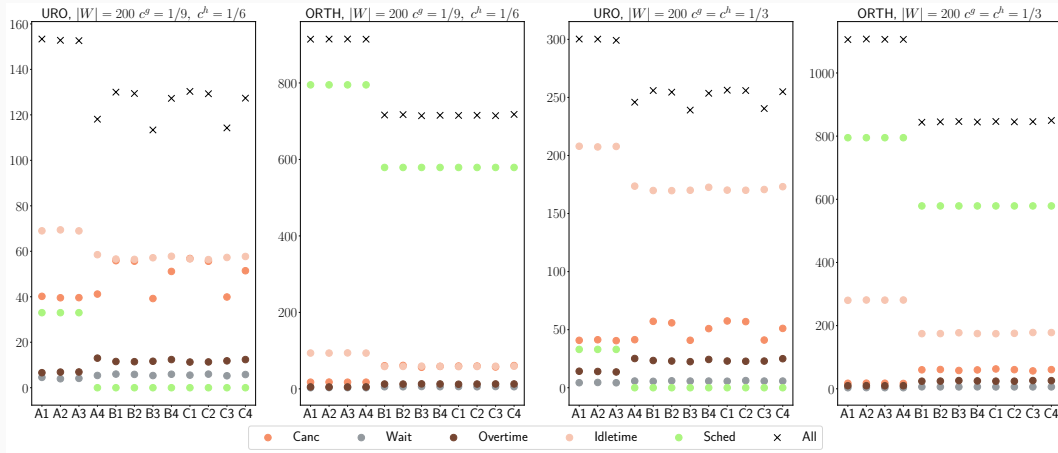
# Scenario Analysis: Parameter Variation

**Robustness level:** A = high ( $\alpha = 0.1$ ), B = medium ( $\alpha = 0.2$ ), C = low ( $\alpha = 0.3$ ).

**Cost balancing:** 1 = none, 2 = cancellations, 3 = both ( $\beta = \nu = 0.5$ ), 4 = waiting times.

## Scenario 1

## Scenario 2



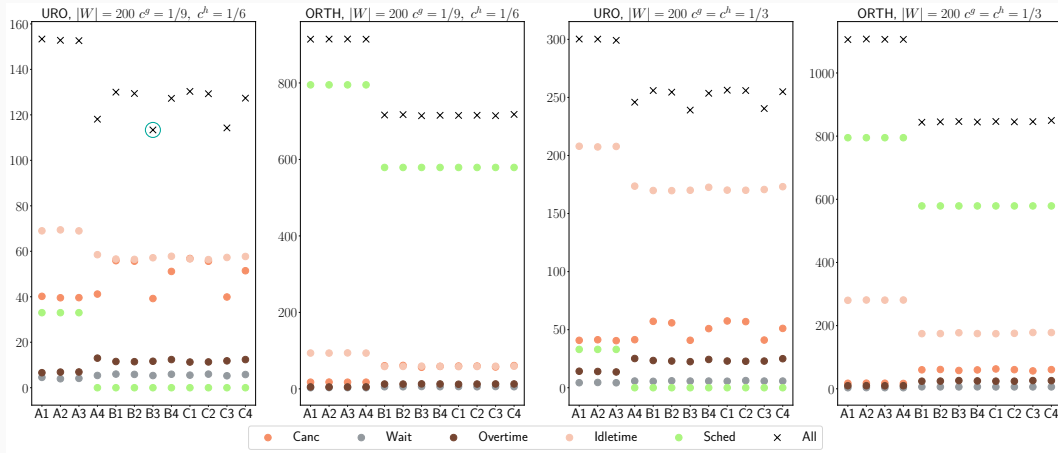
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## Scenario 2

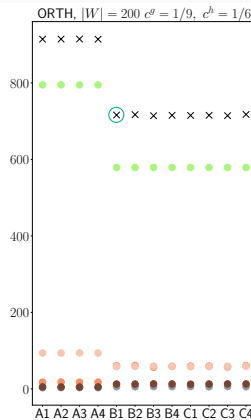
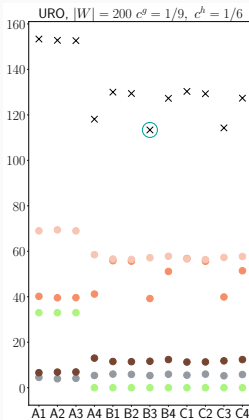


# Scenario Analysis: Parameter Variation

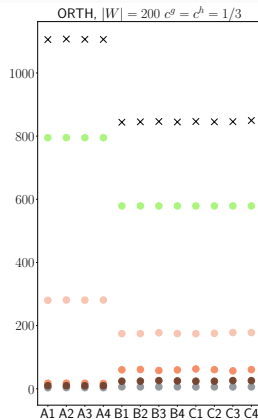
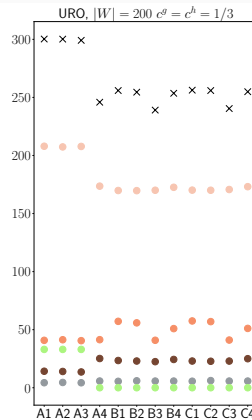
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## Scenario 2



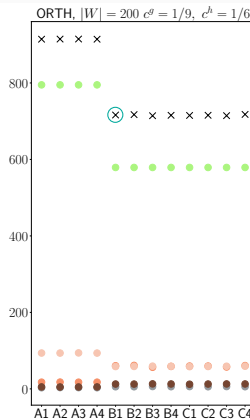
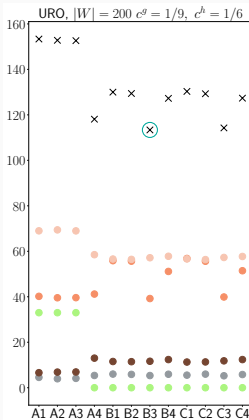
● Canc ● Wait ● Overtime ● Idletime ● Sched × All

# Scenario Analysis: Parameter Variation

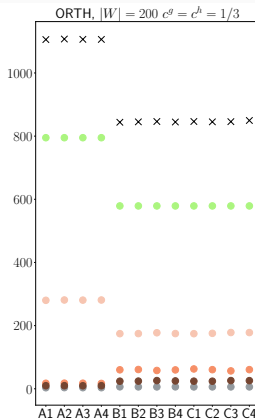
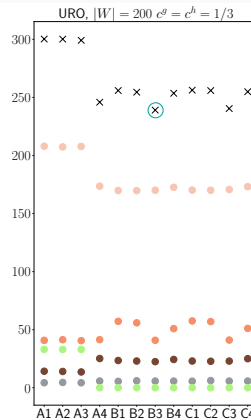
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## Scenario 1



## Scenario 2



● Canc ● Wait ● Overtime ● Idletime ● Sched × All



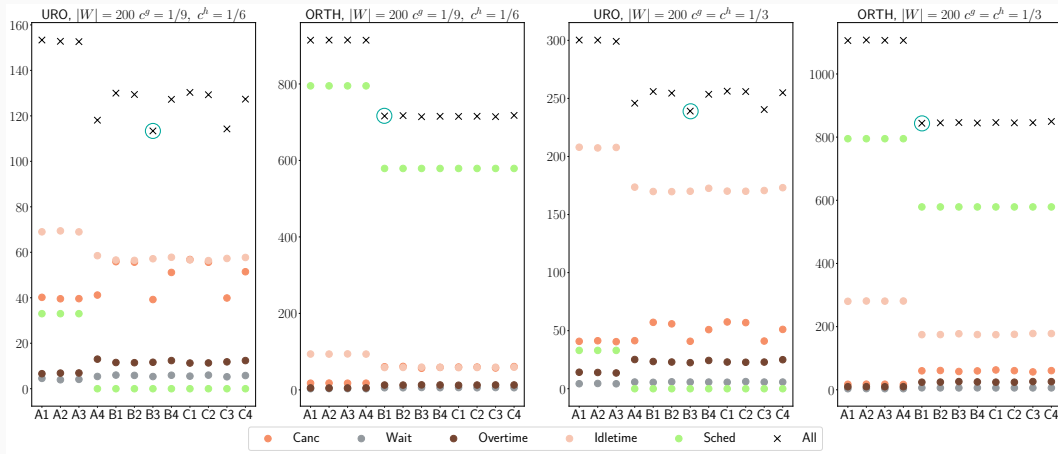
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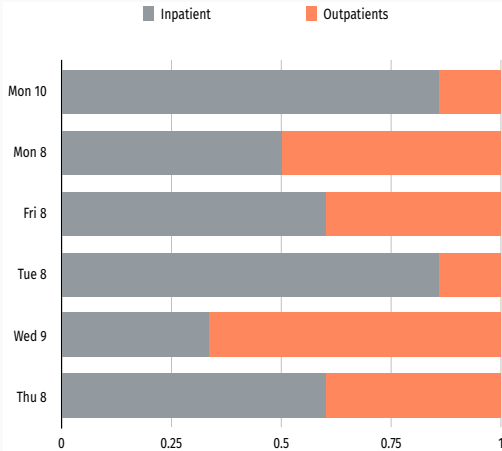
## Scenario 1

## Scenario 2

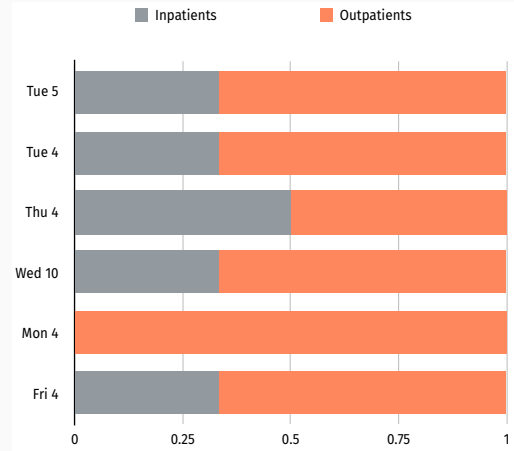


# Scenario Analysis: Inpatients vs Outpatients

## URO

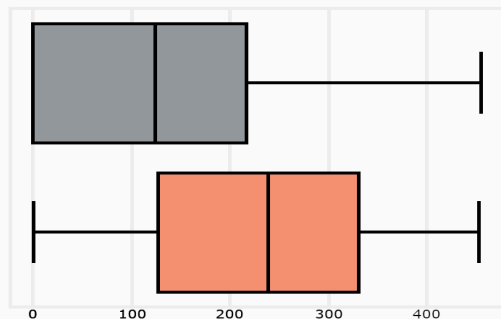


## ORTH

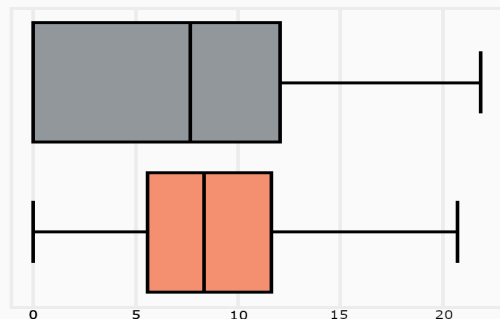


# Scenario Analysis: Inpatients vs Outpatients

Scheduled Start Times (min)



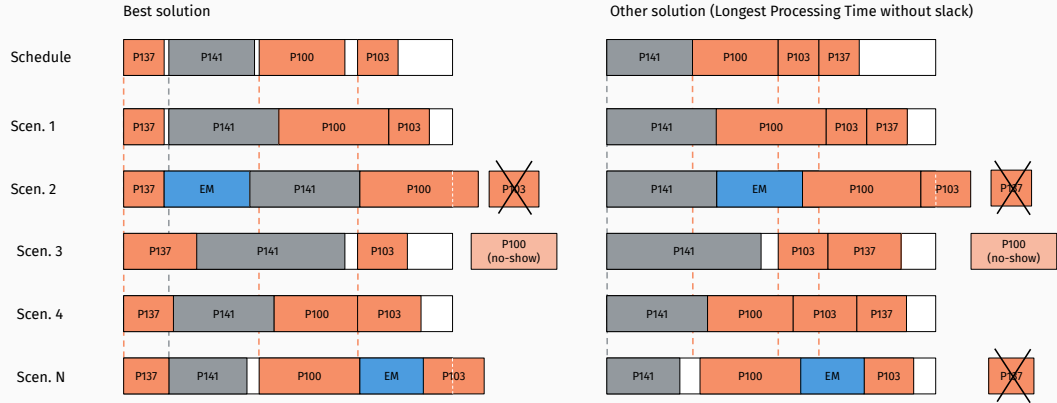
Expected Direct Waiting Time (min)



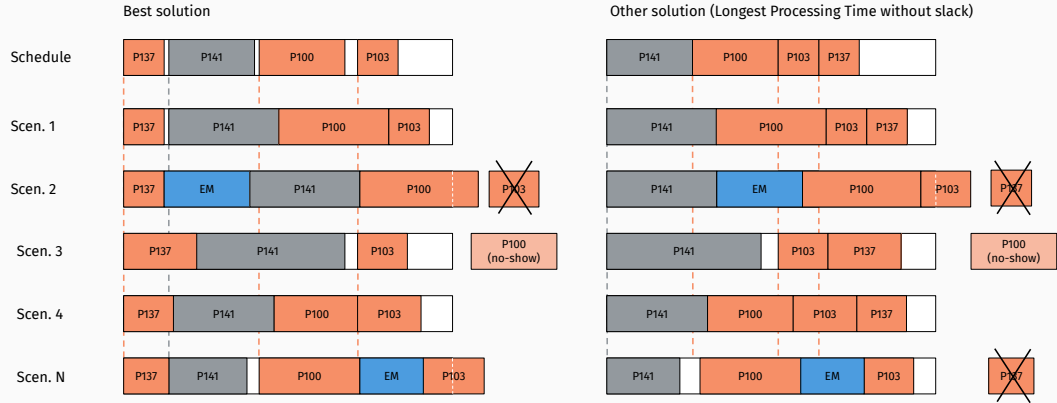
● Outpatients

● Inpatients

# Scheduling Examples



# Scheduling Examples



Video presentation - UI

## Conclusions

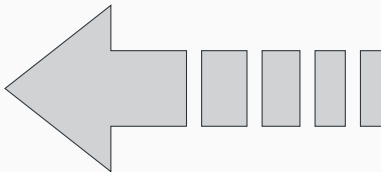
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# Final Remarks & Future Perspectives

Comprehensive approach to deal with different types of patients under uncertainty;

limitations of SAA methodology as soon as the combinatorial and stochastic complexities increase;

general insights: robustness vs. average performance & non-trivial best solutions.



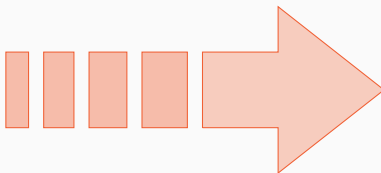
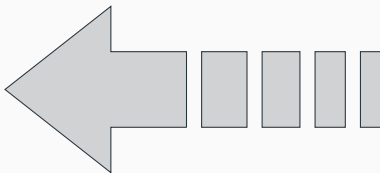


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Comprehensive approach to deal with different types of patients under uncertainty;

limitations of SAA methodology as soon as the combinatorial and stochastic complexities increase;

general insights: robustness vs. average performance & non-trivial best solutions.



Integrating SCI and BRKGA;

Alternative real-time policies:  
stochastic optimization  
+ online optimization;

impact of “robust  
decisions” over time.

*That's all Folks!*

Any Questions?

You can also send me an e-mail at [ambrogio maria.bernardelli01@universitadipavia.it](mailto:ambrogio maria.bernardelli01@universitadipavia.it)