



# **Better decisions for more effective emergency medical care**

**The case of the Portuguese Emergency Medical Service**

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# Agenda



OpLog brief introduction



Emergency medical services (EMSs) and the Portuguese EMS provider (INEM)



Ambulance dispatching and relocation



Integrated staff scheduling



Final remarks

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# Short background

BAC,  
Brussels



Operations  
Research



Maths  
Applied to  
Economics  
and  
Business



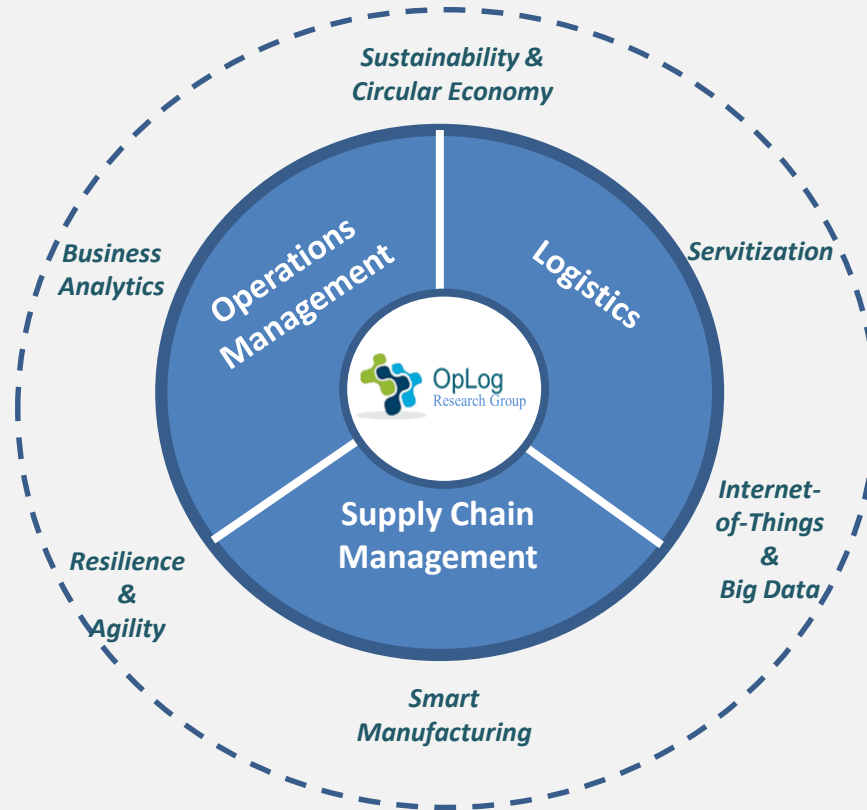
Statistics  
and  
Operations  
Research

Topic: Operating Room  
Planning & Scheduling

Assistant  
Professor  
at IST



# OpLog main research pillars



*Provide support to organizations by developing OR-based methods to inform decision-making.*

# OpLog develops innovative, theoretically sound and demand-driven research ...



... in a closed **collaboration** with national and international:

- Industrial and Service Organizations
- Academic Partners

Targeting

- economic
- environmental and
- social

**drivers**

# OpLog team

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## Post Docs

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Miguel Vieira

+1

**15 PhD Candidates**

**40+ Ongoing MSc Students**



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# What is an EMS?



“a comprehensive system which provides the **arrangements** of **personnel, facilities and equipment** for the **effective, coordinated and timely delivery** of health and safety services to victims of sudden illness or injury”



Simulation, Optimization, Data Science and Artificial Intelligence (AI)



Improve decision making – more effective, coordinated and timely health delivery

# Current international EMS systems evolved from two main models

## Franco-German model *Stay and stabilize*

Main idea



Main provider



Patients treated

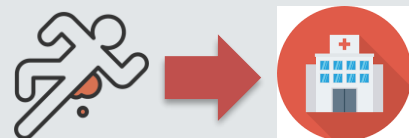


Destination for transported patients







Ward

## Anglo-American model *Scoop and run*



# Why is EMS such an important health service?

- Main goal: provide **timely basic medical care** to victims or emergencies 
- **Prevents** needless mortality or long-term morbidity
- Corresponds to the **pre hospital assistance**
- Has to manage and mobilize **several resources**   
- Aims to serve as many emergencies as possible with an **effective response**

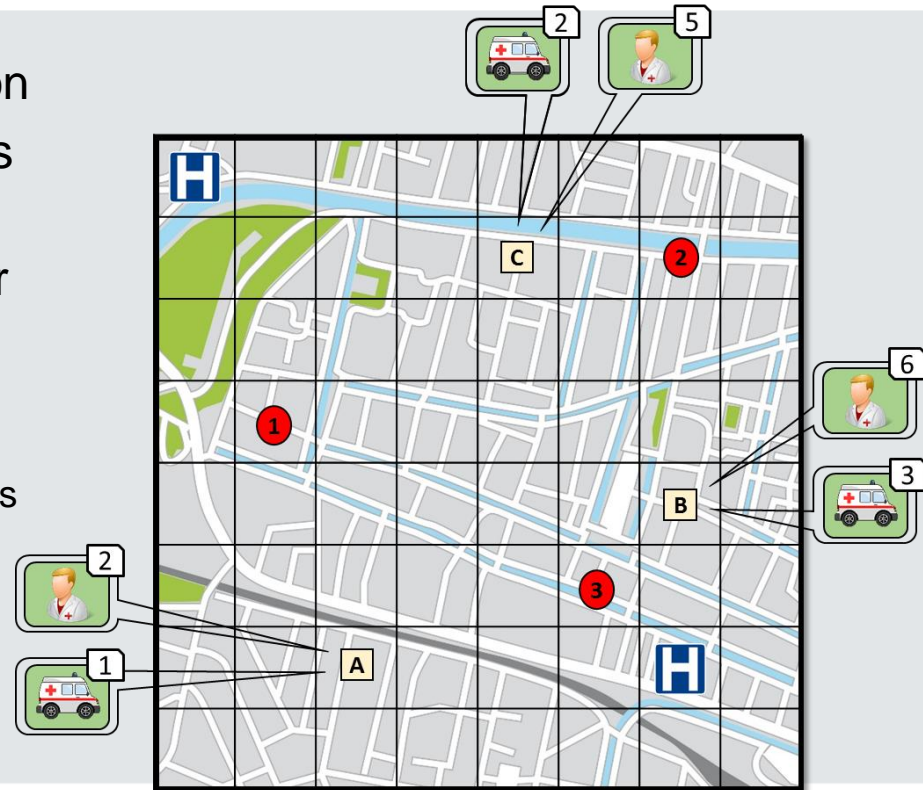


**Response time:** total time between an emergency call received by the system and the moment that an ambulance arrives at the scene



# EMS environment

- **Sub zones:** Basic subdivisions of the region
- **Hospitals:** Health care facility that provides patient treatment
- **Ambulance Bases:** Structures or areas for storage of ambulances
  - **Depots:** A base where ambulances start and end their shift
  - **Potential standby sites:** Sites where ambulances can park during the day while waiting for emergencies
- **Emergency points:** Location where an emergency occurs



# Open questions



- How many?
- Where?
- When?
- How severe?

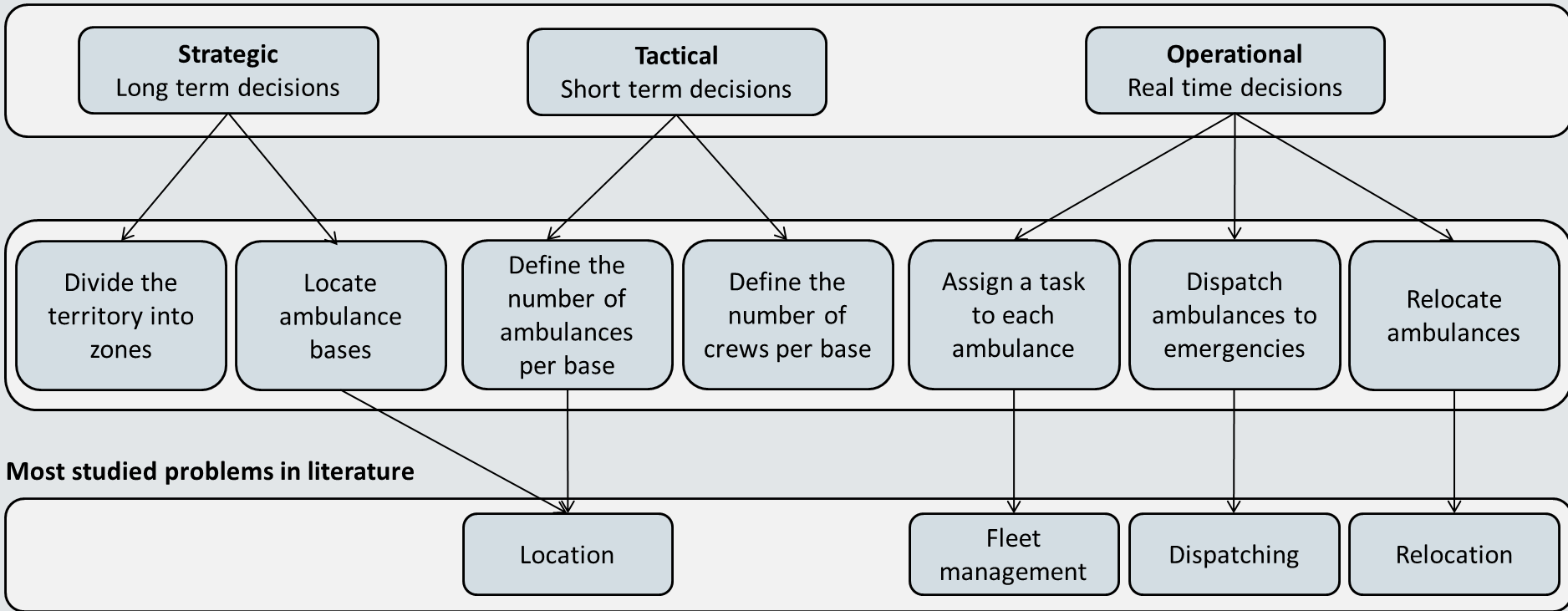


- How many?
- Where?
- When?



- How many?
- Where?
- Should standby sites be used?

# Decision-making process



# INEM: Resources and operational results in the media (August 2018, Público)



SAÚDE

## Presidente do INEM diz que falta de técnicos não afecta socorro

27 de Agosto de 2018

9



INEM

## Sindicato pede abertura de mais concursos para técnicos de emergência pré-hospitalar

RITA MARQUES COSTA

27 de Agosto de 2018

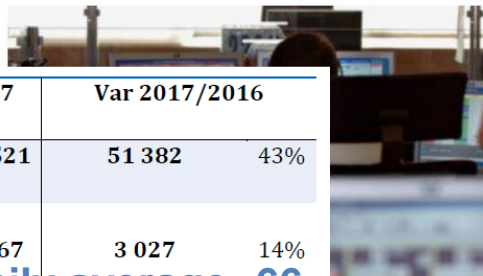


INEM

## INEM prevê renovar até 2021 frota de ambulâncias

26 de Agosto de 2018

2



SAÚDE

## INEM demorou cerca de 30 segundos a atender chamadas nos primeiros sete meses do ano

ANA MAIA

10 de Agosto de 2018



Atividade

2013

2014

2015

2016

2017

Var 2017/2016

Notícias Setor da saúde

102 203

148 469

125 422

120 139

171 521

51 382

43%

Notícias c/ referência INEM  
(imprensa, televisão, rádio e internet)

15 061

22 341

19 203

20 940

23 967

3 027

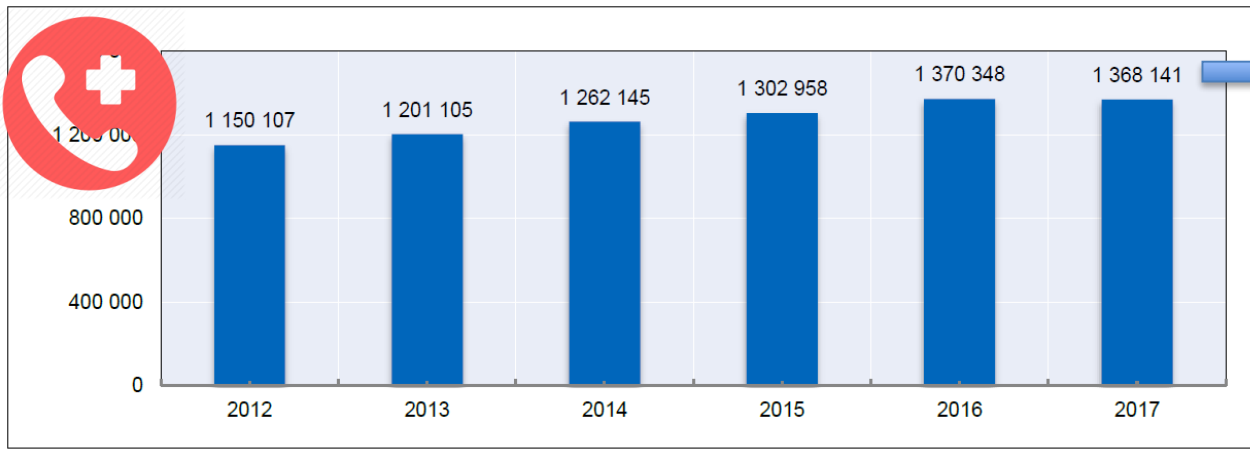
14%


14%

Daily average: 66




# Some figures 2017



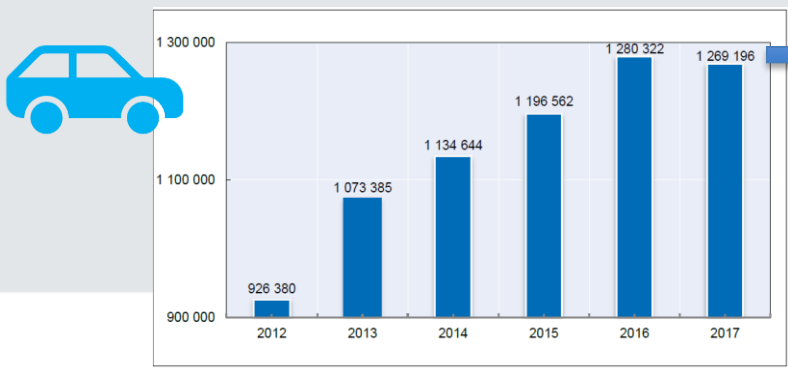
 75.843 (5.5%)

Daily average:

 3.748

 208

 3.477



 93%

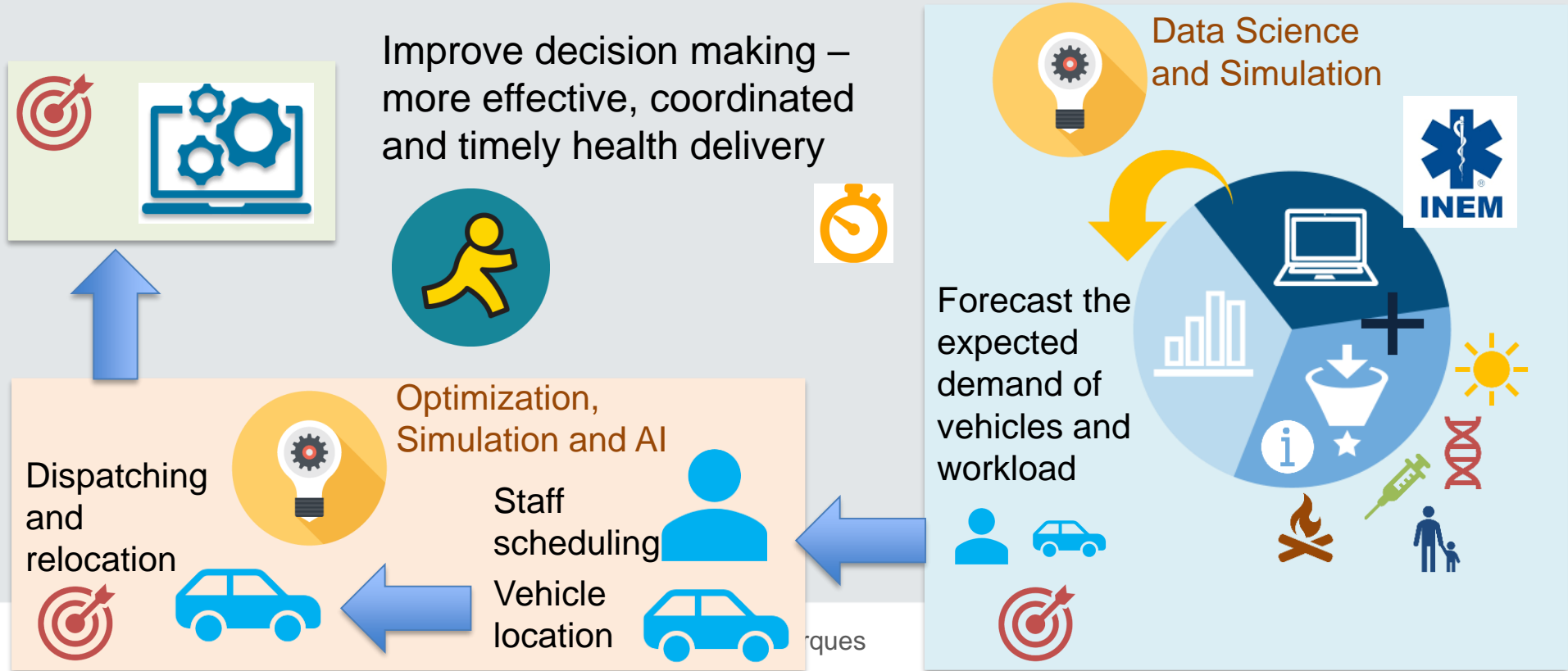
Inês Marques

 966

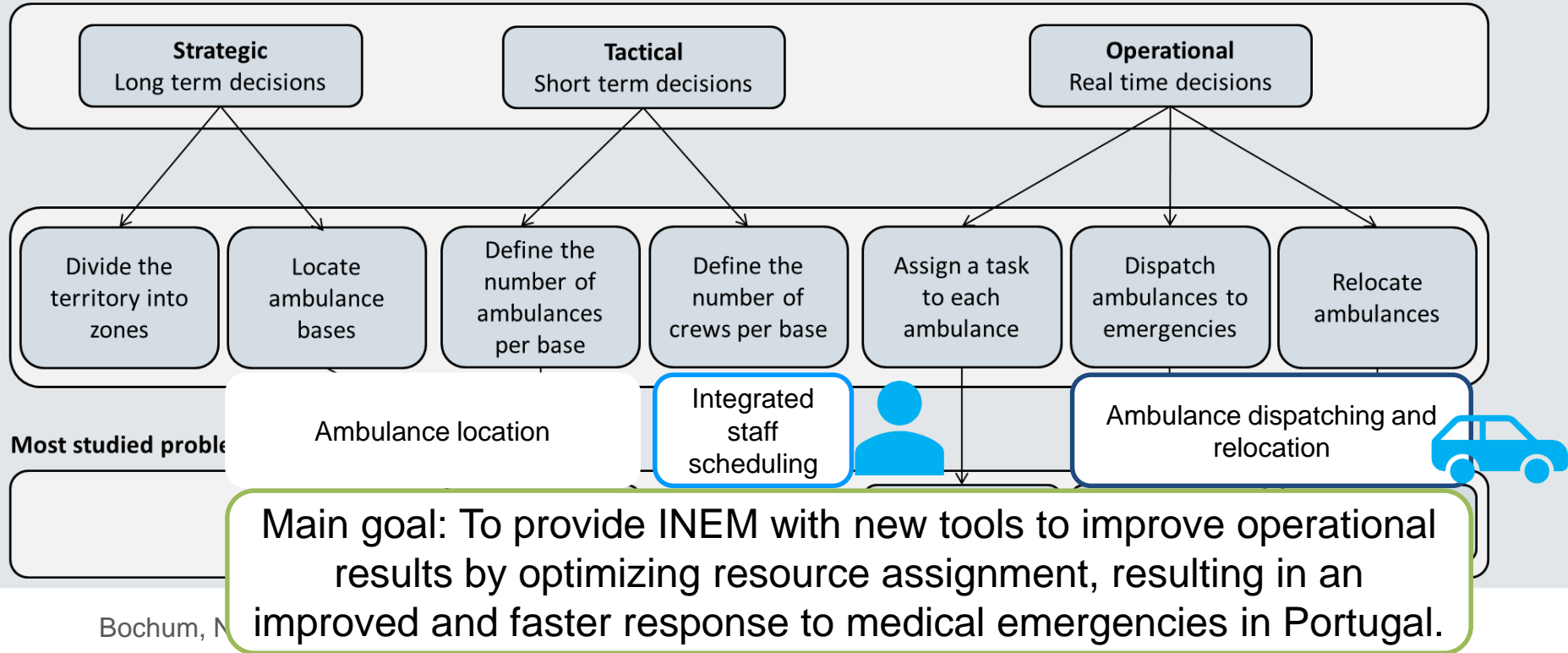
 652



# Data2Help: Data Science for Optimization of EMSs



# Data2Help and the Decision-making process



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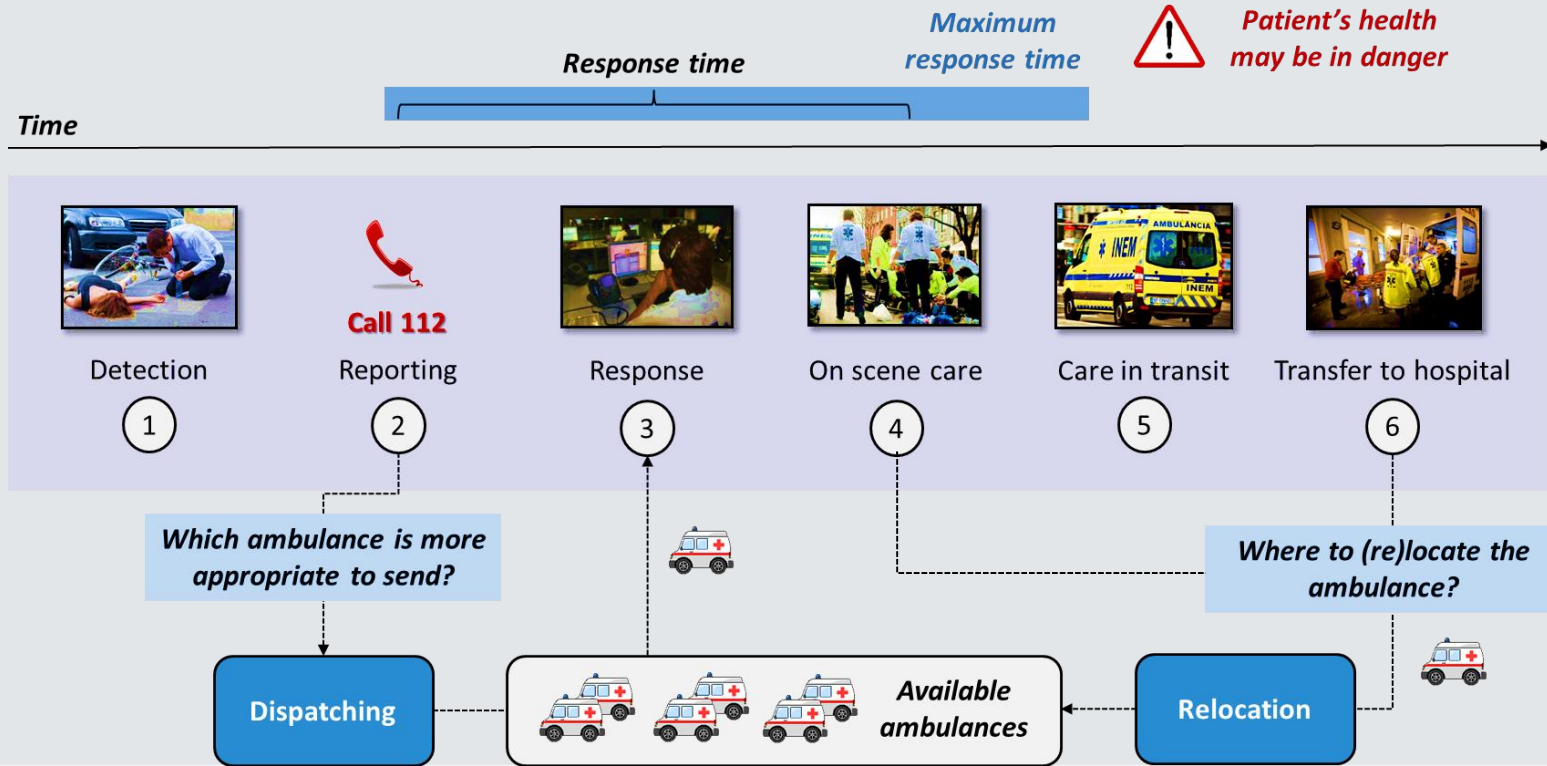


Integrated staff scheduling



Final remarks

# Chain of events















# Motivation

- **Response time is** (sometimes) **too high** affecting patients' health state
  - Effective and efficient emergency response is an **issue that concerns society**
  - Emergency decisions are still based on **staff common sense**
- **Aim: develop optimization tools to dispatch and relocate ambulances**
- analyze the importance of changing the current policies at INEM  
(Dispatch closest available ambulance and relocate ambulances to home base)



# Decisions

-  Hospital
-  Base
-  Ambulance
-  Available ambulance at a base
-  Available ambulance on the road
-  Available ambulance at a hospital
-  Available ambulance at an emergency
-  Busy ambulance
-  Current emergency
-  Previous emergency
-  ---> Route previously defined
-  — Sub-zone limit

**Decision 1**  
Dispatching  
problem

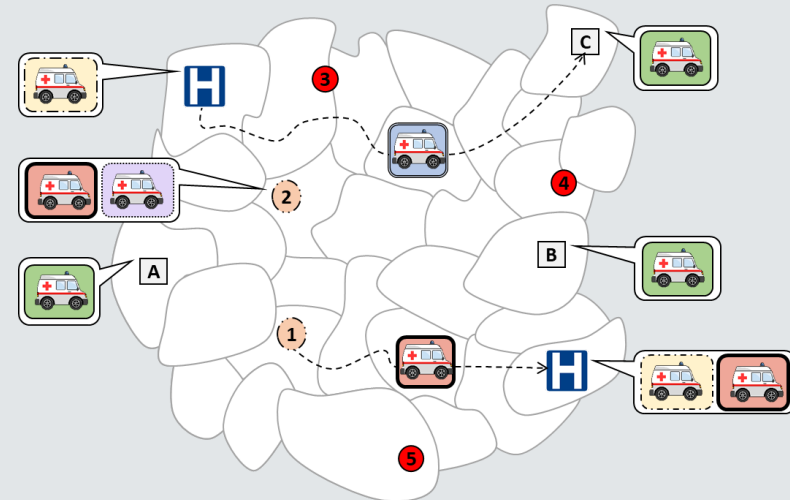
Which (available) ambulance goes to each actual emergency point?

**Decision 2**  
Relocation  
problem I

To which base the available ambulances go after completing the service?

**Decision 3**  
Relocation  
problem II

Are additional relocations between bases needed?



**Main goal:** ensure a good system coverage to provide quick response times to current and future emergencies

# Dynamic problem: coverage

*How to analyze system coverage?*

- **Preparedness** evaluates the ability to serve patients now and in the future
- *Time-preparedness metric*
  - Preparedness for a fleet of available ambulances in the area under study
  - Depends on the historical number of emergency calls and on the travel time
  - increases with available ambulances closer to a sub-zone
  - decreases with the call frequency and the travel time at each moment

# MILP model

**Decision 1**  
Dispatching  
problem

$$x_{ij} = \begin{cases} 1 & \text{if available ambulance } i \text{ is dispatched to emergency } j \\ 0 & \text{otherwise} \end{cases}$$

**Decision 2**  
Relocation  
problem I

**Decision 3**  
Relocation  
problem II

$$y_{il} = \begin{cases} 1 & \text{if available ambulance } i \text{ is assigned to base } l \\ 0 & \text{otherwise} \end{cases}$$

+

$$r_{ij} = \text{extra response time for ambulance } i \text{ to reach emergency } j$$



# MILP model

System capability to handle new emergencies in the future

→ Min system response time for future emergencies

Response time for current emergencies

$$\begin{aligned}
 \min \quad & \sum_{q \in Z} \lambda_q \left( \sum_{l \in B} d_{lq} \delta_{lq}^{min} + INF^{Travel\ time} w_q \right) + \sum_{j \in E} \left( \sigma_j + \sum_{i \in A} d_{ij} x_{ij} \right) \\
 & + P^{Uncovered} \sum_{j \in E} u_j + P^{Response} \sum_{i \in A} \sum_{j \in E} r_{ij} + P^{Target} \sum_{i \in A^{Target}} z_i \quad (2)
 \end{aligned}$$

Number of  
uncovered  
emergencies

Extra response  
times

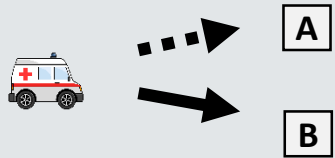
Changes in  
target bases

# MILP model



Uncovered emergency definition

$$\text{s.t.: } u_j = 1 - \sum_{i \in A} x_{ij}, \forall j \in E$$



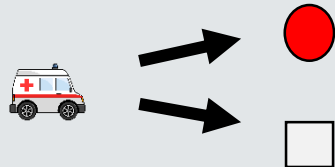
Change target definition

$$z_i = 1 - \left( y_{ib_i} + \sum_{j \in E} x_{ij} \right), \forall i \in (A^B \cup A^R)$$

$$z_i = 0, \forall i \in \{A^B \cup A^R\} \setminus \{A^{Target}\}$$

Maximum number of changes  
in target bases

$$\sum_{i \in A^{Target}} z_i \leq Target^{MAX}$$



Each available ambulance is  
dispatched to an emergency  
or assigned to a base

$$\sum_{j \in E} x_{ij} + \sum_{l \in B} y_{il} = 1 \forall i \in A$$

$$\sum_{i \in A} x_{ij} \leq 1, \forall j \in E$$

# MILP model



Maximum response time

$$\sigma_j + d_{ij} x_{ij} \leq R^{MAX} + r_{ij}, \forall i \in A, j \in E$$


 Whether a base is empty  
 of ambulances or not

$$v_l \leq \sum_{i \in A} y_{il} \leq |A| v_l, \forall l \in B$$


 Minimum time to reach a  
 sub-zone

$$d_{lq} \delta_{lq}^{min} \leq d_{l'q} + M^{Travel\ time} (1 - v_{l'}), \forall l, l' \in B, q \in Z$$

$$\delta_{lq}^{min} \leq v_l, \forall l \in B, q \in Z$$

 Defines the base that reaches the  
 sub-zone in minimum time or no  
 ambulance at all

$$\sum_{l \in B} \delta_{lq}^{min} + w_q = 1, \forall q \in Z$$

 Binary-domain and real-domain  
 variables

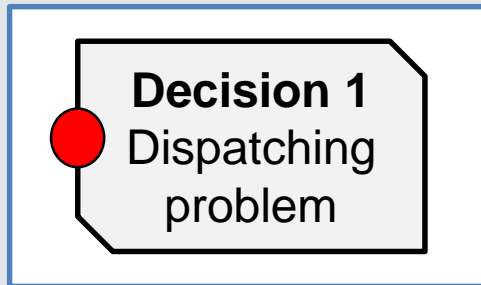
$$x_{ij}, y_{il}, u_j, v_l, z_i, w_q, \delta_{lq}^{min} \in \{0, 1\},$$

$$\forall i \in A, i' \in \{A^B \cup A^R\}, j \in E, l \in B, q \in Z$$

$$r_{ij} \geq 0, \forall i \in A, j \in E$$

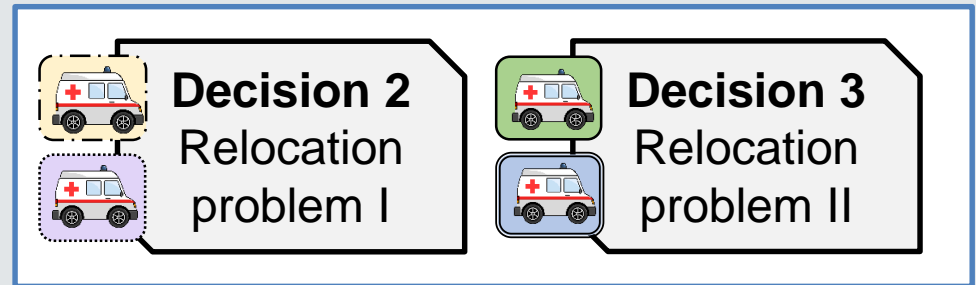
# Heuristic: overview

## Phase 1



needs to be solved within few seconds in order to ensure a quick response time

## Phase 2



can be solved in slightly more time (not too much) as these are not immediate critical decisions

# Phase 1: Dispatching problem

## Phase 1

**Decision 1**  
 Dispatching  
 problem

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### Algorithm 1 Heuristic phase 1 - dispatching decisions

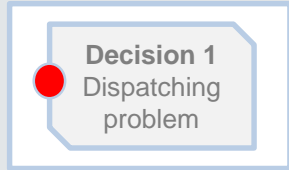
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- 1: **for** all emergencies  $e \in E^*$  **do**
  - 2:     Calculate  $p_{(A \setminus i)}^{(t+1)}$
  - 3:     Dispatch available ambulance  $i'$  such that  $i' = \operatorname{argmax}_{i \in A} \frac{[p_{(A \setminus i)}^{(t+1)}]^\alpha}{1 + d_{ie}}$
  - 4: **end for**
  - 5: \* Emergencies are analyzed by descending order of the corresponding waiting time, i.e. emergencies that occurred earlier are considered first.
- 

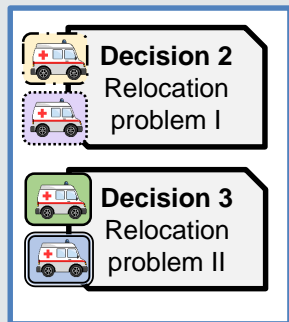
$\alpha$ : calibration factor that weights on preparedness (closest-policy if  $\alpha = 0$ )

# Phase 2: Relocation problem

## Phase 1



## Phase 2



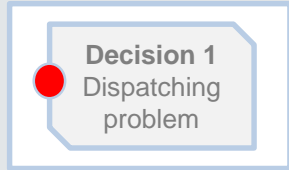
### Algorithm 2 Heuristic phase 2 - relocation decisions

- 1: Define  $S$  as the master solution with an undefined initial cost,  $Cost_S = +\infty$
- 2: Define  $A^{Fixed}$  as the set of available ambulances fixed in  $S$ ,  $A^{Fixed} = \emptyset$
- 3: **while**  $[(A - A^{Fixed}) \neq \emptyset$  **and**
- 4: (number of ambulances that changed target base  $\leq Target^{MAX}$  **or** there are ambulances at an emergency or at a hospital to return to a base) ] **do**
- 5:     **for**  $a \in (A - A^{Fixed})$  **do**
- 6:         Apply **pilot heuristic** (Algorithm 3) to obtain  $Cost_a$  by fixing  $a$
- 7:         Update **master solution**  $S$  if  $Cost_a < Cost_S$
- 8:     **end for**
- 9: **end while**

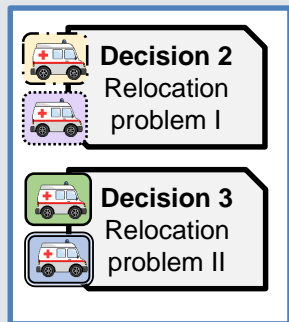
- A **pilot heuristic** is used to generate and compare solutions (tempered greedy meta-heuristic - look-ahead method)
- A **master solution** is extended at the end of each iteration with the best pilot solution

# Phase 2: Relocation problem

## Phase 1



## Phase 2



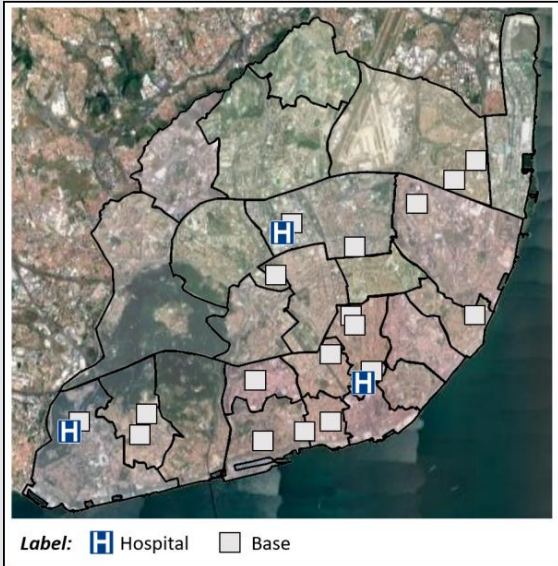
### Algorithm 3 Pilot heuristic

```

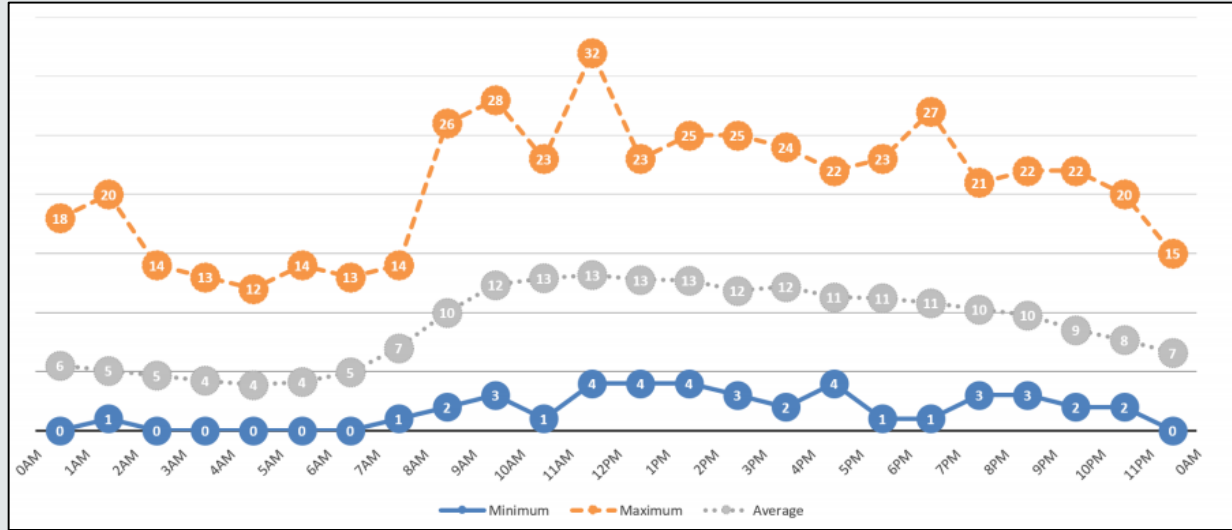
1: Initial ambulance:  $a$ 
2: while [ there are available ambulances to analyze and
3: (number of ambulances that changed target base  $\leq Target^{MAX}$  or there
   are ambulances at an emergency or at a hospital to return to a base) ] do
4:   i) Choose base  $b$  to assign ambulance  $a$ 
5:   if ambulance  $a$  is at an emergency or at a hospital then
6:     Choose between all bases  $b \in B$ , the one that maximizes  $p_A^{(t+1)}$ 
7:   else
8:     Choose between maintaining the current assigned base or changing
     to the closest base, the one that maximizes  $p_A^{(t+1)}$ 
9:   end if
10:  ii) Choose the following ambulance  $a^*$  to analyze
11:     $a^*$  is the furthest available ambulance from  $a$ 
12:     $a = a^*$ 
13:  iii) Update solution cost and other info
14: end while
  
```

- Pilot heuristic: tempered greedy meta-heuristic (look-ahead method)

# Tests: INEM data



Lisbon region divided into sub-zones



Number of emergencies per hour in Lisbon (2nd half of 2016)



# Tests: Instances and scenarios

## Input data (INEM case study)

- **Ambulances:** 37
- **Ambulance bases:** 19
- **Hospitals:** 3
- **Sub-zones:** 24
- **Maximum response time:** 900 seconds

## Rolling-horizon scenario

- **1 day: 288 time periods (1 time period = 5 minutes)**
- **Random number of emergencies with random location at each time period**
- **20% of the emergencies does not require hospital-care**
- **Emergency service takes between 10-40 minutes**
- **1 hour is required to be available for a change to a different target base**

## 50 instances tested

# Tests: Strategies

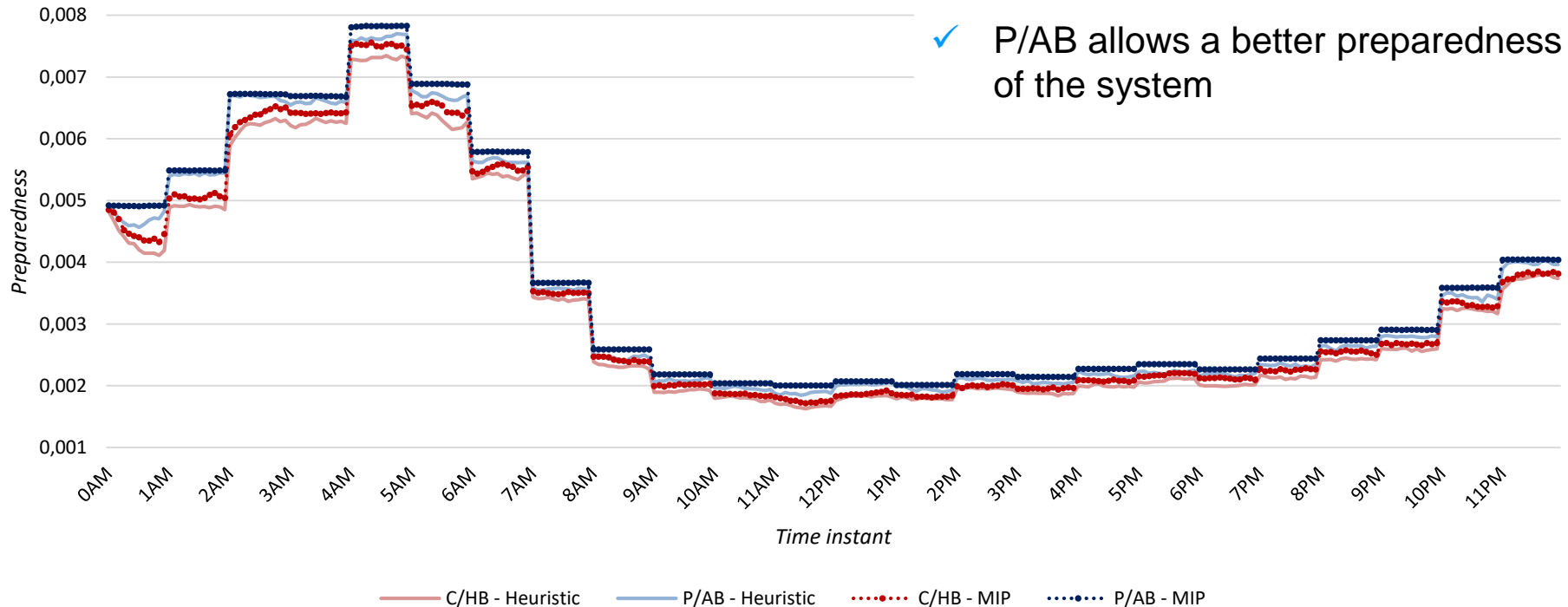
Strategies		Relocation	
		Any base	Home base
Dispatching	Closest (available)		C/HB
	Preparedness	P/AB	

MIP and heuristic adapted to consider INEM policies

MIP and heuristic

# Results: preparedness

*Average preparedness level*

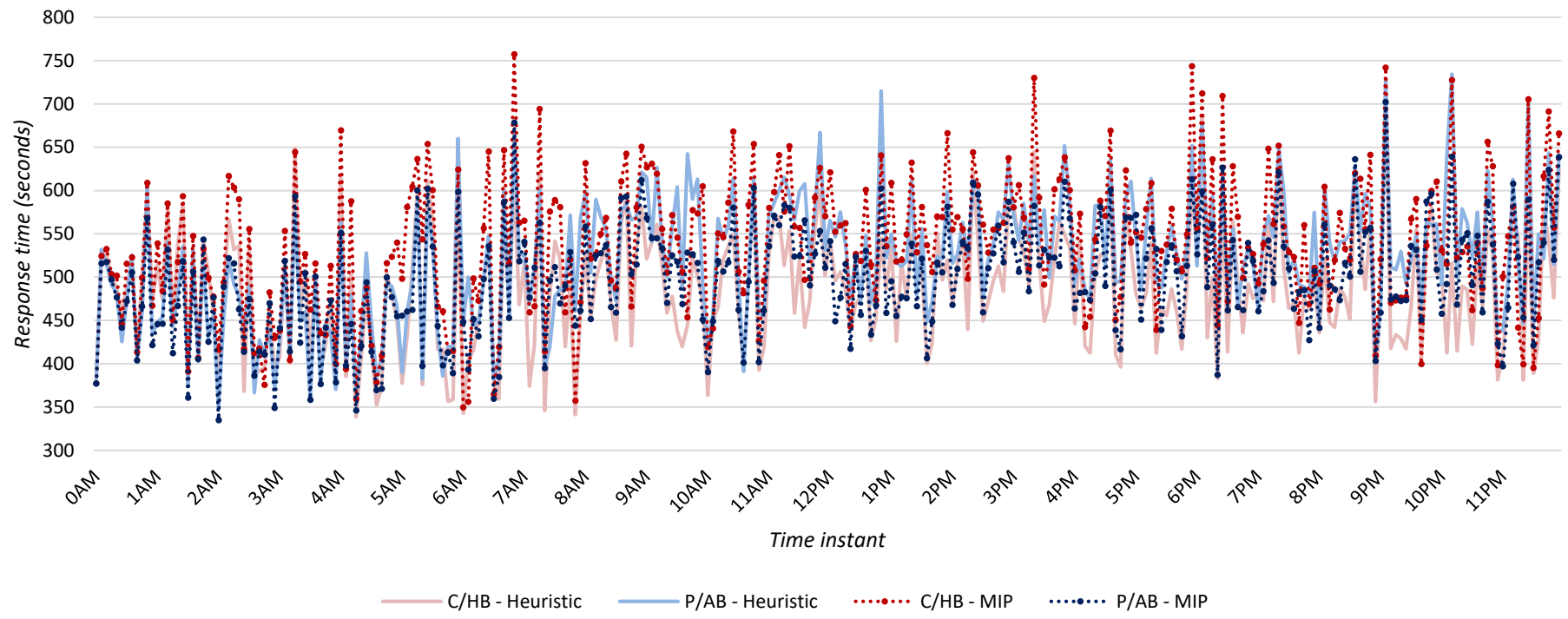


# Results: response time

Average response time



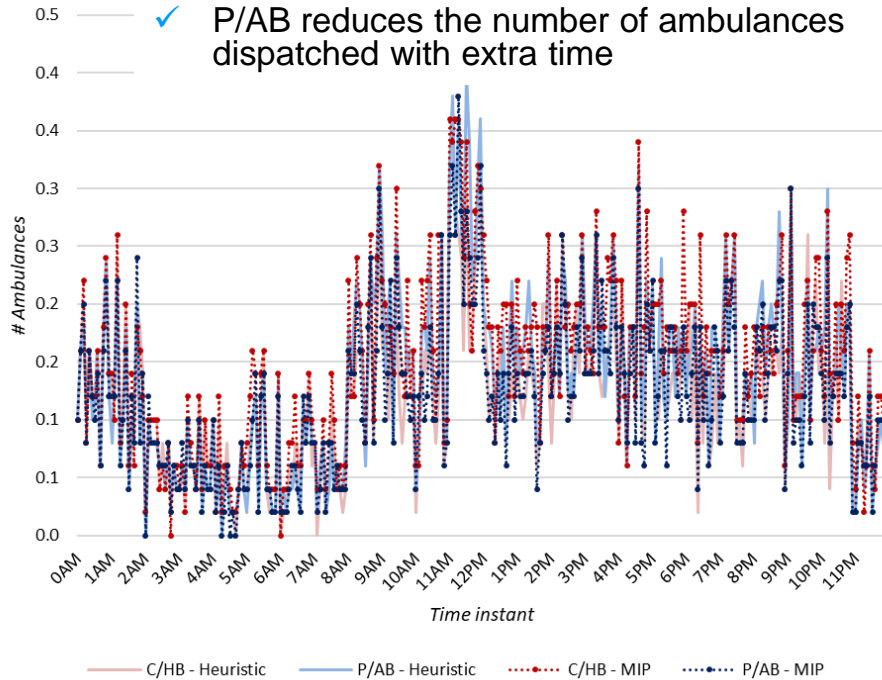
There is no consistent pattern



# Results: extra time

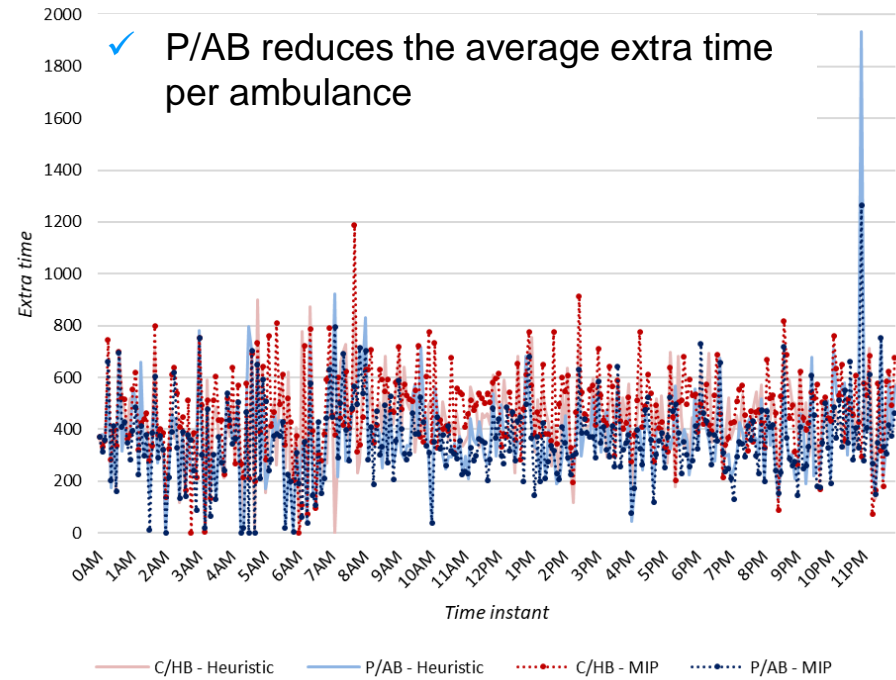
Average # ambulances assigned with extra time

✓ P/AB reduces the number of ambulances dispatched with extra time



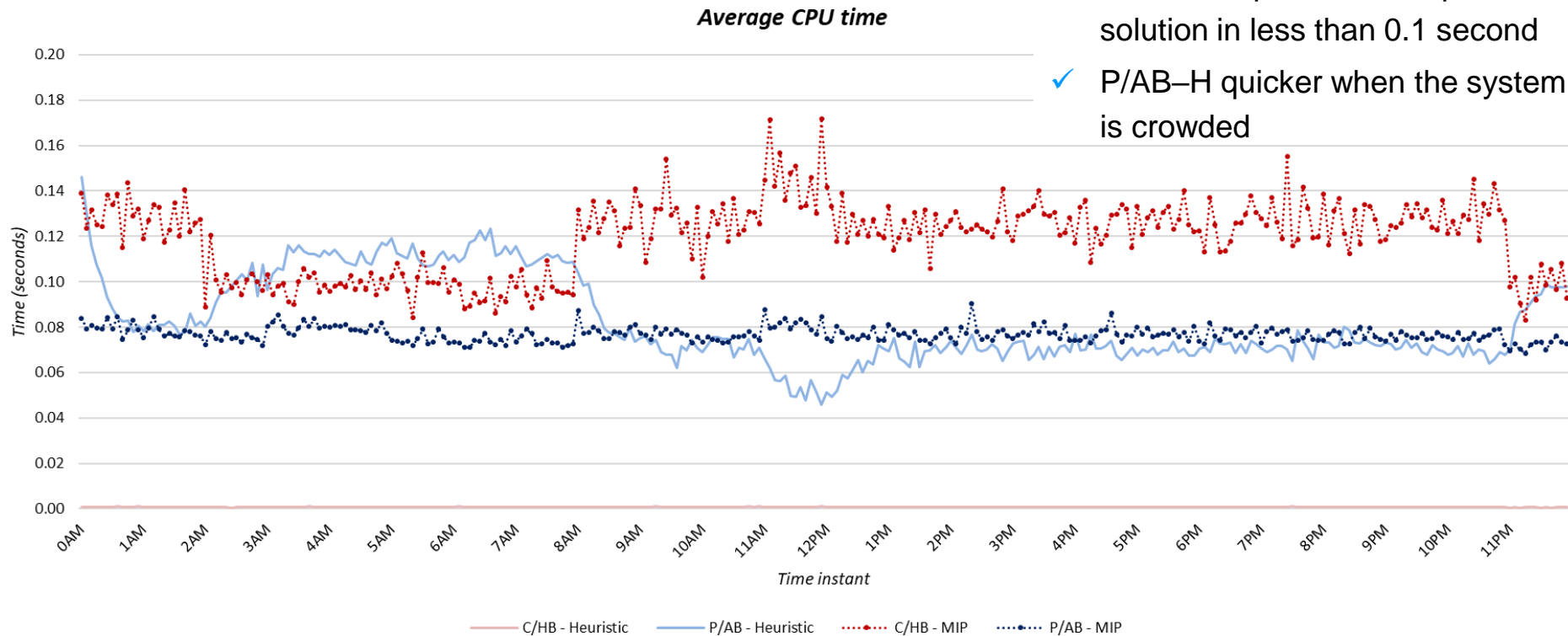
Average extra time per ambulance with extra time

✓ P/AB reduces the average extra time per ambulance



# Results: CPU time

- ✓ P/AB-MIP provides an optimal solution in less than 0.1 second
- ✓ P/AB-H quicker when the system is crowded



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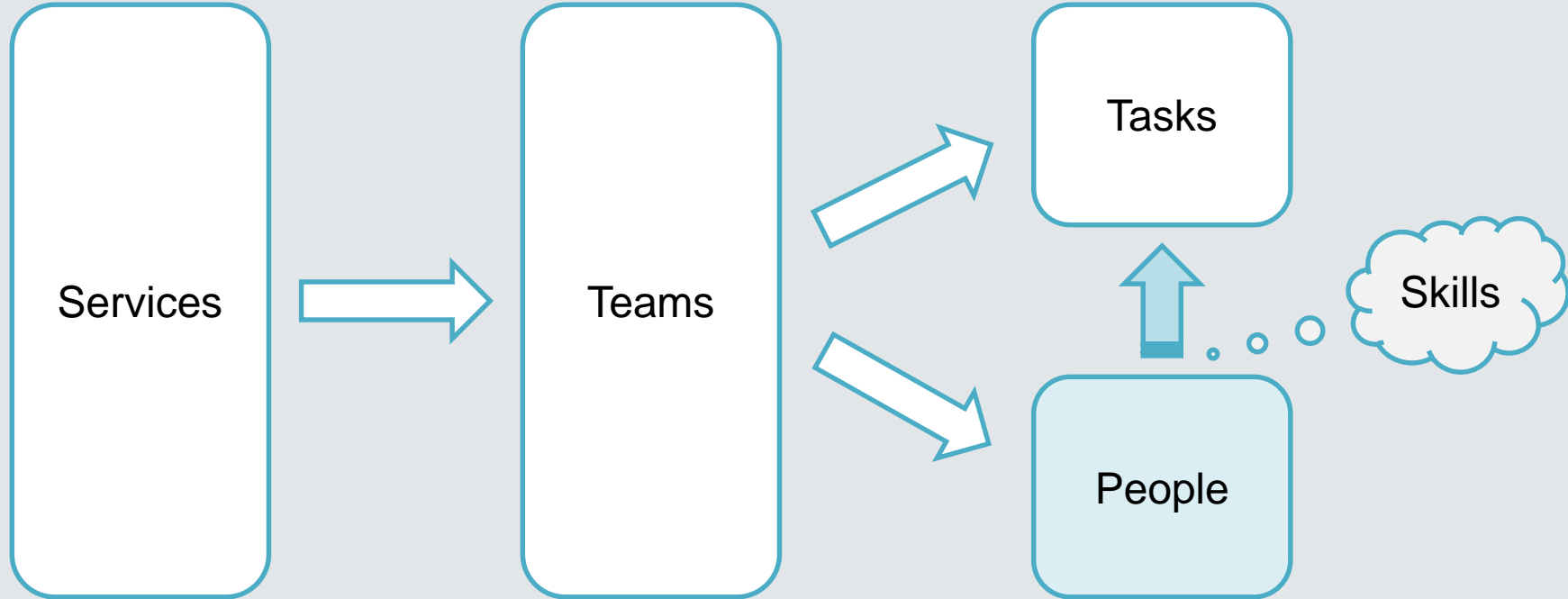
# Motivation

- Quality of the schedule impacts the quality of the emergency care
  - Employee preferences is directly related to on-job performance
  - Timetables are still manually constructed
  - Few research in staff scheduling in EMSs
- **Aim: novel solution approach and automated scheduling tool**
- Provide staff schedules in significantly less time
  - Improve quality and transparency of the schedules
  - Increase employee perception about fairness
  - Focus on functionality of the services and equity among the staff





# Problem statement



# Problem statement

- Services operate 24/7
- Fixed shifts
- Required personnel coverage for each task, day and shift
- Legal regulations
- Organizational and contractual issues
- Integrated staff scheduling for a set of services that share the same workforce



# Problem statement

**MASTER****SUBPROBLEMS**

- Hard constraints

1. Required skills for each assigned task
2. Minimum rest between working shifts
3. Maximum number of consecutive working days
4. Maximum number of consecutive days off
5. Minimum number of Sundays off
6. Minimum number of each shift type

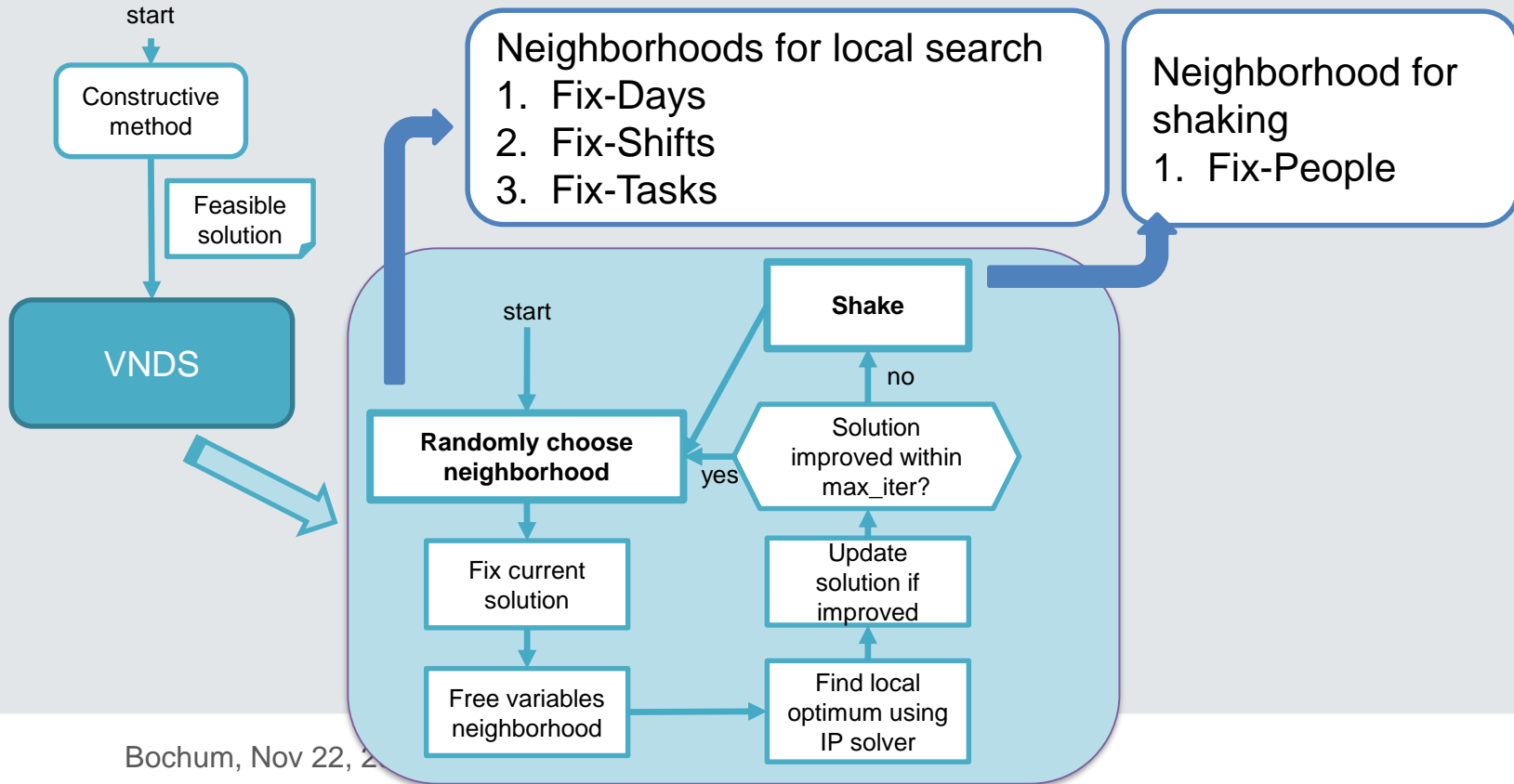
Functionality of  
services

- Soft constraints

1. Coverage requirement (understaffing and overstaffing allowed)
2. Full weekends off
3. Working time equal to contract hours
4. Assign to tasks of own team as much as possible

Equity among  
staff

# Variable Neighborhood Decomposition Search (VNDS)



# Results: case study at INEM

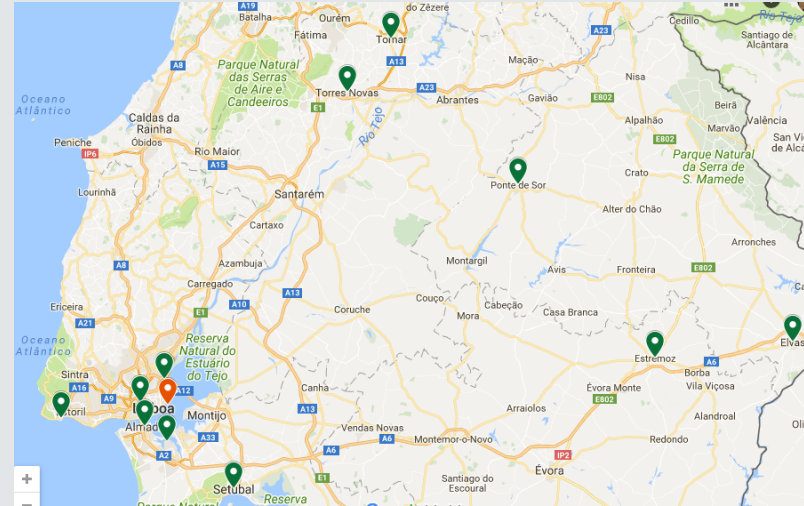


Dispatch center (CODU)

CODU

Emergency Vehicles (EVs)

- Lisboa
- Almada
- Cascais
- Elvas
- Estremoz
- Ponte de Sor
- Sacavém
- Setúbal
- Seixal
- Tomar
- Torres Novas



# Results: case study at INEM

	CODU		EVs					
Task type	CODU Shift Responsible	CODU Task	AEM Driver	AEM Team Responsible	SIV Task	TIP Task	UMIPE Task	MEM Task
Task duration	8 hours							12 hours

- Overall dataset
  - 289 people
  - 22 teams (5 CODU, 17 EVs)
  - 61 tasks (10 CODU, 51 EVs)
  - 28 days (3 shifts per day)

# Results: test sets

- Start from INEM base case (289 people, 61 tasks)
  - Change certain parameter
  - Adjust demands and other requirements
    1. TestMD: 56 days instead of 28
    2. TestMP: 417 people instead of 289
    3. TestLS: 103 tasks instead of 61 (to reduce symmetry)
    4. TestHS: every person can do every task (to maximize symmetry)
- + 15 randomly generated instances

Instance	LP relaxation		CPLEX IP		Diving B		VNDS		
	Time (s)	Opt.	Obj.	Gap	Obj.	Gap	Avg. obj.	SD obj.	Avg. gap
INEM	180	26,442	1,994,304	98.67	76,556	65.46	27,024	42	2.15
INEM MD	572	50,506	2,255,012	97.76	367,952	86.27	51,488	55	1.91
INEM MP	272	39,218	2,821,106	98.53	100,314	60.91	40,797	16	3.87
INEM LS	372	41,386	3,698,754	98.94	191,966	78.44	42,452	136	2.51
INEM HS	304	25,128	441,896	94.31	100,720	75.05	26,062	351	3.58
Test01	505	19,892	253,624	92.16	58,638	66.08	20,389	20	2.44
Test02	1557	61,949	737,112	91.60	298,520	79.25	74,090	6	16.39
Test03	638	47,486	417,742	88.63	214,088	77.82	53,113	191	10.59
Test04	1227	55,322	557,332	90.07	307,248	81.99	61,845	7	10.55
Test05	829	26,711	251,460	89.38	81,446	67.20	27,465	31	2.74
Test06	375	15,140	3,887,199	99.61	32,247	53.05	15,542	4	2.59
Test07	1596	18,337	1,251,654	98.53	125,758	85.42	18,799	115	2.46
Test08	251	10,386	89,688	88.42	41,824	75.17	10,462	0	0.73
Test09	1852	38,398	1,251,960	96.93	121,378	68.36	41,258	21	6.93
Test10	2573	25,120	5,157,815	99.51	110,027	77.17	25,604	195	1.89
Test11	3254	22,065	1,454,516	98.48	144,192	84.70	23,269	20	5.18
Test12	689	15,742	4,022,071	99.61	45,007	65.02	16,501	23	4.60
Test13	563	6452	883,480	99.27	68,164	90.53	6944	66	7.09
Test14	626	14,193	4,580,454	99.69	52,846	73.14	14,625	4	2.95
Test15	659	10,944	2,581,719	99.58	42,821	74.44	11,408	7	4.07

Good performance  
of the VNDS  
heuristic

- Gaps of 0.7 to 16.4 percent w.r.t. LP lowerbound
- Within 1 hour CPU



# Further required developments...

- Extend the scheduling tool
  - Holidays
  - Staff preferences
  - Requests for specific days-off and days-on
  - Uncertainty on unexpected absences
  - Changes between staff members
- Rescheduling in the course of the planning period

# Agenda



OpLog brief introduction



Emergency medical services (EMSs) and the Portuguese EMS provider (INEM)



Ambulance dispatching and relocation



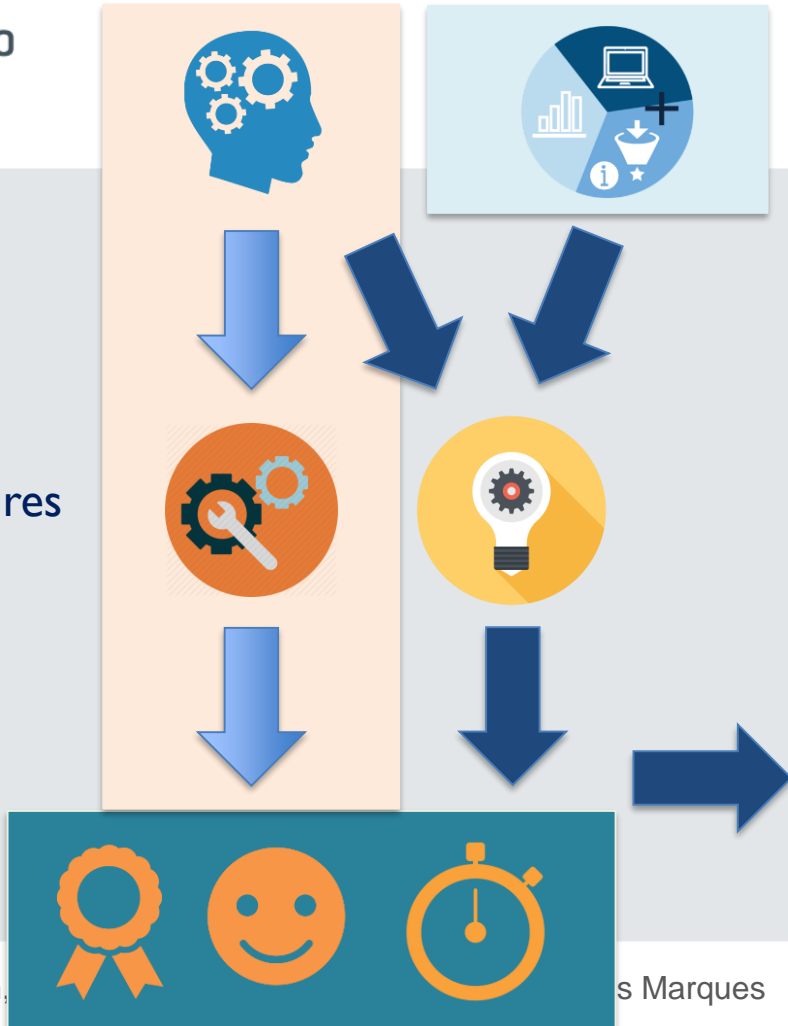
Integrated staff scheduling



Final remarks

# Conclusions and Future research

+  
Real life features  
Uncertainty



**Data2Help: Data Science for Optimization of EMSs**

Improve decision making – more effective, coordinated and timely health delivery

Forecast the expected demand of vehicles and workload

Optimization, Simulation and AI

Dispatching and relocation

Staff scheduling

Vehicle location

Data Science and Simulation

INEM





Integrated staff  
scheduling

Vermuyten H, Rosa JN, Marques I, Beliën J, Barbosa-Póvoa A (2018). Integrated staff scheduling at a medical emergency service: an optimization approach. Expert Systems with Applications 112: 62-76.

**KU LEUVEN**

Ambulance  
dispatching and  
relocation

Carvalho AS, Captivo ME, Marques I. Integrating the ambulance dispatching and relocation problems to maximize system's preparedness. Submitted.

**F C** **Ciências**  
**ULisboa**

Ambulance  
location

Project 2019-2021 - DSAIPA/AI/0044/2018  
Data2Help: Data Science for Optimization of Emergency Medical Services  
Goal: To provide INEM with new tools to improve operational results by optimizing resource assignment, resulting in an improved and faster response to medical emergencies in Portugal.

 **inescid**  
**lisboa**

# Research teams and partners



**Currently:**  
+ 2 MSc students

**For the next 3 years:**  
+ 1 PostDoc  
+ 2 PhD students  
+ 5 MSc students

**KU LEUVEN**



**Ciências  
ULisboa**



- **ImproveOR project**

- i. Developing comprehensive and innovative methods to improve operating room responsiveness to increasing surgical demand and to better coordinate surgical capacity and demand. Decision support tools are to be developed combining optimization approaches (based on multi-objective mathematical programming models, heuristics and simulation) to assist resource capacity planning decisions in the operating room, with structured participatory approaches to capture stakeholders' views and preferences regarding the surgical patient flows and the planning and scheduling of surgeries. The developed methods will be tested and validated in two central hospitals of the Portuguese National Health Service.
- ii. Support in the management of the ImproveOR: Building Decision Support Tools for Improved Operating Room Management project and in the scientific supervision of masters and doctoral students associated with the different research tasks to be developed in the project.

**Monthly remuneration:**

2,128.34€

**Duration:**

3 years

**Candidates with degree abroad:**

Recognition of the degree by the Portuguese Directorate-General for Higher Education

**To appear through:**

ORAHS, APDIO, other OR societies



# **Better decisions for more effective emergency medical care**

**The case of the Portuguese Emergency Medical Service**

**Inês Marques**

Centre for Management Studies, Instituto Superior Técnico

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