



UNIVERSITÀ DEGLI STUDI DI NAPOLI
FEDERICO II

itee_{PhD}
information technology
electrical engineering



DIE
TI

UNI
NA

PhD Student: Fabrizio Tavano

**DEVELOPMENT OF AN EFFICIENT STRATEGY TO LEAD
A TEAM OF ROBOT
WORKING FOR THE SANIFICATION OF THE RAILWAY
STATIONS**

Tutor: Prof. Vincenzo Lippiello

Co-Tutor: Dott. Riccardo Caccavale

Cycle: 35 - Academic year: 2020-21 - PhD Year: Second

My background

- PhD student: Fabrizio Tavano
- DR number: DR993890
- Date of birth: 29/08/1981
- Master Science degree: Electronic Engineering University: Second University of Naples
- Doctoral Cycle: XXXV
- Scholarship type: no scholarship
- Tutor: Prof. Lippiello Vincenzo
- Co-Tutor: Dott. Riccardo Caccavale

Research field of interest

- The field of interest is the decentralized control of autonomous robots collaborating together to solve the same task.
- In order to find the better solution, the following arguments are studied:
 - Reinforcement learning: Deep Q-Networks
 - Model based control technics: Model Predictive Control

Summary of study activities

Title of course (Total number: 12)	Professor\University
SIDRA 2021 PhD Summer School (30 hours), titles: “ Game Theory and Network Systems”, “Modeling and Control of Soft Robotics ”	University of Bologna
MSc course, title: Intelligent robotics	Prof. Alberto Finzi
MSc course, title: Image and Video Processing for Autonomus Driving	Prof. Luisa Verdoliva
MSc course, title: Image Processing for Computer Vision	Prof. Giuseppe Scarpa
MSc course, title: Neural Networks and Deep Learning	Prof. Giuseppe Prevete
MSc course, title: Text Mining	Prof.ssa Flora Amato
MSc course, title: Natural language Processing	Prof. Francesco Cutugno
Module: Statistical Learning	Prof.ssa Anna Corazza
MSc course, title: human-robot interaction	Prof.ssa Silvia Rossi
MSc course, title: Fondamenti di Robotica	Prof. Bruno Siciliano
Data Management	Prof.ssa Flora Amato
Ad hoc course, title: deep learning and computer vision for autonomous systems : focus on drone vision, imaging surveillance and cinematography	Prof. Ioannis Pitas, Aristotle University of Thessaloniki, CELL Center for education and lifelong learning

Some significant titles of followed seminars in this year (Total number: 33)
Seminar title: Artificial intelligence and 5G combined with holographic technology: a new perspective for remote health monitoring, lecturer: Dr. Pietro Ferraro, Dr. Pasquale Memmolo
Optimized Graph Representations for Right-Wing Reddit Community Using Graph Neural Networks , lecturer: Mr Mohammad Diaoulé Diallo, University of Bielefeld
Introduction to Legged robots and examples of IIT’s Dynamic Legged Systems Lab, lecturer: Dr. Claudio Semini, Dr. Michele Focchi,
Introduction to underwater robotics , lecturer: Dr. Claudio Semini, Prof. Gianluca Antonelli
GDPR basics for computer scientists , lecturer: Dr. Rigo Wenning,
Exploiting medical imaging in the era of big data, lecturer: Dr Marco Aiello
Exploiting Deep Learning and Probabilistic Modeling for Behavior Analytics, lecturer: Prof. Giuseppe Manco
Seminar, title: The coming revolution of Data driven Discovery (a fourth Methodological Paradigm of Science), lecturer: Prof. Longo
Seminar title: Sadas Engine, an innovative DBMS for the data warehouse , great performance in the VLDB environment, lecturer: Eng. Luca De Rosa
Approaches to Graph Machine Learning ; Lecturer: Miroslav Cepek–Oracle Labs

Study and training activities-credits earned

	Courses	Seminars	Research	Tutorship	Total
Bimonth 1	22.5	2.4	10	0	34,9
Bimonth 2	6	2	10	0	18
Bimonth 3	0	4.5	10	1 (25 hours)	15,5
Bimonth 4	12	2.8	8	0.6 (15 hours)	23,4
Bimonth 5	30	0	8	0	38
Bimonth 6	0	1	9	0	10
Total	70,5	12,7	55	1,6	139,8
Expected (in 3 years)	30 - 70	10 - 30	80 - 140	0 - 4.8	

Study and training activities-credits earned

Year	Courses	Seminars	Research	Tutorship	TOTAL
	(credits)	(credits)	(credits)	(credits)	
I	44,5 (min 20 - max 40)	11 (min 5 - max 10)	57 (min 10 - max 35)	0 (min 0 - max 1.6)	112,5
II	70,5 (min 10 - max 20)	12,7 (min 5 - max 10)	55 (min 30 - max 45)	1,6 (min 0 - max 1.6)	139,8
TOTAL (ranges ref. in 3 years)	115 (min 30 - max 70)	23,7 (min 10 - max 30)	112 (min 80 - max 140)	1,6 (min 0 - max 4.8)	252,3

Courses for next year

Course	Professor
MSc course, title: Control System Design	Prof. Garone (ULB, Brussels, Belgium)
MSc course, title: Optimization-based Control	Prof. Garone (ULB, Brussels, Belgium)
SIDRA 2022 PhD Summer School	University of Bologna
MSc course, title: Data Mining	Prof. Longo (University of Naples Federico II)

Tutorship

In the period between 1.03.2021- 30.06.2021, I have actively participated to the correction of on-line exercitations and home-works for the training of the students for the following MSc courses:

- Robotic Labs: 20 hours
- Theory of Systems : 20 hours

For a total of 1.6 credits

A Multi-robot Deep Q-Learning Framework for Priority-based Sanitization of Railway Stations

Our needs

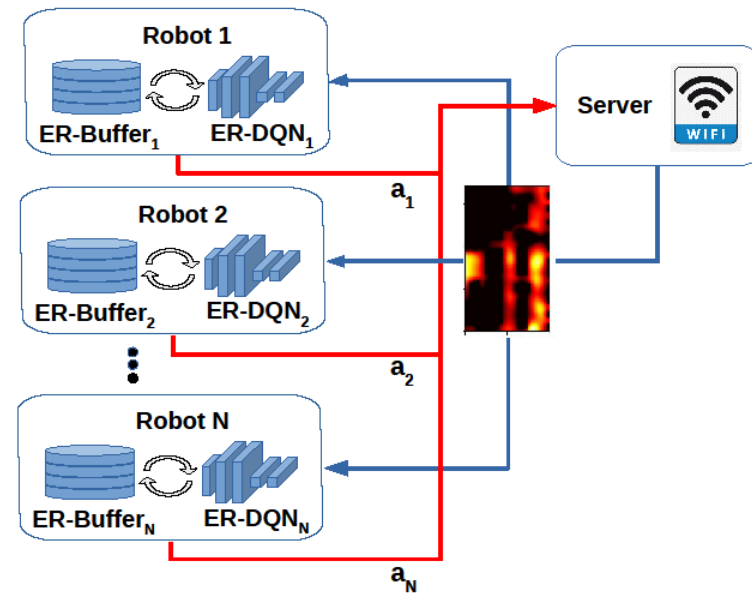
- The role of railway stations in large cities is evolving: the number of services they offer is growing more and more. The stations are no longer simple nodes of the railway network, for the access to the trains as passengers, facilitating the intermodal road-rail exchange. The stations are now location of many other services, they have shops and catering activities, places for recreation and aggregation, environments suitable to organize public events. They are therefore strategic elements for the life of modern cities, which are increasingly popular. Modern society is increasingly open and connected, and the demand for transport is ever increasing. In this context, the railway sector is essential to respond to the new mobility needs.
- On the other hand, the probability of spreading diseases due to the presence of microorganisms (bacteria and viruses) between places of great population density, apparently distant between them, but indeed very connected by modern transport systems, is increasing more and more.
- Furthermore, it is fundamental and strategic that the Railway Transport Company and the Infrastructure Manager Company as Rete Ferroviaria Italiana, are equipped with adequate and modern tools to prevent future diseases from finding opportunities in the stations and on board of trains to spread and reach cities, even distant cities, destinations of the public transport service.



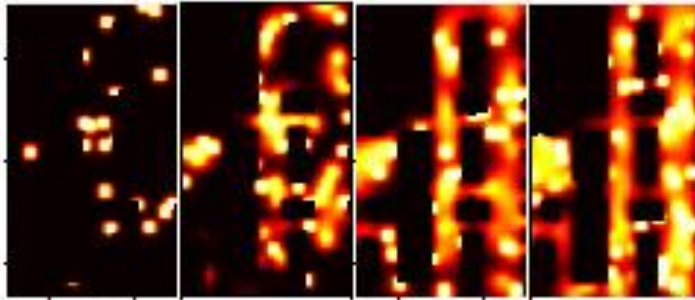
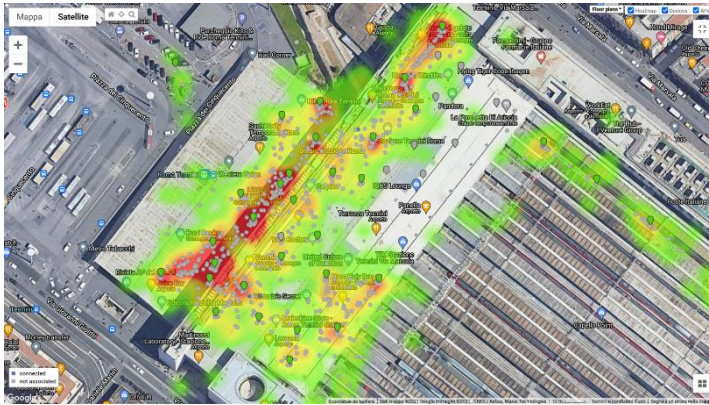
Our Architecture

our solution for the sanitizing of the station is a distributed approach in which a server shares information as an heatmap, and the robots makes autonomous decisions

we have used a neural network for every robot, every robot has the vision of all the environment thanks to an heatmap shared by a server, and in the heatmap there is an indication of the common targets, represented by the peaks of the gaussian functions The state, that represents the input of the CNN is a 2 channel matrix $100 \times 172 \times 2$ where each element has values in $[0,1]$; in the first channel there is a real number representing the risk of the location; each element of the second channel is comprised in $\{0,1\}$, in particular 1 for the robot position, and 0 otherwise.

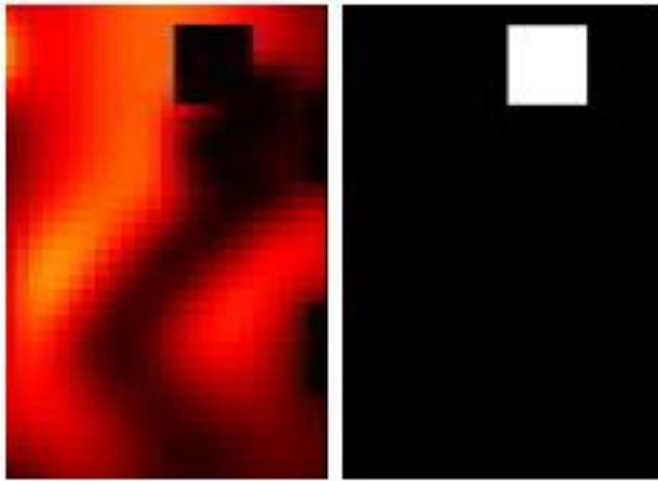


The use of access points in the stations



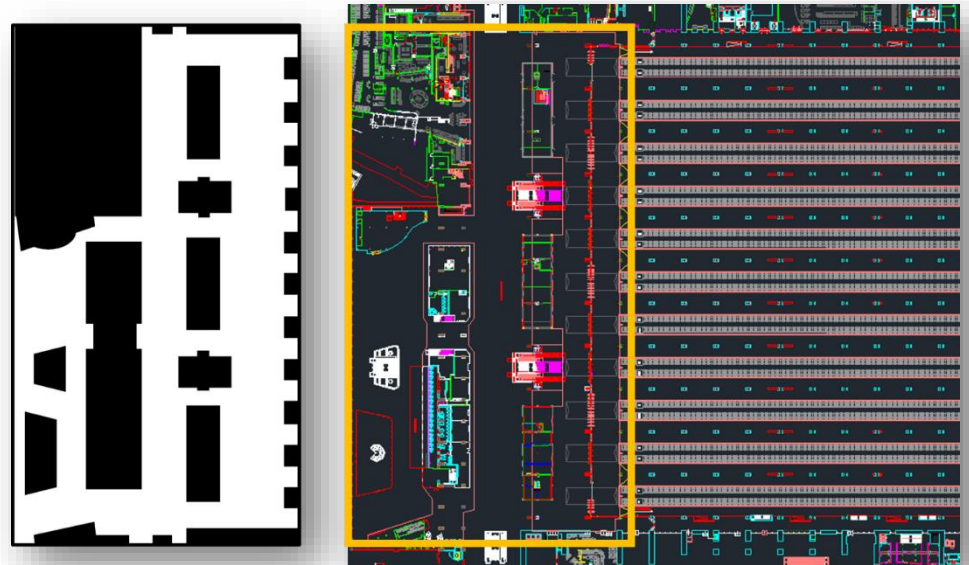
- we will use a server system that will be able to recognize the aggregations of passengers.
- localization of the position of the aggregations is possible by identifying the positions of smartphones thanks to the GPS and the triangulation technique applied to WiFi signal between more than one access point present in the station.
- The team of robots will be driven in an appropriate manner to sanitize the environment in continuous manner during the day.
- The centralized strategy of cooperation will be chosen and optimized thanks to Model predictive control methods and Deep Q-learning methods

The Environment



on the left: the planimetry of the station, and on the right its conversion in an black and white image, preserving the same scale and proportions

Figure at the right: example of the 2-channels matrix representing the priority distribution (left) and position and size of the cleaning range of a single agent (right)



Server-side learning algorithm

On the left the algorithm of the main process (the server) during the training phase. It is described how the main process serves to every robot. at line 5 we set in the heatmap at random positions, a random number of values "1"; the number is approximately 200, the values "1" correspond at the point in which there is the maximum risk of contagion. At line 4 the server receives the starting position of every robot in the heatmap; at line 5 the main process update the heatmap, considering the cleaning action of every robot stating from the knowledge of their positions; in the row 17, the server applies the gaussian filter to the heatmap; in the row 7 the main process sends the updated environment to the robots; the main process at the row 10 verifies if the target in terms of cleaning of heatmap is accomplished

Algorithm 2 Server-side learning algorithm

```
1: procedure SERVER( $k, M$ )
2:   while true do
3:      $done = false$ 
4:      $(x_1, \dots, x_k) = receive\_positions()$ 
5:      $s = generate\_random\_state(M)$ 
6:     for  $stp < max\_step$  do
7:        $send\_heatmap(s)$ 
8:        $(x'_1, \dots, x'_k) = receive\_positions()$ 
9:        $s = update\_state(s, (x'_1, \dots, x'_k))$ 
10:      if  $accomplished(s)$  then
11:         $done = true$ 
12:      end if
13:       $send\_accomplishment(done)$ 
14:      if  $done$  then
15:        break
16:      end if
17:       $s = apply\_filter(s, \mathcal{N}(\mu, \sigma^2), M)$ 
18:    end for
19:  end while
20: end procedure
```


Agent-side learning algorithm

On the right we show the algorithm used for the robots in the training phase. At the beginning in row 4 we do the reset of the initial conditions and we define an initial position for every robot; in the row 5 the robot sends its position to the server, our main-process ; in the row 7 the robot receives the updated environment from the server; in row 8 the robot does a change of position thanks to its neural network ;in row 9 it updates its own actual position in the second channel of the state, and it applies the disinfection on its copy of environment; in row 12, it calculates its own reward and it verifies if the heatmap is cleaned over a fixed threshold

Algorithm 1 Agent-side learning algorithm.

```
1: procedure AGENT( $i, M$ )
2:   while true do
3:      $done = false$ 
4:      $x_i = random\_position()$ 
5:      $send\_position(x_i)$ 
6:     for  $stp < max\_step$  do
7:        $s = receive\_heatmap()$ 
8:        $a_i = select\_action(s, x_i, \epsilon)$ 
9:        $(s'_i, x'_i, r_i) = emulate\_action(s, x_i, a_i, M)$ 
10:       $send\_position(x'_i)$ 
11:       $done = receive\_accomplishment()$ 
12:       $experience\_replay(s, x_i, a_i, r_i, s'_i, x'_i, done)$ 
13:      if  $done$  then
14:        break
15:      end if
16:    end for
17:  end while
18: end procedure
```

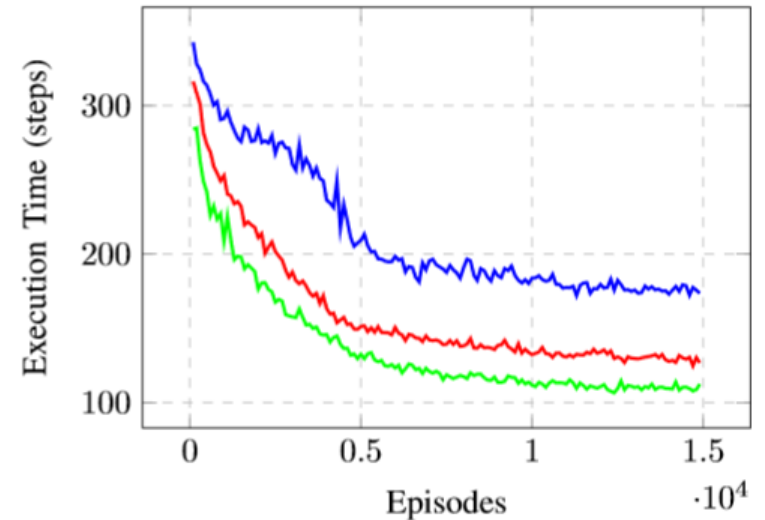
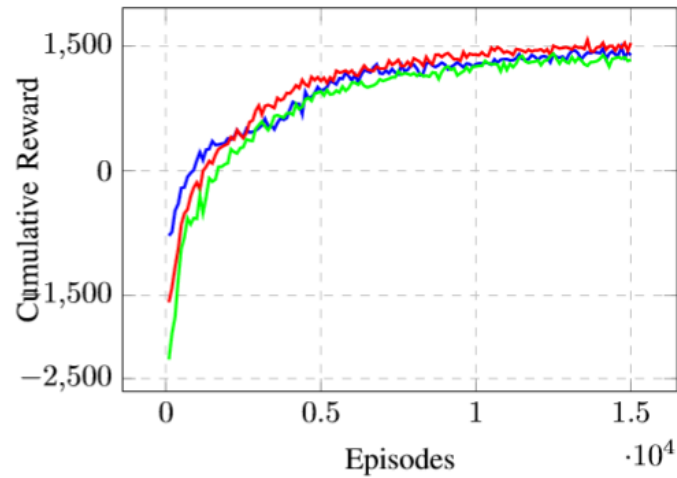
Server-side testing algorithm

On the right side, we briefly describe the algorithm of testing of this solution; the main difference between the case studied in the training, here we change every 15 steps the position of all the clusters of the peoples, positioning them in another random state. Also the number of cluster may vary during the time

Algorithm 3 Server-side testing algorithm

```
1: 30PercMax=MaxClust*30/100
2: MinValue=MaxClust-30PercMax
3: while Run until solved do
4:   NumberClusters=random(MinValue, MaxClust)
5:   server.createClusters(NumberClusters,Environment)
6:   robots.startProcess()
7:   server.receiveNewUpdateFromRobots(Positions)
8:   for timestep in range max episode do
9:     if (timestep is multiple of 15): then
10:       Delete.Clusters[];
11:       NumClusters=random(MinValue;MaxClust)
12:       server.Clusters(NumClusters,Environment)
13:     end if
14:     Heatmap=server.GaussianFilter(Environment)
15:     server.sendEnvironment.toRobots(Heatmap)
16:     server.verifyGoalSatisfied()
17:   end for
18: end while
```

Scalability of the solution



Future studies

- As future research activities, we plan to extend our pilot study by testing the proposed framework in a more realistic scenario, considering more complex robotic models and daily recorded data about people distribution in the station from the WiFi infrastructure and also to design a scaled environment with real robots.
- Furthermore, multi-agent strategies including teams of heterogeneous robots with different cleaning capabilities are currently under investigation. A robot in our experiment, doesn't know the position of the other robots that collaborate with it but their paths. It is also interesting to develop a solution in which robots have different approaches and different instruments for the sanitizing for the environment. Our solution implemented with a dedicated neural network for each robot is studied for this scope.
- It will be compared also the decentralized approach of a Buffer replay DQN, where every robot has its own neural network, with the centralized one, with one network that select the actions to do for every robot.
- A centralized strategy of cooperation will be chosen and optimized thanks to Model predictive control methods in alternative to Q-learning methods. In this manner, it will be possible to do a comparison between the results obtained by the adoption of model-based methods than model-free reinforcement learning algorithm methods.

Products

[P1]

Riccardo Caccavale, Vincenzo Calà, Mirko Ermini, Alberto Finzi, Vincenzo Lippiello and Fabrizio Tavano: title: “A Multi-robot Deep Q-Learning Framework for Priority-based Sanitization of Railway Stations”; AIRO 2021: 8th Italian Workshop on Artificial Intelligence and Robotics of the 20th International Conference of the Italian Association for Artificial Intelligence (AI*IA 2021),online, December 1st-3rd, 2021



UNIVERSITÀ DEGLI STUDI DI NAPOLI
FEDERICO II



A Multi-robot centralized Framework for Priority-based Sanitization of Railway Stations using model predictive control method

Periods abroad and/or in international research institutions: 15.09.2021 - 15.12.2021

Université Libre de Bruxelles, Département : Service d'Automatique et d'Analyse des Systèmes: Prof. Emanuele Garone

Università degli studi di Napoli "Federico II", Dip. DIETI: Prof. Vincenzo Lippiello, Dott Riccardo Caccavale, Fabrizio Tavano

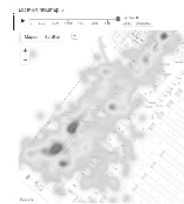


UNIVERSITÀ DEGLI STUDI DI NAPOLI
FEDERICO II

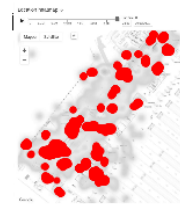
ULB UNIVERSITÉ
 LIBRE
 DE BRUXELLES

The extraction of data

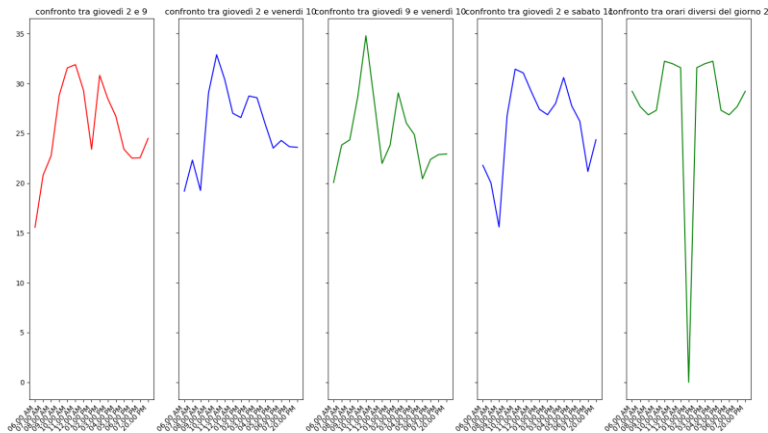
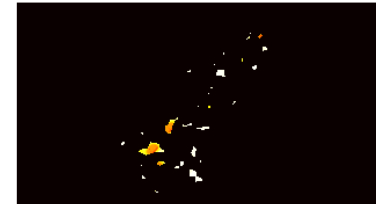
Original



Filter to extract peaks



Extracted peaks



	2sett 06:00 AM	2sett 07:00 AM	2sett 08:00 AM	9sett 06:00 AM	9sett 07:00 AM	9sett 08:00 AM	11sett 06:00 AM	11sett 07:00 AM	11sett 08:00 AM
11sett 08:00 AM	10.527	17.631	9.28	10.527	17.631	9.28	10.527	10.188	0.0
11sett 07:00 AM	4.988	14.649	12.692	4.988	14.649	12.692	4.988	0.0	10.188
11sett 06:00 AM	0.0	14.828	13.145	0.0	14.828	13.145	0.0	4.988	10.527
9sett 08:00 AM	13.145	19.119	0.0	13.145	19.119	0.0	13.145	12.692	9.28
9sett 07:00 AM	14.828	0.0	19.119	14.828	0.0	19.119	14.828	14.649	17.631
9sett 06:00 AM	0.0	14.828	13.145	0.0	14.828	13.145	0.0	4.988	10.527
2sett 08:00 AM	13.145	19.119	0.0	13.145	19.119	0.0	13.145	12.692	9.28
2sett 07:00 AM	14.828	0.0	19.119	14.828	0.0	19.119	14.828	14.649	17.631
2sett 06:00 AM	0.0	14.828	13.145	0.0	14.828	13.145	0.0	4.988	10.527



The Model

$$\hat{x}_z(t + k + 1|t) = \alpha \hat{x}_z(t + k|t) + \sum_z k \hat{h}_z(t + k|t) - \hat{u}_z(t + k|t)$$

$$\hat{y}_z(t + k + 1|t) \in \{0,1\}$$

$$\sum_z^{N_z} \hat{y}_{i,z}(t + k|t) = 1$$

$$\hat{h}_z(t + k + 1|t) = \begin{cases} h_z(t) & k = 0 \\ \hat{h}_z(t + k) & k > 0 \end{cases}$$

$$\hat{y}_z(t + k|t) \leq \sum_z^{N_z} \hat{y}_{i,z}(t + k - 1|t)$$

$$i = 1 \dots, N_R$$

$$z = 1 \dots, N_z$$

$$0 \leq \hat{u}_z(t + k|t) \leq \alpha \hat{x}_z(t + k|t)$$

$$k = 1 \dots, N - 1$$

$$\hat{u}_z(t + k|t) \leq M \sum_z^{N_z} \hat{y}_{i,z}(t + k|t) \quad \text{Mixed Integer Linear Programming (MILP)}$$