

#### A Decision Support Assistant to Operate a Power Grids with Zonal Automatons

Eva Boguslawski, Alessandro Leite, Marc Schoenauer, Matthieu Dussartre, Benjamin Donnot

#### ▶ To cite this version:

Eva Boguslawski, Alessandro Leite, Marc Schoenauer, Matthieu Dussartre, Benjamin Donnot. A Decision Support Assistant to Operate a Power Grids with Zonal Automatons. DTU PES Summer School 2023. Future Energy Systems: Advances in OR and AI., Jun 2023, Lingby, Denmark. . hal-04396136

#### HAL Id: hal-04396136 https://hal.science/hal-04396136

Submitted on 15 Jan 2024

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



Distributed under a Creative Commons Attribution - NoDerivatives 4.0 International License

# A Decision Support Assistant to Operate a Power Re Grids with Zonal Automatons

Eva Boguslawski, Alessandro Leite, Marc Schoenauer, Matthieu Dussartre, Benjamin Donnot RTE, TAU, INRIA, Université Paris-Saclay eva.boguslawski@rte-france.com

### Context

Renewable energy sources are forcing power grid operators to review their grid management strategies

Réseau de Transport d'Électricité (RTE) is developing new adaptive zonal automatons



PARIS-SACLA

- Each automaton monitors a zone of the power grid (up to 40) lines) thanks to an optimization algorithm.
- An automaton can act on the power grid configuration and on the electricity production.
- Each automaton can receive a target path or additional information from operators to decide the appropriate actions to be done. Examples of target path include (un)desirable configurations in future hours or information about other zones.

## **Objective**

Design a decision support assistant for the supervision of zonal automatons able to recommend relevant target paths/information for automatons based on the power grid's configuration and constraints



Decentralized multi-agent reinforcement learning

For example, each agent can be associated with a subset of automatons and their zones. In this context, at each time-step t:

- $\rightarrow$  Each **agent** *i* chooses an **action** according to its observation.
- $\rightarrow$  The **actions** are combined and executed
- $\rightarrow$  A **reward** representing the safety of the hole grid is returned.

### How to emulate a zonal automaton? 4.

To obtain **reasonable computation times** during training, we need a trade-off between speed and realism.

Emulate an automaton through a reinforcement learning agent to:  $\rightarrow$  Operate the grid  $\rightarrow$  Follow a target setpoint

Example of a simulated power grid

#### **Considered method** 3.

### **Reinforcement learning (RL)**

RL approaches are particularly efficient for sequential problems because:

- RL is able to **anticipate** future events

- RL can take into account **uncertainties** 



We designed an agent able to operate a 14-nodes power grid by controlling storage power

				Do Nothing Agen	t RL Agent
Average	number	of	survived	206/288	270/288
time steps per episode				200/200	219/200

Performances of our agent on a 14-nodes simulated grid

### We designed an agent able to follow a storage charge setpoint

We got interesting results about:

- how to penalize deviations from the setpoint.

- the impact of hyperparameters.

- the ability to generalize to other shapes of setpoint which is great as illustrated in the figures below.

Normalized storage charge



Observation

The principle of Reinforcement Learning

### Multi-Agent Reinforcement Learning

**Several agents cooperate** to ensure the safety of the power grid. We think a decentralized approach could greatly **improve** scalability compared to traditional deep learning methods [1].

Comparison of observed charge with a smooth setpoint or a step function setpoint

### References

[1] Karl Johan Åström. Optimal control of markov processes with incomplete state information. Journal of mathematical analysis and applications, 10(1):174-205, 1965.