Applying Reinforcement Learning to IFMIF-DONES HVAC optimisation

Antonio Manjavacas manjavacas@ugr.es Manuel A. Vázquez-Barroso manvazbar@ugr.es

Juan Gómez-Romero jgomez@decsai.ugr.es Francisco Martín-Fuertes francisco.martin-fuertes@ciemat.es









About me...

Antonio Manjavacas

- PhD fellow at University of Granada & IFMIF-DONES (2022–2025)
- "Deep Reinforcement Learning for Generative Design of Safety Elements in IFMIF-DONES"
- Member of SAIL (Sustainable Artificial Intelligence Lab)
 - Reinforcement Learning for HVAC control (see <u>Sinergym</u>)
 - Physics-based models for power prediction
 - Grid operation with Reinforcement Learning



What about fusion?





<image>

2020-2025 Main assembly phase

> 2022 Torus completion*

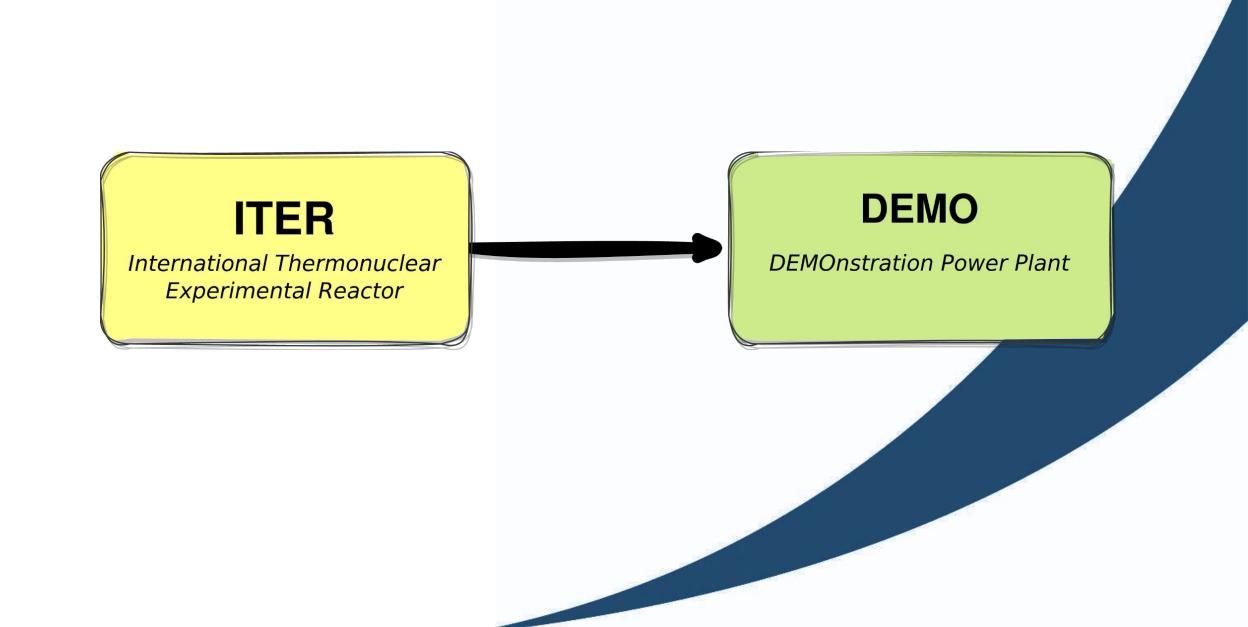
2024 Cryostat closure*

2024-2025 Integrated commissioning phase

> Dec 2025 First Plasma*

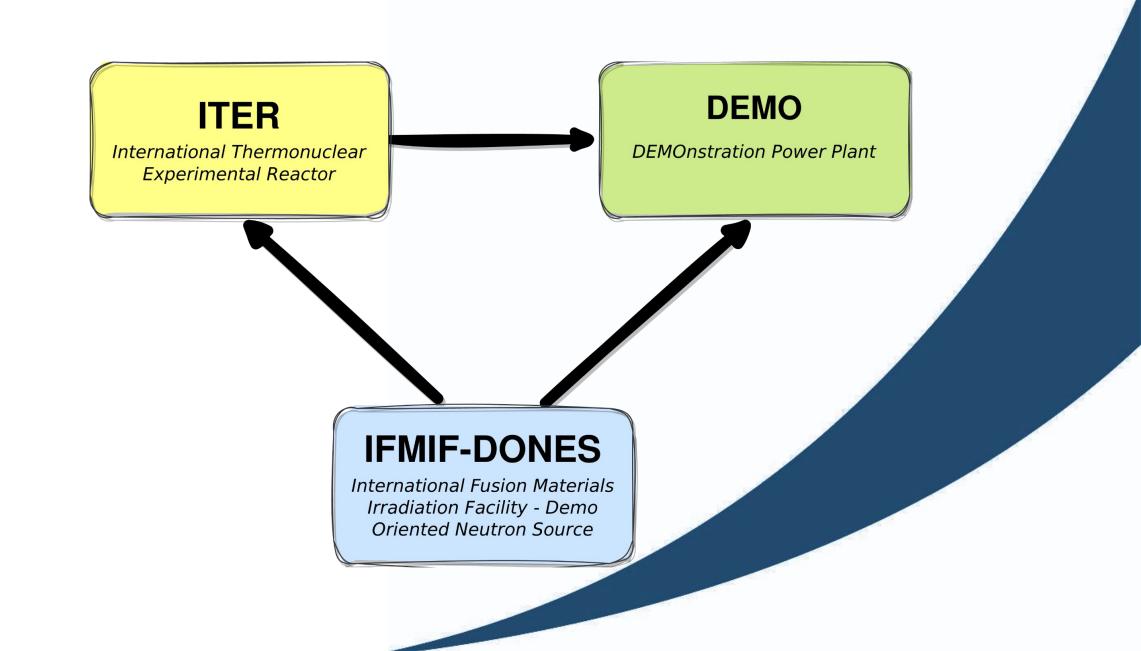
2025-2035 Progressive ramp-up of the machine

2035 Deuterium-Tritium Operation begins



1990s: the materials scientific community promotes the development of a **neutron source** for the qualification of materials to be used in future **fusion reactors (DEMO)**.

• An irradiation facility with material samples being exposed to fusion-like radiation conditions.



IFMIF-DONES

IFMIF-DONES

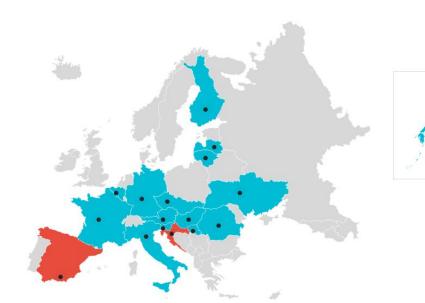
International Fusion Materials Irradiation Facility – DEMO Oriented Neutron Source



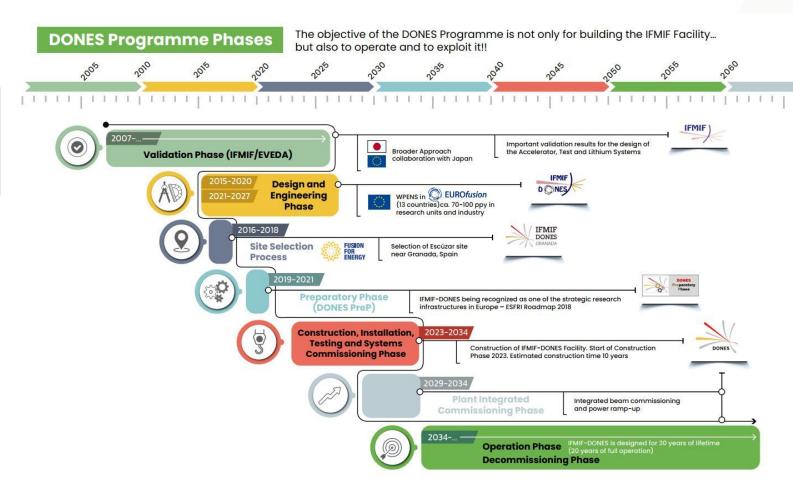


- **Deuteron linear accelerator** with a maximum initial dose of 20 dpa, and up to 50 dpa in its second phase.
- 100 dpa with two 40 MeV and 125 mA beam linear accelerators once the facility is upgraded to **IFMIF**.
- Granada (Spain).

IFMIF-DONES



Partners	Observers		
Spain	Austria	Hungary	Slovakia
Croatia	Belgium	Italy	Slovenia
	Czech Republic	Japan	Ukraine
	Finland	Latvia	Euratom (F4E)
	France	Lithuania	
	Germany	Romania	





BUT

...we have to build them **somewhere**,

BUT

...we have to build them **somewhere**,

...ensuring **safety** conditions,

BUT

...we have to build them **somewhere**,

...ensuring **safety** conditions,

...in an **efficient** way



Potential tritium releases, radiation doses, Li fires...

Potential tritium releases, radiation doses, Li fires...

- **Detritiation** systems
- Inertized rooms → **dynamic confinement**.
- **Purification** and **controlled evacuation** of gases to the environment.

Potential tritium releases, radiation doses, Li fires...

- **Detritiation** systems
- Inertized rooms $\rightarrow \frac{\text{dynamic confinement}}{\text{dynamic confinement}}$.
- **Purification** and **controlled evacuation** of gases to the environment.

Dynamic confinement

ISO 17873

Criteria for the design and operation of ventilation systems for nuclear installations other than nuclear reactors

- **Dynamic confinement** prevents the release of hazardous substances throughout the building, keeping them contained within their respective enclosures.
- It relies on different rooms maintained at different pressures, based on their potential contamination levels.
- Confinement classes: C1, C2, C3, and C4.

Table 3 — Guide to depression values

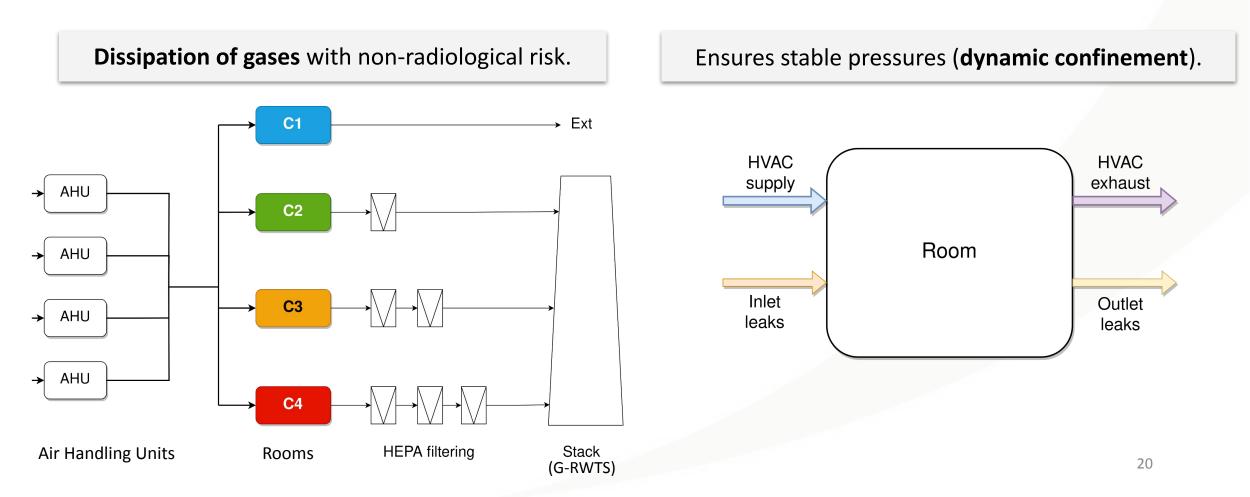
Nature of room or area	Depression value ^a	Containment class
Non-controlled rooms or areas free from contamination	Atmospheric pressure or small overpressure	Unclassified
Supervised areas with low levels of surface or airborne contamination	Less than 60 Pa	C1
C1 should be uncontaminated in normal operations	of all a background of a later becomes a	
Controlled areas with moderate levels of surface or airborne contamination	80 to 100 Pa	C2
Controlled areas with high levels of surface or airborne contamination	120 to 140 Pa	C3
Controlled areas with very high levels of surface or airborne contamination	220 to 300 Pa	C4
Areas which are not accessible except under specific circumstances		
^a Compared to the reference pressure.		



Nuclear HVAC

HVAC

Heating, Ventilation, and Air Conditioning system





Computer code designed to model and simulate **accidents** leading to **radiological** release at nuclear facilities.



A MELCOR model of the IFMIF-DONES main building is under development.



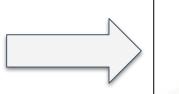
Computer code designed to model and simulate **accidents** leading to **radiological** release at nuclear facilities.



A MELCOR model of the IFMIF-DONES main building is under development.

HVAC design decisions match the design decisions in the simulation model.

Generation / evaluation of design alternatives in the simulated environment



Real system implementation

Reinforcement Learning

- **Deep Reinforcement Learning** has emerged as an outstanding competitor to conventional control methods.
 - HVAC control, particle accelerators, fusion domain.
- MELGYM: a Gymnasium environment for continuous control in MELCOR simulations.



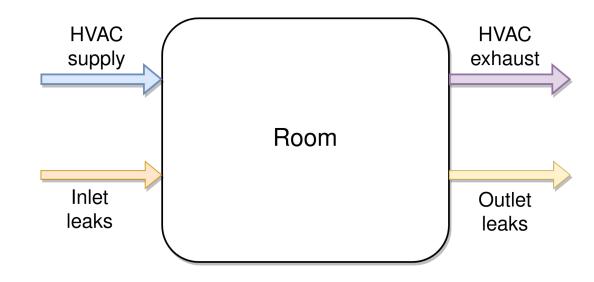
Reinforcement Learning

- **Deep Reinforcement Learning** has emerged as an outstanding competitor to conventional control methods.
 - HVAC control, particle accelerators, fusion domain.
- MELGYM: a Gymnasium environment for continuous control in MELCOR simulations.

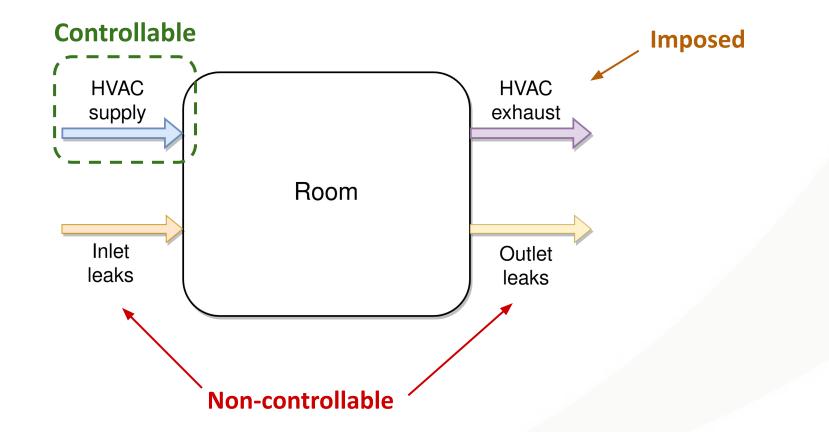


OBJECTIVE: simulate and validate **HVAC** different system design alternatives.

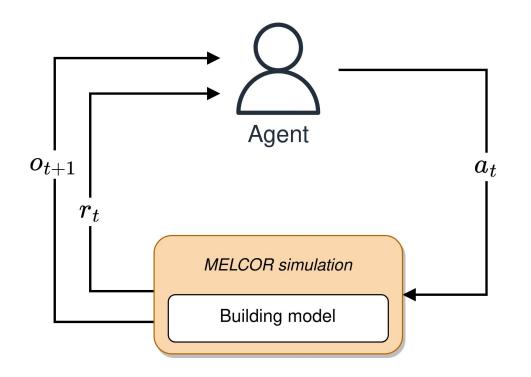
HVAC supply control



HVAC supply control

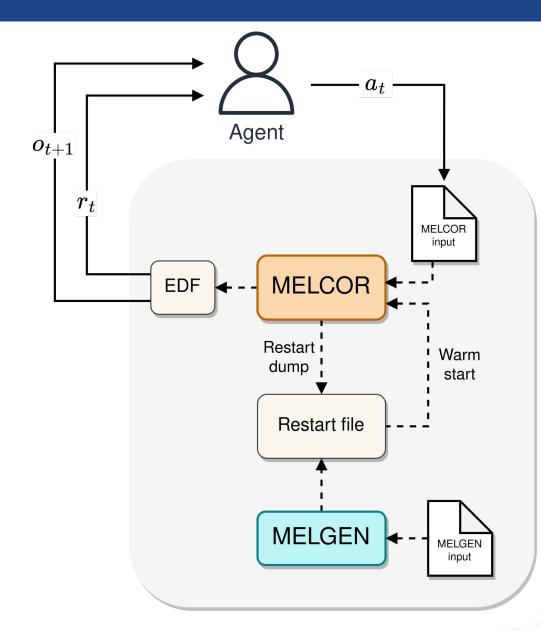


MDP



- **Observations**: current rooms pressures.
- Actions: HVAC air inlet flow rates.
- **Reward**: proximity to target pressure values.
- **Transition**: MELCOR simulation during *T* cycles.

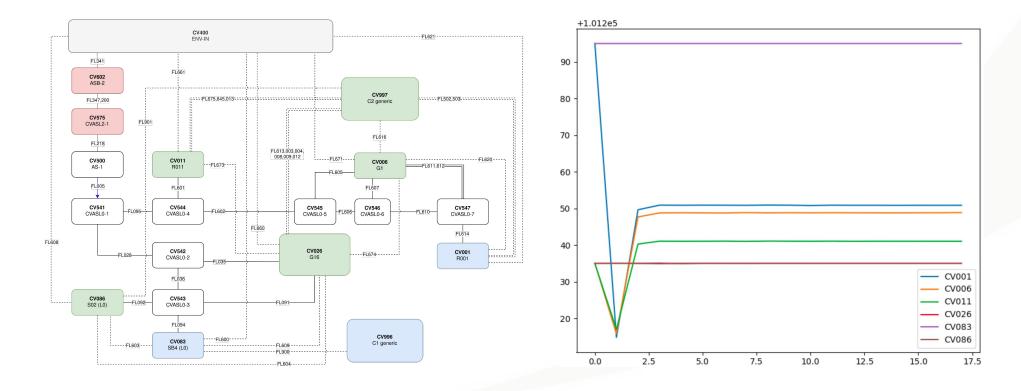
Control loop



- **Observations**: current rooms pressures.
- Actions: HVAC air inlet flow rates.
- **Reward**: proximity to target pressure values.
- **Transition**: MELCOR simulation during *T* cycles.

Preliminary results

- Initial air inlets assessment.
- HVAC power consumption estimates (*in progress*).
- Evaluation of different leak rates in critical rooms.



Applying Reinforcement Learning to IFMIF-DONES HVAC optimisation

Antonio Manjavacas

manjavacas@ugr.es

Manuel A. Vázquez-Barroso manvazbar@ugr.es

Juan Gómez-Romero jgomez@decsai.ugr.es Francisco Martín-Fuertes francisco.martin-fuertes@ciemat.es



