



Machine learning helps control tokamak plasmas

Controlling the magnetic field confining the plasma inside a tokamak is not an easy feat. To create different configurations of the plasma, the magnetic field is adjusted by fine tuning tens of magnetic coils (30 in the case of Korea Superconducting Tokamak Advanced Research (KSTAR), pictured). This adjustment is a well-understood, yet challenging, optimization problem. Writing in *Nature*, Jonas Degraeve and colleagues show that reinforcement learning — a type of machine learning where agents try to find the best action in order to maximize a reward — can

solve this problem and help create a variety of plasma configurations, including ones that hadn't been achieved before.

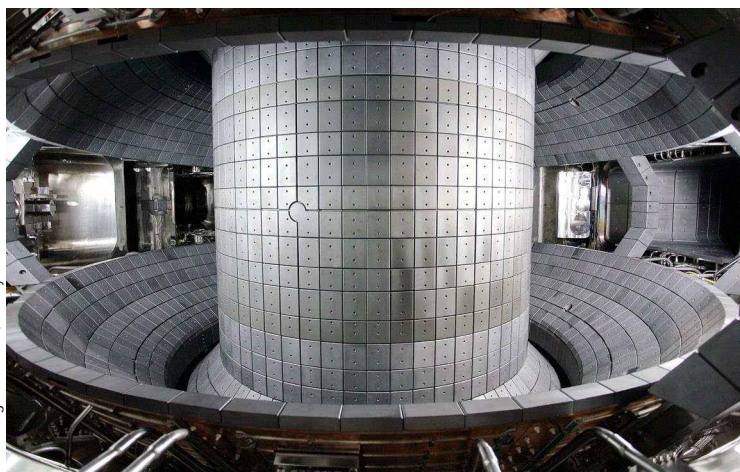
Once an objective for the experiment is given (for example, the type of plasma configuration that is desired), the reinforcement learning algorithm finds the optimal 'control policy' by interacting with a tokamak simulation. This simulation is accurate enough from a physics perspective, while staying relatively computationally cheap. The control policy is then directly implemented in real time on the

tokamak hardware, that of Tokamak à Configuration Variable, the fusion reactor at the École Polytechnique Fédérale de Lausanne. The desired plasma configurations, some simple, some more complex and an unusual one with two separate plasmas ('droplets'), were successfully realized in experiments. These configurations can be realized with traditional control methods, but the reinforcement learning approach is more straightforward. Moreover, this architecture could be swiftly implemented in new tokamaks.

Reinforcement learning has already been explored for the control and optimization of particle accelerators and light sources, so the study from Degraeve et al. adds to the growing machine learning toolbox for experimental physics.

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ORIGINAL ARTICLE Degraeve, J. et al. Magnetic control of tokamak plasmas through deep reinforcement learning. *Nature* <https://doi.org/10.1038/s41586-021-04301-9> (2022)

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