

# ML for nuclear fusion: plasma diagnosis, prediction, and control

Jaemin Seo

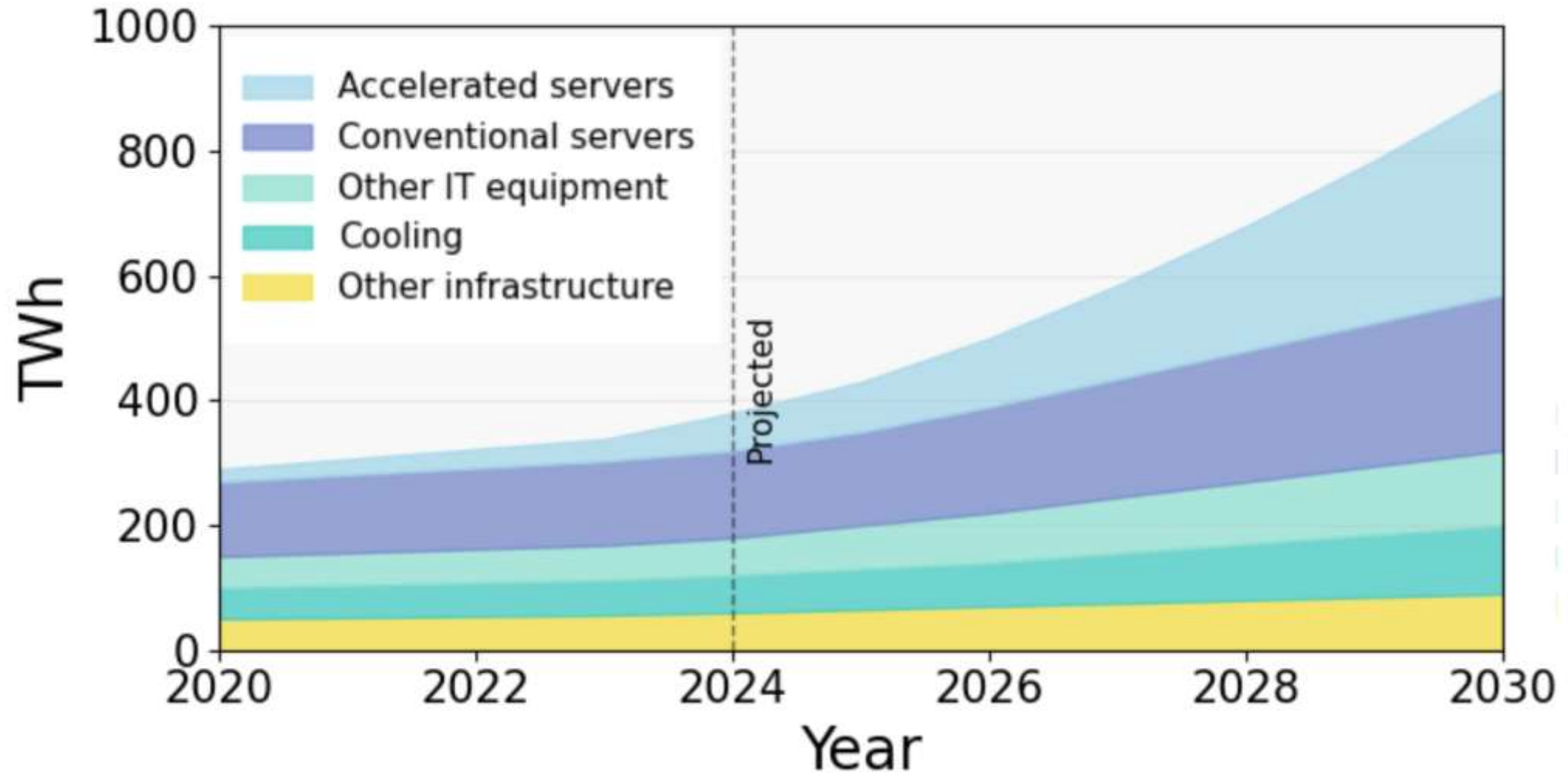
Department of Physics, Chung-Ang University

[jseo@cau.ac.kr](mailto:jseo@cau.ac.kr)

KIAS CAINS 2026 Winter Workshop  
2026-01-10

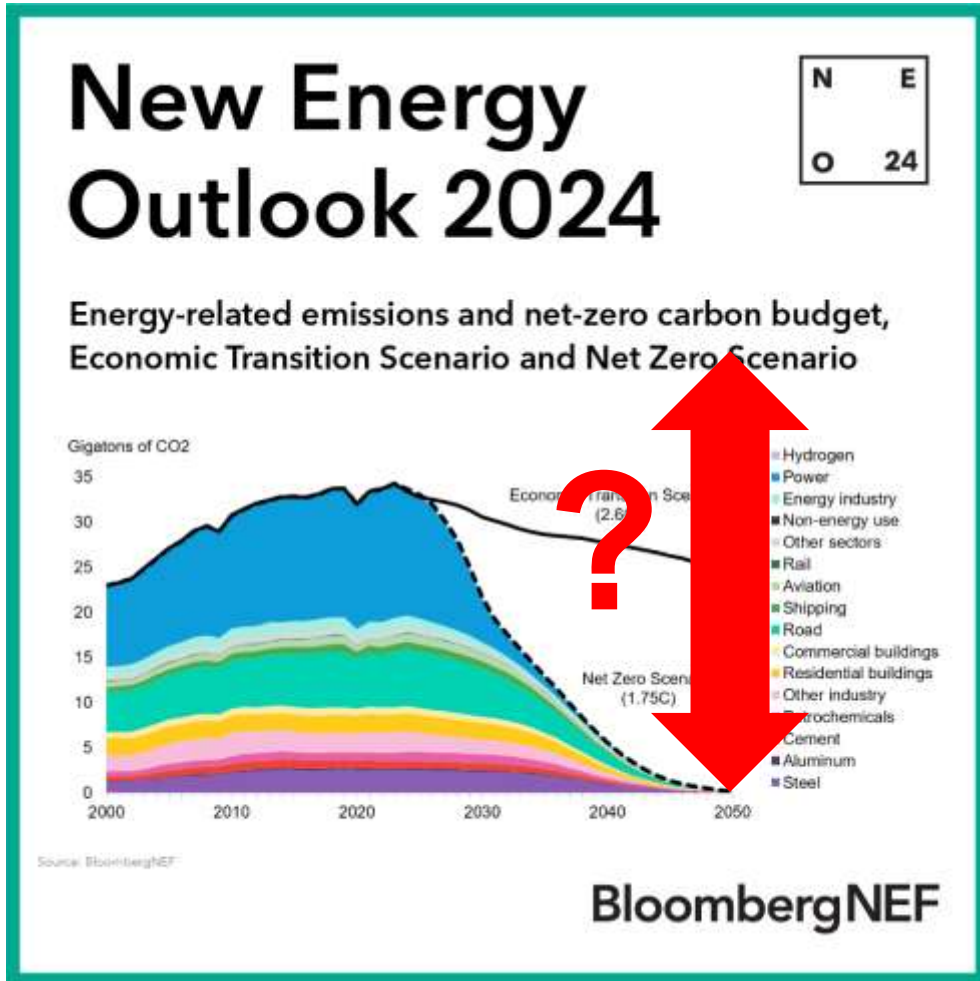


# AI & Energy Consumption



<https://www.iea.org/data-and-statistics/charts/global-data-centre-electricity-consumption-by-equipment-base-case-2020-2030>

# Net Zero 2050



**We cannot use fossil fuels!**

**We need a new energy source:**

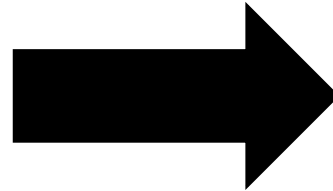
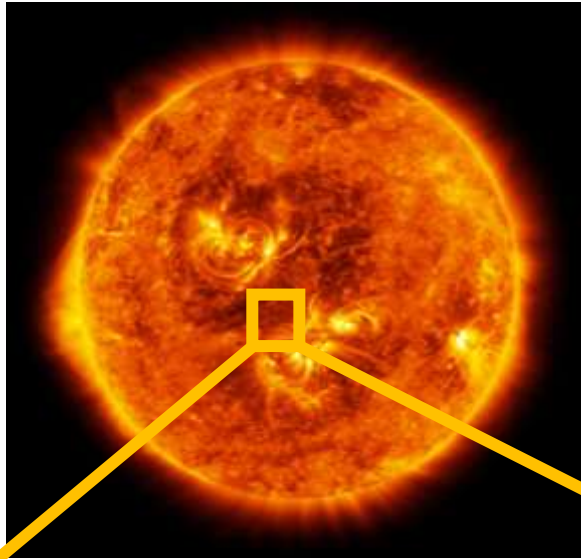


**“Nuclear Fusion” or “Artificial Sun”**

# Contents

- **Introduction: What is nuclear fusion & plasma?**
- **ML for plasma diagnosis**
- **ML for plasma prediction**
- **ML for plasma control**
- **Summary**

# Artificial Sun?



Can we make it on Earth?

# Sun = Plasma = Ionized gas



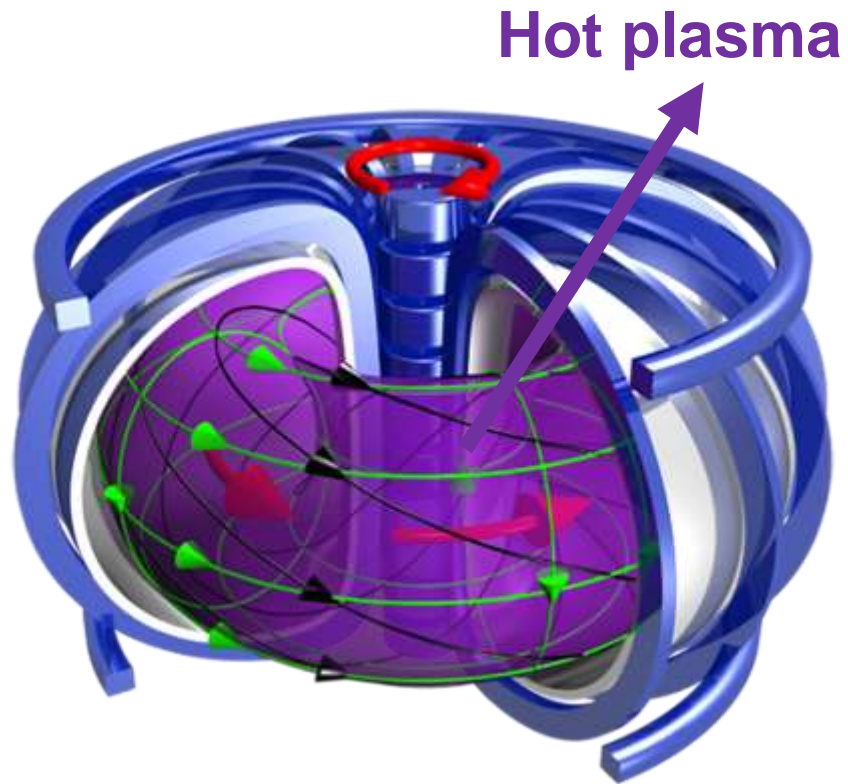
It is an “electrically conducting gas.”

For nuclear fusion,  
>100 million °C needed!

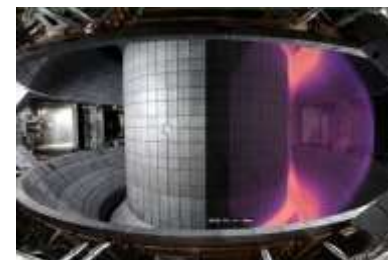
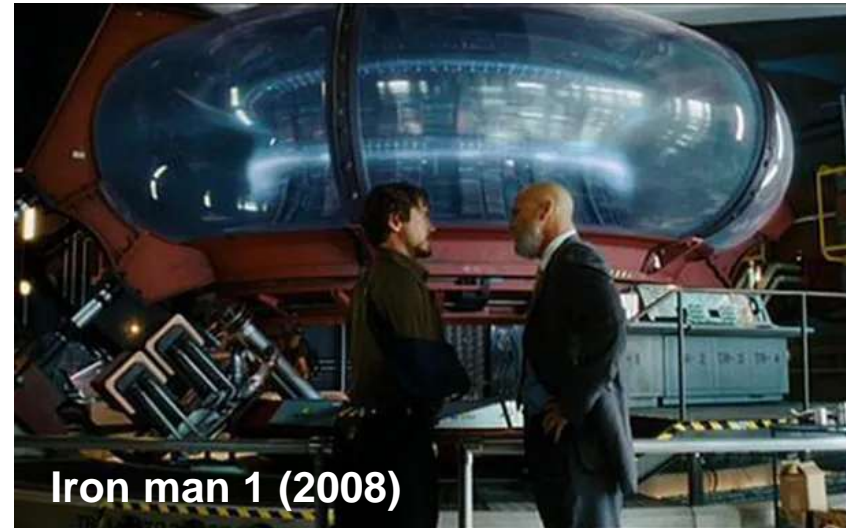


The 4<sup>th</sup> state of matter, plasma

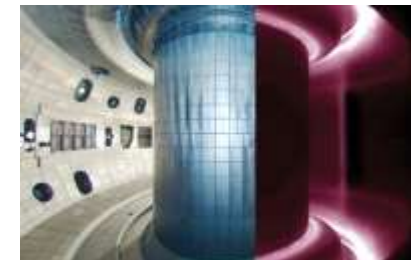
# Tokamak: a "Sun Container"



Confines the plasma with magnetic field!



KSTAR (Korea)



DIII-D (US)



ITER (International)

# For fusion energy,

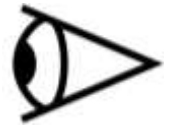
- We need to maintain the high-pressure plasma with stability.
- So we need

**(1) High-temperature**

**(2) Predict the plasma dynamics**

**(3) Control the plasma**

**“Plasma diagnosis”**



**“Plasma prediction”**



**“Plasma control”**



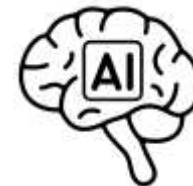
# Let's use ML for nuclear fusion!

- By using ML,

(1) How to diagnose the plasma

(2) How to predict the plasma

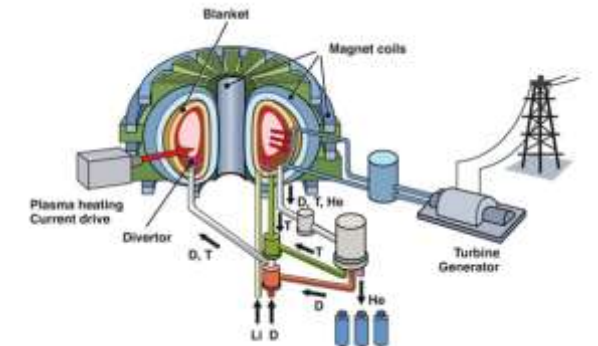
(3) How to control the plasma



Unstable plasma



Stable energy source



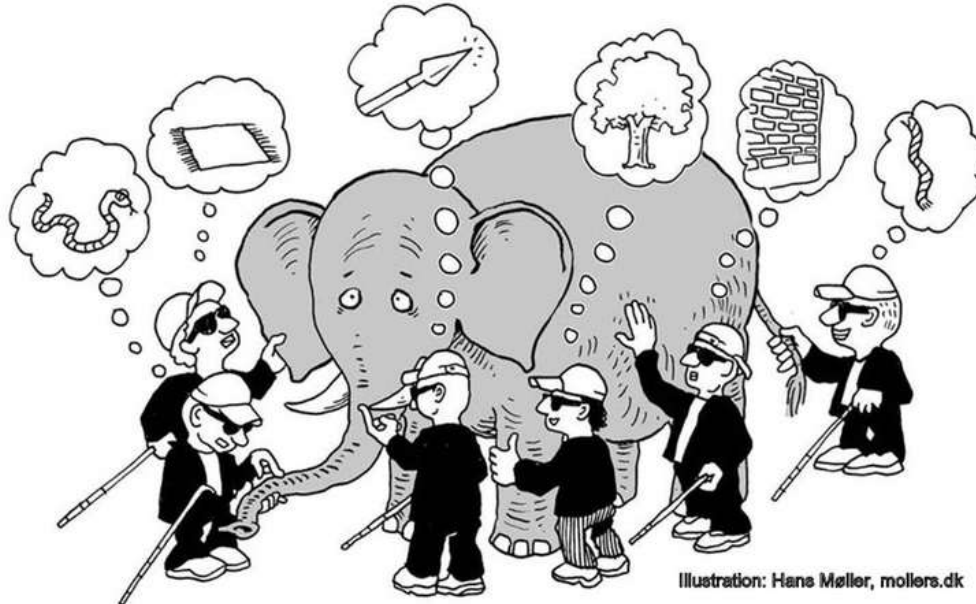
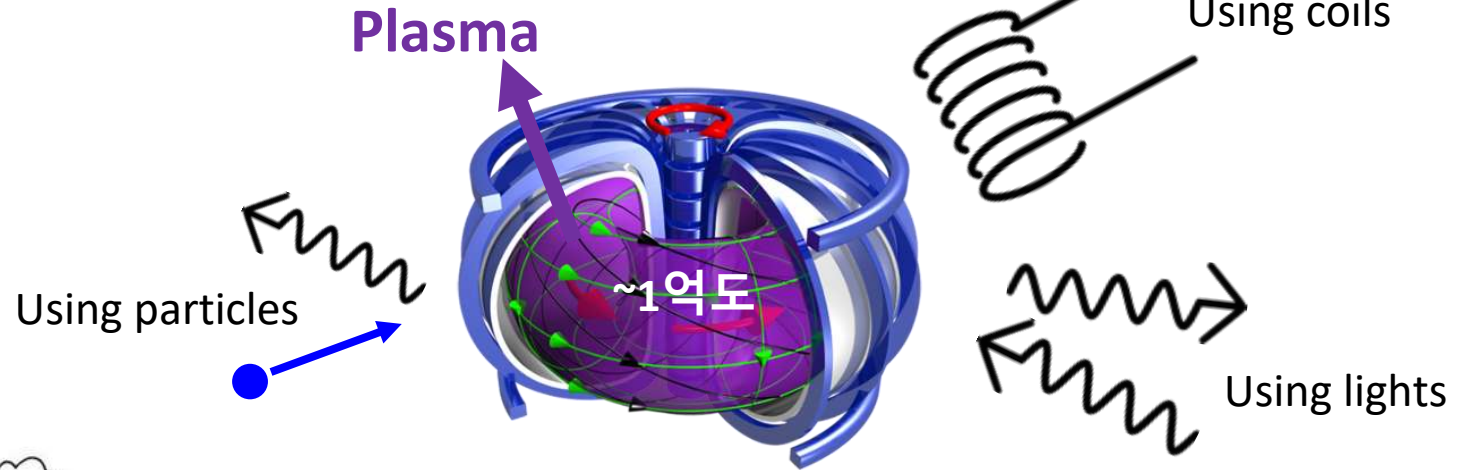
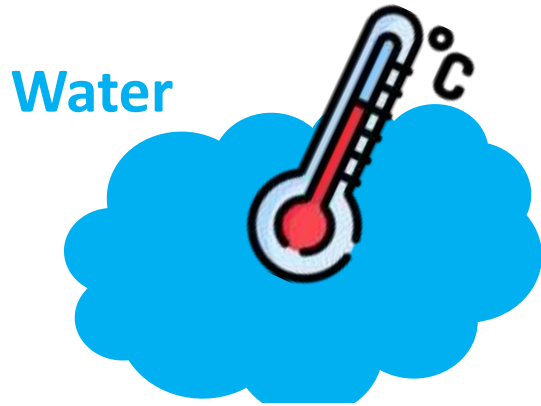
Fusion commercialization!

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# Plasma "diagnosis?"

- How to measure the temperature of hot plasma?



Diagnostics, not measurement!

**Inverse &  
ill-posed problem!**

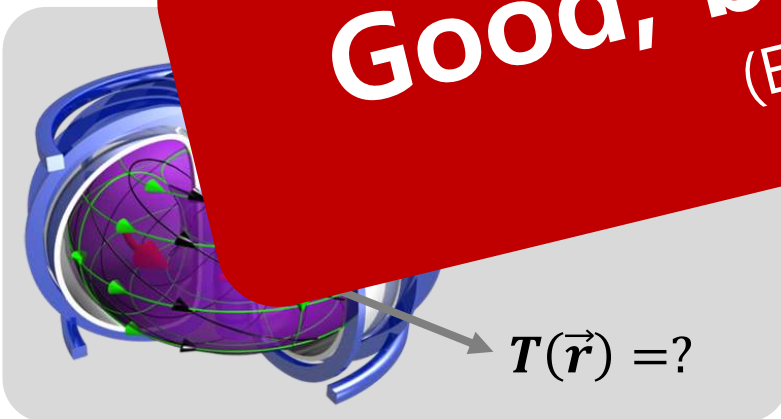
# A simple way: Data-driven diagnosis

- In the past,

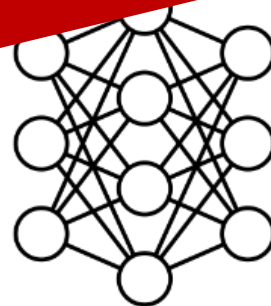
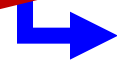


plasma profiles  
(temperature, mag field, ...)

**Good, but often physically invalid!**  
(Energy conservation, force balance, ...)



$$T(\vec{r}) = ?$$

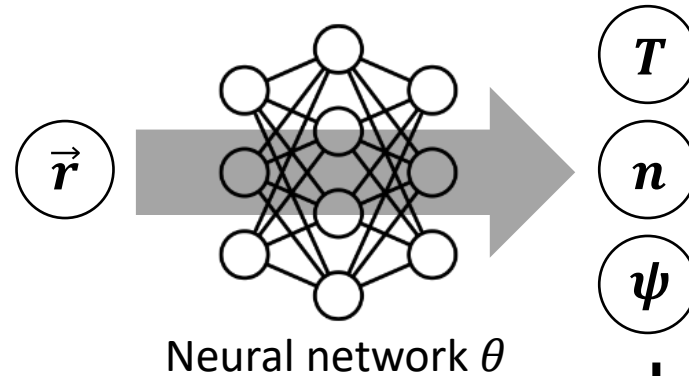
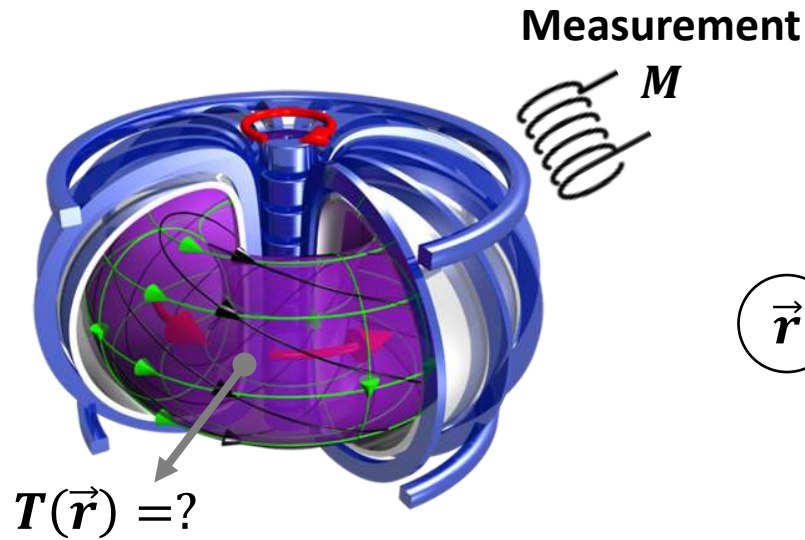


Neural network  $\theta$

- Data-driven

R Shousha\*, J Seo\* et al, Nucl. Fusion 64 (2024) 026006

# Another way: Physics-informed NN (PINN)



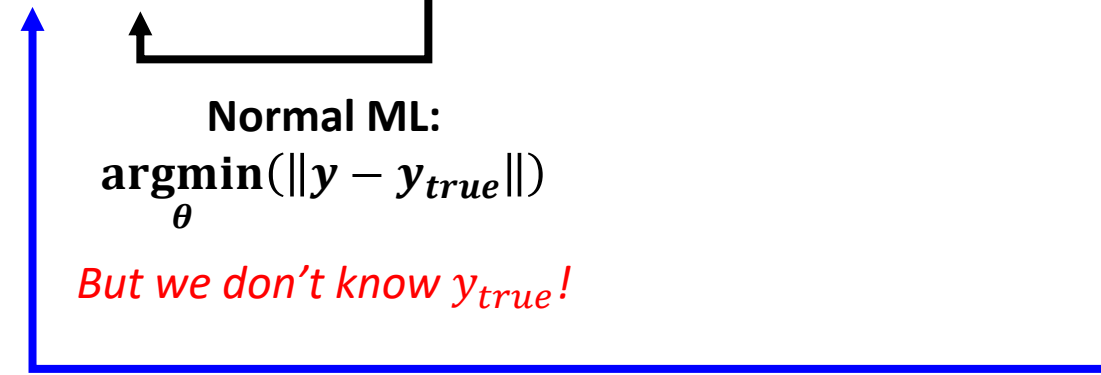
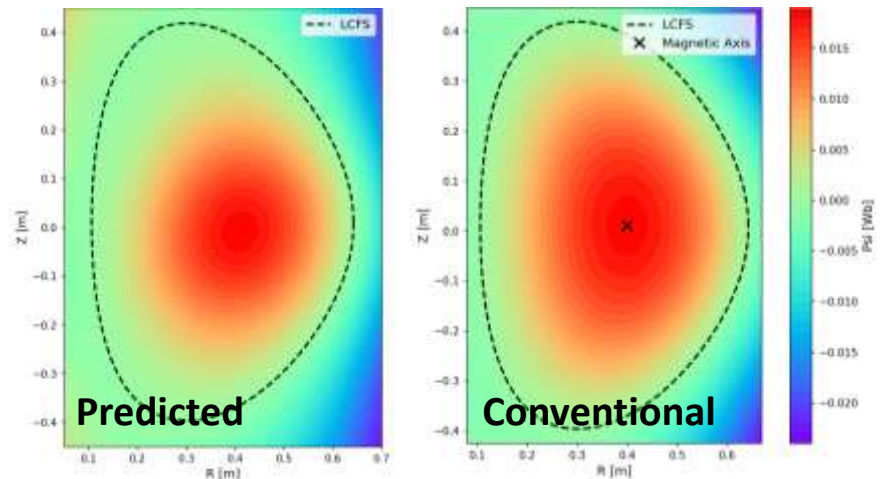
$$\frac{\partial^2 \psi}{\partial r^2} - \frac{1}{r} \frac{\partial \psi}{\partial r} + \frac{\partial^2 \psi}{\partial z^2} = -\mu_0 r^2 \frac{dp}{d\psi} - \frac{1}{2} \frac{dF^2}{d\psi},$$

1. How well it satisfies physics laws

$$L_{phys} = \|f(T, n, \psi; \vec{r})\|$$

2. How well its synthetic measurements match the true measurements

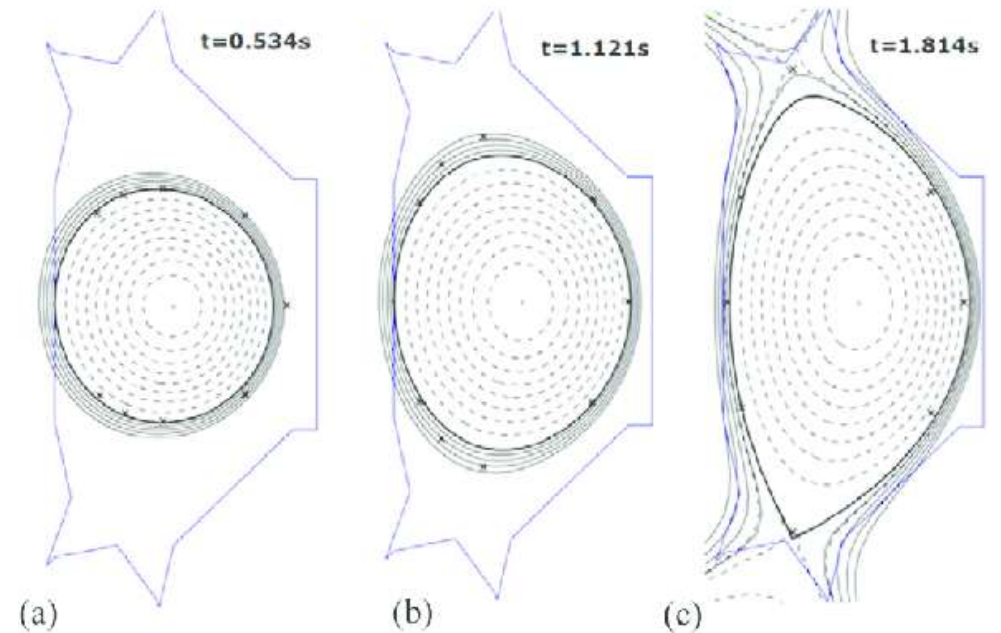
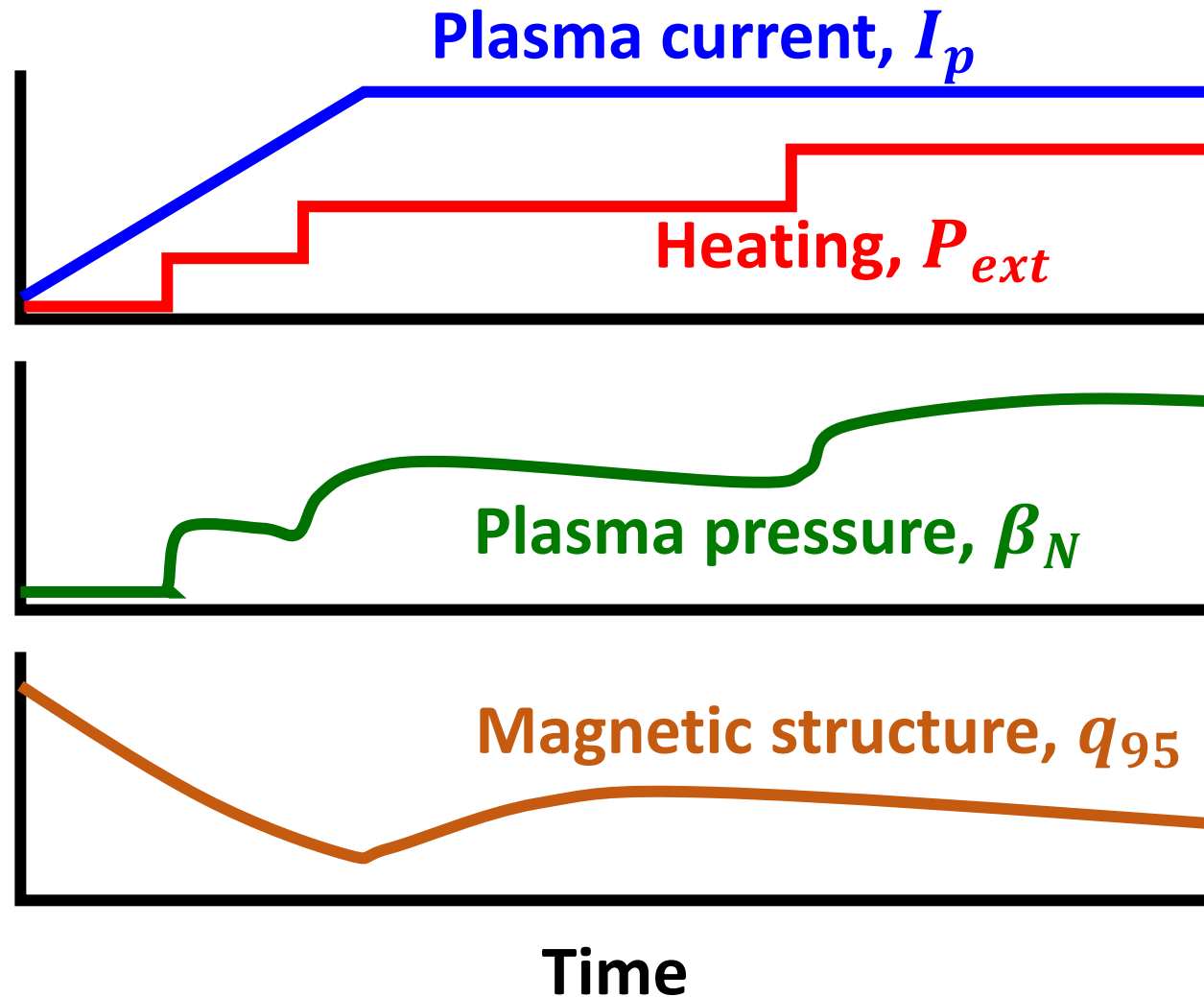
$$L_{meas} = \|M(y) - M_{true}\|$$



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# Typical plasma discharge in tokamak



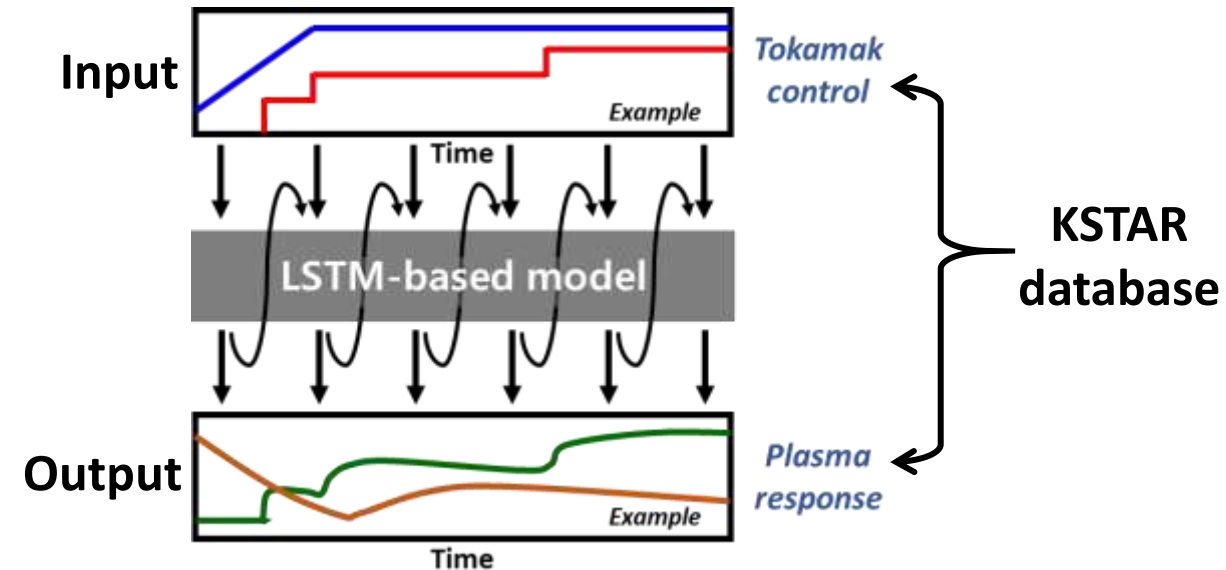
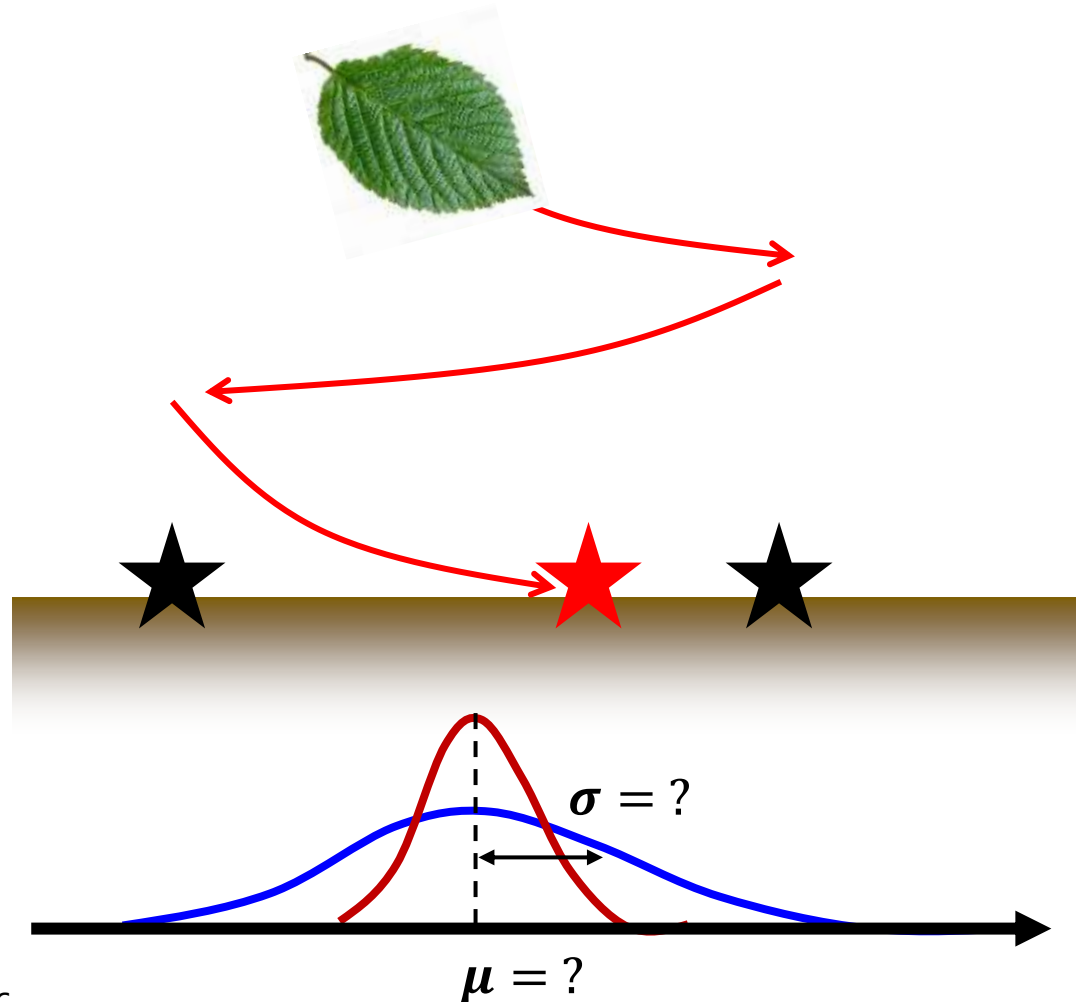
Plasma shape evolution

# Difficulty of plasma prediction?

J Seo et al, Nucl. Fusion 61 (2021) 106010

J Seo et al, Nucl. Fusion 62 (2022) 086049

- Predict the “average” dynamics ( $\mu(t)$ )  
: **LSTM or transformer**
- Predict the “uncertainty” of the dynamics ( $\sigma(t)$ )  
: **Bayesian or ensemble**

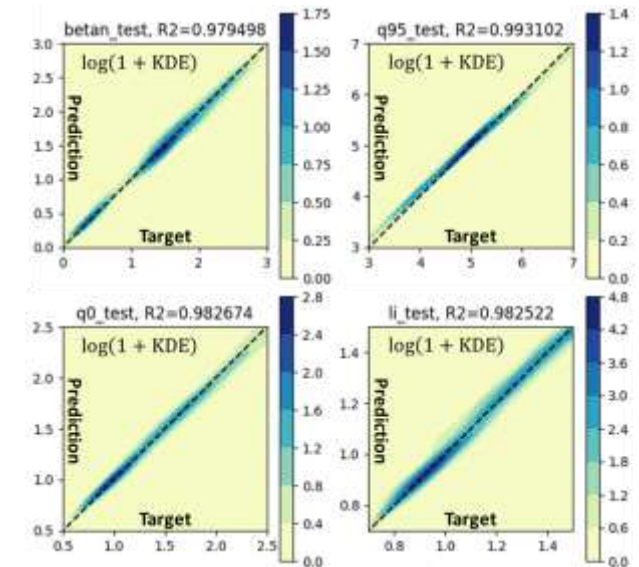
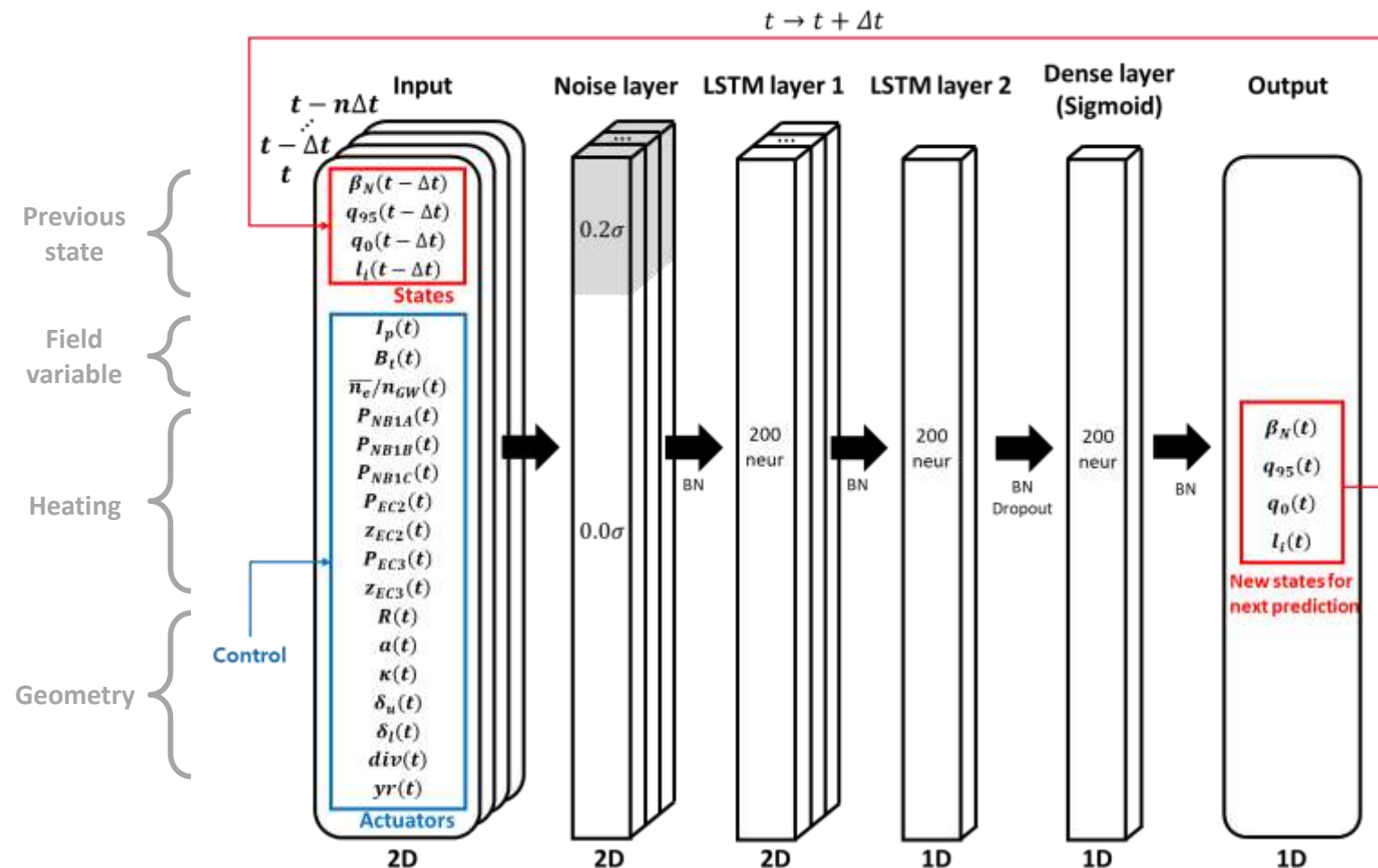


# Data-driven plasma prediction

J Seo et al, Nucl. Fusion 61 (2021) 106010

J Seo et al, Nucl. Fusion 62 (2022) 086049

- NN architecture & training with KSTAR database



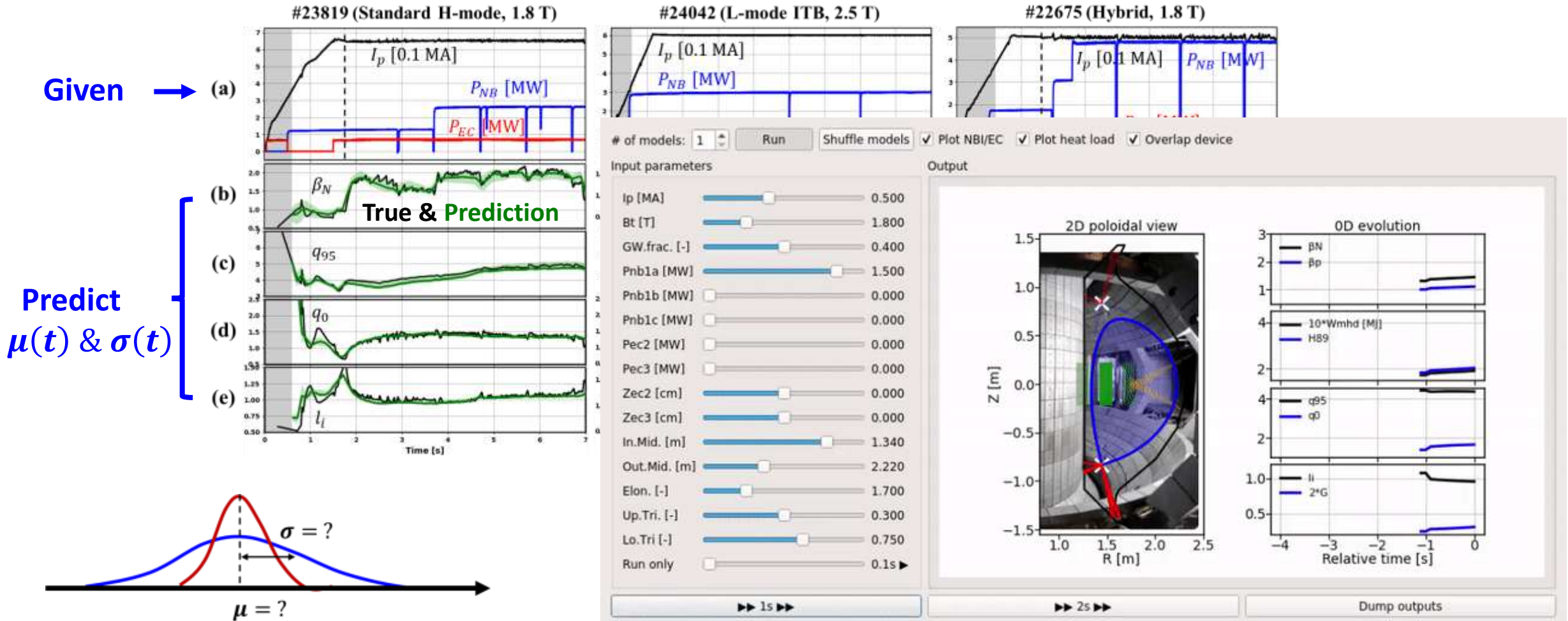
Regression plot for testset

# Data-driven plasma prediction

J Seo et al, Nucl. Fusion 61 (2021) 106010

J Seo et al, Nucl. Fusion 62 (2022) 086049

- Plasma parameter prediction in KSTAR



# Contents

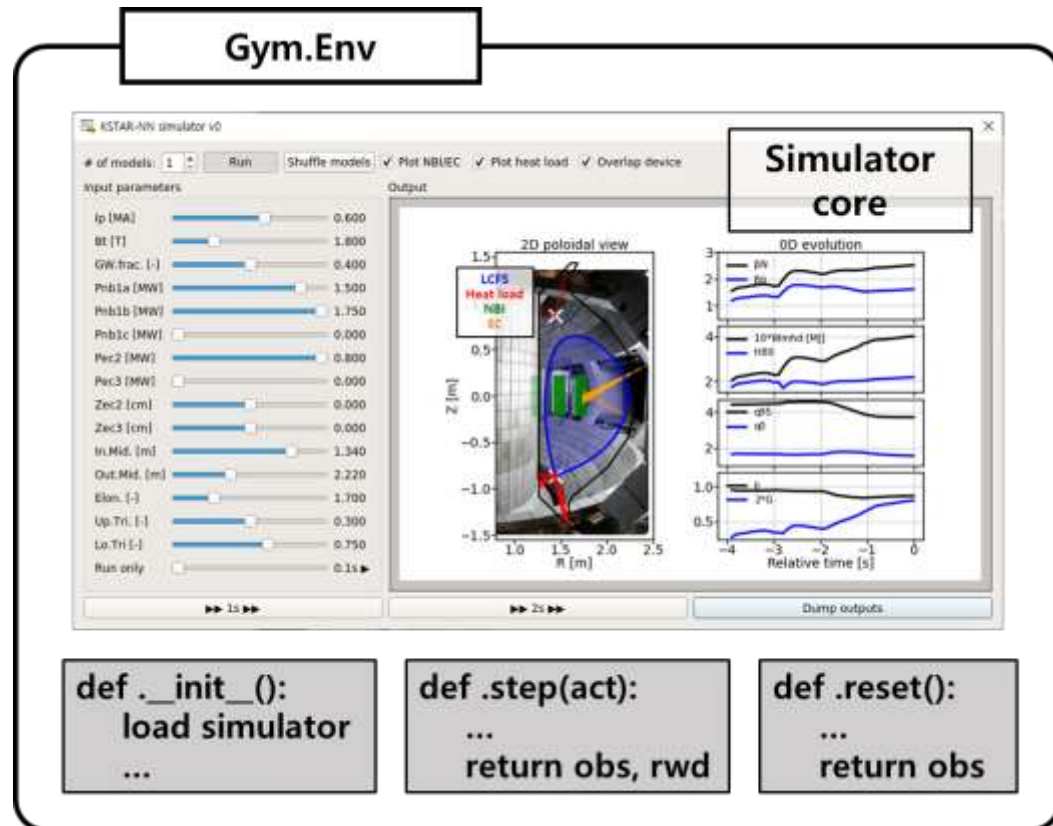
- Introduction: What is nuclear fusion & plasma?
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# We can predict, then we can control!

J Seo et al, Nucl. Fusion 61 (2021) 106010

J Seo et al, Nucl. Fusion 62 (2022) 086049

- Reinforcement learning for plasma control



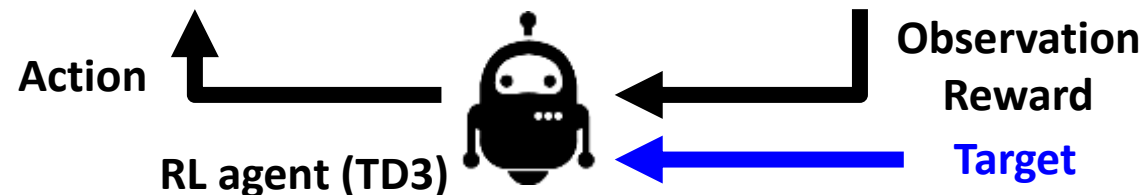
What we need to do is defining act, obs, and rwd!

- Action:** Control variables
- Observation:** Plasma response & target
- Reward:** Higher if the plasma is close to the target

**Action:**  $\{I_p, \kappa, \delta_u, \delta_l, R_{in}, R_{out}\}$

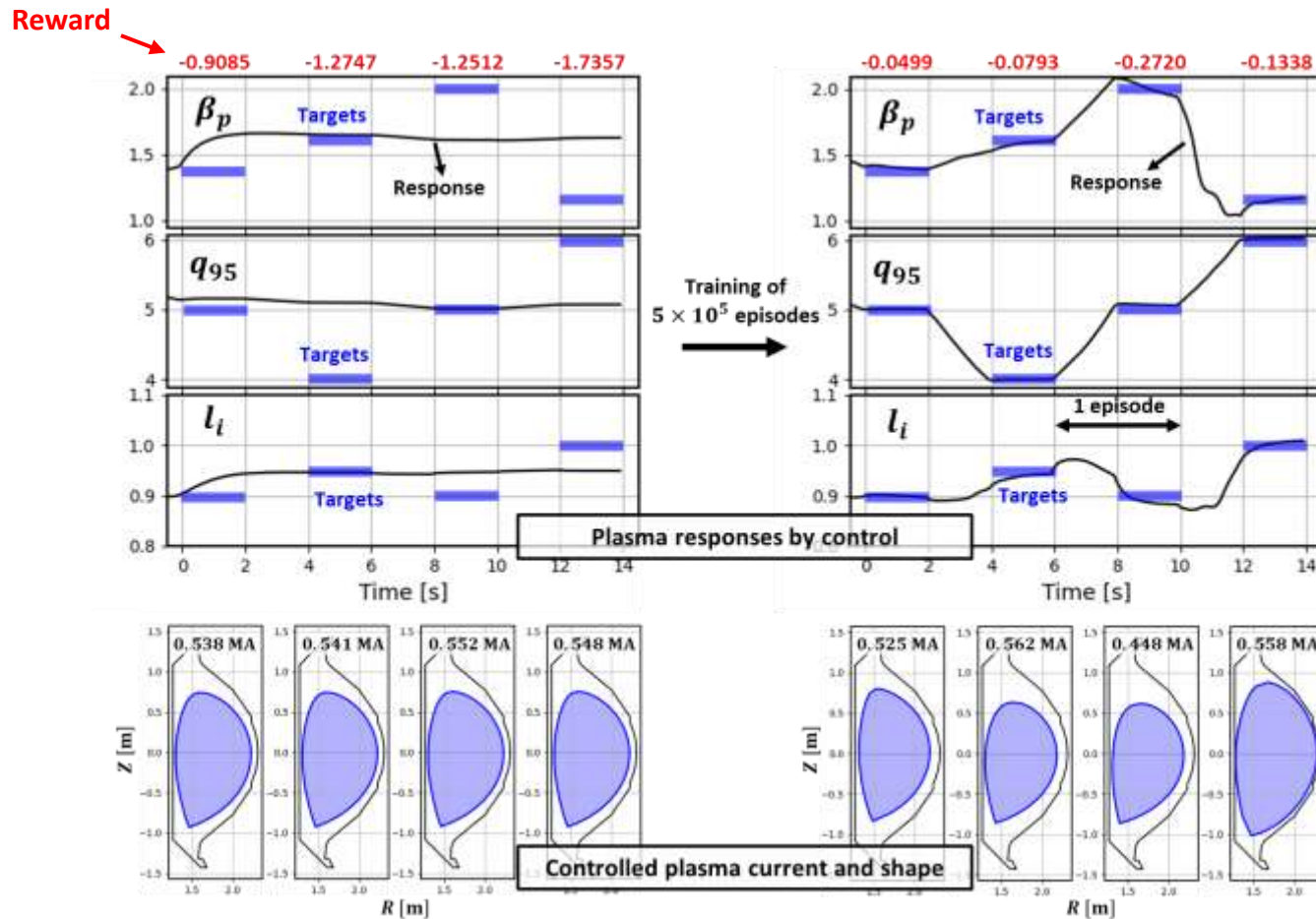
**Observation:** Previous action,  $\{\beta_p, q_{95}, l_i\}_{old}, \{\beta_p, q_{95}, l_i\}_{target}, \{P_{NB}'s\}$

**Reward:**  $-\text{RMS}\left(\frac{y-y_{target}}{\epsilon_y}\right)_{y=\beta_p, q_{95}, l_i}$



# Plasma control with RL

- Training in the simulation



Before training

After training



- After enough training, the AI determines reasonable solution of  $I_p$  and the boundary shape to reach the target of multiple parameters.

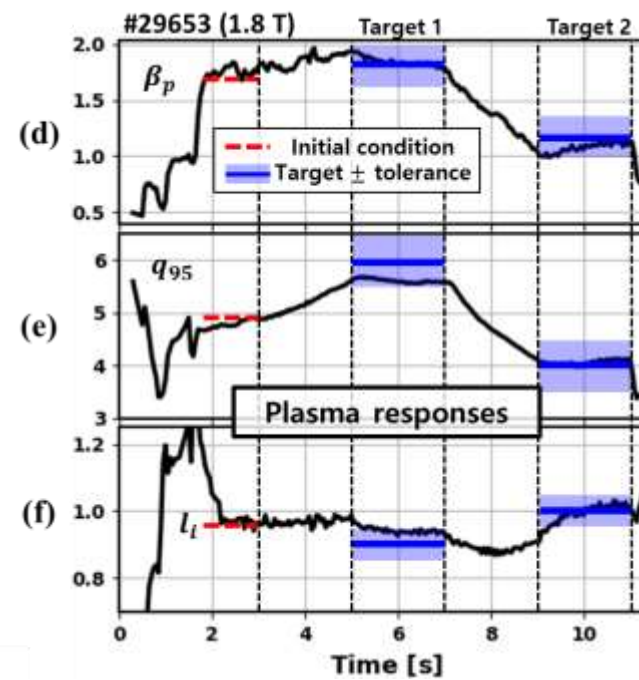
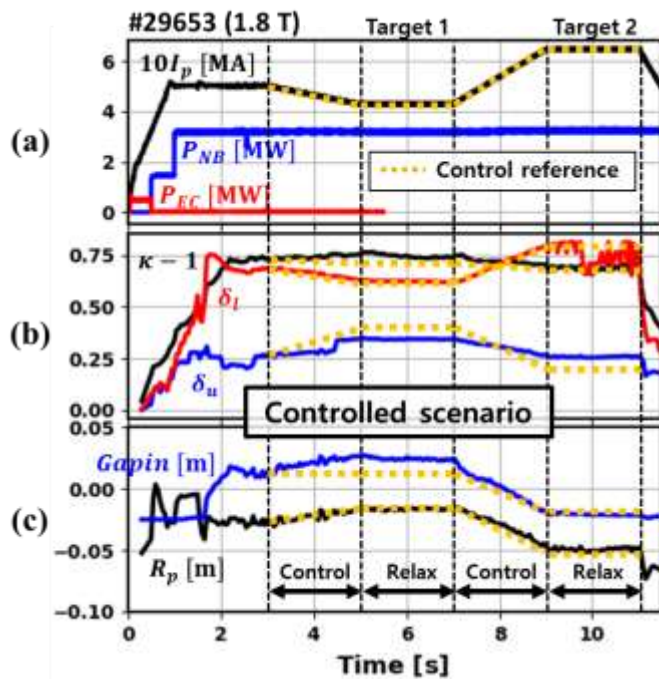
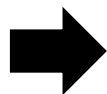
[https://github.com/jaem-seo/AI\\_tokamak\\_control](https://github.com/jaem-seo/AI_tokamak_control)

# Plasma control with RL

- Validation in the KSTAR experiment

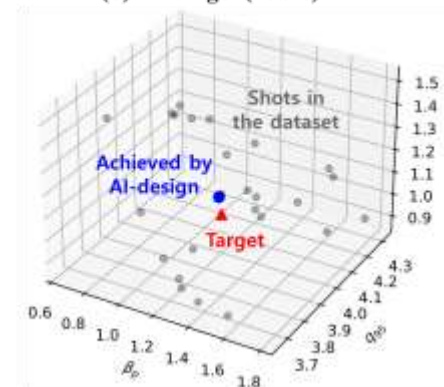
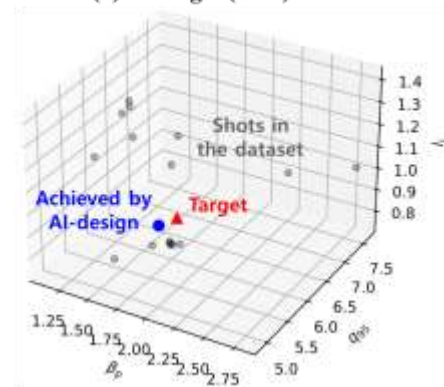
Target setting

$$(\beta_p, q_{95}, l_i) = (1.8, 6.0, 0.9) \text{ \& } (1.2, 4.0, 1.0)$$



(a) 1<sup>st</sup> target (6.9 s)

(b) 2<sup>nd</sup> target (10.9 s)

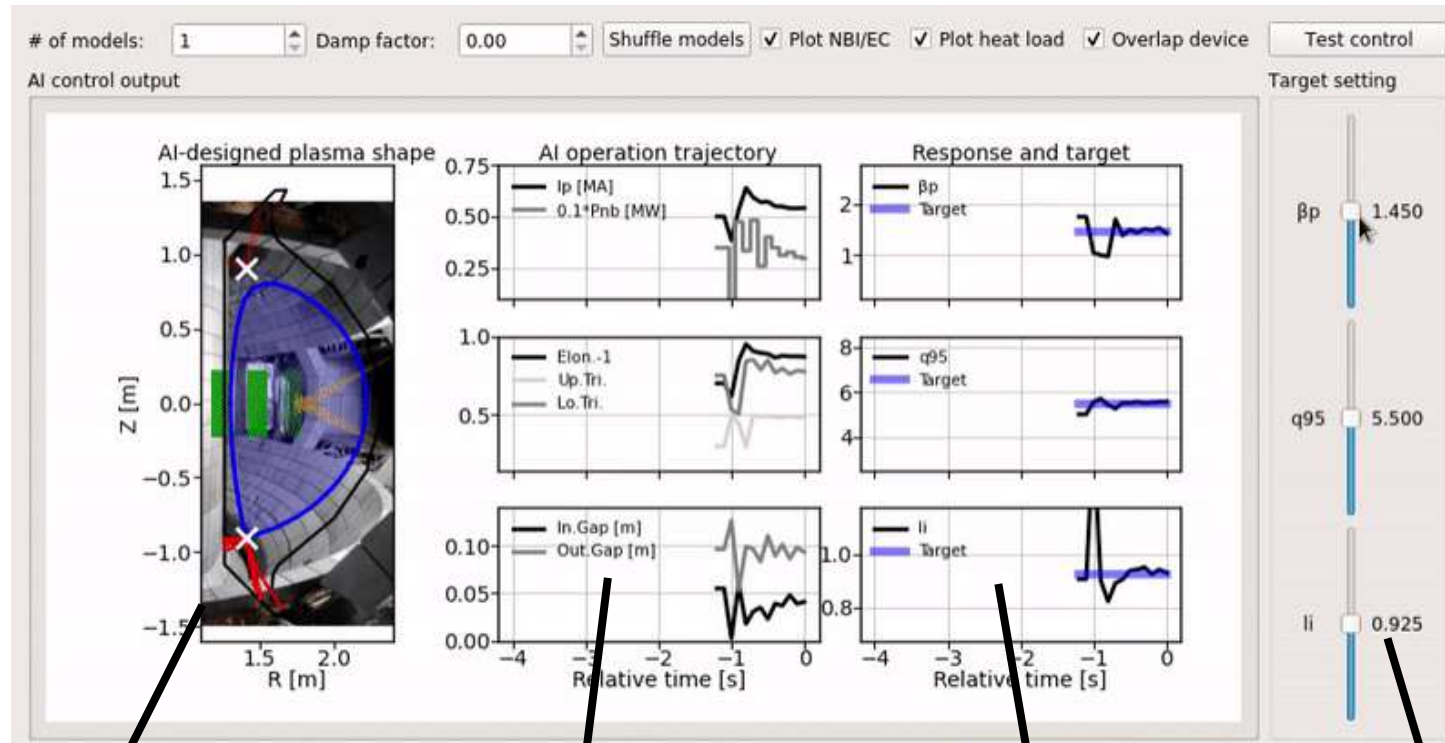


- The plasma response followed the preset targets.
  - However, actual shape controls were not perfect.
- Control/diagnostics uncertainty should be reflected later.

# Plasma control with RL

- Now, let's try **real-time control**

[https://github.com/jaem-seo/AI\\_tokamak\\_control](https://github.com/jaem-seo/AI_tokamak_control)



AI-determined  
plasma shape

AI-determined  
command signals

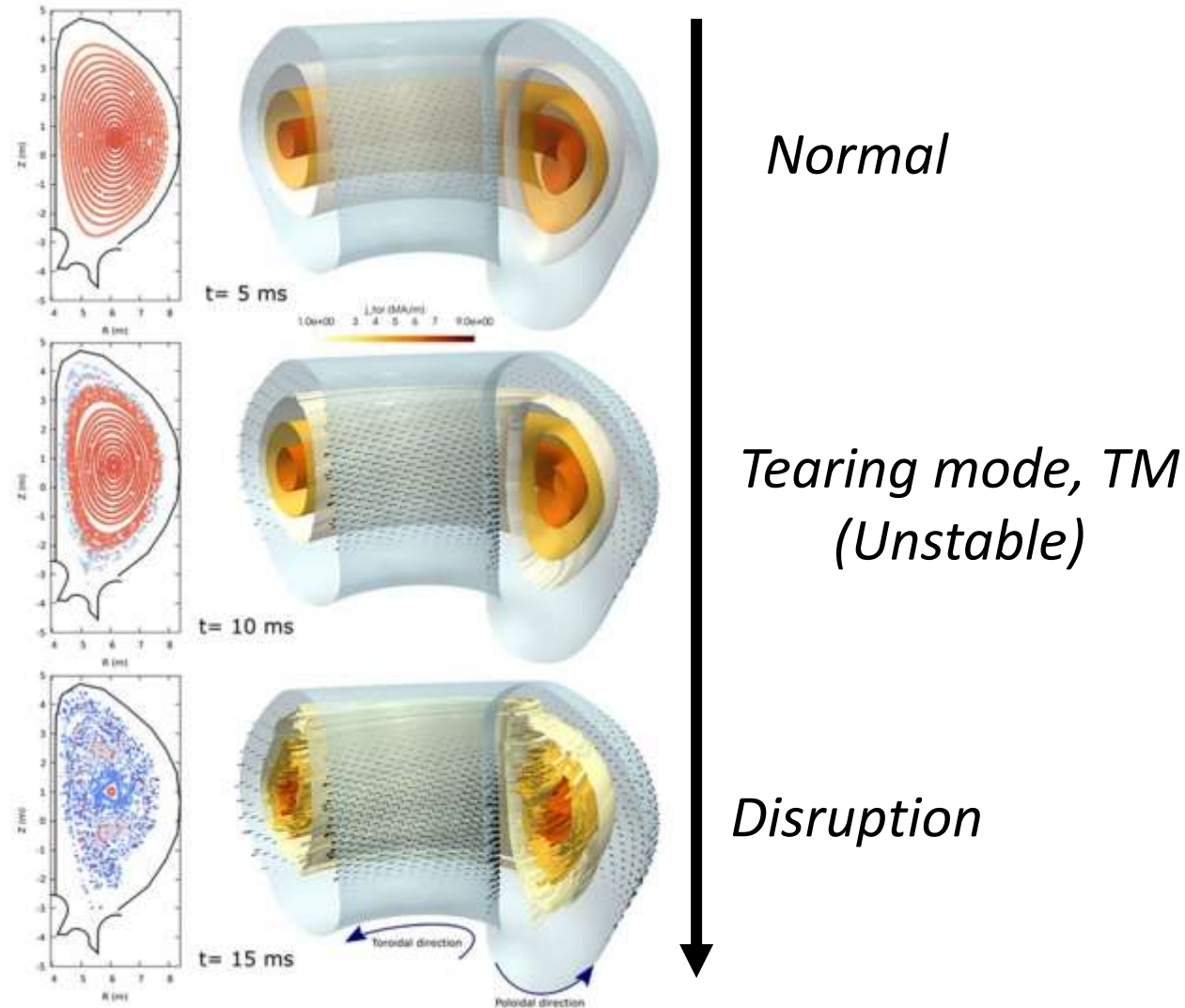
Target & plasma  
response

Target setting

- After RL training, the AI can control the command signals for  $\{I_p, P_{NB}, \text{shape}\}$  to **track the time-varying targets**.
- It seems reasonable ( $\beta_p$  vs NBI /  $q_{95}$  vs shape).
- Will be tested with different RL settings and constraints.

# Plasma instability

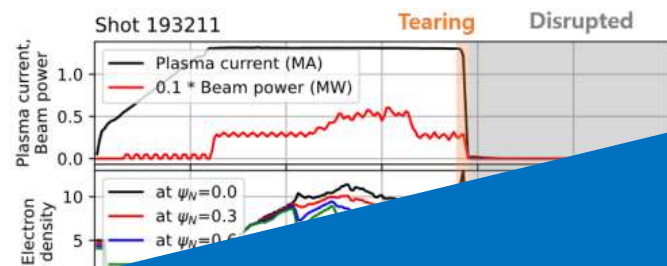
- Although we can control the plasma, the plasma can become **very unstable!**



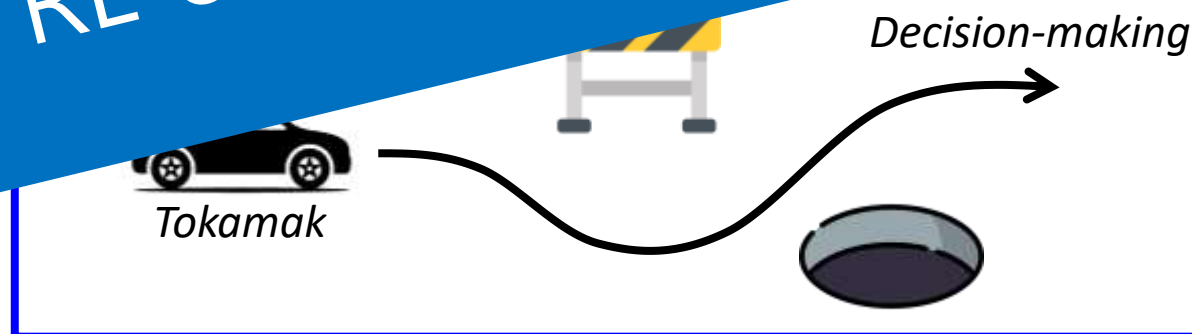
# Avoiding plasma instability with ML

- Let's use two different AIs: **prediction & control**

**AI 1:** Predicting future TM possibility



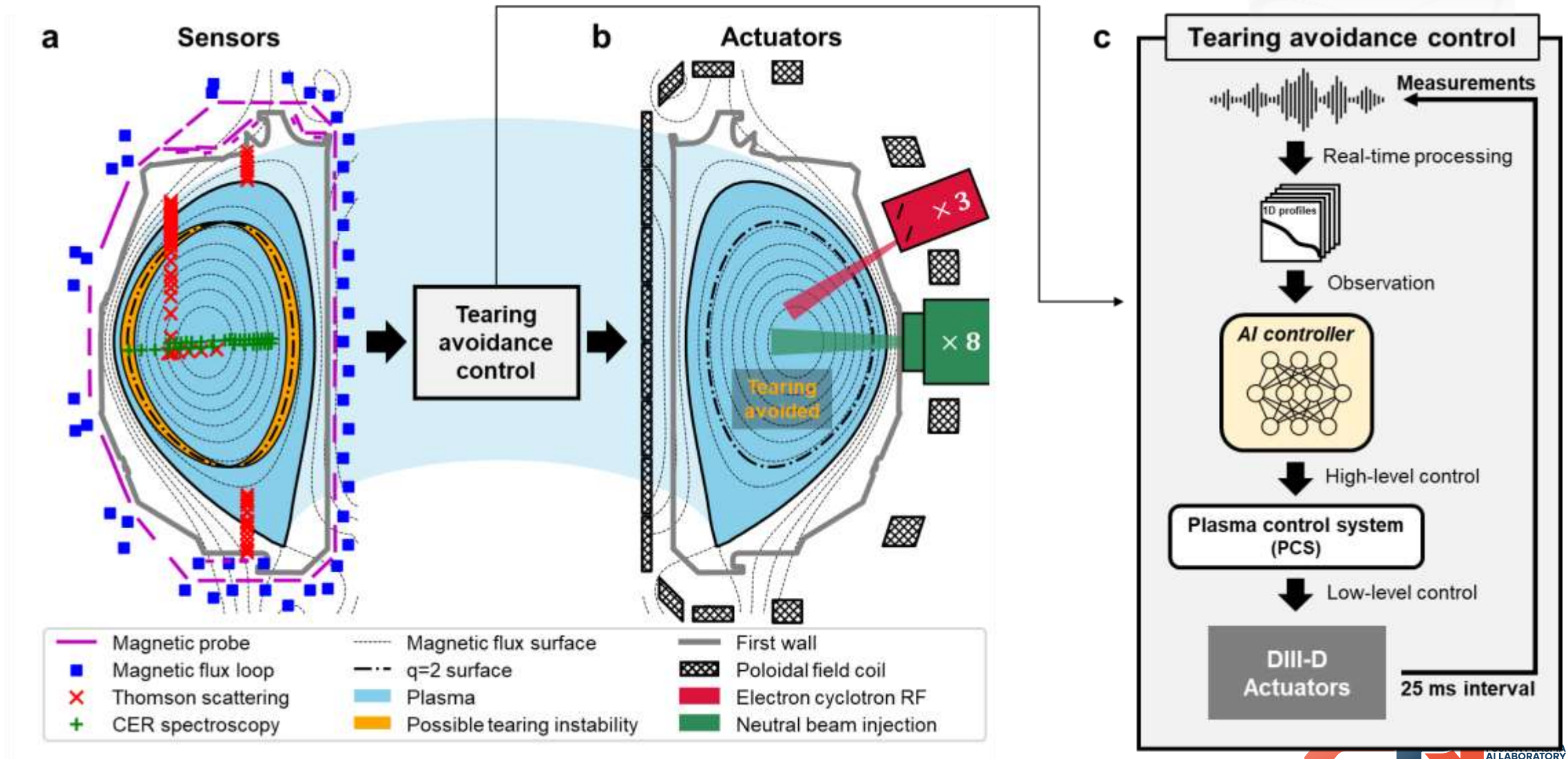
**Same strategy as before**  
(Making a surrogate model  
& training RL control agent)



# Avoiding plasma instability with ML

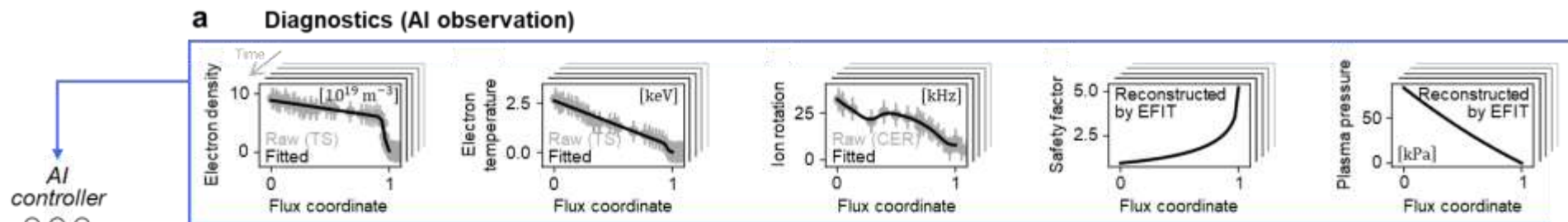


- Implementation in DIII-D

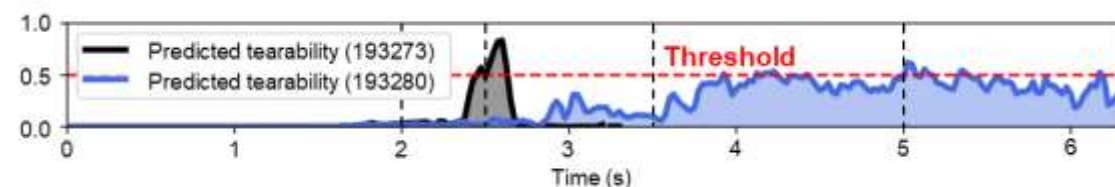
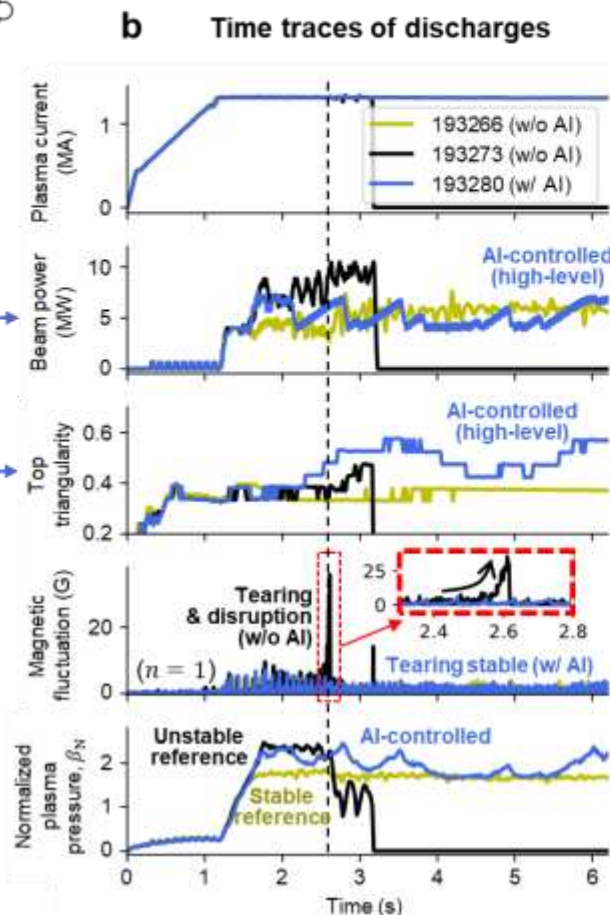


# Avoiding plasma instability with ML

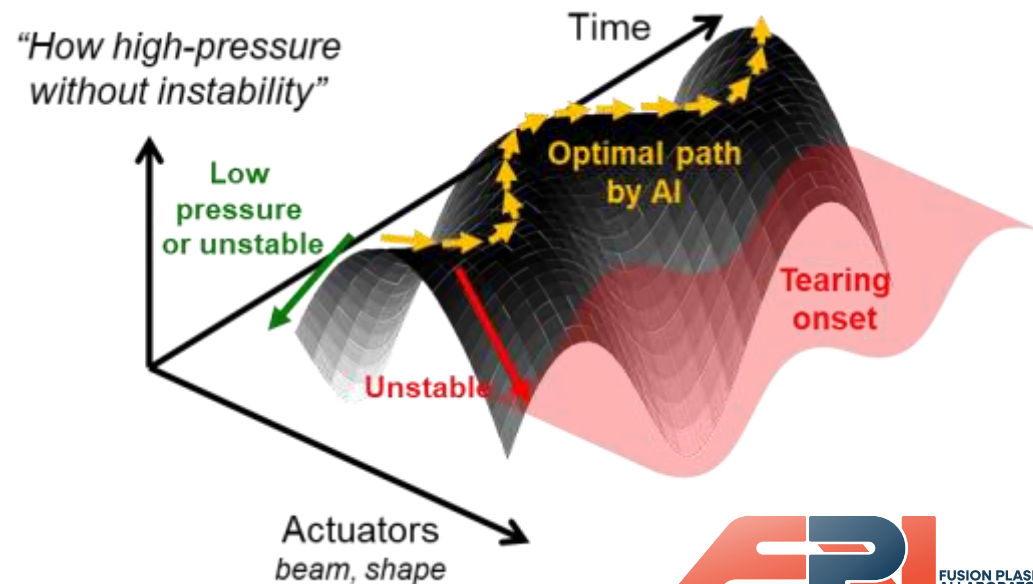
What AI sees →  
Kinetic profiles



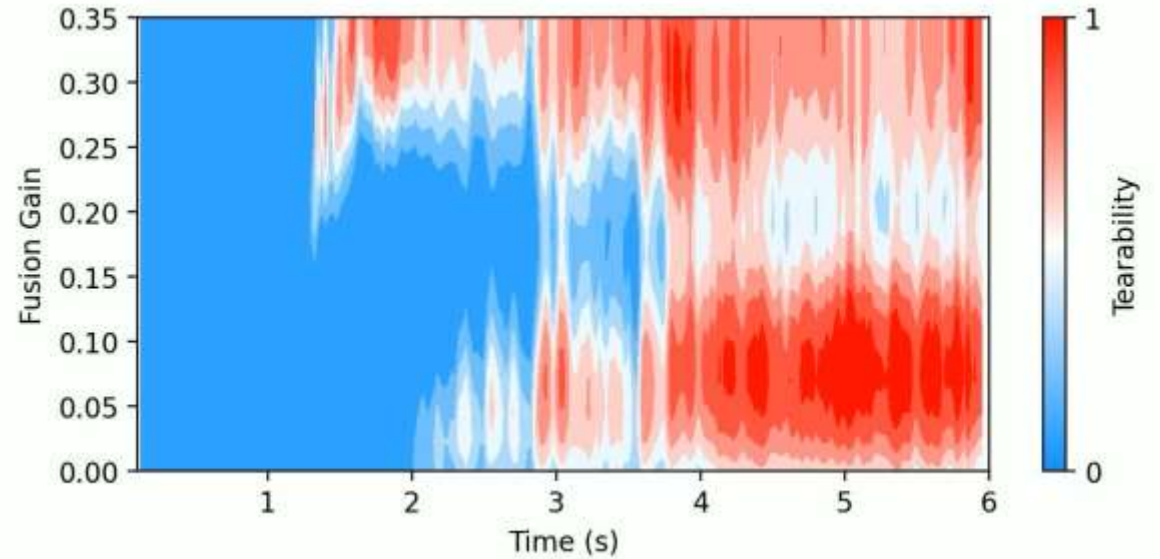
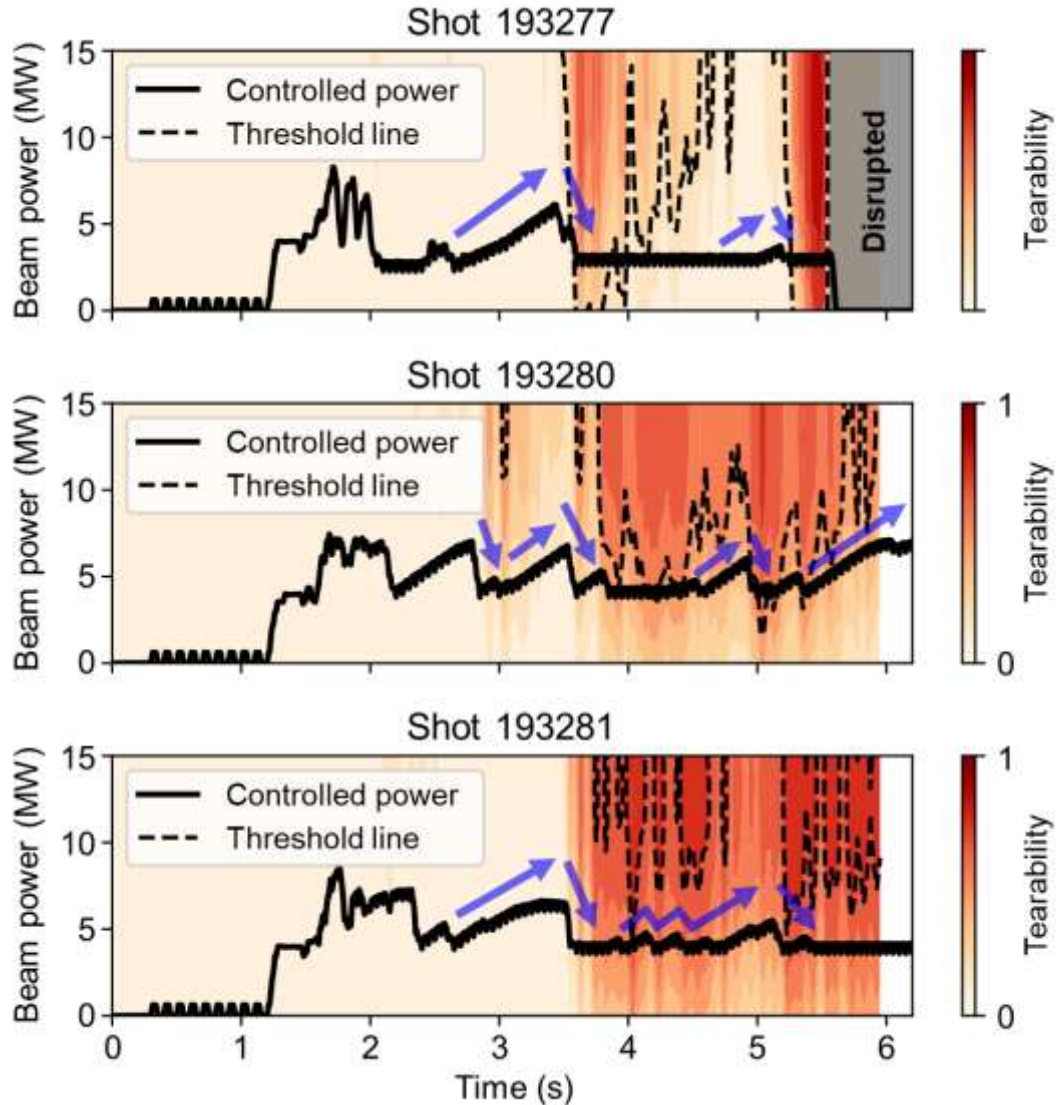
What AI handles →  
 $P_{NB}$  &  $\delta_t$



What we want →  
Higher  $\beta_N$  w/o TM



# Avoiding plasma instability with ML



# Contents

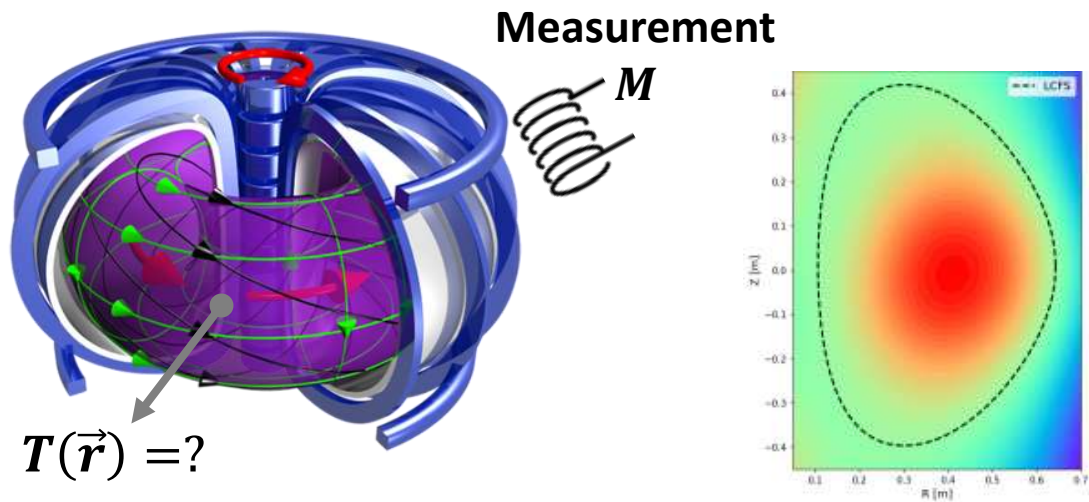
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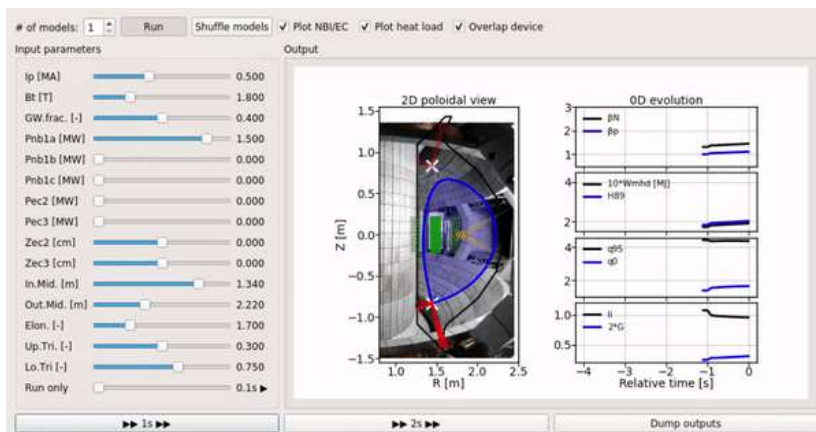
[1] R Shousha\*, J Seo\* et al, Nucl. Fusion 64 (2024) 026006  
 [2] J Seo et al, Nucl. Fusion 61 (2021) 106010

[3] J Seo et al, Nucl. Fusion 62 (2022) 086049  
 [4] J Seo et al, Nature 626 (2024) 746

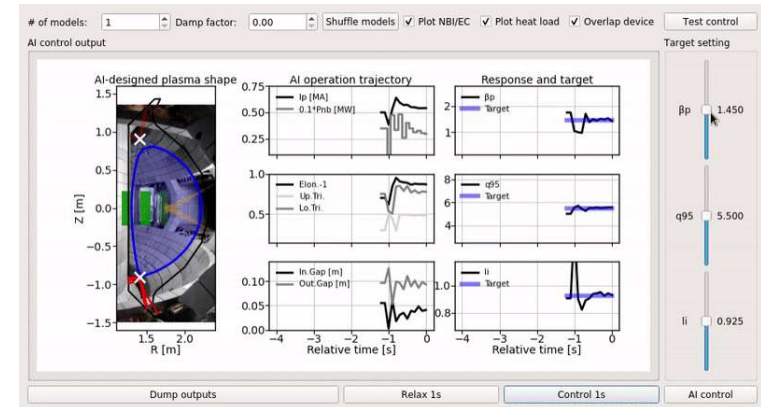
## 1. ML-based plasma diagnosis



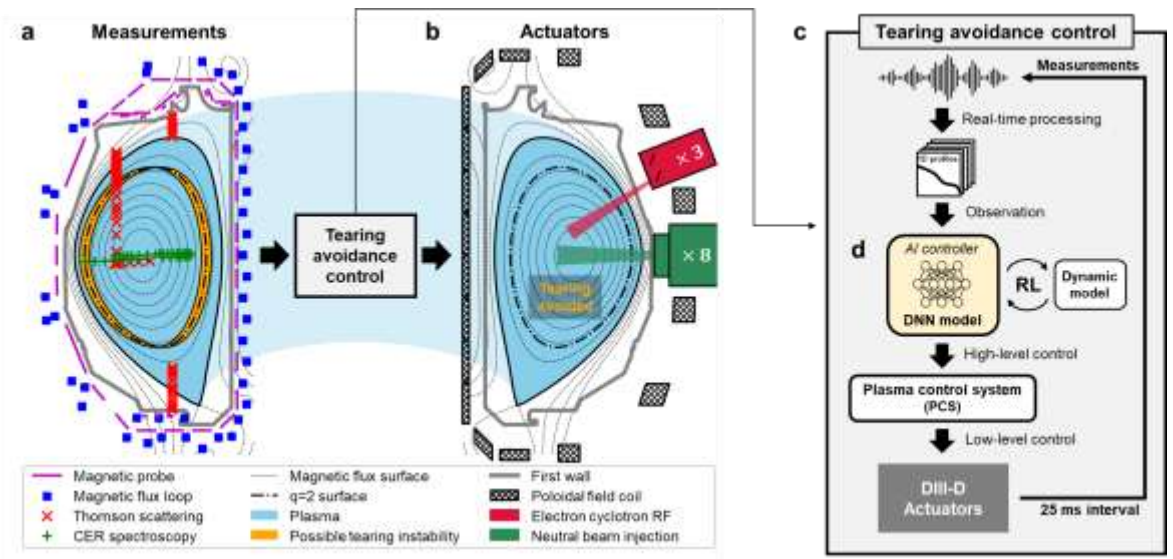
## 2. Data-driven plasma simulator



## 3. RL-based tokamak controller



## 4. RL-based instability avoidance





AI

Thank you for  
your attention!

