

# AI/ML in Fusion Energy

Tyler Ellis, Ph.D.

# CFS on a path to deliver commercial fusion energy



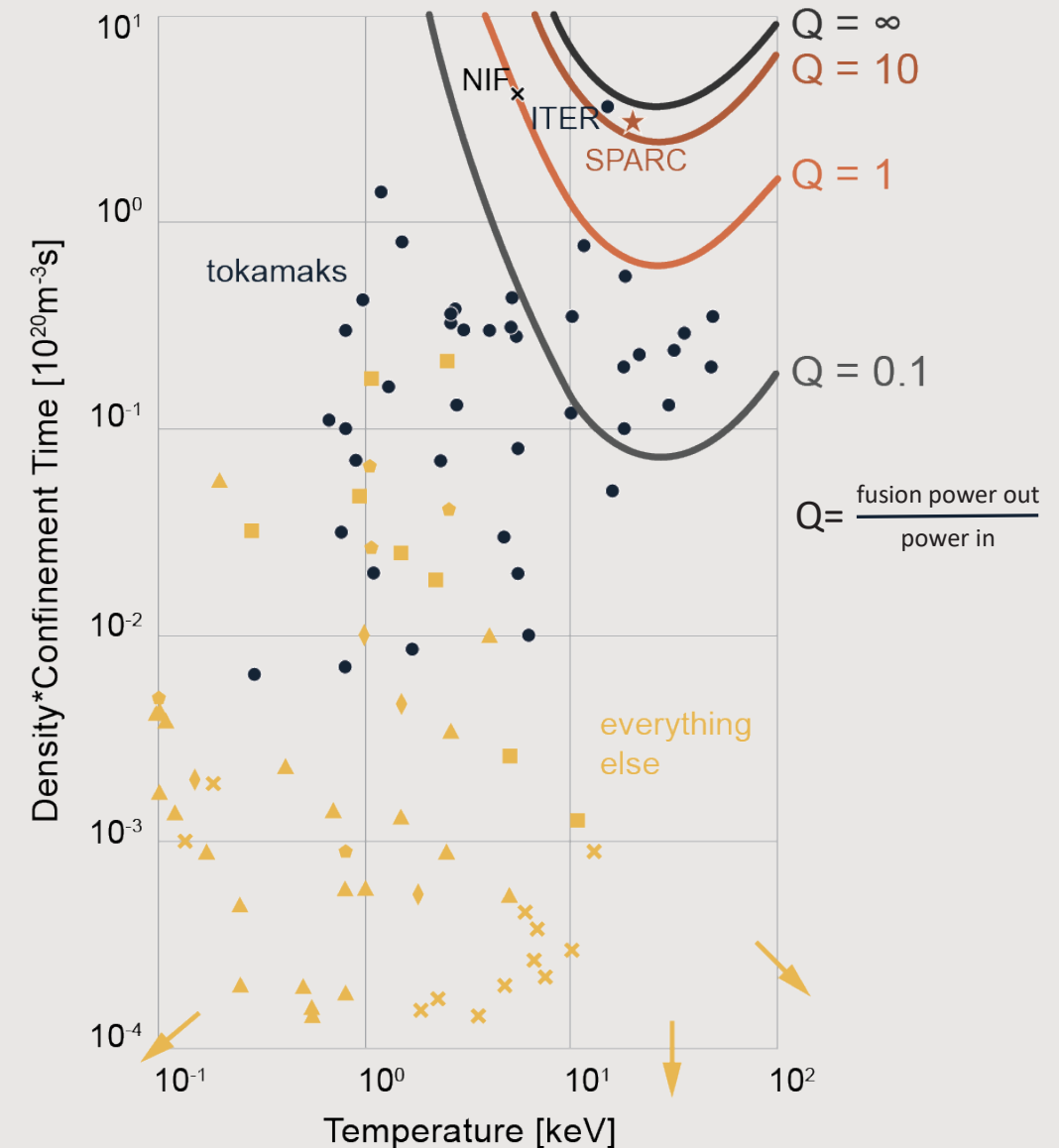
- CFS founded in 2018, spun out of MIT
- Raised more than \$2 billion
- Built a high caliber, diverse team
- Over 900 employees



# On the verge of commercially relevant fusion power



- On the cusp of a key milestone of net energy gain, more energy out than in ( $Q > 1$ )
- Machines called “tokamaks” are closest
- >150 tokamaks have been built worldwide
- Magnets hold and insulate the plasma and very high magnetic fields make tokamaks smaller





# SPARC, first commercially relevant fusion machine



Validated  
approach

*J. Plasma Phys.* (2020), vol. 86, 865860502 © The Author(s), 2020.  
Published by Cambridge University Press  
This is an Open Access article, distributed under the terms of the Creative Commons  
Attribution-NonCommercial-NoDerivatives licence (<http://creativecommons.org/licenses/by-nc-nd/4.0/>), which  
permits non-commercial re-use, distribution, and reproduction in any medium, provided the original work is  
unaltered and is properly cited. The written permission of Cambridge University Press must be obtained for  
commercial re-use or in order to create a derivative work.  
doi:10.1017/S0022377820001257

## Overview of the SPARC tokamak

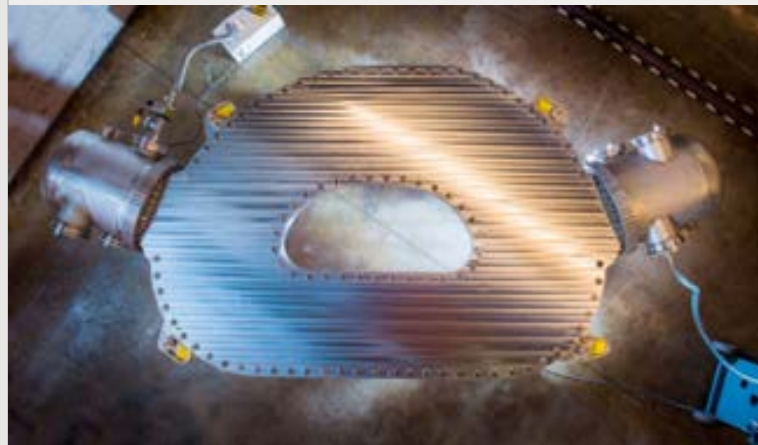
A. J. Creely<sup>1</sup>, M. J. Greenwald<sup>2</sup>, S. B. Ballinger<sup>2</sup>, D. Brunner<sup>1</sup>, J. Canik<sup>1</sup>,  
J. Doody<sup>2</sup>, T. Fülöp<sup>2</sup>, D. T. Garnier<sup>2</sup>, R. Granetz<sup>2</sup>, T. K. Gray<sup>1</sup>, C. Holland<sup>1</sup>,  
N. T. Howard<sup>2</sup>, J. W. Hughes<sup>2</sup>, J. H. Irby<sup>2</sup>, V. A. Izzo<sup>2</sup>, G. J. Kramer<sup>2</sup>,  
A. Q. Kuang<sup>2</sup>, B. LaBombard<sup>2</sup>, Y. Lin<sup>2</sup>, B. Lipschultz<sup>2</sup>, N. C. Logan<sup>2</sup>,  
J. D. Lore<sup>1</sup>, E. S. Marmar<sup>1</sup>, K. Montes<sup>2</sup>, R. T. Mumgaard<sup>1</sup>, C. Paz-Soldán<sup>2</sup>,  
C. Rea<sup>2</sup>, M. L. Reinke<sup>1</sup>, P. Rodríguez-Fernández<sup>2</sup>, K. Särkimäki<sup>2</sup>,  
F. Sciortino<sup>2</sup>, S. D. Scott<sup>1</sup>, A. Snicker<sup>2</sup>, P. B. Snyder<sup>2</sup>, B. N. Sorbom<sup>1</sup>,  
R. Sweeney<sup>1</sup>, R. A. Tinguey<sup>2</sup>, E. A. Tolman<sup>2</sup>, M. Umansky<sup>1</sup>, O. Vallhagen<sup>2</sup>,  
J. Varje<sup>2</sup>, D. G. Whyte<sup>2</sup>, J. C. Wright<sup>2</sup>, S. J. Wukitch<sup>2</sup>, J. Zhu<sup>2</sup>  
and the SPARC Team<sup>1,2</sup>

<sup>1</sup>Commonwealth Fusion Systems, Cambridge, MA, USA

<sup>2</sup>Plasma Science and Fusion Center, Massachusetts Institute of Technology, Cambridge, MA, USA

Peer reviewed  
publications

Demonstrated  
technology



Built world's strongest  
superconducting magnets

Accelerated  
construction



We are building  
it now



# Construction of SPARC Facility in Devens, Massachusetts





# CFS magnet factory in Devens, Massachusetts

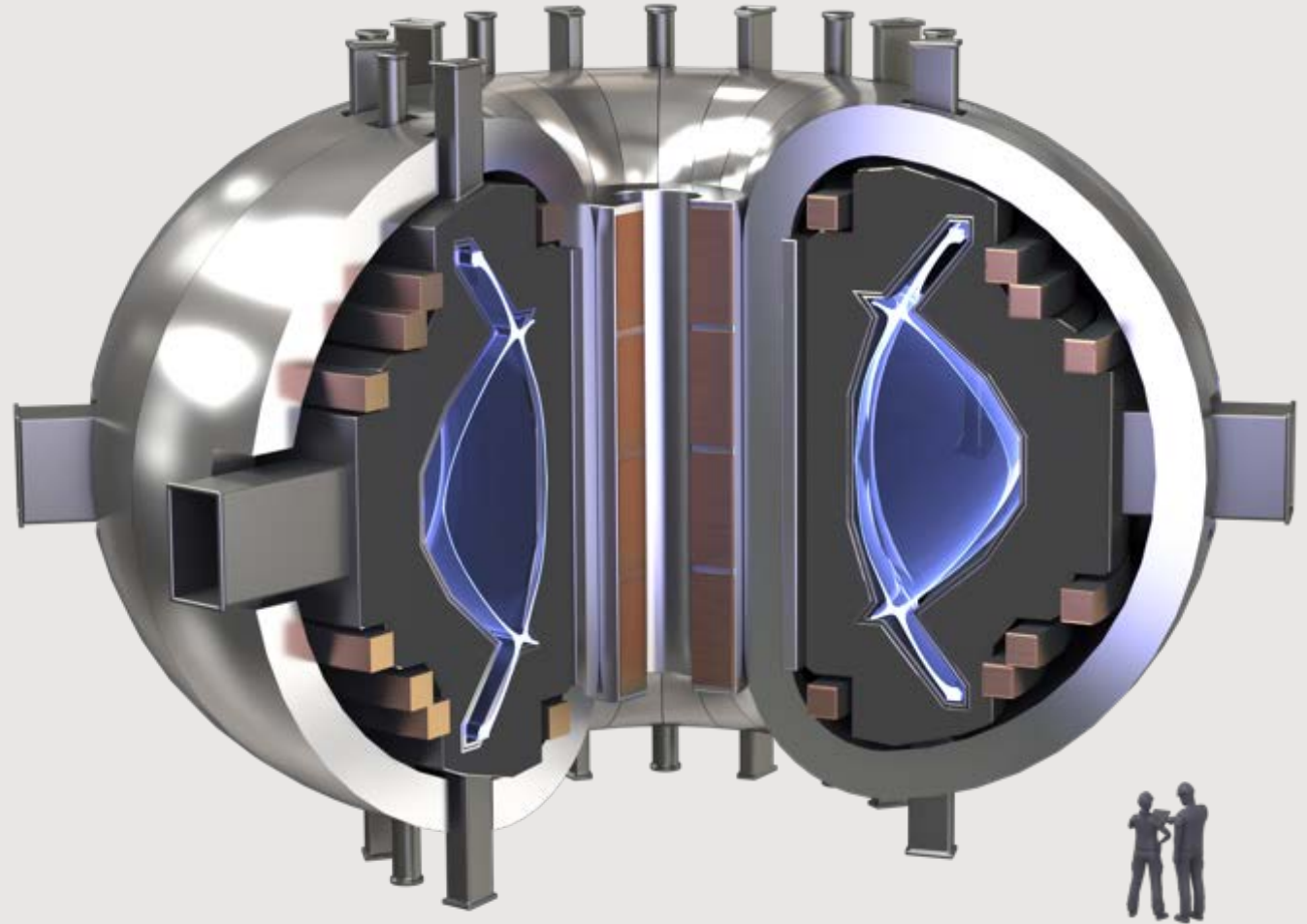




# ARC, world's first commercial fusion power plant



- Validation from SPARC platform that de-risks:
  - Economics  
using SPARC costs and supply chain
  - Performance  
using SPARC operations to optimize
  - Technology  
using SPARC and innovative R&D pathways in parallel
- Preparing to construct ARC, site search underway



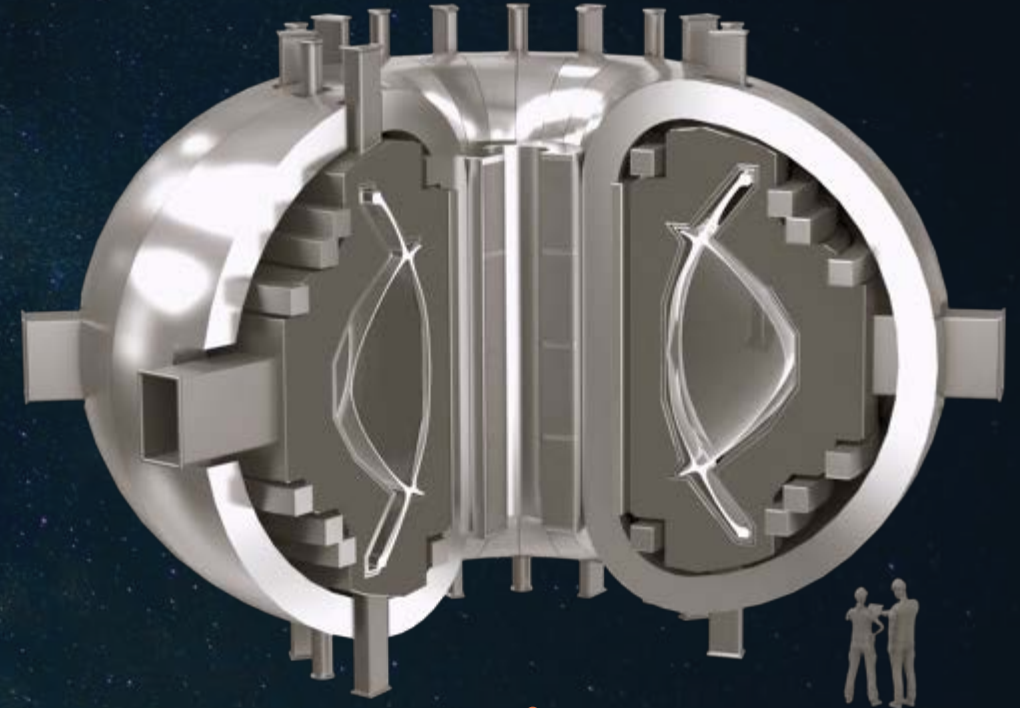
# Risk retirement in concrete steps



R&D

Commercial demo

Commercial powerplant



Physics  
COMPLETED

Magnet tech  
COMPLETED

SPARC  
UNDER CONSTRUCTION

ARC  
EARLY 2030s





# U.S. Regulatory Treatment for Fusion

- The optimal approach to regulate fusion was discussed over two years through 10 NRC-led public meetings with wide stakeholder input.
- In March 2023, the NRC Commission unanimously voted in favor of regulating fusion energy facilities under a byproduct materials approach (10 CFR 30).
- The ADVANCE Act, signed into law July 9, 2024, codified this regulatory treatment for fusion into the Atomic Energy Act.



*NRC Commissioner meeting on fusion regulations*



# AI/ML in Fusion

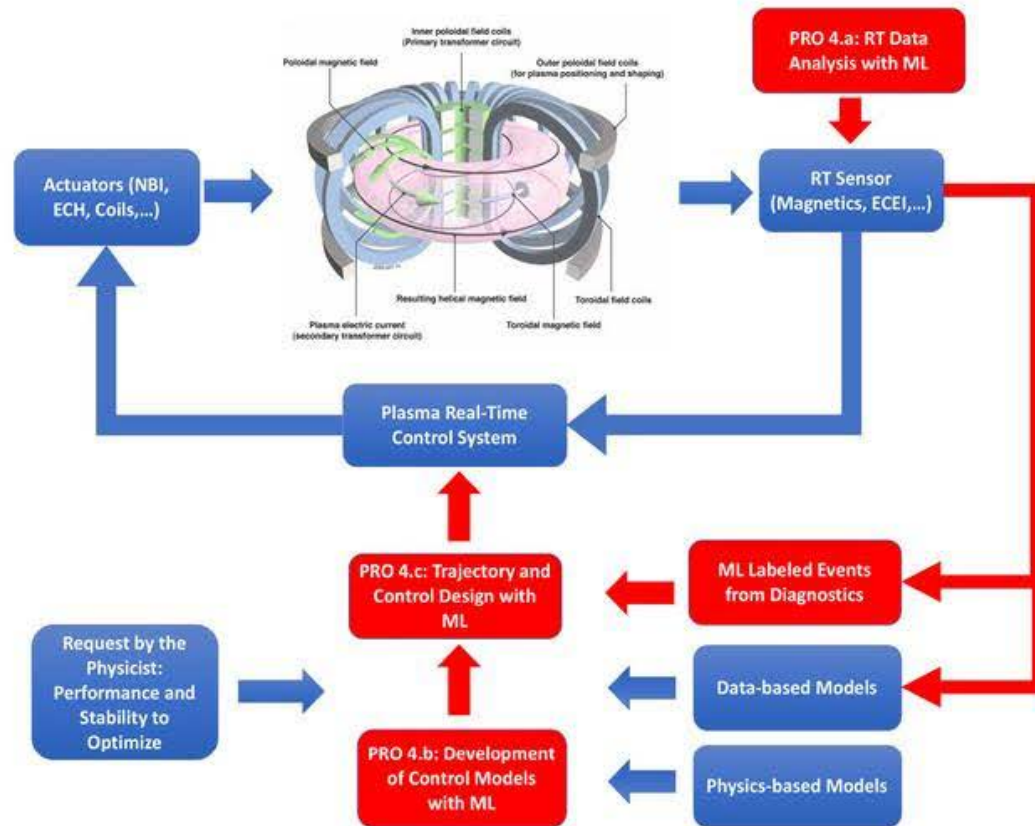
- AI/ML has been applied to fusion in three major ways:
  1. AI/ML can better compare datasets across different existing fusion machines since they have taken data in different ways. This would provide more data that could further improve operational models.
  2. AI/ML can develop enhanced plasma control models that enhance uptime operations. For example, the development of ML algorithms that actively monitor for disruptions.
  3. AI/ML can help the discovery of surrogate/reduced models in several areas like prediction of turbulent transport.
- Basically, AI/ML can be a means to speed up the pathway of fusion commercialization and enhance uptime operations but it's not likely to enable breakthroughs in the final design of mature fusion machine concepts.



# Summary of main AI/ML applications for fusion



## Data drives fusion experiments' design, simulation, analysis, control and optimization



Machine Learning has a key role in:

- advancing science discovery;
- accelerating simulations via surrogates;
- predictive modeling;
- trajectory planning and control;
- database and data analysis;
- ...

Adapted from D. Humphreys et al. "Advancing Fusion With Machine Learning" DOE Workshop (2020)

# Active fusion research initiatives in AI/ML at MIT PSFC



- MIT Plasma Science and Fusion Center is a [Collaborating Centre](#) on AI in Fusion and Plasma Science for IAEA and has research groups studying:
  - Training of sequence-to-sequence models that predict the onset of disruptions <sup>1,2</sup>
  - Addressing plasma dynamical uncertainty by training control policies on parallel physics simulators <sup>3</sup>
  - ML models to accelerate plasma simulations to both aid in the discovery of new physics/optimized designs as well as allow more accurate models to potentially be used inside control systems <sup>4,5</sup>



[1] Keith, Zander, et al. "Risk-Aware Framework Development for Disruption Prediction: Alcator C-Mod and DIII-D Survival Analysis." Journal of Fusion Energy 43.1 (2024): 21.

[2] Zhu, J. X., et al. "Integrated deep learning framework for unstable event identification and disruption prediction of tokamak plasmas." Nuclear Fusion 63.4 (2023): 046009.

[3] Wang, Allen M., et al. "Active Disruption Avoidance and Trajectory Design for Tokamak Ramp-downs with Neural Differential Equations and Reinforcement Learning." arXiv preprint arXiv:2402.09387 (2024).

[4] Rodriguez-Fernandez, P., et al. "[Enhancing predictive capabilities in fusion burning plasmas through surrogate-based optimization in core transport solvers](#)." Nucl. Fusion 64 076034. (2024).

[5] Rodriguez-Fernandez, P., et al. "[Core performance predictions in projected SPARC first-campaign plasmas with nonlinear CGYRO](#)." Physics of Plasmas. 31 062501 (2024).





Commonwealth  
Fusion Systems