

Applications and opportunities for fast ML in fusion science and engineering

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Fast Machine Learning for Science Workshop
Southern Methodist University
October 4, 2022



Outline

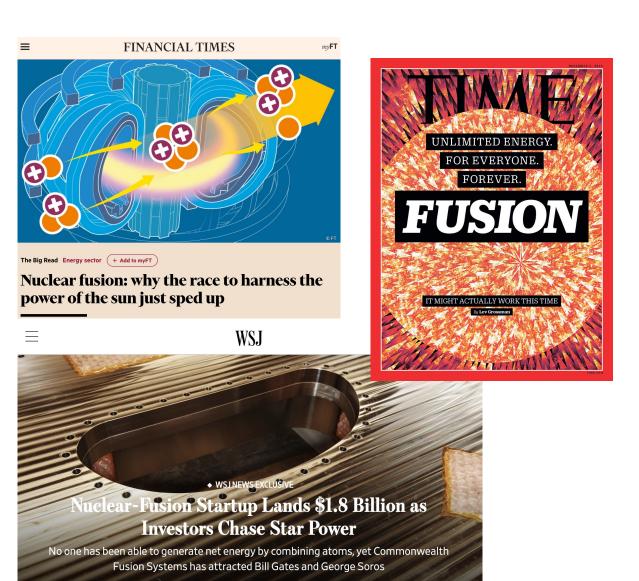


- Motivation and context
 - Public and private investment in fusion energy has dramatically increased
 - Complexity of fusion plasmas → large gap to necessary latencies
- Offline applications and opportunities in fusion energy
 - Design and engineering: Lightweight models for optimization
- Online applications and opportunities in fusion energy
 - Sensor fusion: Sparse, external, nonlinearly coupled
 - Active control: Tight latency (< ms), advanced control
- Summary



You may have heard about fusion in the news over the last few years







NUCLEAR-FUSION ENERGY RECORD

The Joint European Torus has doubled the record for the amount of energy made from fusing atoms. s being built in southern France. Tritium is a are, radioactive isotope of hydrogen; when it produce many more neutrons than do reac ions between deuterium particles alone. That prepare the machine for the onslaught, Tritium

In an experiment on 21 December 2021, IET's double the 21.7 megajoules released in 1997 over around 4 seconds. Although the 1997 ower', that spike lasted for only a fraction of second, and the experiment's average power ernanda Rimini, a plasma scientist at the CCFE who oversaw last year's experimenta campaign. The improvement took 20 years ware upgrades that included replacing the tokamak's inner wall to waste less fuel, she says

Producing the energy over a number of heating, cooling and movement happening inside the plasma that will be crucial to run ITER, says Rimini.

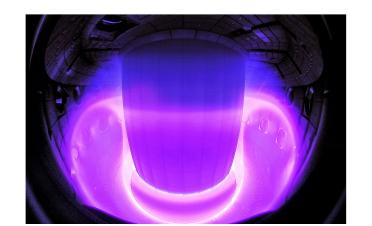
Five seconds "is a big deal", adds Proll. "It is really, really impressive Last year, the US Department of Energy's



WIRED BACKCHANNEL BUSINESS CULTURE SEAR IDEAS SCIENCE SECURITY

DeepMind Has Trained an AI to Control Nuclear Fusion

The Google-backed firm taught a reinforcement learning algorithm to control the fiery plasma inside a tokamak nuclear fusion reactor





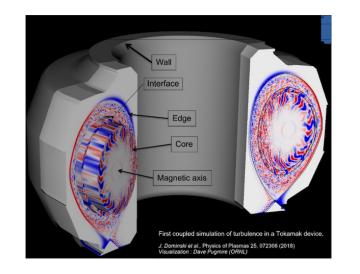
Ok, so what changed?

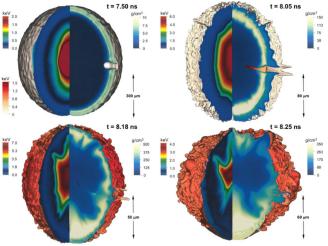


- Computing power
 - Simulations are crucial to understanding fusion plasmas
 - Many models have only recently become tractable









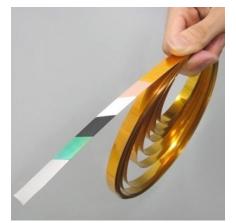


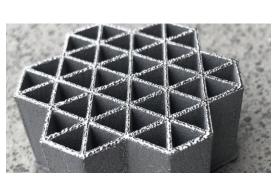
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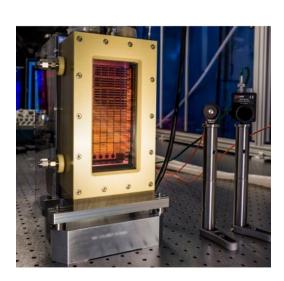


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 - Simulations are crucial to understanding fusion plasmas
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- New technology
 - HTS magnets, diode lasers, etc.
 - Advanced manufacturing, power electronics, etc.











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- New technology
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- Global scientific consensus on readiness
 - Private sector interest (> \$4B of investment)
 - Renewed public sector interest







Source: Fusion Industry Association (2022)

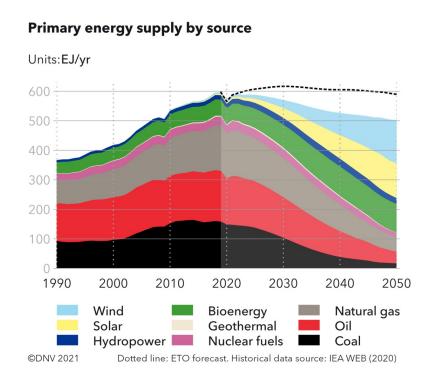


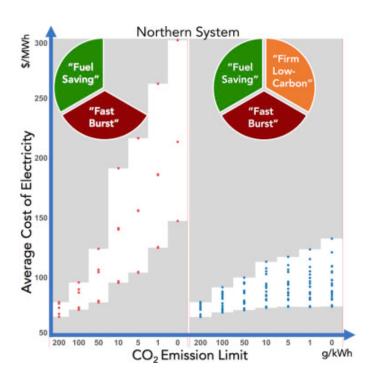


Fusion is now being actively pursued to accelerate access to a zero-carbon future



- Provide a dispatchable source of carbon-free energy
 - Reduce storage/overbuild requirements to cover seasonality
 - High-quality process heat for non-electrical applications
- Enables a paradigm shift to continued growth in energy usage







Many private and government projects are now aiming for fusion demonstrations in the near term



- Both private and public projects are actively building fusion demonstrations
 - Global enterprise with billions of \$/year in total investment
- A variety of methods and designs are being pursued
 - Public \rightarrow lower risk; Private \rightarrow higher risk
 - Net-energy demonstrations as early 2023







^{2- &}quot;Final report of the committee on a strategic plan of U.S. burning plasma research," The National Academies Press (2019)

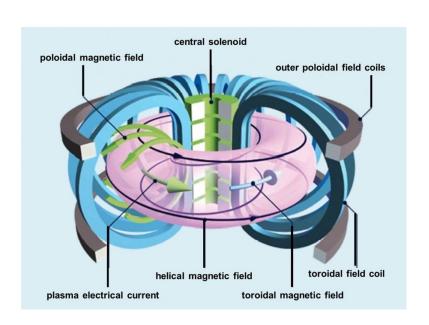
^{3- &}quot;European Research Roadmap to the Realisation of Fusion Energy," EUROfusion (2018)

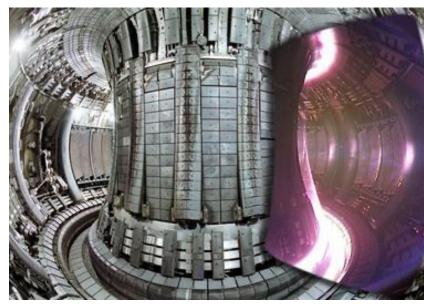


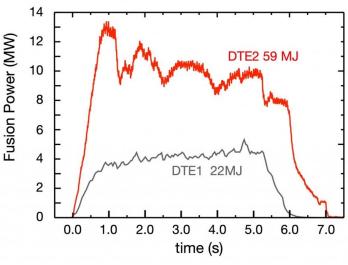
This talk will focus on considerations for the tokamak magnetic confinement concept



- "Magnetic confinement" is the leading technique
 - Accounts for bulk of private and public investment
- The tokamak is the leading configuration
 - Has demonstrated large amounts of fusion power
- Considerations are common to all fusion concepts with different weights







WEDNÉSDAY, NOVEMBER 9, 1994

Experimental Fusion Reactor At Princeton Sets a Record

By The Assemied Press

An experimental fusion reactor a Princeton University has set anothe world record by generating 10.7 mi lion watts of power, considered crucial step toward developing ner commercial fusion reactors.

ommercial fusion reactors.
The one-second burst of energy
rom the Tokamak Fusion Test Rector, enough to power 2,000 to 3,000
omes momentarily, was "a major
inector of the university's Plasma
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hone interview on Saturday, in
fasy, the reactor produced 9.2 milon watts of energy.

"The T.F.T.R. team has demontrated successful use of practical sion fuels, exceeded fusion power oals and reached world-record temeratures," Dr. Davidson said. Officials of the laboratory pre-

Officials of the laboratory p sented the results yesterday in M neapolis at the annual meeting of i Division of Plasma Physics of i American Physical Society. Fusion, the process that now

Fusion, the process that power the Sun, occurs on Earth when special forms of lightweight atoms lik hydrogen gas are slammed togethen at very high temperatures an fused, releasing energy. Fission which powers existing commercia nuclear reactors, involves breakin apart very heavy atoms like uranium but produces far more danger ous radioactive byproducts than fusion.

The Tokamak reactor at P ton, designed in 1976, was inten produce 10 million watts of he reactor set its first record in becomber when it produced the quivalent of 6.4 million watts of

Dr. Davidson said the results will p scientists and engineers design

e reactors.

the better information that we brovide as input to the design of facilities, the better their dewill be, and the sooner practical in will be developed and the attractive the power source be." Dr. Davidson said. "This is

The reactor, operated by the Plasshylises Laboratory, is schedd to be shu down and dismantled sy fall. Some of its auxiliary says are expected to be reused for next planned experimental reacat the laboratory, which is concred the country's leading center fusion research.

ts replacement, the \$750 million kamak Physics Experiment, is der design and is scheduled to be

A third experimental reactor, the billion International Thermonusar Experimental Reactor, is a int project of Russia, Japan, the pited States and a European coaliin. It is scheduled to be ready for creation by 2005.

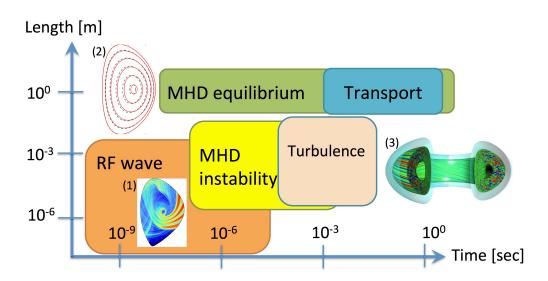
Company News: Tuesday through Saturo Business Day

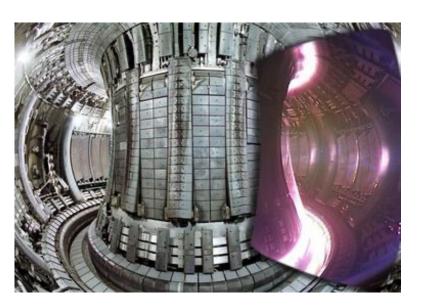


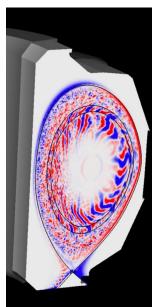
Fusion plasmas are a complex system that challenges understanding, prediction and control



- Multiscale nonlinear dynamics
 - Coupled multispecies system + E-M
 - Turbulence driven by steep gradients
- First-principles models are rarely tractable for online applications and design optimization
 - Even simplified/linearized models too slow
- Diagnostic access is extremely limited
 - Almost exclusively external diagnostics (eg. optical)
- Sensor fusion required for most signals
 - Deconvolve and/or localize measurements
- Strict latency requirements (ms or faster)





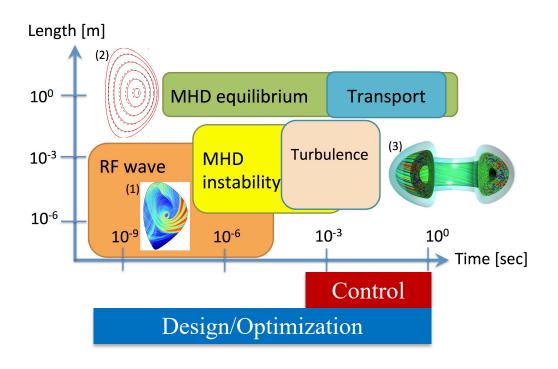




Machine learning holds promise to help accelerate the realization of fusion energy



- Commercial systems have different needs for models used in design and control
 - No more research "compromises"
 - Predict first not model after
- Models must be fast
- Models must be accurate
- Models must be robust
- Fertile ground for this community!





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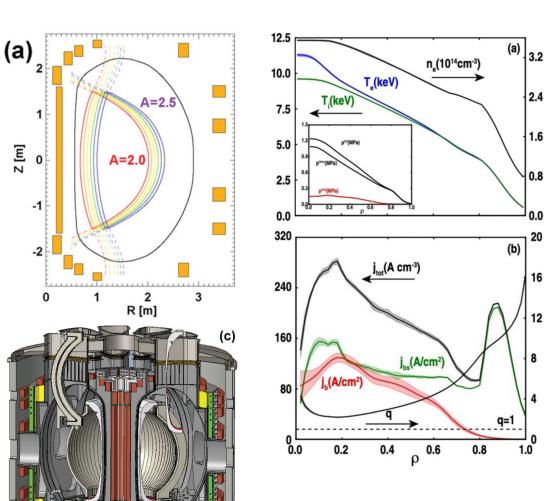




Design point optimization is critical to performance and economics



- Fusion production depends on *T* and *n* profiles
 - Must be in equilibrium (fast and slow)
- Energy deposition and transport is complex
 - Sources: fusion products, external heating, etc.
 - Sinks: turbulent transport, radiation, etc.
 - Many complex bifurcations/phase transitions
- Core and surrounding system are coupled
 - Magnets set plasma shape
 - RF waves refract through plasma
- Large high-dimensional optimization space
 - Blanket required for energy and fuel cycle
 - Maintenance needs to be considered

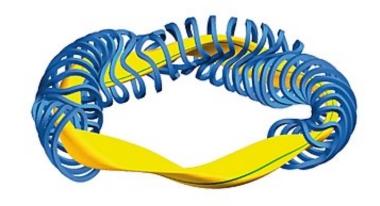


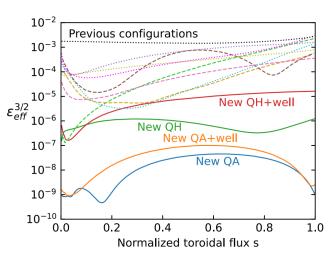


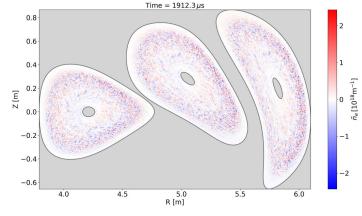
The stellarator concept focuses heavily on confinement optimization through design



- By introducing 3D fields stellarators can confine fuel without current flowing in the plasma
 - High-dimensionality optimization space
- Transport of heat and mass is very sensitive to magnetic field geometry
 - Neo-classical losses: particle drifts
 - Plasma turbulence: global nonlinear problem
- Machine learning approaches are being investigated to accelerate optimization
 - Transport calculation (or proxy) required





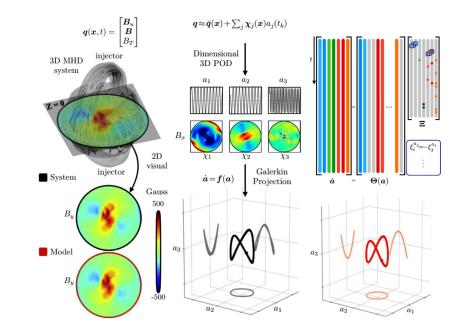


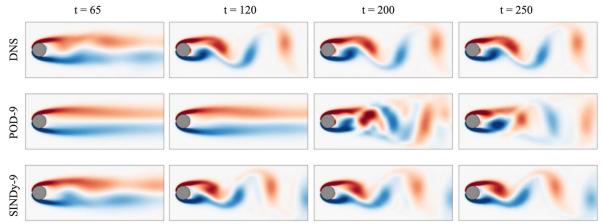


Model-discovery methods can build fast reduced-order models of nonlinear plasmas



- Reduced-order methods have potential for developing fast surrogate models
 - ODE, ML, etc.
- The SINDy method has shown promise on reproducing nonlinear plasma dynamics
 - Build nonlinear ODEs directly from data
 - Naturally supports physics-informed constraints
- Physics-informed constraints can significantly improve model quality
 - Reduce required training data
 - Enforce local or global stability







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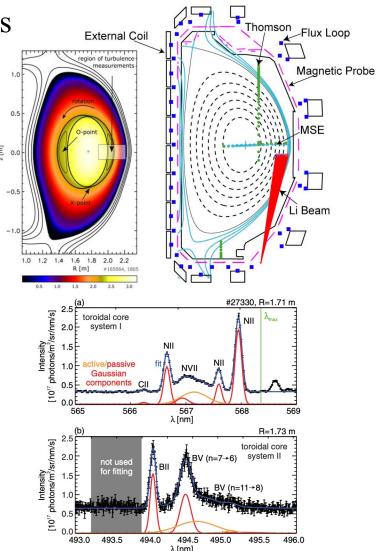


Building a consistent picture of equilibria and dynamics from diagnostics is complicated



• Sparse diagnostic set necessitates integration of multiple signals

- Plasma state must be reconstructed in time
- Even fewer signals in future reactors
- Optical and other EM-based diagnostics are complex
 - Nonlocal measurements: equilibrium-dependent sightlines
 - Convolve many fundamental quantities (T, n)
 - Complex spectrographic landscapes
- Still early days for applying machine learning to this area
 - Restrict ourselves to things we can observe directly (next section)
 - Lump this step in to a larger NN (following section)
- Accurate sensor fusion can make some control easier
 - Ability to work directly with primary variables





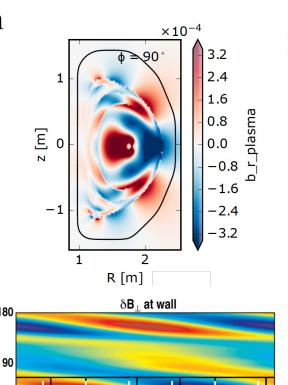
Active control of instabilities in tokamaks requires low-latency pipelines

Poloidal angle θ (Deg.)

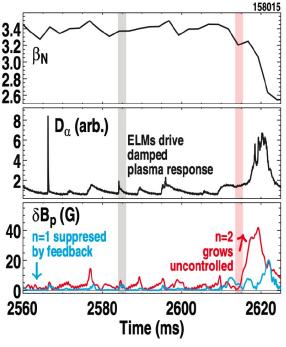


• Instabilities can degrade performance in fusion plasmas

- Avoided by reduced performance
- Suppressed by active control
- Active control requires a low-latency pipeline
 - Mode growth rates are on ms timescales
 - GPUs are frequently used in community
- Plasma response can be complex, motivating optimal control
 - Opportunity for ML system response models



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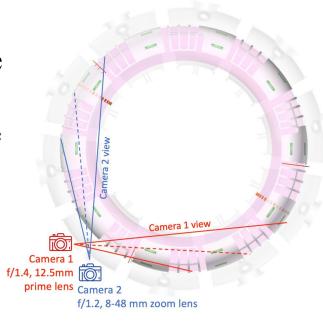


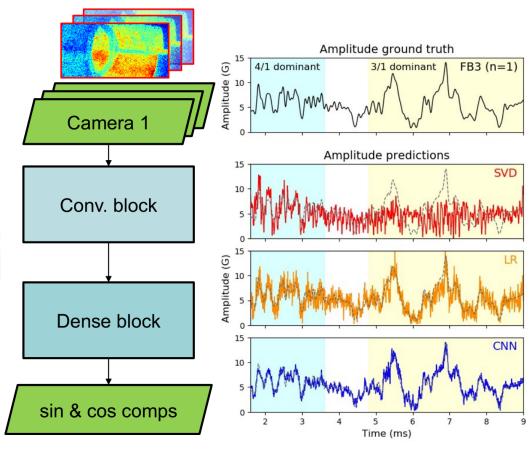


We are now working with this community to leverage tools for low-latency ML



- Diagnostic access will become more restrictive in reactor environments
 - Remote observations only, complex emission functions
- ML applies a promising path to fast mapping from signals to desired quantities
- We are working to use optical cameras to observe mode phase/amplitude
 - Convolves mode with edge conditions, reflections, etc.
 - HLS4ML enables implementation on camera FPGA



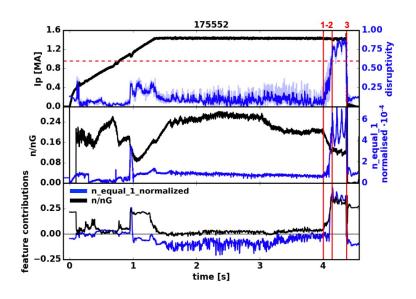


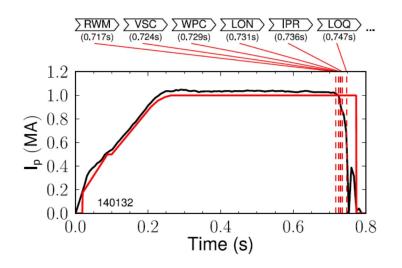


Accurate prediction of plasma disruptions is a significant open question for tokamaks



- A disruption is a rapid termination of plasma current
 - Caused by a range of phenomena: Instabilities, loss of control, etc.
 - Can damage device → more frequent maintenance
- Avoidance and/or mitigation requires long lead time
 - Causality is complex: nonlinearity, multi-event chains
- Machine learning approaches have shown promise
 - Classification: random forest, NN, etc.
 - Online usage requires robust, low-latency implementations
- Very low false negative tolerance $(\rightarrow 0)$
 - Very asymmetric datasets
 - Want as close to day 1 as possible: transfer or virtual learning



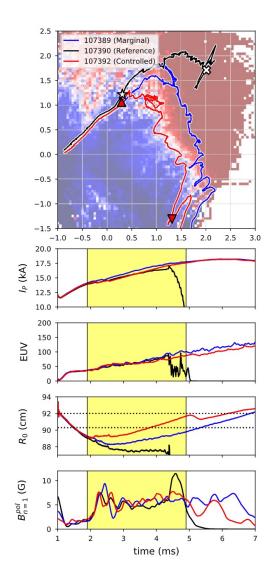


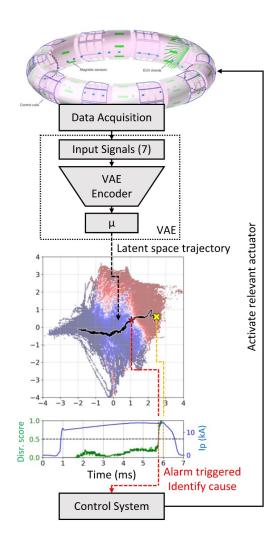


A VAE-based method has shown promise for both detection and avoidance of disruptions



- Diagnostics form a high-dimensional observation space
 - Important dynamics are expected to behave with shared low dimension structure
- A VAE was trained with a 2D latent space representation from 7D input data
- Successfully demonstrated the prediction and avoidance of disruptions
 - Stability boundaries were identified in latent space
 - Local gradients used to identify actuator outputs
- Further study is underway
 - More signals, larger devices



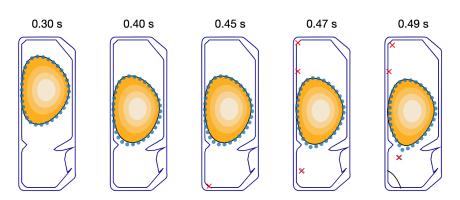


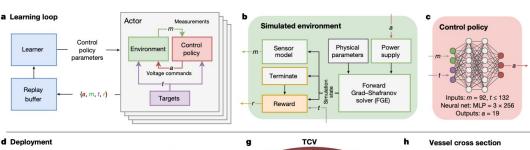


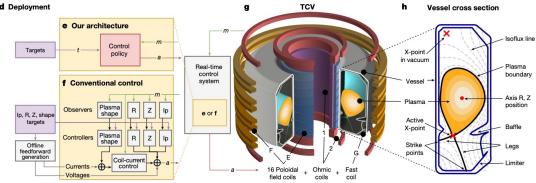
Reinforcement learning was recently demonstrated to build an end-to-end controller



- EPFL in collaboration with Google demonstrated a NN-based controller built using reinforcement learning
 - Successfully controlled real plasmas in TCV
 - Performed better than existing hand-tuned controller
- Requires fast, accurate training environment
 - Approaches optimized for data paucity
 - Possible application for reduced-order models
- Online implementation using CPU
 - < 10 ms latency requirements
 - Other applications require lower latency
 - Additional diagnostics require more throughput









Summary



- The world is now working to realize fusion as an energy source in the near term
 - Public and private sectors are moving together
- Many applications within the fusion space require fast nonlinear models
 - Design optimization: Fast surrogate models and interpolation over high-dimensional datasets
 - Sensor fusion: Integrate multiple signals with nonlinear dependence into unified state
 - Active control: Surrogate models for system response and/or end-to-end controllers
- Machine learning methods can (will?) play an integral part in fusion's realization
 - Models need to be fast, accurate and robust
 - Hardware pipelines will likely be required to satisfy latency requirements
- Interested? Talk to your local fusion scientist or I can point you in the right direction



Funding and collaboration acknowledgements



Funding for this work provided by:





Work performed in collaboration with:













Thank you for your attention



Questions?