



# PROCEEDINGS OF THE WORKSHOP ENABLING TECHNOLOGIES FOR DIGITAL TWIN APPLICATIONS FOR ADVANCED REACTORS AND PLANT MODERNIZATION

Virtual Workshop September 14-16,2021

Date Published:

Prepared by:

- J. Carlson
- D. Eskins
- R. Gascot
- R. Iyengar
- C. Ulmer

U.S. Nuclear Regulatory Commission

Research Information Letter
Office of Nuclear Regulatory Research

#### **Disclaimer**

This report was prepared as an account of work sponsored by an agency of the U.S. Government. Neither the U.S. Government nor any agency thereof, nor any employee, makes any warranty, expressed or implied, or assumes any legal liability or responsibility for any third party's use, or the results of such use, of any information, apparatus, product, or process disclosed in this publication, or represents that its use by such third party complies with applicable law.

This report does not contain or imply legally binding requirements. Nor does this report establish or modify any regulatory guidance or positions of the U.S. Nuclear Regulator Commission. This report is not binding on the Commission	rt y



#### **EXECUTIVE SUMMARY**

The Office of Nuclear Regulatory Research (RES) at the U.S. Nuclear Regulatory Commission (NRC) has initiated a future focused research project to assess the regulatory viability of digital twins for nuclear power plants. The objectives of this project are to:

- Understand the current state of the technology and potential applications for the nuclear industry,
- · Identify and evaluate technical issues that could benefit from regulatory guidance, and
- Develop infrastructure to support regulatory decisions associated with digital twins.

As a follow-on to the workshop hosted in December 2020 (ML21083A132), RES sponsored the Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization 2021 Online Workshop. The workshop was hosted by Idaho National Laboratory (INL) in collaboration with Oak Ridge National Laboratory (ORNL), the Department of Energy's (DOE) Advanced Research Projects Agency-Energy (ARPA-E), and the Electric Power Research Institute (EPRI) and was held September 14-16, 2021.

The 3-day workshop was composed of five technical and panel sessions with 29 presenters from a wide range of national and international organizations, including universities, national laboratories, government agencies, nuclear vendors, nuclear industry, advanced reactor developers, and digital twin developers. With 324 participants from across the globe, the workshop provided a forum for nuclear industry and digital twin stakeholders to discuss the application of digital twins and digital twin enabling technologies such as advanced sensors and instrumentation, data analytics, machine learning and artificial intelligence in the current light water reactor (LWR) fleet and advanced reactor designs. The workshop also included an overview of the next steps toward regulatory realization of digital twins in the nuclear industry.

The workshop had two main purposes: (1) to review and exchange information on the current applications of digital twin enabling technologies, and (2) to identify necessary steps toward regulatory realization of digital twins. The workshop sessions covered the following topics: industry applications to digital twins in nuclear, advanced sensors and instrumentations, use cases of digital twin enabling technologies in nuclear power plants, digital twin enabling technologies in advanced reactor applications, and steps toward regulatory realization of digital twins.

On the first day of the workshop, Tuesday, September 14, 2021, Mr. Ray Furstenau, Director of RES, opened the workshop with introductory remarks and moderated a panel session on industry applications of digital twins in nuclear and Dr. Hasan Charkas from EPRI moderated a technical session on advanced sensors and instrumentations. On the second day of the workshop, Wednesday, September 15, 2021, Dr. Gene Carpenter representing Advanced Research Projects Agency - Energy (ARPA-E) moderated a technical session on use cases of digital twins enabling technologies in nuclear power plants and Ms. Angela Buford, Office of Nuclear Reactor Regulation (NRR), NRC, moderated a technical session on digital twin enabling technologies in advanced reactor applications. On the third day of the workshop, Thursday, September 16, 2021, Mr. Eric Benner, NRR, moderated a panel session on steps toward regulatory realization of digital twins and delivered the workshop closing remarks.

The following are some major takeaways from the workshop:

#### **Technical Challenges/Opportunities**

- The nuclear industry and national laboratories have demonstrated interest and are
  pursuing the use of digital twin technologies and have realized advanced capabilities in
  preventive maintenance optimization, work order data analysis, anomaly detection and
  diagnosis, and real-time radiation monitoring.
- There is significant interest/effort in the development of advanced sensor technologies and applications including wireless communication, multi-modal sensing, condition-based monitoring and maintenance, and operation in harsh environments, especially those introduced by advanced reactor designs (e.g., extreme temperature, radiation, corrosivity).
- Many advanced reactor developers are designing plants integrated with digital twins (DTs) throughout their lifecycle to facilitate improved decision making and greater operational flexibilities (e.g., potential dynamic operating envelope).
- Challenges exist in the following areas: real-time reduced-order or surrogate models, data production and integration, virtual prototyping, autonomous control, and sensor requirements.

#### Regulatory Challenges/Opportunities

- There are three main categories of potential DT use: 1) use by industry for inherent benefits (e.g., improved design, construction, operations and maintenance), 2) use by industry as a tool for regulatory compliance (e.g., licensing submittals, safety analysis), and 3) use as an NRC regulatory tool (e.g., shared source of plant information, enabler of iterative design approvals and just-in-time regulation).
- Industry and regulators need to develop agreed upon guidance and frameworks for acceptance of DT applications that is consistent, explicit, and enables the use of DTs as an additional avenue for meeting the intent of existing regulations.
- One approach to building confidence in DT technology an important aspect for acceptance and adoption of DTs – is pioneering DT applications with non-safety components or systems and demonstrating acceptable performance prior to safety-related applications.
- DTs have the potential to enhance NRC inspection activities, including automated regulatory compliance testing and on-demand access to high-fidelity plant information.

All presentations slides from this workshop are available in the NRC's Agencywide Documents Access and Management System, under Accession No. <u>ML21342A121</u>.



#### **TABLE OF CONTENTS**

E	XECUTIVE S	UMMARY	V
1	Day 1 Pres	entations	1
•		Session 1: Industry Applications of Digital Twins in Nuclear	
	1.1.1	Innovation Journey to a Brighter Future	
	1.1.2	EPRI's Digital Twin Related Activities for Nuclear Applications	2
	1.1.3	Westinghouse Perspectives	
	1.1.4	Exelon Nuclear Innovation Projects Overview	
	1.1.5	ARPA-E GEMINA Portfolio and Digital Twins	
	1.1.6	Digital Twin Monitoring for Advanced Reactors and Plant Modernization	2
		1 2: Advanced Sensors and Instrumentations	
	1.2.1	Advanced Sensors and Instrumentation for Digital Twin Applications	
	1.2.1	Non-intrusive Temperature and Pressure Wireless Sensor and	
	1.2.2	Transceiver System for Extreme Environment Applications	3
	1.2.3	Data Analytics and Remote Monitoring Integration	
	1.2.4	Digital Twin Impact on I&C Systems Development for Xe-100	
	1.2.5	Online Monitoring (OLM) Implementation to Extend Transmitter	⊤
	1.2.0	Calibration Intervals in Nuclear Facilities	4
2	Day 2 Pres	entations	5
_		n: Use Cases of Digital Twin Enabling Technologies in Nuclear Power	
		t	5
	2.1.1	Integrated Risk-informed Condition-based Maintenance Capability and	
		Automated Platform	5
	2.1.2	Al Driven Scalable Condition-based Predictive Maintenance Strategy	
	2.1.3	Thermal Performance Management for Nuclear Power Plant with Digital	
		Twins	6
	2.1.4	Digital Twin of a Real-time Radiation Monitoring Network	
	2.1.5	Non-destructive Examination (NDE) 4.0 and ML for In-Service	
		Inspections	6
	2.2 Session	n 2: Digital Twin Enabling Technologies in Advanced Reactors	
		ications	7
	2.2.1	Xe-100 Digital Twin Technologies Overview	7
	2.2.2	Humble Al for Reliable Machine Learning-Based Health Twin	8
	2.2.3	Enabling Technologies for Digital Twins Applications for the KP-FHR	
	2.2.4	High-Fidelity Digital Twins for BWRX-300 Critical Systems	
	2.2.5	Molten Salt Loop Development Acceleration with Disturbed Single	
		Crystal Harsh Environment Optical Fiber-Sensors	9
	2.2.6	Digital Twin to Production Reactors, The Simulation Continuum	
		,	
3	Day 3 Pres	entations	10
	3.1 Panel S	Session: Steps Toward Regulatory Realization of Digital Twins	10
	3.1.1	Digital Twins – Regulatory Viability	
	3.1.2	Xe-100 Licensing Perspectives: Steps Toward Realization of Digital	
		Twins	11
	3.1.3	Using Digital Twins to Support Regulations	11
	3.1.4	Nuclear Energy Institute (NEI) Perspectives on Digital Twins	
	3.1.5	Kairos Perspective	

3.1.6	Westinghouse Perspective	12
APPENDIX A	Workshop Attendees	A-1
APPENDIX B I	Presentation Slides	B-1

Page intentionally left blank

#### 1 DAY 1 PRESENTATIONS

#### 1.1 Panel Session 1: Industry Applications of Digital Twins in Nuclear

In this panel, representatives from the nuclear industry provided a perspective on how digital twins technology has been implemented or may be implemented. The panel provided insights of the potential challenges and benefits of the identified digital twins enabling technologies and introduced additional important enabling technologies. Potential industry strategies for accommodating these technologies within the nuclear lifecycle and any new competencies or technical disciplines needed to support digital twin technologies were discussed as well.

Participants on this session presented the following initiatives/perspectives about the industry applications of digital twins in nuclear:

- Xcel Energy discussed their Cap Intelligent Advisor (artificial intelligence (AI) and machine learning (ML)) and Digital Ops Factory programs which facilitates the search, entry, and analysis of data.
- The Electric Power Research Institute (EPRI) presented an update of their ongoing project that aims to explore benefits, challenges, and potential use cases for advanced reactors and will establish industry guidelines, best practices, and recommendations for implementing digital twins in advanced reactor life cycle management.
- Westinghouse Electric Company discussed practical aspects that can be implemented with the correct use of the digital twin technologies.
- Exelon Corporation presented their remote monitoring project. This initiative uses wireless sensors to support plant monitoring.
- The Department of Energy (DOE) Advanced Research Projects Agency-Energy (ARPA-E) presented several initiatives that demonstrate that the use of digital twin enabling technologies can decrease operations and maintenance (O&M) labor cost and map temperatures of reactors components, among other benefits.
- Metroscope introduced their software that diagnosed equipment problems for all types of plant auxiliary systems.

The presentations slides for Day 1 can be found <u>here</u> and in the Agency Documents Access and Management System (ADAMS) under <u>ML21342A122</u>.

#### **Presentations**

#### 1.1.1 Innovation Journey to a Brighter Future

Patrick Burke, Vice President of Nuclear Strategy Xcel Energy

<u>Presentation Overview</u>: This presentation described several initiatives that Xcel Energy is working on to implement the use of ML/AI to facilitate the search, entry, and analysis of operational data. Other potential efforts were discussed.

#### 1.1.2 EPRI's Digital Twin Related Activities for Nuclear Applications

Craig Stover, Program Manager Advanced Nuclear Technology EPRI

<u>Presentation Overview</u>: During this presentation EPRI discussed in more details the four top innovation areas of interest at this moment. Those are ML and big data, reliable data sharing, digital twinning, and advanced manufacturing. All these areas have great potentials for several possible use cases in the existing and next generation of nuclear power plants.

#### 1.1.3 Westinghouse Perspectives

Scott Sidener, Consulting Engineer, Digital Innovation Westinghouse Electric Company

<u>Presentation Overview</u>: There was no presentation from Westinghouse. Scott Sidener discussed the company's perspectives on digital twin technologies.

#### 1.1.4 Exelon Nuclear Innovation Projects Overview

Rick Szoch and Tim Alvey Exelon Corporation

<u>Presentation Overview</u>: Exelon Corporation presented their Innovation Culture and Digital Transformation initiatives. Remote monitoring is their main effort currently, and this project creates a new wireless infrastructure that provides a method to utilize wireless sensors.

#### 1.1.5 ARPA-E GEMINA Portfolio and Digital Twins

Charalampos Andreades, Technology to Market Advisor ARPA-E

<u>Presentation Overview</u>: ARPA-E introduced the Generating Electricity Managed by Intelligent Nuclear Assets (GEMINA) program and highlighted benefits that digital twins can provide to the industry, including the enabling technologies/capabilities. Dr. Andreades also discussed initiatives that are utilizing digital twin-related technologies.

#### 1.1.6 Digital Twin Monitoring for Advanced Reactors and Plant Modernization

Aurélien Schwartz Metroscope

<u>Presentation Overview</u>: Metroscope presented their software, which consists of a trusted and reliable digital twin-based program that can diagnose equipment problems for all types of plant auxiliary systems.

#### 1.2 Session 2: Advanced Sensors and Instrumentations

This session includes presentations focusing on advanced sensors and instrumentation with an emphasis on digital twin applications. The presentations demonstrated the development of new sensors and the greater digital integration of existing sensors. A recurring focus included continuous monitoring control system integration. There is a general goal to move to condition-based monitoring. It was noted that advanced sensors still require testing in the more extreme environments found in advanced reactors. Finally, it was shown how virtual sensors with a digital twin can be used to optimize instrumentation and controls (I&C) systems during the design phase.

Presenters in this session discussed the following use cases of digital twins:

- Iterative plant design using digital twins
- Digital twin virtual sensors to optimize instrumentation

Challenge identified by presenters:

• Validation of novel sensor performance under extreme conditions

#### **Presentations**

#### 1.2.1 Advanced Sensors and Instrumentation for Digital Twin Applications

Pattrick Calderoni, National Technical Director, Advanced Sensors and Instrumentation Manager, Measurement Science Department INL

<u>Presentation Overview</u>: The development efforts for advanced sensors were outlined. Advanced sensors include multi-point and multi-modal sensors that must be built to withstand the reactor environment. Advanced sensors will be integrated with control systems to provide real-time data and feedback. Specific sensors were discussed with the predominant feature being high-temperature operation, including in-core neutron flux, high-temperature thermocouples, ultrasonic sensors for temperature measurement and structural health monitoring, and optical fiber-sensors. Wireless technologies and power harvesting were also discussed

# 1.2.2 Non-intrusive Temperature and Pressure Wireless Sensor and Transceiver System for Extreme Environment Applications

Jorge Carvajal, Fellow Engineer Westinghouse Electric Company

<u>Presentation Overview</u>: Ongoing sensor development efforts were reviewed. Wireless sensors for temperature and pressure measurement were presented for use inside sealed fuel rods and inside seal dry storage steel cannisters. The fuel rod sensors would also measure pellet elongation and operate passively. These sensors provide real-time and continuous data. It is suggested that these sensors could be used to increase data availability to accelerate fuel qualification and technical analysis. Efforts are ongoing to test sensors at high temperature and in radiation environments.

#### 1.2.3 Data Analytics and Remote Monitoring Integration

Molly Strasser Xcel Energy

<u>Presentation Overview</u>: Xcel Energy presented their process for moving to a condition-based monitoring and maintenance scheme. Wi-Fi was installed to connect with sensors and smart devices in the hands of personnel. New sensors were added, and existing analogue gauges were integrated into the digital network. Sensing includes vibration, gauge readers, void monitoring, remote radiation mapping, valve position indication, acoustic monitoring for switchgear and transformers, and continuous thermal imaging. Advanced models were used for processing of continuously collected data.

#### 1.2.4 Digital Twin Impact on I&C Systems Development for Xe-100

Matthew Hertel, Senior Nuclear I&C Engineer X-enegy

<u>Presentation Overview</u>: The development of I&C systems for Xe-100 was presented. A digital twin was used in lieu of a physical plant. I&C system design was optimized using an iterative process using specifications, transient analysis, and physical systems.

## 1.2.5 Online Monitoring (OLM) Implementation to Extend Transmitter Calibration Intervals in Nuclear Facilities

Brent Shumaker and H.M. Hashemian Analysis and Measurement Services Corporation (AMS)

<u>Presentation Overview</u>: Online monitoring for transmitter calibration was presented. It was presented that transmitters were calibrated at every outage, but typically don't drift in that period. Instead, a condition-based maintenance system using online monitoring was implemented to only calibrate those sensors that had drifted. This system was implemented in Sizewell B since 2005, Vogtle units 1 and 2 since 2018, and the Advanced Test Reactor since 2015.

#### 2 DAY 2 PRESENTATIONS

# 2.1 <u>Session: Use Cases of Digital Twin Enabling Technologies in Nuclear Power</u> <u>Plant</u>

Even though digital twin technologies have not been fully implemented in the nuclear industry, they are gaining momentum and support. At this moment there are several projects ongoing to implement this technology into the industry. From risk-informed approaches to optimize equipment maintenance to a specific software that use a combination of digital twin and AI to provide a diagnosis, there are several successful cases in the industry where this technology has been deployed. In this session, participants presented concrete examples and use cases of digital twin technologies in the nuclear industry.

Presenters in this session discussed the following use cases of digital twins:

- Preventive maintenance optimization
- Work order data analysis
- Identification of fault signatures
- Development of an automated digital platform
- Detection of high-pressure heater leak
- Detection of condensate collector tank leak
- Detection of condenser losses
- Real-time radiation monitoring network (this case is still in the experimental phase)

#### Challenges identified by presenters:

- Lack of synchronization
- · Lack of good quality data for different fault modes
- Data imbalance
- Lack of model generalization
- Qualification for ML process

The presentations slides for Day 2 can be found here and in the ADAMS under ML21342A123.

#### **Presentations**

## 2.1.1 Integrated Risk-informed Condition-based Maintenance Capability and Automated Platform

Matthew Yarlett, Project Engineer Westinghouse Electric Company

<u>Presentation Overview:</u> Presentation of the research results of the project between PKMJ Technical Services, INL, and Public Services Enterprise & Group (PSEG) based on development of condition-based monitoring models and integrating those models into a digital platform for use by the nuclear industry. This project integrated advancements in online monitoring and data analytics techniques with advanced risk assessment methodologies. Preventive maintenance optimization, work order data analysis, and

identification of fault signatures, among others, were presented as successful accomplishments of the developed technique.

#### 2.1.2 Al Driven Scalable Condition-based Predictive Maintenance Strategy

Koushik A. Manjunatha, PhD Staff Research Scientist INL

<u>Presentation Overview</u>: Introduction of a decentralized AI approach using heterogeneous data from a nuclear power plant (NPP) asset to deploy condition-based predictive maintenance strategies. The approach is simple and scalable across different assets at the plant site and across the nuclear fleet. Predictive maintenance models, fault signature identification, fault probability were some examples of results that can be obtained using the technique. This presentation also highlighted several challenges that lead to inaccurate model interpretations.

#### 2.1.3 Thermal Performance Management for Nuclear Power Plant with Digital Twins

Christophe Duquennoy, PhD, Nuclear Fleet Thermal Performance Expert Électricité de France S.A. (EDF)

<u>Presentation Overview</u>: This presentation highlighted use cases on EDF experience. The Metroscope software, which combines a digital twin of the process with AI to perform a diagnosis in operations, is the program utilized by EDF. Early detection of condenser collector tank leaks, condenser loses, and heaters tube rupture are examples of diagnosis activities predicted by the software.

#### 2.1.4 Digital Twin of a Real-time Radiation Monitoring Network

Richard McGrath, Principal Technical Leader in Radiation Safety Group EPRI

<u>Presentation Overview</u>: This presentation explained the ongoing results of the EPRI's project regarding the use of a digital twin as a real-time radiation monitoring network. This EPRI NextGen RP Project have the potential of optimize the way radiation protection is performed in NPP. The project consists in two phases: Phase 1 - "Demonstration of geophysical application for analyzing radiological survey data", and Phase 2 - "Apply machine learning to geophysical radiological survey application".

#### 2.1.5 Non-destructive Examination (NDE) 4.0 and ML for In-Service Inspections

likka Virkkunen, Professor Aalto University

<u>Presentation Overview</u>: An introduction to a new system that increases the reliability of the use of NDE for inspections through the application of ML. This new procedure can be integrated to existing NDE methods and allow connected systems to aggregate data. Tools like edge computing and ML are vital for the implementation. The biggest hurdle is the qualification of the system.

# 2.2 <u>Session 2: Digital Twin Enabling Technologies in Advanced Reactors</u> Applications

The NRC is preparing to review and regulate a new generation of advanced non-light water reactors, and this session covers the intersection between digital twin enabling technologies and advanced reactor applications. Many companies are planning to use digital twins not only for operation and maintenance, but also in the design, licensing, construction, and decommissioning phases of the NPP lifecycle. Several different approaches to the use of digital twin technologies have been presented, each involving iterations between experiments, simulations, prototypes, and digital twin models. To fully realize the benefits of digital twins, developers need high-quality training data, high-speed surrogate and reduced-order models that can run in real time, and considerable numbers of legacy and advanced sensors to provide the necessary information to the models so that the digital twin can inform predictive maintenance and optimize operational efficiency.

Presenters in this session discussed the following use cases of digital twins:

- Anomaly detection with ML
- Informing maintenance & security in design space
- Detection (fault or anomaly detection), diagnosis (place faults in classes), and health estimation and forecasting (includes performance prediction)
- Health evaluation and analysis in real-time, semi-autonomous control room, operational reliability and diagnostics, and safety hazard intervention and limiting defense (detect and mitigate cyber threats)
- Recognize and address mechanical and thermal fatigue failure modes which drive O&M activities and costs
- Use of DT to fully model all aspects of the system, from physics to controls to properties controlling the device

#### Challenges identified by presenters:

- Dealing with the statistically impressive results of ML, which might be individually unreliable
- Lack of good quality data for training algorithms and building the DT
- Integration of heterogeneous models in a DT
- Uncertainty quantification methodology
- Getting right balance between digital and physical models in the development process

#### **Presentations**

#### 2.2.1 Xe-100 Digital Twin Technologies Overview

Ian Davis, Senior Digital Twin System Engineer X-energy

<u>Presentation Overview</u>: X-energy is using digital twin technologies to support the ongoing development of its advanced high-temperature gas-cooled reactor (HTGR) design and plans to use these technologies to support future operation and maintenance of Xe-100 sites. Some of the inherent characteristics of this advanced reactor concept, such as the robustness of the tri-structural isotropic (TRISO) particles that contain the fuel, low power density of the core, and strong negative temperature coefficient of reactivity, reduce safety concerns and offer

opportunities for implementation of new technologies. Some of the digital twin tools include three-dimensional models that can be explored with augmented reality/virtual reality (AR/VR) and coupled to the detailed operator training simulator, a plant historian (includes dashboards and visualizations), and custom ML/Al models for aspects such as conducting predictive maintenance and optimizing operational efficiency.

#### 2.2.2 Humble AI for Reliable Machine Learning-Based Health Twin

Dr. Nurali Virani, Lead Scientist General Electric Company (GE)

Presentation Overview: Machine learning-based health twins can be used for fault detection, diagnosis, or health estimations of physical systems and components. A key aspect of using health twins-based automation for critical industrial infrastructure is the development of characterization regions of reliability and trust as safety and performance are paramount. GE has developed an AI program referred to as "Humble AI" that is aware of its own competence and improves its competence via learning. The crucial element of this program is the model competence evaluation, which analyzes model inputs, model internal representations, and model outputs, and identifies regions of trust, overlap/ambiguity, and extrapolation to get justification-based reliability. The AI was tested using Tennessee Eastman Process simulation data and achieved an overall accuracy of 76% on all the data it was shown; however, on 56% of the data, the AI program had an accuracy over 99%. This highlights the concept that an AI could be used to automate certain aspects of operation and maintenance (where the confidence and accuracy is high) and request human assistance outside of that region of confidence.

#### 2.2.3 Enabling Technologies for Digital Twins Applications for the KP-FHR

Anthonie Cilliers, Senior Management Kairos Power

<u>Presentation Overview</u>: Kairos Power plans to use robust digital twin technology in several aspects of the Kairos Power Fluoride Salt-Cooled High-Temperature Reactor (KP-FHR). Among these are systems for the following functions: safety hazard intervention and event limiting defense (KP-Shield) to provide a passive, robust, reliable safety shutdown capability; operational reliability and diagnostics (KP-Sword) for active plant control; health evaluation and analysis in real-time (KP-Heart) to provide intelligent health monitoring; semi-autonomous industrial grade human machine interface technology (KP-Sight) for a semi-autonomous control room. Kairos power will carry out the development process as follows: 1) small test facilities, 2) large test facilities, 3) prototypical facilities, and 4) commercial facilities. A key aspect of the design is to minimize the safety envelope using state-based plant information.

#### 2.2.4 High-Fidelity Digital Twins for BWRX-300 Critical Systems

Emilio Baglietto, Associate Professor of Nuclear Science and Engineering Massachusetts Institute of Technology (MIT)

<u>Presentation Overview</u>: Discussion of the use of high-fidelity digital models using computational fluid dynamics (CFD) to inform maintenance and operational decisions. This is especially useful in new applications that do not have a rich database from which to draw. MIT

has been able to demonstrate reduction of operating uncertainty through high-fidelity simulations and accurately predict velocity and temperature fluctuations responsible for fatigue. The STRUCT program developed by MIT was also able to capture complex phenomena driven by the formation and interaction of large turbulent structures that are strongly non-linear and not prone to "lumping" and generalization. Additionally, MIT's program demonstrated accelerations between 50 and 100 times compared to traditional large eddy simulations. These high-fidelity CFD models can be used to create the surrogate models used by a digital twin.

# 2.2.5 Molten Salt Loop Development Acceleration with Disturbed Single Crystal Harsh Environment Optical Fiber-Sensors

Michael Buric, Staff Scientist National Energy Technology Laboratory (NETL)

<u>Presentation Overview</u>: Single crystal optical fibers such as  $Y_3Al_5O_{12}$  (YAG) or sapphire are used for distributed temperature sensing to map high-radiation and/or high-temperature environments like liquid-fueled molten salt reactors (LFMSRs). The single crystal optical fiber technology can extend into nuclear harsh environments and provide data to not only guide reactor design and improvement through thermal efficiency, but also inform LFMSR transient response. This technology can be used to gather thousands of data points to map reactor coolant temperatures or other parameters, and preliminary testing indicates accuracy with temperatures up to  $1000^{\circ}$ C and a standoff distance up to 50 feet.

#### 2.2.6 Digital Twin to Production Reactors, The Simulation Continuum

Bob Urberger, Chief Software Engineer Radiant

Roger Chin, Software Architect Radiant

<u>Presentation Overview</u>: Radiant plans on using a digital twin as a common tool between regulators and developers to ensure common sources of information for aspects relevant to them. Throughout the development process, Radiant is advancing their design by using Nuclear Energy Advanced Modeling and Simulation (NEAMS) tools to run high-fidelity multiphysics reactor simulations, and then using the NEAMS results to create reduced- order models for their digital twin software. They iteratively use the high-fidelity NEAMS tools, reduced-order models, hardware-in-the-loop simulations, and subscale or full-scale prototypes to refine both their models and design. Radiant will be able to demonstrate the safety of various operations by using a digital twin run, hybrid simulation run, or full prototype run.

#### 3 DAY 3 PRESENTATIONS

#### 3.1 Panel Session: Steps Toward Regulatory Realization of Digital Twins

This panel is focused on the intersection between digital twins enabling technologies and regulatory activities. Panelists provided their unique perspectives and insights on how digital twin technology may be employed as a tool for both industry regulatory compliance and perhaps the NRC itself as well as insights into the regulatory outcomes, challenges, resources, and gaps, especially those that are unique or novel, associated with digital twin technologies and areas where regulatory processes should be focused to accommodate these technologies.

Presenters in this session discussed the following use cases of digital twins:

- Support visualization with linkage to 3D models with AR/VR, operator simulator training, access to plant historical data, and the development of Al/ML models
- Simulator certification, human factor evaluation, and operator workload reduction
- Support optimized security staffing via security analysis and what-if security scenarios
- Support analysis submitted to NRC for human factor evaluation and staffing
- Standardize internal documentation, provide visualizations, and automate analysis
- Visualization and creation of "virtual sensors" to provide greater insights into the actual plant state
- Enable efficient and lower risk design with an iterative design process, shifts to virtual design space, and hardware-in-the-loop development and testing
- Facilitate the licensing process by structuring documentation and submittal information
- Prediction of future operational states using faster-than-real-time simulation
- Provide greater operational and regulatory flexibility by calculation of a dynamic operating envelope
- Operational anomaly detection
- Use of AI/ML for plant control functions, event prediction, equipment remaining useful life estimates, and sensor drift detection
- Provide common, rich data for industry and regulators

#### Challenges identified by presenters:

- Establishment of an appropriate regulatory guidance for approval of digital twin technologies and applications such as AI/ML control systems, autonomous systems, dynamic operating envelopes, reduced-order and multi-domain models, and reduced plant safety footprints
- Implement a streamlined regulatory process and appropriate common information interface to enable rapid regulatory response to plant design changes
- Determine appropriate verification, validation, and uncertainty quantification processes for digital twin models

The presentations slides for day three can be found here and in the ADAMS under ML21342A124.

#### **Presentations**

#### 3.1.1 Digital Twins - Regulatory Viability

Jeremy Bowen, Deputy Director, Division of Engineering, Office of Nuclear Regulatory Research NRC

<u>Presentation Overview</u>: Discussion of NRC's ongoing digital twin research activities which include technical preparedness, regulatory readiness, assessment of standards, and communication and knowledge management; digital twin project completed activities, publications, and takeaways thus far, and active and future project tasks.

#### 3.1.2 Xe-100 Licensing Perspectives: Steps Toward Realization of Digital Twins

Tom Braudt, Licensing Engineer X-energy

Steve Vaughn, Licensing Engineer X-energy

<u>Presentation Overview</u>: Discussion of Xe-100 digital twin tools and their uses for digital control and monitoring, predictive maintenance, high-fidelity simulator development, human factor engineering, operator workload reduction, training, dose reduction, security, fire detection, and to support plant state awareness in different production modes; plans to use DT help analyze operator workload and staffing methodology for topical report to be submitted to the NRC; and plans to use Al/ML for control, event prediction, equipment remaining useful life estimates, and sensor drift detection.

#### 3.1.3 Using Digital Twins to Support Regulations

Paul Keutelian, Radiant

<u>Presentation Overview</u>: Discussion of Radiant's vision to make nuclear portable and use DTs to maximize the speed of design iteration; support regulatory intents to protect personnel, the environment, and hardware, and prove the protections; dynamically assess risk and risk-informed decisions; act as a common source of information for both regulators and developers; and standardize internal documentation, provide visualizations, and automate analysis.

#### 3.1.4 Nuclear Energy Institute (NEI) Perspectives on Digital Twins

James Slider, Technical Advisor NEI

<u>Presentation Overview</u>: Discussion of NEI's purpose and organization; the importance of a common language for DTs; challenges presented by DT model complexity, real-time inputs, and NRC acceptance and usage for regulatory decisions; NEI's promotion of advanced ideas and best practices within the industry; and NEI's goal to work with industry and the NRC to

realize the benefits of DT technologies by establishing a predictable regulatory framework for approving DT applications and protecting public health and safety.

#### 3.1.5 Kairos Perspective

Anthonie Cilliers, Senior Manager Instrumentation, Controls and Electrical Kairos Power

<u>Presentation Overview</u>: Discussion of the Kairos definition of a DT; DT uses to provide virtual plant sensors, operations databases, support for operator training, and data analytics; demonstrations of smaller subsystems such as a molten salt coolant loop and a virtual counterpart to support a faster iterative design process; how a DT facilitates the intersection between regulatory and design spaces and speeds the licensing process; operational phase use of a DT to predict future plant conditions, create a dynamic operating envelope, and detect anomalies; design phase use to DT to more accurately define safety margins and reduce safety-related footprint within a plant; and use of a DT to reduce design risks.

#### 3.1.6 Westinghouse Perspective

Brian Golchert, Principal Engineer Westinghouse

<u>Presentation Overview</u>: Discussion of costs associated with DT including lack of guidance for related NRC submissions for techniques such as reduced-order modeling and coupling of single-domain models; the need to establish a DT business case; use of DT to support hardware-in-the-loop; use of DT in place of prototypes to reduce design, development, and regulatory costs; and the need for industry and the NRC to develop guidance needed to implement DTs.

Page intentionally left blank

### APPENDIX A WORKSHOP ATTENDEES

First Name	Last Name	Email Address	
Mohammad	Abdo	mohammad.abdo@inl.gov	
Kamal	Abdulraheem	kamalabdulraheem@gmail.com	
Chethan	Acharya	ckachary@southernco.com	
Vivek	Agarwal	vivek.agarwal@inl.gov	
Indarta	Aji	indartaaji@gmail.com	
Ahmed	Alshehhi	ahmedal565@gmail.com	
Tim	Alvey	tim.alvey@exeloncorp.com	
Harry	Andreades	charalampos.andreades@hq.doe.gov	
Michela	Angelucci	michela.angelucci@phd.unipi.it	
Kwame	Ansah	ansahkwame466@gmail.com	
Todd	Anselmi	todd.anselmi@inl.gov	
Thompson	Appah	appahtompson@gmail.com	
Jeffrey	Arndt	arndtjl@westinghouse.com	
Steven	Arndt	arndtsa@ornl.gov	
Dushyant	Arora	dushyant.arora@siemens.com	
Oussama	Ashy	ashyo@ws-corp.com	
Paridhi	Athe	pathe@ncsu.edu	
Md Samdani	Azad	samdaniazad@konkuk.ac.kr	
Vittorio	Badalassi	badalassiv@ornl.gov	
Jin Whan	Bae	baej@ornl.gov	
Emilio	Baglietto	emiliob@mit.edu	
Nicholas	Baldasaro	nick@hoplite.ai	
Han	Bao	han.bao@inl.gov	
Sergiu	Basturescu	Sergiu.Basturescu@nrc.gov	
Melissa	Bates	melissa.bates@nuclear.energy.gov	
Randall	Belles	bellesrj@ornl.gov	
Eric	Benner	eric.benner@nrc.gov	
Jacob	Benz	jacob.benz@pnnl.gov	
Mounia	Berdai	mounia.berdai@cnsc-ccsn.gc.ca	
Satyan	Bhongale	sbhongale@x-energy.com	
Harry	Bonilla-	hbonilla@iastate.edu	
Tanner	Boone	tanner.boone@nrc.gov	
Katarzyna	Borowiec	borowieck@ornl.gov	
Jyoti	Bose	jyoti.bose@alithya.com	
Jeremy	Bowen	jeremy.bowen@nrc.gov	
Thomas	Braudt	tbraudt@x-energy.com	
Alexander	Brazalovich	abrazalovich@x-energy.com	
Michael	Breach	michael.breach@nrc.gov	
Jeren	Browning	jeren.browning@inl.gov	
Logan	Browning	logan.browning@inl.gov	
John	Buchanan	jbuchanan@dekabatteries.com	
Angela	Buford	Angela.Buford@nrc.gov	
Michael	Buric	michael.buric@netl.doe.gov	

Pat	Burke	Patrick.B.Burke@xcelenergy.com	
Troy	Burnett	troy.burnett@inl.gov	
Rob	Burns	rob@arthur.ai	
Jonathon	Burstein	jdburste@bechtel.com	
Scott	Bussey	scott.bussey@nrc.gov	
Dirk	Cairns-	dirk.cairns-gallimore@nuclear.energy.gov	
Pattrick	Calderoni	pattrick.calderoni@inl.gov	
Clevin	Canales	ccanales@x-energy.com	
Salvatore	Cancemi	salvatore.cancemi@phd.unipi.it	
Brent	Capell	bcapell@epri.com	
Jesse	Carlson	jesse.carlson@nrc.gov	
Gene	Carpenter	gene.carpenter@hq.doe.gov	
Jorge	Carvajal	carvajjv@westinghouse.com	
Arindam	Chakraborty	achakraborty@viascorp.com	
Alvin	Chan	alvin.chan@opg.com	
Hasan	Charkas	hcharkas@epri.com	
Jaydev	Chauhan	jaydev.chauhan@opg.com	
Yifeng	Che	yfche@mit.edu	
Danny	Chien	npc1@nrc.gov	
Roger	Chin	mroizo40@hotmail.com	
Roger	Chin	roger@radiantnuclear.com	
Helene	Chini	chinih@westinghouse.com	
Hangbok	Choi	Hangbok.Choi@ga.com	
Anthonie	Cilliers	cilliers@kairospower.com	
Stephanie	Coffin	stephanie.coffin@nrc.gov	
Christopher	Cook	christopher.cook@nrc.gov	
Justin	Coury	justin.coury@nrc.gov	
Christopher	Crosby	ccrosby@osisoft.com	
Brad	Crotts	bradley.crotts@orano.group	
Amy	Cubbage	amy.cubbage@nrc.gov	
Samir	Darbali	samir.darbali@nrc.gov	
lan	Davis	idavis@x-energy.com	
Niyera	Davoodian	ndavoodian@gmail.com	
Grigorios	Delipei	gkdelipe@ncsu.edu	
Matt	Dennis	matthew.dennis@nrc.gov	
David	Desaulniers	david.desaulniers@nrc.gov	
Hadja Fanta	Diakhaby	nalahadja@gmail.com	
Xiaoxu	Diao	diao.38@osu.edu	
Nam	Dinh	ntdinh@ncsu.edu	
Elvis	Dominguez	dominguezoee@ornl.gov	
Valentin	Drouet	valentin.drouet@metroscope.tech	
Trevor	Dudley	drtdudley@mozweli.com	
Christophe	Duquennoy	christophe.duquennoy@edf.fr	
Carmen	Dykes	carmen.dykes@nrc.gov	
Derek	Ebeling-Koning	ebelind@westinghouse.com	
Shannon	Eggers	shannon.eggers@inl.gov	
Robert	England	robert.england@inl.gov	

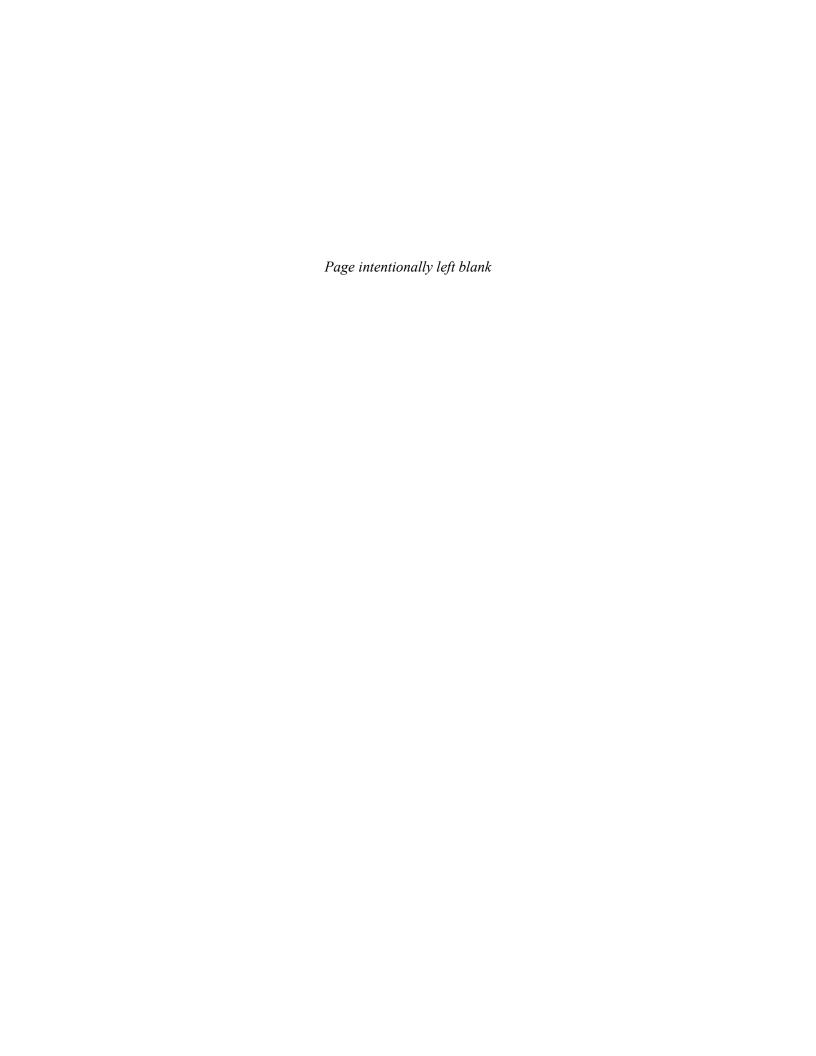
Doug	Eskins	doug.eskins@nrc.gov	
Kale	Evans	kalejevans@gmail.com	
Nathan	Faith	Nathan.Faith@ExelonCorp.com	
Amjad	Farah	amjad.farah@opg.com	
Mario	Fernandez	mario.fernandez@nrc.gov	
William	Ferrell	will@ams-corp.com	
Matthew	Ferri	mattferri@gmail.com	
Leo	Fifield	leo.fifield@pnnl.gov	
Eric	Focht	eric.focht@nrc.gov	
James D	Freels	freelsjd@gmail.com	
Raymond	Furstenau	raymond.furstenau@nrc.gov	
Pratik	Gandhi	pratik.gandhi@npxinnovation.ca	
Alex	Garrison	alex@radiantnuclear.com	
Marisol	Garrouste	mgarrou@umich.edu	
Ramon	Gascot	ramon.gascot@nrc.gov	
Lou	Gaussa	gaussalw@westinghouse.com	
Debraj	Ghosh	dghosh@iisc.ac.in	
Anders	Gilbertson	anders.gilbertson@nrc.gov	
James	Godwin	drjamesgodwin30@gmail.com	
Brian	Golchert	golchebm@westinghouse.com	
Gregory	Golding	greggolding@moltexenergy.com	
Carlos	Gonzalez	Carlos.Gonzalez@nrc.gov	
Nicholas	Goss	nicholas.goss@westinghouse.com	
Fred	Grant	ffgrant@sgh.com	
Scott	Greenwood	greenwoodms@ornl.gov	
Donna	Guillen	Donna.Guillen@inl.gov	
Anil	Gurgen	agurgen@ncsu.edu	
Alexandria	Haddad	awhadda@sandia.gov	
Andrew	Hahn	ashahn@sandia.gov	
Botros	Hanna	bn@nmsu.edu	
Leroy	Hardin	roy.hardin@gmail.com	
Leroy	Hardin	roy.hardin@nrc.gov	
Brennan	Harris	brennan.harris@inl.gov	
Kurt	Harris	kurt.harris@flibe-energy.com	
Robert	Harwood	robert.harwood@slingshotsimulations.co.uk	
Alex	Hashemian	alex@ams-corp.com	
Hash	Hashemian	hash@ams-corp.com	
Trey	Hathaway	Alfred.Hathaway@nrc.gov	
Gale	Hauck	hauckge@ornl.gov	
Jeff	Hawkins	jeffhawkins@jhawkconsulting.com	
Robert	Hayes	rbhayes@ncsu.edu	
Joe	Heit	joe.heit@aveva.com	
Eric	Helm	eric.helm@framatome.com	
David	Henderson	david.henderson@nuclear.energy.gov	
Peter	Henkes	phenkes@wisc.edu	
Richard	Henry	richard.henry@opg.com	
Raul	Hernandez	Raul.hernandez@nrc.gov	

Matthew	Hertel	MHertel@x-energy.com
Dan	Hoang	dan.hoang@nrc.gov
Alec	Holla	alec.holla@npxinnovation.ca
Brooks	Holland	brooks.holland@inl.gov
Zachary	Hollcraft	zachary.hollcraft@nrc.gov
Philip	Honnold	phonnol@sandia.gov
Loren	Howe	loren.howe@nrc.gov
Timothy	Huddleston	timothy.huddleston@inl.gov
Nathanael	Hudson	Nathanael.Hudson@nrc.gov
Lauren	Hughes	lhughes@wpainc.com
John	Hughey	john.hughey@nrc.gov
Clyde	Huibregtse	huibregc@oklo.com
Amy	Hull	amy.hull@nrc.gov
Matthew	Humberstone	Matthew.Humberstone@nrc.gov
Eman	Ibrahim	eman.ibrahim@canada.ca
Mesfin	Ibrahim	mesfin.ibrahim@connect.polyu.hk
Dan	Isaacs	dan@omg.org
Raj	Iyengar	raj.iyengar@nrc.gov
Prashant	Jain	jainpk@ornl.gov
Nicholas	Jameson	nicholas.jameson@inl.gov
Patty	Jehle	Patricia.Jehle@nrc.gov
Mike	Jenkinson	mike.jenkinson@siemens.com
Bob	Jewart	robert.jewart@inl.gov
Daniel	Ju	daniel.ju@nrc.gov
Chul Hwan	Jung	chulhwan.jung@cnsc-ccsn.gc.ca
Takanori	Kajihara	kajihara@tamu.edu
Aris	Kalafatis	aris.kalafatis@opg.com
Min-Tsung	Kao	kaom@ornl.gov
Fuad	Kassab Junior	fuad.kassab@usp.br
Maxine	Keefe	maxine.keefe@nrc.gov
Paul	Keutelian	paul@radiantnuclear.com
Genghis	Khan	khan@ge.com
Hamed	Khodadadi	hakhodadadi1986@gmail.com
Anya	Kim	anya.kim@nrc.gov
Paul	Klein	paul.klein@nrc.gov
Brendan	Kochunas	bkochuna@umich.edu
Andrea	Kock	alk@nrc.gov
Alan	Konkal	alan.konkal@nrc.gov
Ben	Kosbab	bdkosbab@sgh.com
Ashish	Kotwal	ashish.kotwal@und.edu
Robert	Krawczak	krawczrk@westinghouse.com
Roman	Kuchma	wonderwouker@ukr.net
Vineet	Kumar	kumarv@ornl.gov
Jonathan	Kyle	jonathan.kyle@ansys.com
Wilson	Lam	wilsonlam.cns@gmail.com
Jeffrey	Lane	lanejw@zachrynuclear.com
John C	Lane	jcl1@nrc.gov

Kyoung	Lee	leeko@ornl.gov	
David	Lefrancois	david.lefrancois@alithya.com	
John	Lehning	jxl4@nrc.gov	
Matthew	Levasseur	mplevasseur@bwxt.com	
Binghui	Li	binghui.li@inl.gov	
Jun	Liao	liaoj@westinghouse.com	
Bruce	Lin	bruce.lin@nrc.gov	
Linyu	Lin	linyu.lin@inl.gov	
Yong Chang	Liu	liuyongchang@gmail.com	
Deleah	Lockridge	lockridgedv@ornl.gov	
Christopher	Lohse	christopher.lohse@inl.gov	
Cihang	Lu	cihanglu@gmail.com	
Louise	Lund	Louise.Lund@nrc.gov	
Lee	Maccarone	lmaccar@sandia.gov	
Shah	Malik	Shah.Malik@nrc.gov	
Koushik	Manjunatha	koushik.manjunatha@inl.gov	
Koushik	Manjunatha	koush91@gmail.com	
CS	Manohar	manohar@iisc.ac.in	
Jonathan	Marcano	Jonathan.Marcano@nrc.gov	
Josh	May	josh@radiantnuclear.com	
Richard	Mcgrath	RMCGRATH@EPRI.COM	
Noreddine	Mesmous	noreddine.mesmous@cnsc-ccsn.gc.ca	
Ernest	Mileta	Ernest.Mileta@opg.com	
Jessie	Milligan-Taylor	Jessica.milligan-taylor@cnsc-ccsn.gc.ca	
Marwan	Mohamed	marwan.mohamed@inl.gov	
Ricardo	Moreno	ricardo.morenoescudero@ge.com	
Jawad	Moussa	Jmoussa@unm.edu	
Alewyn	Mouton	alewyn.mouton@opg.com	
Raheel	Naqvi	Raheel.naqvi@opg.com	
Curt	Nehrkorn	curt.nehrkorn@hq.doe.gov	
Scott	Nelson	nelsonsw@ornl.gov	
Carl	Neuschaefer	chneusch@aol.com	
Thien	Nguyen	thien.duy.ng@gmail.com	
Daniel	Nichols	daniel.nichols@nuclear.energy.gov	
Mirela	Nitoi	mirela.nitoi@nuclear.ro	
Kerstun	Norman	Kerstun.Norman@nrc.gov	
Alistair	Norris	alistair.norris@jacobs.com	
Jesus M	Nunez	jesus@nuclearalternativeproject.org	
Bill	Obaker	obakerwr@westinghouse.com	
Joe	Oncken	joseph.oncken@inl.gov	
Attendee	One	yadav.vaibhav@gmail.com	
Ekaterina	Paladi	Ekaterina.paladi@metroscope.tech	
Pallavi	Pandey	pallavip@iisc.ac.in	
Nithin	Panicker	panickerns@ornl.gov	
Sara	Perez-Martin	sara.perez@kit.edu	
Eternity	Perry	eternity@ams-corp.com	
Angelica	Petrovic	angelica.petrovic@inl.gov	

Jeffrey	Poehler	jeffrey.poehler@nrc.gov	
Joseph	Poisson	jpoisso@entergy.com	
Cosmin	Popescu	valentin.popescu@cne.ro	
Steve	Prescott	Steven.Prescott@inl.gov	
Dylan	Prevost (Doe)	dylan.prevost@nuclear.energy.gov	
Ivan	Price	irprice@sandia.gov	
Craig	Primer	craig.primer@inl.gov	
Anthony	Qualantone	aqualantone@x-energy.com	
Alexandre	Quertamp	alexandre.quertamp@metroscope.tech	
Mihaela	Quirk	mihaela.quirk@hq.doe.gov	
Brandon	R	ricebc@inl.gov	
Majdi	Radaideh	radaideh@mit.edu	
Jean	Ragusa	jean.ragusa@tamu.edu	
Pradeep	Ramuhalli	ramuhallip@ornl.gov	
Bob	Randall	bob.randall@nrc.gov	
Ravi	Raveendra	rraveendra@cometacoustics.com	
Wendy	Reed	wendy.reed@nrc.gov	
Seyed	Reihani	sreihani@illinois.edu	
Mehdi	Reisi Fard	mehdi.reisifard@nrc.gov	
Florencia	Renteria	florenciaren@gmail.com	
Gustavo	Reyes (InI)	gustavo.reyes@inl.gov	
Alex	Rhodes	alex.rhodes79@outlook.com	
Daniel	Rosas	daniel.rosas@opg.com	
Cormac	Ryan	cormac.ryan@aveva.com	
Will	S	coinbird@gmail.com	
Dagistan	Sahin	dagistan.sahin@nist.gov	
Osman	Sahin	osman.celikten@nist.gov	
Michele	Sampson	michele.sampson@nrc.gov	
Erica	Sanchez	erica.sanchez@inl.gov	
Daniel	Sandoval	drsando@sandia.gov	
Suman	Saurav	sumanjiseie@gmail.com	
Abhinac	Saxena	asaxena@ge.com	
Abhinav	Saxena	asaxena@ge.com	
Randall	Schmidt	randy.schmidt@exeloncorp.com	
Paul	Schuck	paul.schuck@inl.gov	
Aurelien	Schwartz	aurelien.schwartz@metroscope.tech	
Garry	Schwarz	garry.schwarz@cnsc-ccsn.gc.ca	
Ting-Leung	Sham	tingleung.sham@inl.gov	
Neil	Sheehan	Neil.Sheehan@NRC.GOV	
Brent	Shumaker	brent@ams-corp.com	
Scott	Sidener	sidenese@westinghouse.com	
Paul	Sirianni	sirianpm@westinghouse.com	
Alexandra	Siwy	Alexandra.Siwy@nrc.gov	
James	Slider	jes@nei.org	
Curtis	Smith	curtis.smith@inl.gov	
Mohamed	Soliman	mohamed_saied66666@mail.ru	
Sharon	Soogrim	sharon.soogrim@nrc.gov	

Christopher	Spirito	christopher.spirito@inl.gov	
Antoanela	Stoica	antoanela.stoica@cne.ro	
Craig	Stover	cstover@epri.com	
Molly	Strasser	Molly.J.Strasser@xcelenergy.com	
Cheng	Sun	cheng.sun@inl.gov	
Xiaodong	Sun	xdsun@umich.edu	
Richard	Szoch	richard.szoch@exeloncorp.com	
Emre	Tatli	tatlie@westinghouse.com	
Nazila	Tehrani	nazila.tehrani@nrc.gov	
Keith	Tetter	Keith.Tetter@nrc.gov	
James	Tompkins	james@radiantnuclear.com	
Ricardo	Torres	ricardo.torres@nrc.gov	
Robert	Tregoning	robert.tregoning@nrc.gov	
Panagiotis	Tsilifis	panagiotis.tsilifis@ge.com	
Bogdan	Tutuianu	bogdan.tutuianu@cne.ro	
Мо	Uddin	mouddin009@gmail.com	
Мо	Uddin	muddin@structint.com	
Rizwan	Uddin	rizwan@illinois.edu	
Christopher	Ulmer	christopher.ulmer@nrc.gov	
Troy	Unruh	troy.unruh@inl.gov	
Bob	Urberger	bob@radiantnuclear.com	
Johannes	Van Der Watt	johannes.vanderwatt@und.edu	
Stephen	Vaughn	svaughn@x-energy.com	
Justin	Vazquez	Justin.Vazquez@nrc.gov	
Swetha	Veeraraghavan	swethav@iisc.ac.in	
Rattehalli	Vijay	rattehalli.vijay@unnpp.gov	
Purna	Vindhya	purnavindhya@iisc.ac.in	
Nurali	Virani	nurali.virani@ge.com	
likka	Virkkunen	iikka.virkkunen@aalto.fi	
Cody	Walker	cody.walker@inl.gov	
William	Walsh	william.walsh@nuclear.energy.gov	
Congjian	Wang	congjian.wang@inl.gov	
Guanyi	Wang	guanyi.wang@anl.gov	
Weijun	Wang	weijun.wang@nrc.gov	
Justin	Weinmeister	weinmeistejr@ornl.gov	
Cindy	Wellenbrock	ckwellenbrock@gmail.com	
Timothy	West	timothy.west@inl.gov	
Chad	Wilhelm	cmwilhel@bechtel.com	
Katherine	Wilsdon	katherine.wilsdon@inl.gov	
Paul	Witherell	paul.witherell@nist.gov	
Brian	Wittick	brian.wittick@nrc.gov	
Jennifer	Wong	jennifer.wong@opg.com	
Vaibhav	Yadav	vaibhav.yadav@inl.gov	
Xingyue	Yang	xingyue.yang@inl.gov	
Matthew	Yarlett	Matthew.Yarlett@Westinghouse.com	
Jordan	Zenhenko	jordan.zenhenko@opg.com	
Jack	Zhao	jack.zhao@nrc.gov	



#### **APPENDIX B PRESENTATION SLIDES**











# **Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization**

# Ray Furstenau

**Director, Office of Nuclear Regulatory Research** 

September 14-16, 2021



# Future Focused Research (FFR)

 FFR program supports the NRC vision of becoming a modern, risk- informed regulator

## **Objectives**



- Close technical gaps ahead of <u>regulatory needs</u>
- Support transformative, innovative ideas
- Follow trends across industry and federal agencies
- Engage with industry, public, government and university communities
- Build new in-house capabilities that will attract and retain top talent

#### **Process**



- Open to ideas from across the agency
- Senior Level Staff panel reviews & makes project recommendations
- Monitor and communicate progress via program reviews and seminars

# **Workshop Overview**

Day 1 Tuesday, September 14 <sup>th</sup>		Day 2 Wednesday, September 15 <sup>th</sup>		Day 3 Thursday, September 16 <sup>th</sup>	
11:00 - 11:15	Introduction and Opening Remarks: NRC	11:00 - 12:45	Technical Session	11:00 - 12:45	Panel Session
11:15 - 12:45	Panel Session  Nuclear Industry Applications of  Digital Twins		Use Cases of Digital Twin Enabling Technologies in Nuclear Power Plants		Steps Toward Regulatory Realization of Digital Twins
12:45	Break	12:45	Break	12:45	Closing Remarks: NRC
2:00 - 3:45	Technical Session  Advanced Sensors & Instrumentations	2:00 - 4:00	Technical Session  Digital Twin Enabling Technologies in Advanced  Reactor Applications		
3:45	Adjourn	4:00	Adjourn		











# **Nuclear Industry Applications of Digital Twins**

Opening Panel Session - September 14, 2021

**Moderator: Ray Furstenau, NRC** 

Patrick Burke, Xcel Energy
Craig Stover, EPRI
Scott Sidener, Westinghouse
Rick Szoch & Tim Alvey, Exelon
Harry Andreades, ARPA-E
Aurélien Schwartz, Metroscope





### INNOVATION JOURNEY TO A BRIGHTER ENERGY FUTURE

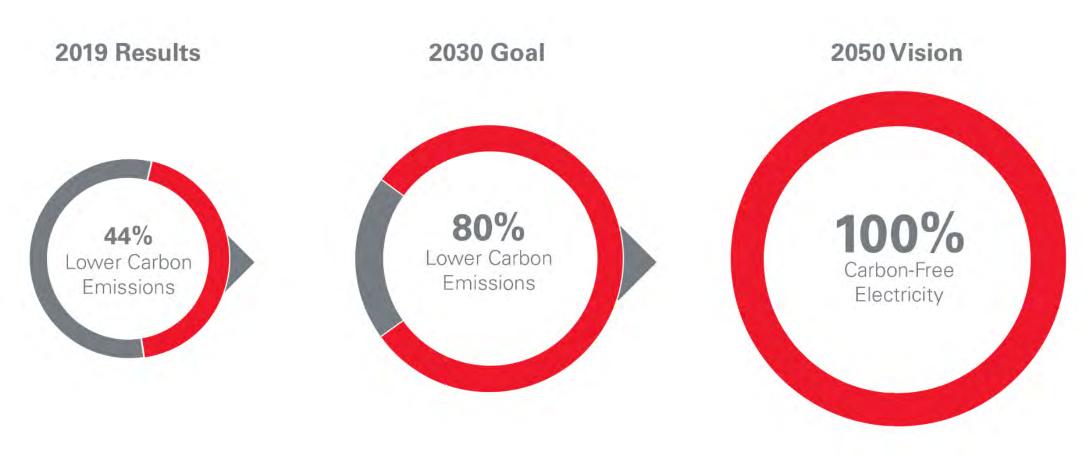
**Patrick Burke** 

**Vice President Nuclear Strategy** 

September 14, 2021

### Leading the Clean Energy Transition

A bold vision for a carbon-free future Monticello and Prairie Island Nuclear Plants

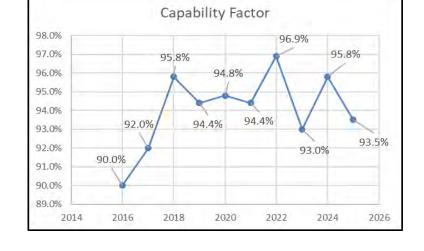


Company-wide emissions reductions from the electricity serving our customers, compared to 2005

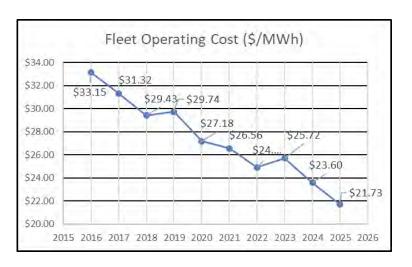
### Safe, Reliable, and Cost Effective Operation Through Innovation and Technology

 Safe - High safety metrics & ratings (NRC, INPO)

Reliable - Capability Factors 90% to 95%



Cost Effective 33\$/Mw to 26\$/Mw



### **Track Record of Innovation**

- Cap Intelligent Advisor (AI/ML) and Digital Ops Factory
  - Search, Entry, Analytics
- Automation and Work Management integration GE APM
  - Wi FI and Sensors and Remote Monitoring
- Early adopters for Organizational Transformation Fleet Services Model (Operate & Maintain),
- Risk based initiatives, TSTF-425 & 505 and 10CFR50.69
- Security Innovation and Modification
- Outage Improvements (Drones, Communications, etc.)
- Hydrogen production DOE demonstration project (HTSE)
- Flexible Operation Integration with Wind
- Exploring operational services with NuScale

### **Future State**

- High levels of Safe and Sustainable performance through Technology
- Highly Skilled Multi functional workers that are data and digital fluent
- Services Organization across multiple units leveraging remote technologies
- Automation of analytics and reporting
  - Compliance Validation
  - Daily Ops reports risk based priorities
  - Real time equipment performance monitoring
  - Reporting automation (MSPI, etc)
- Integrated operations (Flex, H2, Grid Support, Storage)

## **Xcel** Energy®

© 2020 Xcel Energy

# EPRI's Digital Twins Related Activities for Nuclear Applications

**An Overview** 

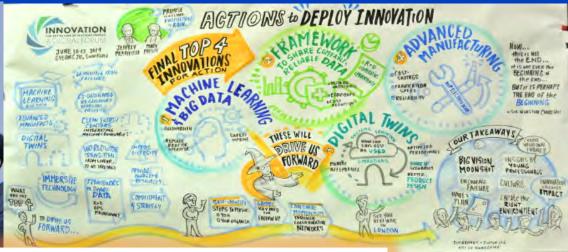
Craig Stover
Program Manager
Advanced Nuclear Technology (ANT)

September 14, 2021













### JUNE 10-12, 2019 GYEONGJU, SOUTH KOREA



















### Collaboration to Accelerate the Top 4 Innovations









Framework to Share Comparable, Reliable Data

Digital Twinning





EPRI Contact: Thiago Seuaciuc-Osorio tseuaciuc-osorio@epri.com



EPRI Contact:
Rob Austin
raustin@epri.com



EPRI Contact:
Hasan Charkas,
hcharkas@epri.com



EPRI Contact:
David Gandy,
dgandy@epri.com





### Challenge

#### Who should attend?

EPRI and its co-organizers are inviting top innovators and influencers from around the world:

- Movers
- **Shakers**
- **Activists**
- Mirrors
- Super-nodes Simply healthy restless people



#### Attendees will leave with ways to:

- Drive innovation using a new network
- Lead a cultural shift in their organizations.
- Inspire and on-board others into the innovation movement for responding to climate change, deep decarbonization, and the energy challenge for the future of nuclear energy.











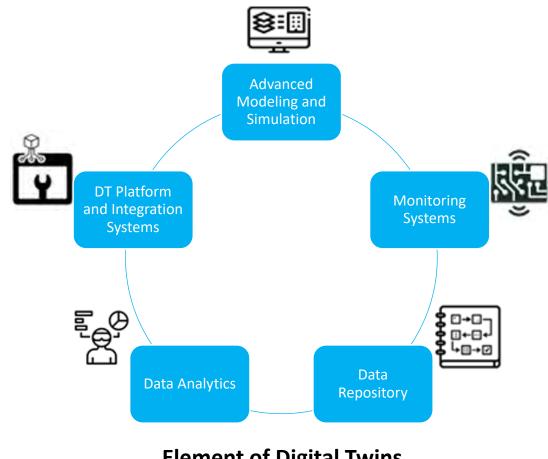
### Diversity

Q1/Q2 2022



### **EPRI Digital Twin Engineering Overview**

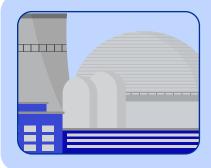
- Formed an internal cross-cutting team for collaboration
- Developed two technical insights document (3002020014) and (3002022555)
- EPRI Digital Twins information video released in December 2020 and another one should be released soon.
- Near term the team is working on the following:
  - What impact do DT applications have on nuclear power plant construction, operation, maintenance and decommissioning?
  - What DT applications can be deployable in the near future?



**Element of Digital Twins** 

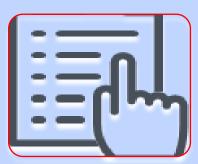


### Digital Twins (DTs) for Advanced Reactors (ARs)



### **Objective**

- Explore benefits, challenges and potential use cases for AR applications
- Establish industry guidelines, best practices and recommendations for implementing DTs in ARs life cycle management



#### **Status**

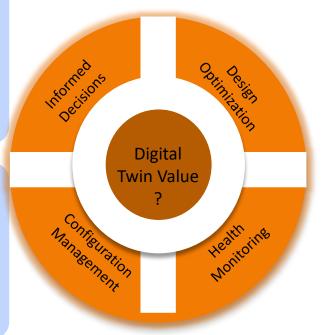
- Summarized use cases for various stages of ARs life cycle
- Developed a framework of DT project phases
- Selected use cases to further develop DT diagrams for them and to understand needed details to deploy them.



### **Next Steps**

- Summarize experiences and recommendations
- Publish in 2022







### Digital twins use cases....what are we finding so far?

- The sky's the limit!. Its important to assess
  - Technology readiness
  - Cost and Value
  - Scalability
  - Regulatory acceptance
  - Applicability
- Enabling technologies like AI and ML as well as data analytics are important for ensuring successful DT applications



The key question: where does it make sense to use DTs? (Technology readiness, cost benefits, and priorities)

#### Description

#### Real-Time Construction Sequence Optimization and Front-Running Simulations

- A DT based on BIM that integrates 3D geometry, construction sequencing information, cost model, schedule, and so forth, to visualize and simulate detailed work sequencing.
- During construction phase, use the DT to perform real-time optimization and front-running simulations informed by monitoring of schedule- and cost-critical parameters (for example, concrete curing; health, efficiency, and productivity of construction equipment; quantities of materials delivered and installed; current weather and forecast; environmental/health hazards; geolocation and biometrics of personnel; status of work products and inspections; supply chain data and predictions; billings from suppliers).
- Real-time tracking of critical parameters enables identification and mitigation of suboptimal construction operations.
- Front-running simulations provide real-time analysis of the likely consequences/outcomes of
  construction project events and deviations from the plan; provide construction management with
  high-fidelity information to support optimal responses to construction deviations.

#### 3D Mapping and Augmented Reality Visualization of Ambient Radiation Levels

- Integrate radiation monitoring sensor data with 3D building models to provide real-time, in situ mapping of ambient radiation levels throughout plant.
- Equip plant personnel with mobile augmented reality equipment integrated with the ambient
  radiation maps and personnel location tracking to enable real-time visualization of radiation
  levels in the field, for example, safety glasses with projected images indicating radiacative
  "hot spots" to improve implementation of "as low as reasonably achievable" (ALARA) dose
  management principles.
- Use pathfinding algorithms to guide personnel through optimal routes to their destination to minimize personnel dose exposure.
- Compare real-fine measurements from distributed sensors in plant and dosimeters worn by plant
  personnel with historical data to provide event-driven alarms and notifications of unexpected and
  new hot spots, which could be indicative of system faults, leaks, or malfunctions.
- Couple personnel location tracking with contamination monitors to facilitate tracking, tracing, and deanup of contamination sources.

#### Target Benefits

- Improved efficiency of construction operations.
- Optimized costs and schedule.
- Early warning of personnel health hazards during construction for improved safety.
- Informed ad-hoc decisionmaking based on knowledge of likely consequences of unplanned events that occur during construction and potential outcomes of mitigating actions.
- Enhanced personnel safety, reduced dosage.
- Decreased personnel effort for performing radiation surveys manually.





### **SCOTT SIDENER**

CONSULTING ENGINEER, DIGITAL INNOVATION

WESTINGHOUSE ELECTRIC COMPANY

### **Exelon Nuclear Innovation Projects Overview**



### **Innovation Culture and Digital Transformation**

#### DATA-DRIVEN INSIGHTS HIGH VALUE DECISIONS

Extend nExus capability to include digital twin and mobile capabilities

Data strategy and Big Data Platform capabilities

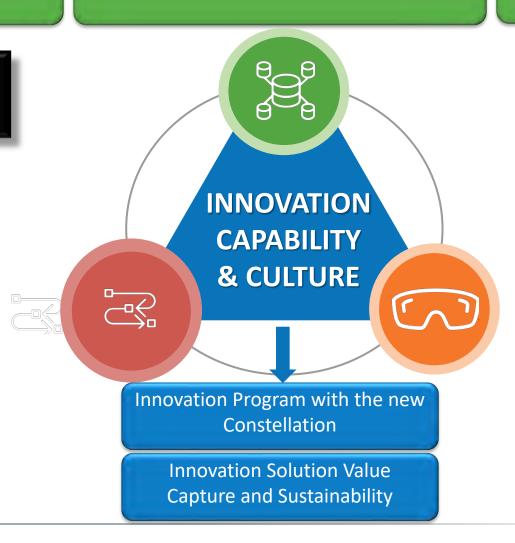
Data analytics and AI/ML to achieve value through data-driven decision-making

Direction for 2022 and beyond

### DIGITAL PROCESS TRANSFORMATION

Implement process transformations to enable annual savings

Deliver insights and savings through workflow automation with Al & ML



### CONNECTED MOBILE WORKFORCE

Expand Use of Smart Procedures

**Future of Work Instructions** 



## We Have a Burning Platform!!

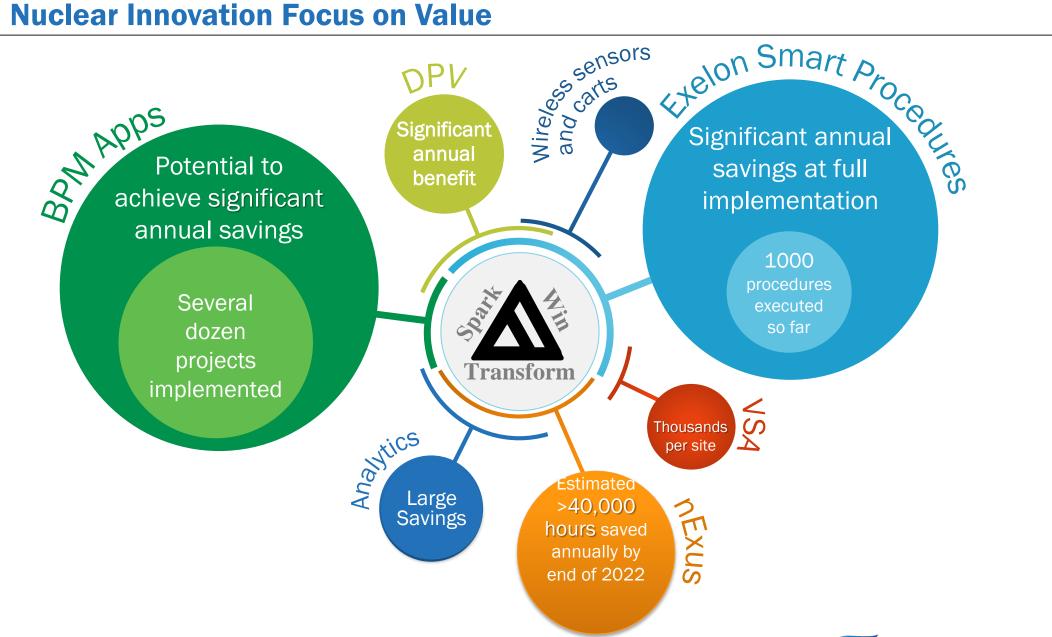
- Competition with natural gas
- Aging plants
- Plant shutdowns due to economics
- Domestic nuclear construction costs
- Extension of operating life cycles
- Extension of operating life cycles



Maintaining our domestic nuclear fleet is a matter of national security



#### **Exelon Nuclear Innovation Focus on Value**





### **Exelon Nuclear Innovation - Remote Monitoring**

#### Wireless Sensors

Provides a cost-effective method to enable new innovative ideas

#### **Executive Summary:**

**Problem –** The high cost of running wires and using traditional install is cost prohibited. Most data is collected manually by operator rounds

#### Solution -State of art communications infrastructure

- Wireless infrastructure that provides a cost-effective method to utilize wireless sensors and enable new innovative ideas and predictive technologies
- Low-cost wireless sensors (design once-install many) installed to improve predictive technology and facilitate elimination of high-cost time-based maintenance
- Reduce the time that it takes from an idea to implementation

#### Wireless Sensors - Benefits

#### Eliminate unplanned downtime

Identify defects in rotating machines before they happen and eliminate downtime

#### Real time visibility into health

Provide real-time visibility into machine health; can replace walk around data collection

#### Monitor inaccessible machinery

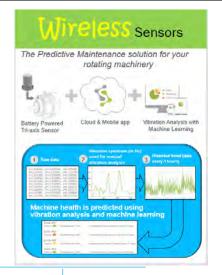
Continuous, safe and consistent data collection for remote assets and hazardous locations, reducing exposure

#### Dose savings

Reduction in Ops & Maintenance rounds

#### Multi parametric PdM

Leverage multi-parament sensor data to get a holistic view into machine health



#### **Innovative Improvement:**

### How did the <u>old</u> system work in terms of people, process, and tools?

Traditional sensors required wires. This process can take years to design, plan & install. This process can also cost millions of dollars to complete. This process is very slow & costly, and it never gets done.

### How does the <u>new</u> solution work in terms of people, process, tools?

Wireless sensors allow us to rapidly deploy sensors where needed at a much lower cost. These sensors allow us to use design once & install many concept. Wireless sensors can be installed both temporary or permanently installations. They can also be portable or mobile.

#### **New Capabilities:**

 Wireless sensors are expected to define and transition to a future state where maintenance on consequential equipment is accomplished "just-in-time". This is enabled via state-of-the-art diagnostic and analytics tools, wireless technologies, and an altered work management strategy.

#### **Application:**

- Wireless sensors to support plant monitoring included licensed and unlicensed radio frequency
- Cellular devices to support plant communication
- Connecting devices & sensors to Predix APM

#### Value:

There are numerous benefits to the ability to connect Internet of Things devices to the corporate network. Ability to utilize a variety of wireless sensors in the field. The ability to capture and transmit image data back to a centralized server where it can be used by a variety of applications.

#### **Other Uses or Potential Enhancements:**

- Maintenance frequency optimization based on procedure results
- Schedule optimization based on real time task duration
- Connecting procedures to live plant data



### **Exelon Nuclear Innovation - Remote Monitoring**

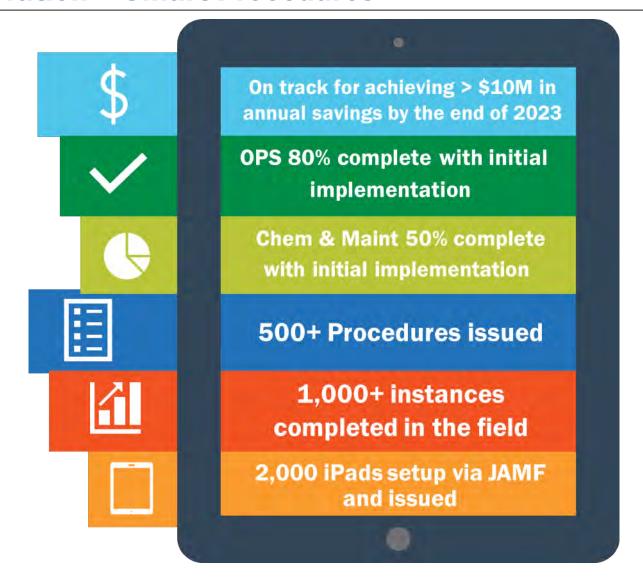


- Personnel able to perform monitoring from their desk.
- Ability for event-triggered actions: Upon an event, predefined actions can be performed, including camera movement and email notification.
- Cart setup can be potentially stored in the Power Lab and can be sent for simple assembly and setup at the site with hotspot.
- Potential for cost saving by reducing personnel burden and ability to monitor hotspot behavior.





### **Exelon Nuclear Innovation – Smart Procedures**





### **Exelon Nuclear Innovation Analytic Projects**

### **Potential Opportunities**

- Full automation of resource intensive processes
- Improved trending and data insight
- Enhanced predictive capabilities
- Process optimization and forecasting

#### **Desired Results**

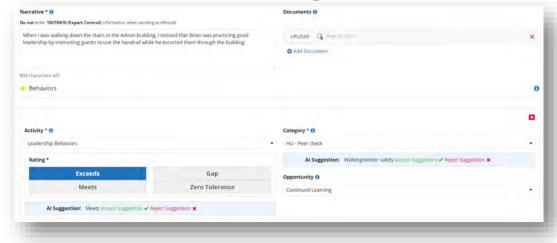
- Reduced operating costs
- Improved accuracy of decision making
- Improve workforce quality of life
- More time working on the "right stuff"



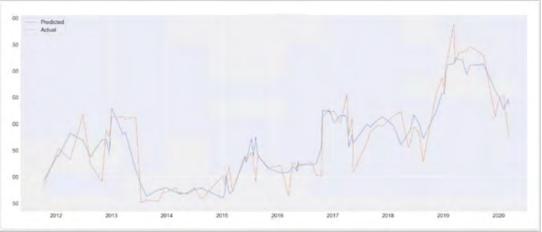


### **Exelon Nuclear Innovation Analytic Projects**

**Observation Categorization** 



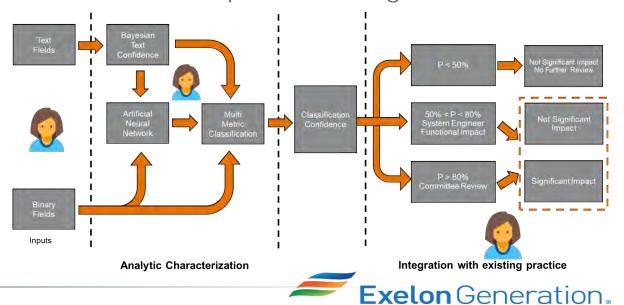
Chemistry Sampling Analytic



#### Initial License Training Throughput Optimization



#### Condition Report Screening Automation







### **ARPA-E GEMINA Portfolio and Digital Twins**

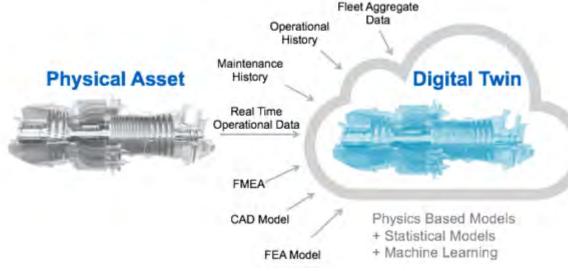
Dr. Harry Andreades
Technology-to-Market Advisor (support contractor to ARPA-E)
Contact: Dr. Jenifer Shafer, Program Director, jenifer.shafer@hq.doe.gov

December 7, 2021

### **Defining Digital Twins**

Mapping of <u>physical asset models</u> in a digital platform where a <u>virtual digital</u> replica is created

- Consists of three basic building blocks:
  - 1. 3D models
  - 2. Simulators
  - 3. PLM platform to centralize, organize, and manage data



 Continuous updating (sensors) and real-time data analysis to model physical asset

### Digital twins provide a range of benefits

- Allow for optimal operations and condition-based maintenance
- Time travel: Allow for manipulation of DT for scenario and what-if analysis without disturbing physical asset (continuously updated data goes beyond static picture)
  - This can also apply during design phase prior to physical asset launch
- Enables:
  - Rapid design iterations and optimization
  - Remote operations
  - Autonomous power plants
  - Fleet management
  - Performance improvement

### Where are DTs applied currently?

Oil & Gas, Gas Turbines (and Combined Cycles), Wind Power, Hydro, etc.















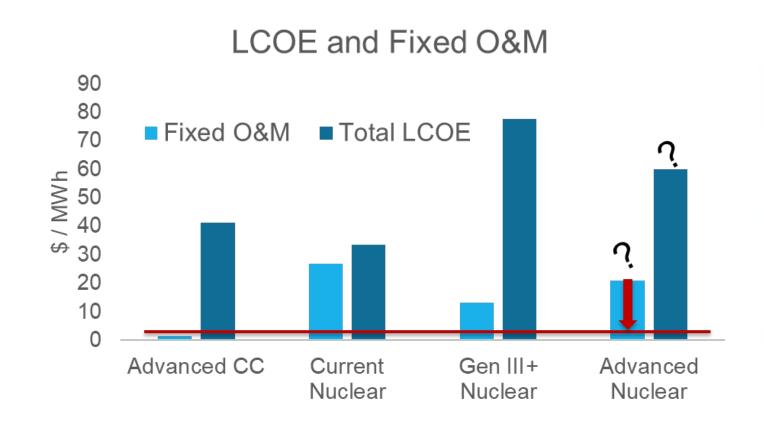


### GEMINA Generating Electricity Managed By Intelligent Nuclear Assets

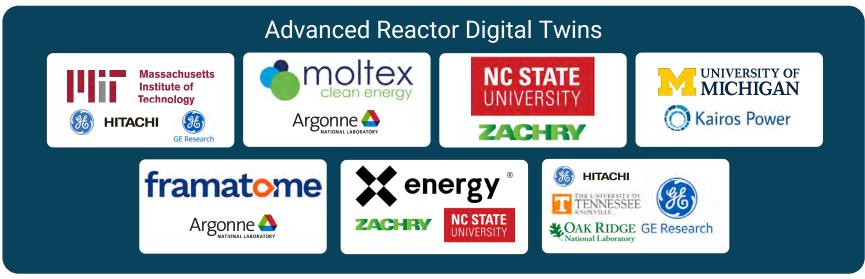
Goal: Develop the tools and cost basis for ARs to achieve fixed O&M costs of \$2/MWh without shifting costs to other parts of LCOE

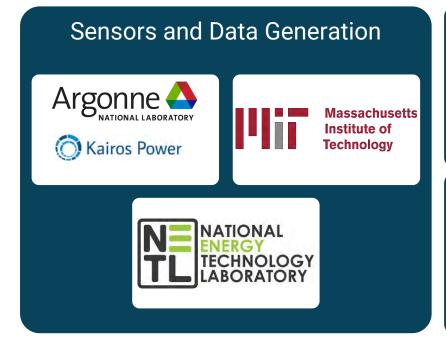
Awardee teams are developing the following for one or more of the most promising AR designs:

- Digital twins for advanced reactor systems
- Relevant cyber physical systems
- O&M approaches for advanced reactors
- Cost models and design updates



## ARPA-E teams are building digital tools for ARs and building blocks for DTs

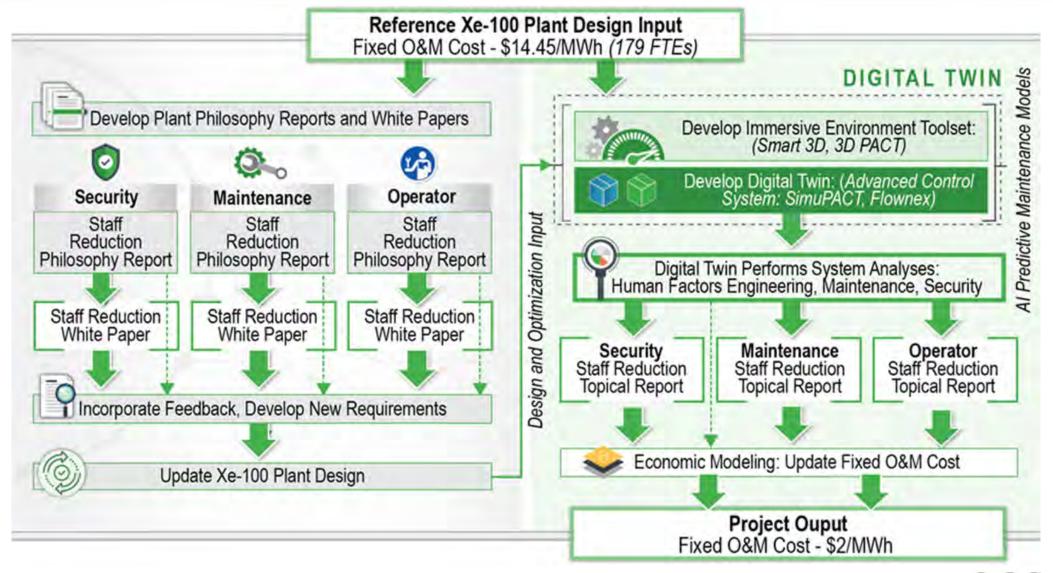








## **X-energy:** Advanced Operation & Maintenance Techniques Implemented in the Xe-100 Plant Digital Twin to Reduce Fixed O&M Cost



## **GE – Research:** AI-ENABLED PREDICTIVE MAINTENANCE DIGITAL TWINS FOR ADVANCED NUCLEAR REACTORS

*Summary* 

#### **Program Impact**

Al-enabled predictive maintenance to ↓ O&M labor costs

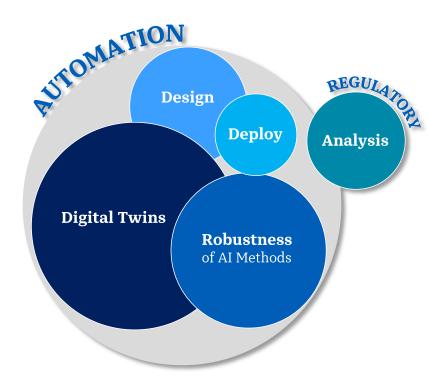
from \$15/MWh to \$3/MWh in an Advanced Nuclear Reactor



Program Targets		
Metric	From	То
Automation  ↓ labor costs	None	Automated workorders $\downarrow$ Planning staff Online calibration $\downarrow$ Tech and admin staff
Predictive Maintenance  ↓ labor & mat'l	Alarms	$\downarrow$ Forced outages & trips Al-driven predictive algorithms $\rightarrow$ $\downarrow$ Labor headcount
Trust	Human	Humble & explainable AI quantify uncertainty to establish trust in the models & encourage automation

#### **Technology Summary**

- Reactor Operations Physics-informed machine learning, sensor optimization
- ▶ Reactor Health Causal, humble & explainable AI for predictive maintenance
- **Decision Making** Autonomous risk-informed decisions for reconfiguration & maintenance



## **U Michigan:** Project SAFARI: Secure Automation for Advanced Reactor Innovation

### GOAL

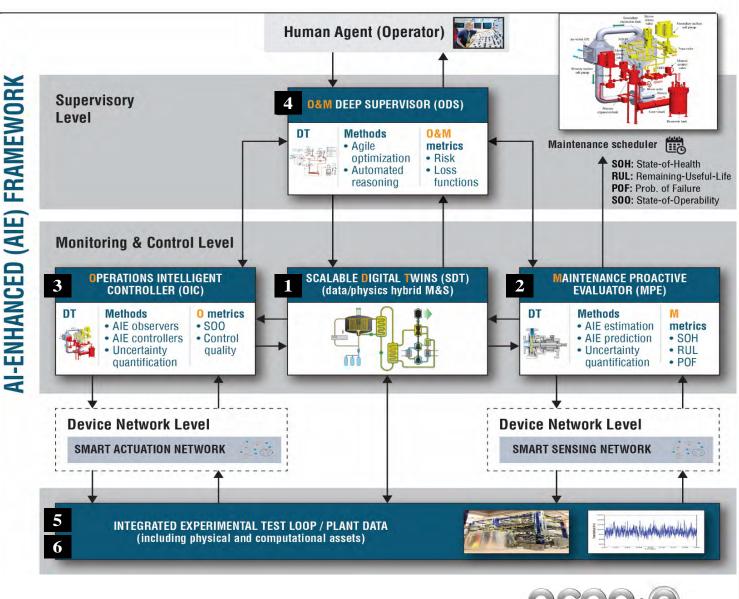
Reduce NPP O&M costs by delivering a capability which will enable smart functionalities in advanced reactor systems including:

- Autonomous, flexible operations
- Predictive maintenance
- Agile Design
- System and sensors optimization

#### **DEMO** Kairos FHR

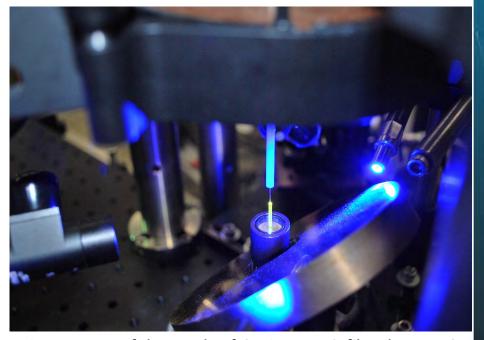
#### **END PRODUCT**

Physics-based, data-enabled, modular and scalable capability that can be extended and applied to any reactor technology



### **NETL:** DISTRIBUTED MOLTEN SALT LOOP SINGLE-CRYSTAL HARSH-ENVIRONMENT OPTICAL FIBER-SENSORS

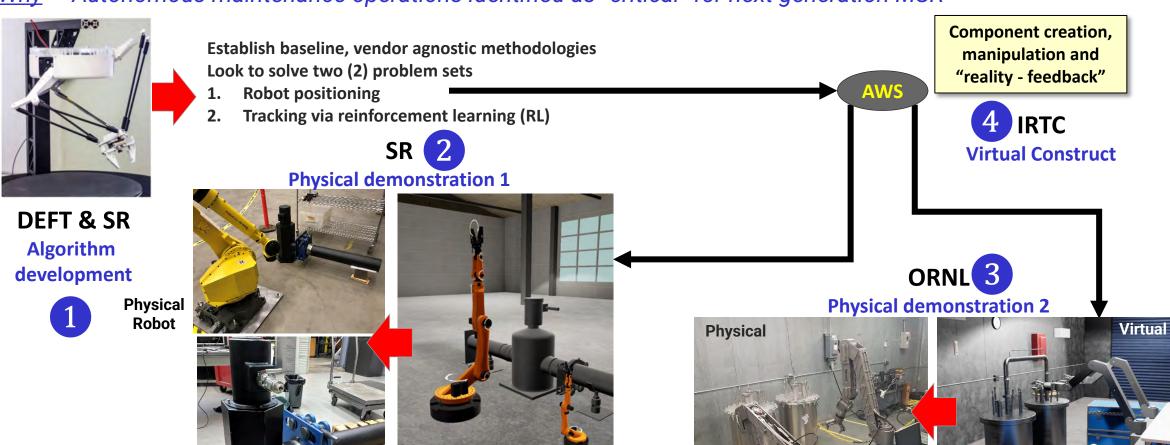
- Introducing fully-distributed sensing to MSRs
- Growing new cladded single-crystal optical fibers for molten-salt environments
- Gathering thousands of data-points to map reactor coolant-path temperatures or other parameters
- Mapping in-core temperature distributions
- Next-gen sensing replaces single-point sensors like thermocouples
- Providing data to guide reactor design and improvement through thermal efficiency



First successful growth of Cerium YAG fiber by LHPG

### **SRI:** ML FOR AUTOMATED MAINTENANCE OF FUTURE MSR

Create an integrated software and algorithm architecture to "teach" automated systems from simulated data <u>Why</u> - Autonomous maintenance operations identified as "critical" for next generation MSR

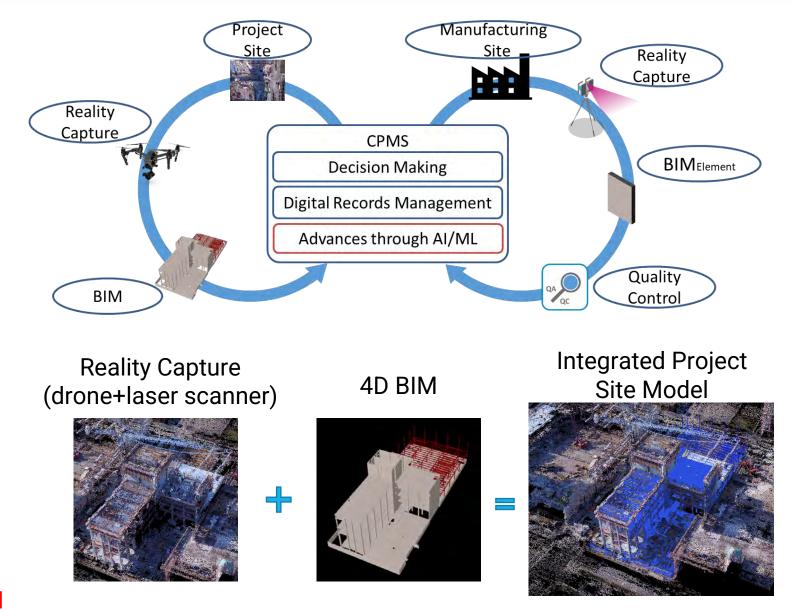


Virtual

replicates

Physical maintenance equipment

## **NSCU:** A Data-driven Approach to High Precision Construction and Reduced Overnight Cost and Schedule



# **EPRI:** BUILD-TO-REPLACE: A NEW PARADIGM FOR REDUCING ADVANCED REACTOR O&M COSTS

- Identify representative SSCs for evaluation of design life implications for cost and performance associated with licensing, construction, operation, maintenance, and decommissioning
- Identify at least two reference advanced reactor (AR) designs to establish baseline O&M cost, enveloping multiple missions and a broad plant parameter envelope
  - Target: small modular light-water reactor and high-temperature gas-cooled reactor as mature technologies,
  - Aspirational goal: extend to a molten salt reactor design
- Define scenarios for reduced SSC lifetimes to evaluate impacts on O&M costs against other categories, including construction and decommissioning

Metric	State of the Art	Proposed
Fixed O&M cost	\$20/MWh for light-water reactor (LWR) fleet	< \$10/MWh for more mature ARs < \$5/MWh for less mature ARs
		< ψυ/ivivii ioi iess mature Aixs
Design life for major plant components	30 – 40 years for PWR steam generator	Major SSCs < 15 years
,	60+ years for RPV	No life limiting SSCs

#### **SUMMARY**

- ARPA-E has a complimentary fission portfolio which targets both capital and operating costs reductions for making advanced reactors commercially competitive
- Digital twins and their enabling technology can be disruptive in changing the design, construction, and operating phases for advanced reactors
- Multidisciplinary capabilities are needed for successful implementation of DTs
- ARPA-E has a strong technology-to-market focus and encourages/enables performers to focus on commercial relevance and commercialization aspects of their technology
- Performer material is from unrestricted/publicly-available info from the annual ARPA-E Nuclear Program Review meeting, and can be found at <a href="https://arpa-e.energy.gov/2021-annual-nuclear-review-meeting">https://arpa-e.energy.gov/2021-annual-nuclear-review-meeting</a>.





# Digital Twin Monitoring for Advanced Reactors and Plant Modernization

Webinar - 09/14/2021

Aurélien Schwartz

Software company founded in Paris in 2018, with offices in Germany and the USA Patented technology originating from the R&D Center of EDF constituted of more than 2000 researchers



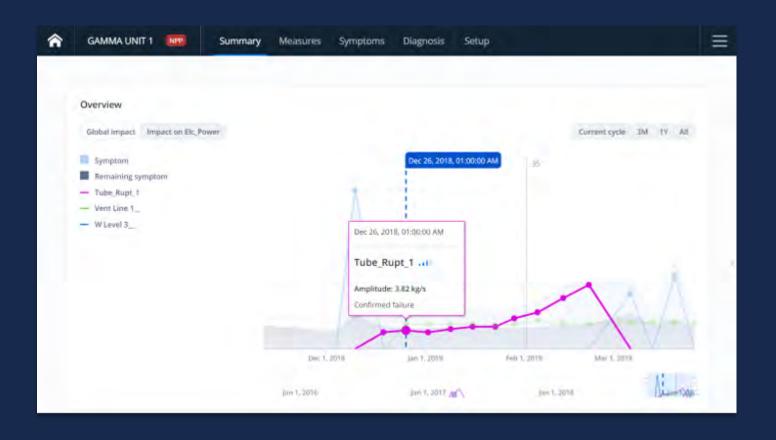


# Starting with a usecase

Tube rupture in a High Pressure Feedwater Reheater (Blayais, France)

Automatically detected on Dec 23 with a magnitude of 3kg/s

Fixed at the end of Feb with a magnitude of the leak of 15 kg/s



# Key principles

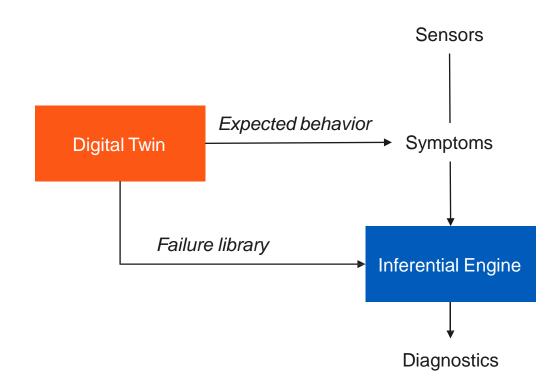
First, we use a model to build live symptoms for the plant

```
measurements - expected values = symptoms
```

We then use a failure library embedded in the digital twin to preform a root cause analysis.

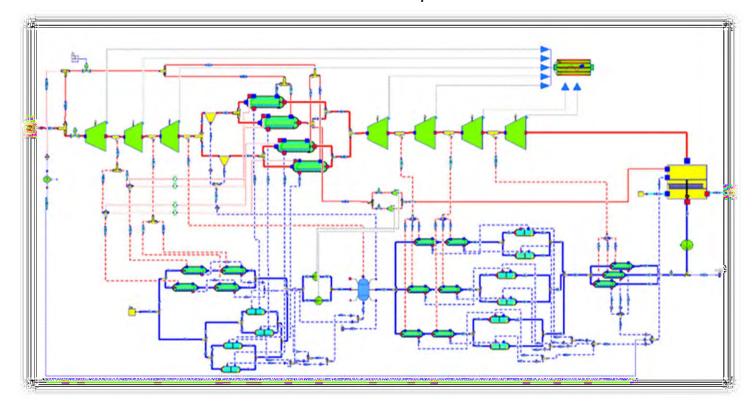
```
diagnostics = P(failures | symptoms)
```

Failures are classified by magnitude, impact and likelihood.



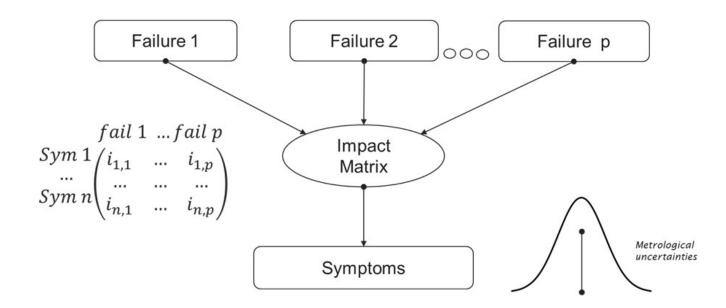
# Physical based model

The model of a NPP is composed of around 12 000 equations and variable declarations. It simulates both the nominal behaviour and failures of the plant



# Inferential engine

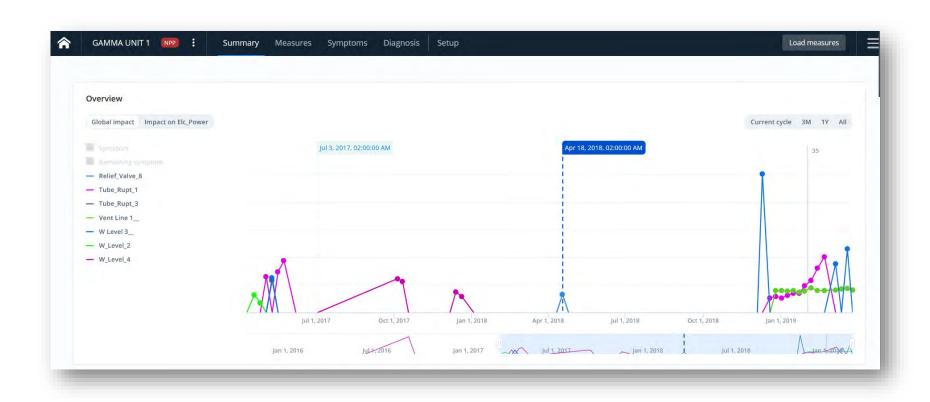
The inferential engine performs a root causes analysis of underlying problems, through the analysis of the symptoms. It solves the following probability: *P* (failure 1, failure 2, ... failure p | Symptoms)



# The software

Software UX designed with operating teams

Reliability of 90% observed in 2020 - 2000 GWh energy loss detected - Over 300 users



# Technological landscape

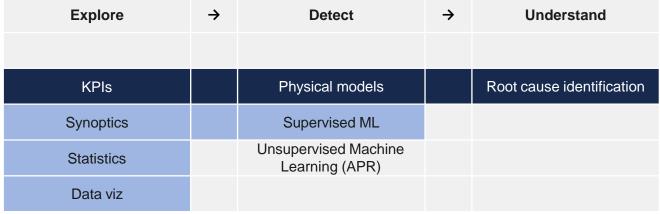
Monitoring of power plants aims at understanding any problem the plant during the operations

Soon we think that Physical Models and ML approaches will feed a unique diagnostic engine, making the best out of available process data

Į
i

Under

development



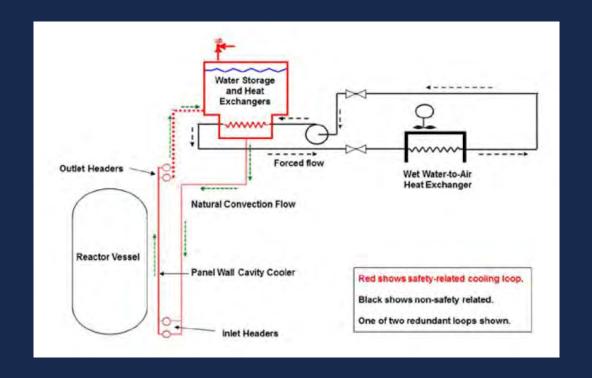


Production ready

# Innovation with ARPA-E

Asset Performance and Reliability for the HTGR Reactor Cavity Cooling System Using Metroscope

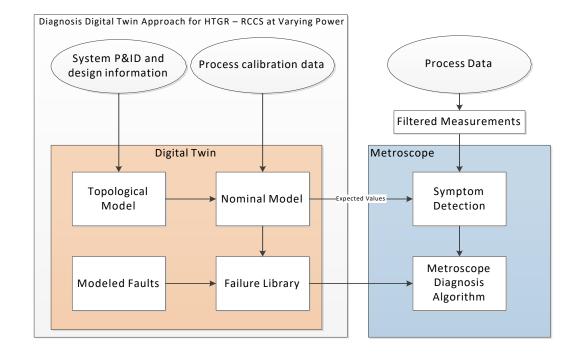
"If we change the digital twin, we change the diagnostics"



## In a nutshell

Opportunity: Create trusted, reliable
Digital Twin-based diagnoses of equipment
problems for all types of plant auxiliary
systems – beginning with advanced
reactors

**Success:** Achieve a generalized approach to digital twins that can anticipate new faults over a variety of system modes and can be deployed in a commercial offer with demonstrated ROI



# Contributors



Builds the digital twin and preforms the research work



Provides diagnostics technology and software platform



Provides cyber-digital asset test data and technical consultation



Eric Helm Principal Investigator



Mary Beth Baker Project Manager



**Todd Matthews**TH Modeling and
Application Developer



Pascal Brocheny TH Modeling

## Other current focus

Tomorrow Christophe Duquennoy will be speaking about diversification on cooling tower and return of experience at EDF

Digital twin are already ready for Gas Plants, Diesels Plants but also common industrial assets such as Data Center cooling processes

Over 50GW and 60 industrial units equipped worldwide

# metroscepe

Thank you for your attention











### **Advanced Sensors and Instrumentations**

**Technical Session – September 14, 2021** 

**Moderator: Hasan Charkas, EPRI** 

Pattrick Calderoni, INL
Jorge Carvajal, Westinghouse
Molly Strasser, Xcel Energy
Matt Hertel, X-energy
Hash Hashemian & Brent Shumaker, AMS



February 10, 2021

#### Pattrick Calderoni - INL

National Technical Director, Advanced Sensors and Instrumentation Manager, Measurement Science Department

# Advanced Sensors and Instrumentation for Digital Twin applications

Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization September 14, 2021



### **DOE Advanced Sensors and Instrumentation program**

#### **Mission**

Address **critical technology gaps** for monitoring and controlling existing and advanced **reactors** and supporting **fuel cycle** development



Reliable, cost-effective, real-time, accurate, and high-resolution measurement of the performance of existing and advanced reactors core and plant systems



Resilient, real-time transmission of sufficient amount of data for online monitoring and advanced data analytics

#### **Vision**

ASI research results in advanced sensors and I&C technologies that are qualified, validated, and ready to be adopted by the nuclear industry



Machine learning and artificial intelligence processes to enable semi-autonomous operation and maintenance by design



Enable near real-time control of plant or experiments process variables to enhance performance

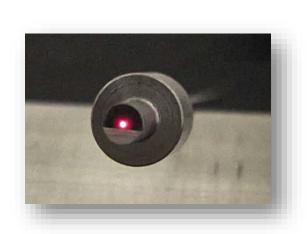
### **Program objectives FY22-25**

#### Sensors for advanced reactors

- Develop advanced sensors (multi-mode; multi-point/distributed; miniature size and limited or no penetrations) and supporting technology (rad-hard electronics, wireless communication, power harvesting) for nuclear instrumentation
- Demonstrate nuclear instrumentation performance in conditions relevant to advanced reactors (including irradiation)
- Establish a supply chain for advanced reactor instrumentation (fabrication and services)

#### • **Instrumentation** for irradiation experiments

- Provide real time instrumentation and passive monitors to measure local operational parameters (neutron flux, temperature, pressure, mechanical solicitations) in TREAT, ATR, HFIR and MITR experiments
- Develop methods to characterize nuclear fuel and material properties (thermal conductivity, microstructure, mechanical behavior) during irradiation

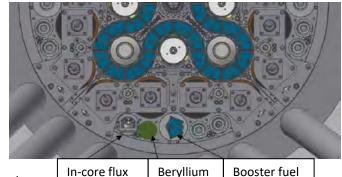






Instrumented capsule for LWR fuel safety test in TREAT

Neutron flux sensors (SPND, fission chambers and dosimetry) deployed in ATRC to characterize the I-loop booster performance



element

sensors

### **Program objectives FY22-25**

#### Digital Technology for advanced reactors

- Integrate advanced sensors and instrumentation in Nuclear Digital Twins (NDT) with Hardware in the Loop simulation for the phased demonstration of performance-based control algorithms to enable autonomous operation
- Develop condition monitoring technologies for anomaly detection, diagnostics, prognostics, and decision making that can operate on streaming data
- Develop modeling and simulation tools for communication technologies to support integration with control systems

autonomous operation of advanced reactors **AUTOMATED CONTROL PERFORMANCE MONITORING** Supervisory algorithm Physics-based pattern recognition **OFF-NORMAL EQUIPMENT HEALTH MONITORING DECISION MAKING** Automated reasoning diagnostics Reinforcement Learning **MAINTENANCE SHEDULING** Markov Process optimization

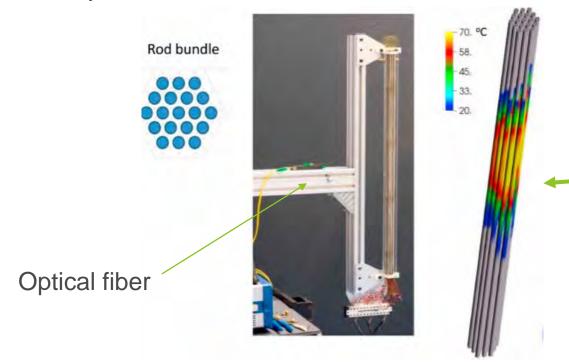
A logical progression towards sensor-based

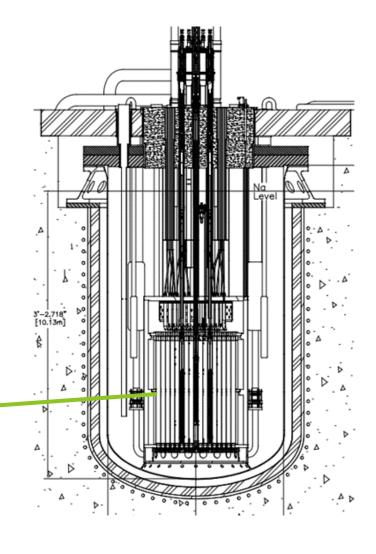
### **Sensing modalities**

#### In-core

 Multi-point and distributed measurements of process variable fields

 Core-wide estimation of temperature, power, approach to safety limits etc.



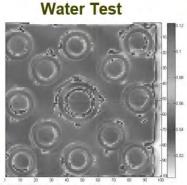


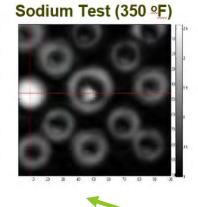
### **Sensing modalities**

#### In-vessel

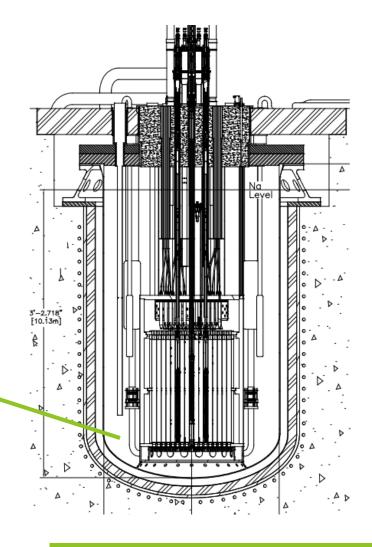
Imaging opaque environments for in-service inspection







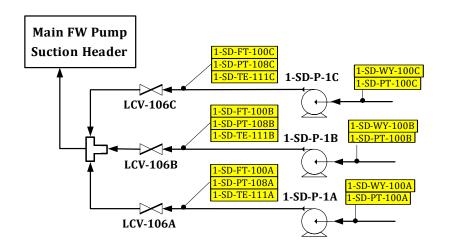


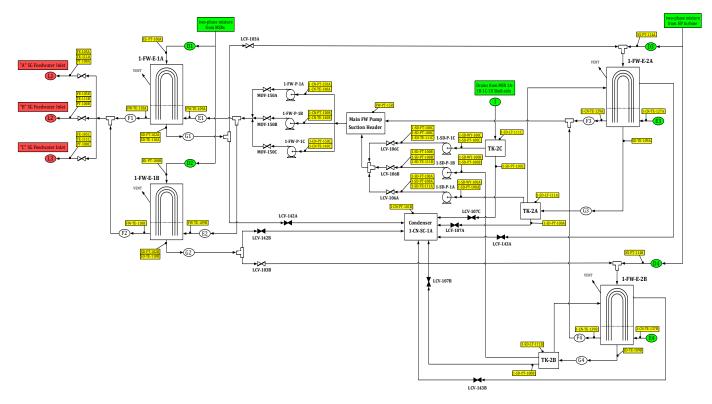


### **Sensing modalities**

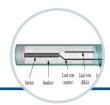
#### **Ex-vessel**

 Plant-wide sensing combined with process models for full state awareness



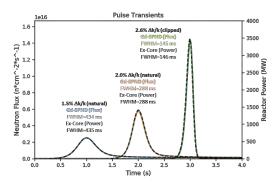


#### Technology demonstration – in-core neutron flux sensors

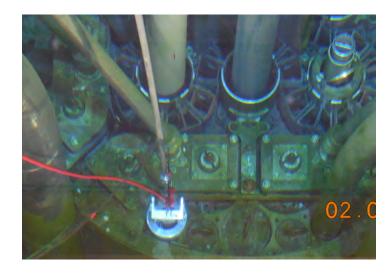


- Neutron and gamma detection
- Fast (Hf, Gd) and slow (Rh, Vd) time response
- Established design and fabrication process at INL
- Calibration and temperature compensation development in NRAD (summer 2021)
- Performance demonstration in TREAT, AGR5/6/7, ATRC, and MITR (FY22)

#### Self Power Detectors



TREAT pulse transient with Gd- and Hf-SPNDs compared to an ex-core detector.

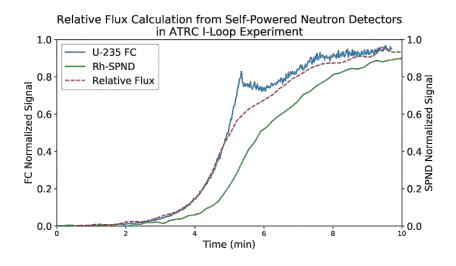


Test rig was installed in ATRC I-13 position on 2/11/2021 and irradiated for six hours.

- VTR project on fast spectrum SPND design optimization using ORNL Geant4 code for Ta emitter
- NRAD test include fission chambers from CEA (Loic Barbot 2 months visit) and Photonis

#### **ATR-C test objectives:**

- Testing instruments in representative environments (SPND, FC, dosimeters),
- Developing key domestic expertise for incore instrumentation,
- Supporting characterization of test positions (Iloop booster)



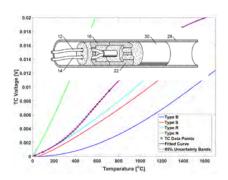
#### Technology demonstration – TCs for fuel and cladding materials

<0.5 seconds



- Mo-Nb junction for high temperature applications (1600 C) and low drift under neutron irradiation
- Performance demonstration in AGR5/6/7 – highest temperature ever recorded in pile without drift (1482 C)
- Design optimization: corrosive environments, multi-point detection
- Commercialization: TCF with Idaho Labs Corp, ASTM standard and industrial qualification at AMS

#### High Temperature Irradiation Resistant (HTIR) thermocouple



HTIR response compared with standard types

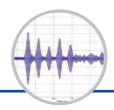
Table 1: Summary of performance parameters for the HTIR-TC

**Response Time** 

Performance	Performance Requirement	Performance	
Parameter	Fuel Test Application	Requirement	
		Stand-Alone Application	
Temperature	Room Temperature - 1600°C	Room Temperature -	
Range		1600°C	
Accuracy	Not Specified	±1%	
Drift	3% for 4.5 x 10 <sup>21</sup> nvt	3% for 4.5 x 10 <sup>21</sup> nvt	
	(thermal)	(thermal)	
Life	4.5 x 10 <sup>21</sup> nvt (thermal), or 10	18 months or 4.5 x 10 <sup>21</sup>	
	thermal shocks (room	nvt (thermal)	
	temperature to 1600°C)		
Mechanical			
Ruggedness:			
Rugged Junction	Rugged mechanical junction	Rugged mechanical	
	design	junction design	
Bend Radius	Minimum of 0.5 inch	Minimum of 0.5 inch	
Thermal Shock	5 sudden startups and 5	100°C/hr	
	sudden shutdowns—each	No.	
	causing a thermal shock on		1/2-
	the order of room temperature		
	up to 1600°C		

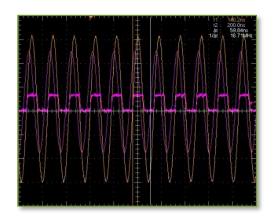
<0.5 seconds

#### Technology demonstration – acoustic sensors



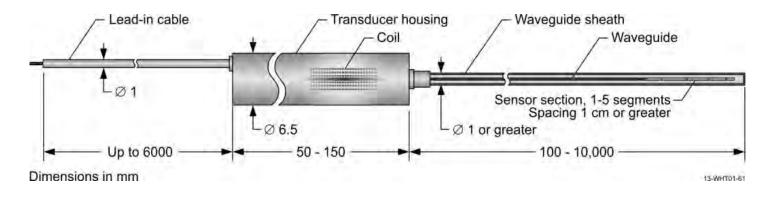
- Ultrasound based sensors enable distributed temperature measurements up to 2200°C
- INL had demonstrated the reliability of magnetostrictive material transducers under irradiation
- Current research focuses on waveguide design optimization and unfolding signal response of distributed measurements

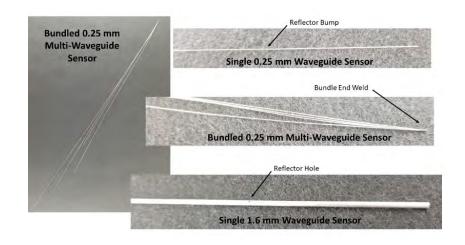
#### Ultrasonic Thermometers



UT operational window defined in high temperature furnace (up to 2200 C):

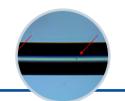
- Sheathed multi-waveguide materials: 316-stainless steel, lanthanated molybdenum, and zircaloy-4
- Non sheathed single waveguide materials: 316-stainless steel, molybdenum and tungsten





- UT performance modeling and design optimization is ongoing
- UT demonstration continues in irradiation experiments (DISECT in BR2, TREAT, MITR)
- Consolidating work on Surface Acoustic Wave sensors development based on radiation resistance materials (AIN, LiNbO)

#### Technology demonstration – optical fiber sensors



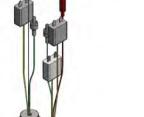
- Advanced sensor configuration and interrogation techniques to measure:
  - Distributed temperature, strain and vibration
  - Fission gas pressure and composition
- Engineering solutions for sensor packaging, pressure feeds
- Active compensation techniques for OF sensors operating in radiation environments

Optical Fibers

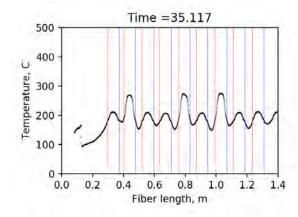


Prototype optical fiber pressure sensor based on Fabry-Perot interferometry

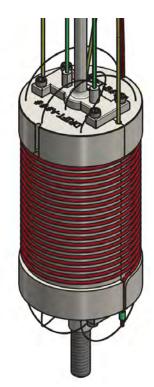
Optical Frequency Domain Reflectometry (OFDR) for temperature mapping of TREAT heat sink

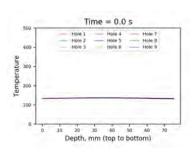


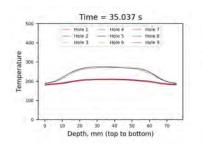
Distributed Temperature Sensing (DTS) for DRIFT experiment in TREAT: temperature profile along the length of a **single** fiber

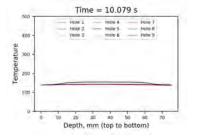


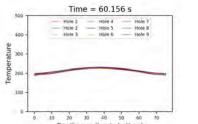












Spatially resolved time dependance:

- Black traces are radially closer to fuel
- Excellent symmetry
- Effects of heat sink become more important after 1 min

#### Technology development



- Ultrasonic transducers and sensor systems operating in extreme conditions with long operational life
- The use of Z-cut LiNbO3 piezoelements coated with Cr/Au thin film allows high temperature operation (800 C) and rad tolerant
- Broadband transducer (bandwidth up to 30 MHz) now being coupled with temperature, pressure, flow and SHM acoustic emission sensors

**Radiation Endurance Ultrasonic** Transducer, X-wave Innovation Inc



REUT prototype, (left) Version 1, (right) Version 2, and (up) CAD model of Version 3

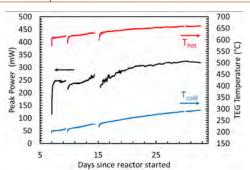
Courtesy of Uday Singh and Dan Xiang, SBIR grants recipients



- High-temperature & high-powerdensity capability for in-core or invessel power harvesting
- Technology development through ASI funded project at the University of Notre Dame (Yanliang Zhang)
- Performance demonstration in MITR though NSUF funded project

#### Thermoelectric generators (TEGs) for power harvesting

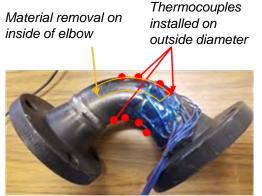
KEY METRICS	Half- Heusler TEG	COMMERCIAL TEG
Power density (W/cm²)	2-3	0.2-0.3
Max Operating temp	650 °C	250 °C





- Automated technology coupled with advanced data analytics for assessing the health of pipes in nuclear power plants as the pipe material degrades due to corrosion
- Combines innovations in materials for sensing both chemical and mechanical degradation with statistical algorithms based on Bayesian modeling.

#### **Diagnostics and Prognostics of Corrosion Processes in Pipes**



D. Adams, Vanderbilt University

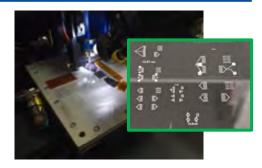
Y. Zhang. University of Notre Dame

V. Agarwal, Idaho National Laboratory

- courtesy of Tim McIntyre

- Passive wireless sensor technology based on a network of digitally printed radio frequency (RF) surface acoustic wave (SAW) sensors
- Enable multi-point and multi-mode sensing (temperature, hydrogen gas, voltage, and current)
- Demonstration included two-antenna relay for communication / sensing through RF-opaque materials

**Direct Digital Printing of Passive Wireless Sensors** 



Printed SAWs developed and fabricated at ORNL





# Non-Intrusive Temperature and Pressure Wireless Sensor and Transceiver System for Extreme Environment Applications

Jorge Carvajal, Fellow Engineer September 2021

Presented at "Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization" workshop



### Sensor Applications and Benefits



- Non-intrusive real-time data
- Accelerates new fuel development
- Accelerates new fuel licensing
- Increase operating margins at plants
- Enhances instrumentation for DOE test reactors





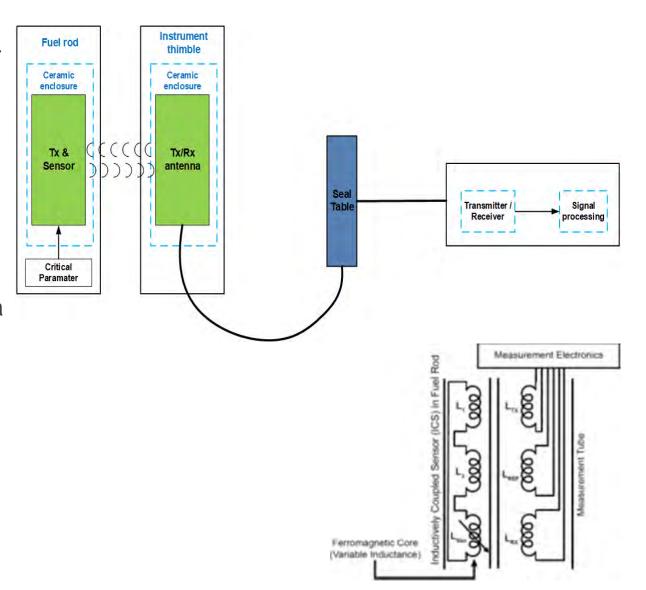
- Both Commercial & DOE canisters
- Accelerates licensing of casks
- Improved efficiency of interim storage
- Develop basis for dry storage of new fuels
- Utility fuel handling no longer reliant on conservative models:
  - Fuel can be removed from pool earlier
  - Reduced drying cycles
- Increased heat load margins in storage
- Enables canister surveillance monitoring



- · Applications requiring wireless real-time surveillance in extreme conditions
- Nuclear safeguards
- Nuclear defense
- Gas pipelines
- Gas storage
- · Etc.

### In-Rod Sensor System Introduction

- Wireless sensor located in a fuel rod provides realtime data
  - centerline fuel pellet temperature
  - pellet elongation
  - rod internal pressure
- Facilitates licensing of new fuel products
- Enhances plant operation through improved utilization of margins
- Similar to in-rod data collection techniques used in test reactors such as Halden but with wireless data transmission
- Delta in the coupling amplitude is proportional to measurement of interest
- U.S. patent application serial No. 16/214445 and 16/564150



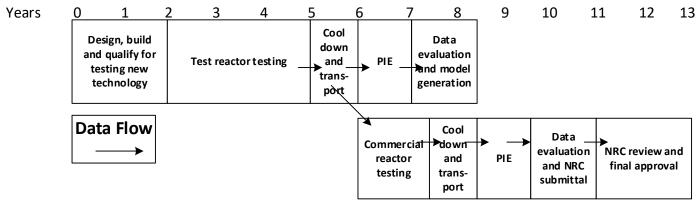


#### Benefits

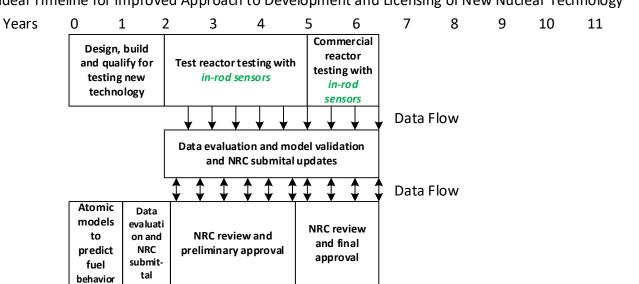
- Non-intrusive real-time data instead of the typical "cook and look" significantly accelerates development
- New approach significantly reduces time for licensing, increases reliability of models and therefore licensing decisions
- Significant enhancement to in-core instrumentation for test reactors, especially National Laboratory Reactors (e.g. INL ATR and ORNL HFIR)



Ideal Timeline for Current Approach to Development and Licensing of New Nuclear Technology



Ideal Timeline for Improved Approach to Development and Licensing of New Nuclear Technology

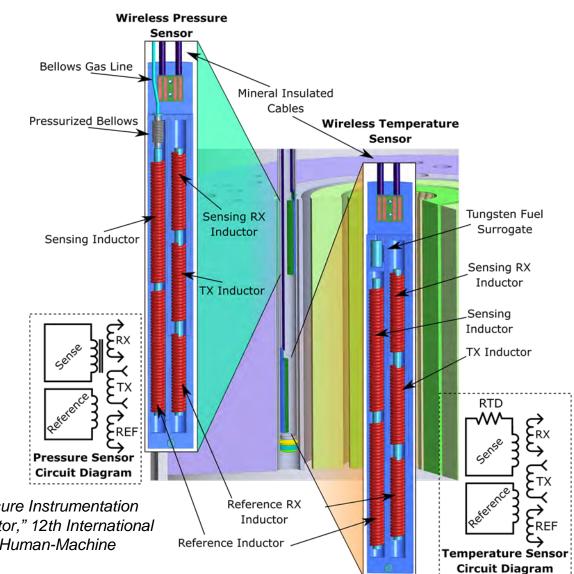


### In-Rod Sensor System Next Step

- ORNL HFIR Irradiation & PIE (NSUF project)
  - Temperature & Pressure sensor to be installed in HFIR late 2021
  - Expected temperature range 300°C 500°C
  - 5 cycles, total dose ~3X10<sup>21</sup> n/cm<sup>2</sup>
  - Fuel surrogate simulates temp source
  - External gas injection simulates rod internal pressure
  - Continuous data collection
- WEC Long Term Temperature test
  - Two sensors to be installed in autoclave at prototypical fuel rod pressures
  - Held at ~400°C for 6 12 months
  - Continuous data collection

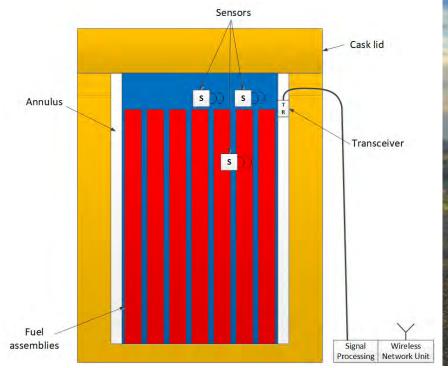


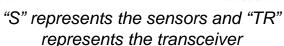
P. Mulligan, "In-Core Neutron Flux, Temperature, And Pressure Instrumentation For The Wire-21 Experiment In The High Flux Isotope Reactor," 12th International Conference On Nuclear Plant Instrumentation, Control, and Human-Machine Interface Technologies (NPIC & HMIT), June 2021.



### Dry-Cask Sensor System Introduction

- Technology derived from the Westinghouse Accident Tolerant Fuel (ATF) In-rod sensor development
- Wireless sensor located inside dry cask steel canister provides real time data such as temperature and pressure
- Sensor does not require penetrations to the canister
- Multiple sensors can be interrogated simultaneously
- Sensor lifetime > 40 years (based on dry cask 3.5 Grad 40-year TID estimate).
- Long term maximum operating temperature 425°C
- U.S. patent application serial No. 16/448706









#### Customer Value – Utilities, NRC, DOE

- Online data potentially enables the reduction in time and cost of spent fuel management. Utilities are no longer reliant on overly conservative peak clad temperature (PCT) models, which prescribe:
  - When fuel is removed from spent fuel pool
  - Heat load margins in dry storage canisters
  - Drying cycle duration
  - Canister susceptibility to degradation (e.g. chloride induced stress corrosion cracking, etc.)
- Accelerates dry storage and transportation of advanced fuels
  - NRC will require data to develop technical basis of dry storage of new materials
  - Synergy with high enrichment, high burnup, and ATF
- Dry cask inventory at utility is **growing** given the lack of a permanent repository location



### Dry-Cask Sensor System Next Step

#### WEC/EPRI Joint Sensor Demonstration Project Objectives:

Accelerate development and qualification of remote wireless sensors for use in extreme environments, e.g. high radiation fields and temperatures

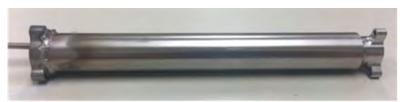
- Phase I: Establish Design Criteria
  - Develop Sensors Performance Requirements Document
  - Identify Key Objectives, Functional Criteria, Survivability Needs, etc.
    - Separate Dry Storage, Transportation, and Disposal Considerations
- Phase II: Laboratory Scale Testing and Qualifications
- Phase III: Full Scale Testing and Validation
  - Canister Mockups and Loaded Canister Demonstrations



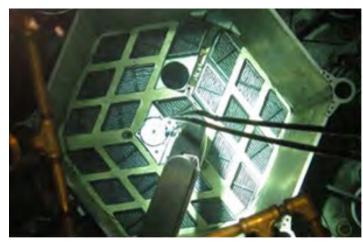
### MITR Test (April 2019) Power cycling & temperature measurement results



Ceramic enclosure housing sensor components

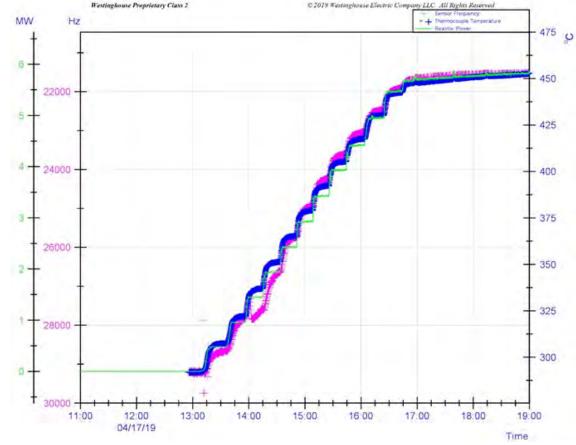


SS enclosure holds ceramic enclosure within autoclave PWR loop



MITR Core with water loop position open

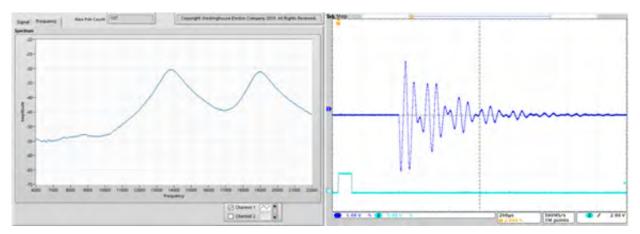
Neutron flux =  $1e14 \text{ n/cm}^2/\text{sec}$ Coolant temperature = 300°C Component max temperature (gamma heating) ~ 500°C



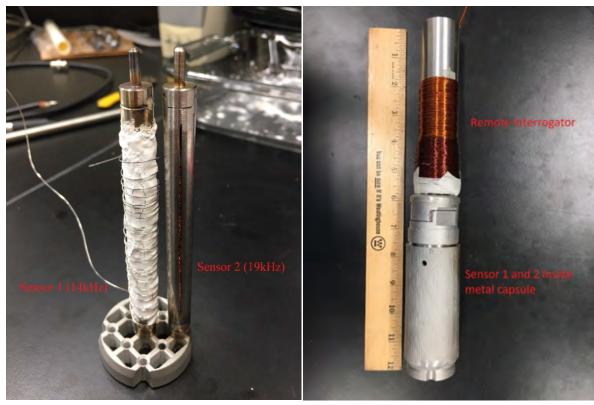


### Functional Test – Multiple Sensors

- Two resonant sensor circuits assembled in a metallic enclosures and remotely interrogated
- Received signal cannot be resolved in time domain. FFT necessary to resolve signal



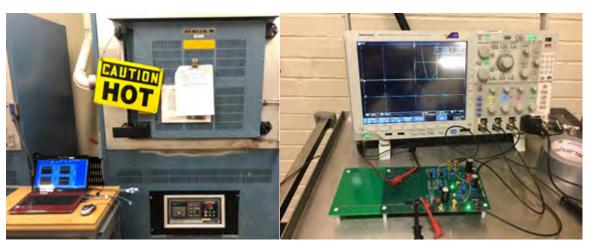
Multiple sensors in the frequency (left) and time (right) domain



Sensors at two unique frequencies with remote interrogator



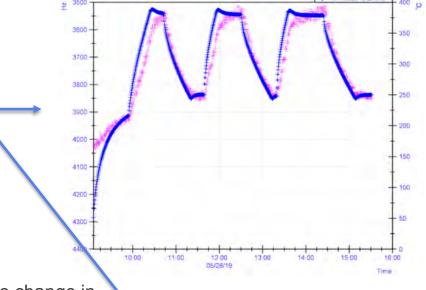
### Temperature & Gamma Irradiation Test



High temperature oven & data acquisition system

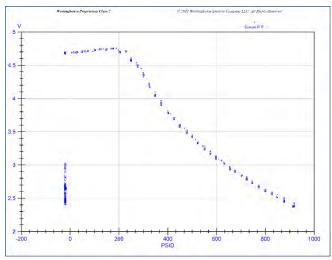


Gamma testing: Clean Hot Cell exterior (left) and interior (right)



-Minimal to no change in inductance values

-Slight increase in inductor's series resistance



Sensor signal vs. pressure

Blue - TC

Purple - Sensor



## Thank you!

Jorge Carvajal, carvajjv@westinghouse.com



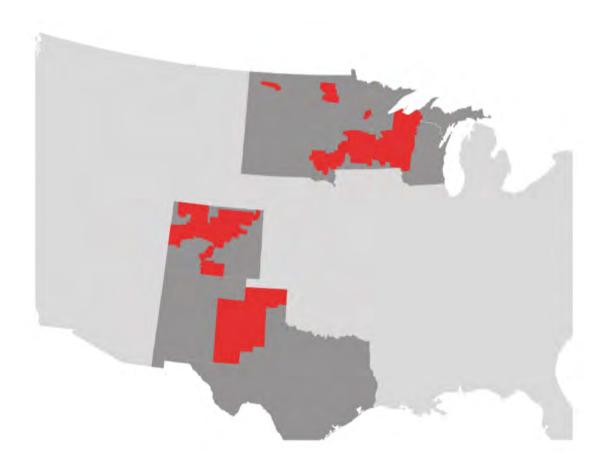


Data Analytics and Remote Monitoring Integration

**Molly Straser** 

## **Xcel Energy Company Overview**

- 3.7 million electric customers
- 2.1 million natural gas customers



© 2021 Xcel Energy





# Monitoring & Diagnostics (M&D)

Utilizing advanced analytics to proactively identify equipment risk, leading to reduced maintenance expense and unplanned outage impact, while enabling condition-based maintenance practices and operational excellence.

Advanced Pattern Recognition (APR) predictive analytics identify statistical deviations of modeled operating parameters.



# Xcel Energy Monitoring & Diagnostics (M&D)

2014: Inception - 7 plant pilot program

2016: Major thermal units included:14 plants, 35 units

2017: Expansion to wind, 3 nuclear units

2020: 13,135 MW thermal generation

 846 MW Wind generation, 2000 MW Wind EOY 2020

Predictive Analytics - Mechanical condition monitoring

Failure modes limited by plant instrumentation
 \$29M avoided and hard savings through Q1 2020

>4000 actionable advisories

### M&D Success & Results



#### **Operational Excellence**

- Excessive desuperheat spray leading to HRSG tube damage
- Feedwater regulator closed resulting in low fuel gas temp – damage prevented
- Excessive feedwater heater draining low levels, erratic levels
- Poor condenser performance, efficiency impact



#### **Avoided Cost Examples**

- Wind turbine gearbox numerous early gear defects avoiding gearbox replacement ~ \$350k per event
- Air heater guide bearing temperature increase, lube oil supply problem corrected
- Steam turbine vibration changes, balancing prior to forced event
- Fan bearing temperature increase, cooler operation corrected preventing failure
- Boiler acoustic leak indication, operation mitigation until scheduled outage



#### **Direct Savings Examples**

- Major maintenance deferrals, known good condition and performance, i.e., BFP overhaul elimination ~ \$250k
- Capital budget reduction for wind turbine gearbox replacements, early fault identification and known condition
- Condition based maintenance known good condition allows for delay or elimination of scheduled or calendar-based maintenance – expansion priority



## **Nuclear Focus**

# VERIFYING COMPLIANCE THROUGH TECHNOLOGY

#### CORRECTIVE ACTION PROGRAM

Improved CAP screening tools with the help of Idaho National Laboratory reduces the resource burden

#### PREVENTIVE MAINTENANCE

Maintain equipment at the optimal time using data trends from new wireless sensors and advances in machine learning

#### **OPERATOR ROUNDS**

Critical data for trending plant performance with wireless sensors and machine learning application developed with USA

#### INFRASTRUCTURE

Permanent Wi-Fi in plants serves as backbone for technological improvements

#### STAKEHOLDER AUDITS & INSPECTIONS

New process for automated data sharing supports inspections and key performance indicators

### TECHNICAL SPECIFICATION SURVEILLANCES

Increased public safety by monitoring plant equipment with wireless sensors ensures continual compliance with technical specifications





# Xcel Energy Nuclear Innovation: Sensor Infrastructure

#### **Mechanical Sensors**

- Vibration Sensors
- Wireless Gauge Readers
- Void Monitoring
- Remote Radiation Mapping
- Valve Position Indication

#### **Electrical Sensors**

- EPRI Acoustic Monitors for Transformers
- EPRI Disconnect Switch Monitor
- Continuous Thermal Imaging

#### **Wi-Fi Devices**

AR Headsets / iPhones / Tablets

# Xcel Energy Nuclear Innovation: USA Advanced Remote Monitoring

- Xcel is working with INL to begin development of a method to streamline current pain points in the M&D architecture
- Part of a larger initiative with collaboration with USA plants, as well as Idaho National Lab
  - Standardized Monitoring and Diagnostics (M&D) Software Platform
  - Automatic thermal performance and fire detection using image/video recognition tied into cyber compliant systems
  - Beginning to automate operator round data collection
  - Transformer and cycle isolation monitoring
  - Begins to apply machine learning to monitoring limits (Xcel collaboratively with INL)



### **Questions?**

Molly Strasser

Nuclear Innovation Manager

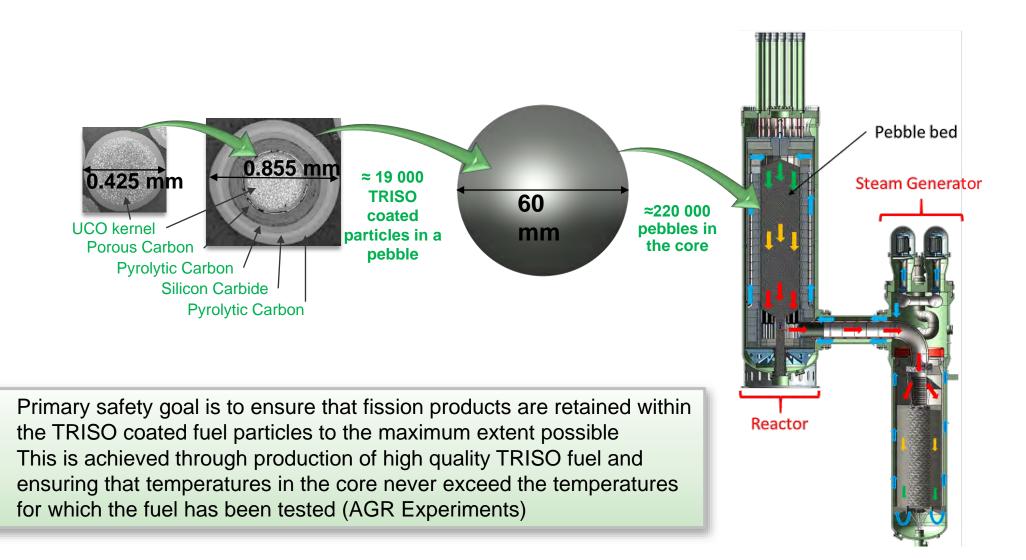
Molly.J.Strasser@xcelenergy.com

# 2021 Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization





#### **UCO TRISO Particle – Primary Fission Product Barrier**

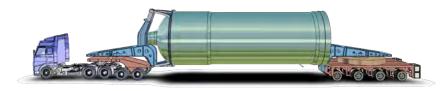


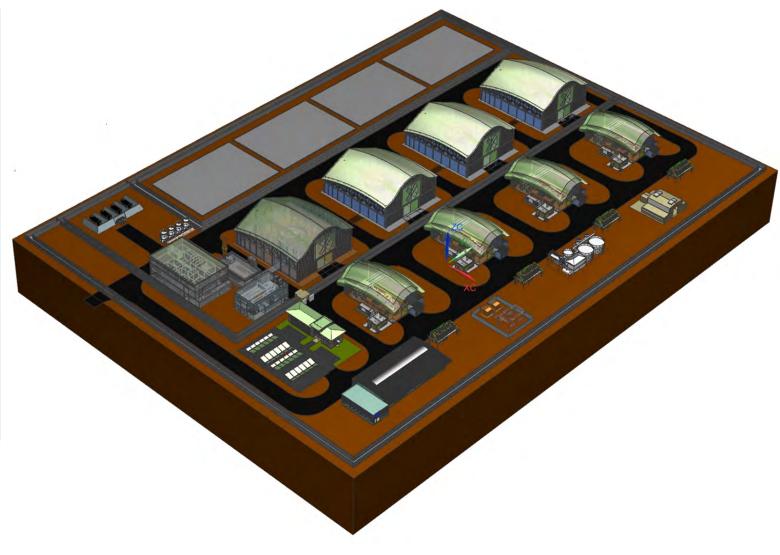


#### **Background: Xe-100 Plant Overview**

Standard X-energy plant have 4 Reactors - 4 Turbines producing 320 MWe, attributes include:

- 200MWth/80MWe Per Module
- Process heat applications Proven intrinsically safe
- Meltdown proof
- Walk-away safe
- Modular construction
- Requires less time to construct (2.5-4 years)
- Road transportable for diverse
- geographic areas
  Uses factory-produced components
  Load-following to 40% power within 15 minutes
- Continuous fueling; resilient on-site fuel storage

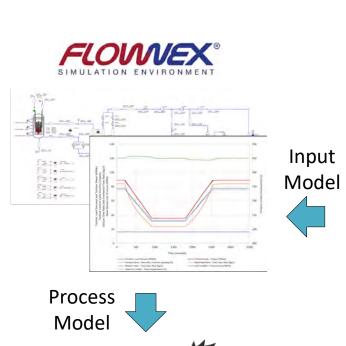








#### **Digital Twin I&C Development Cycle**



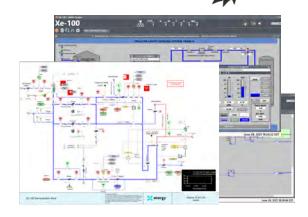
#### **I&C Systems Design**



Initial Design Start Here



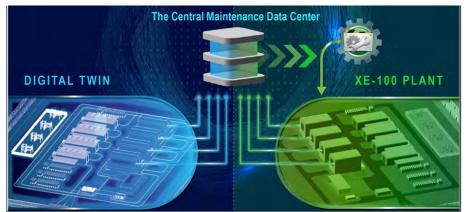




SimGenics



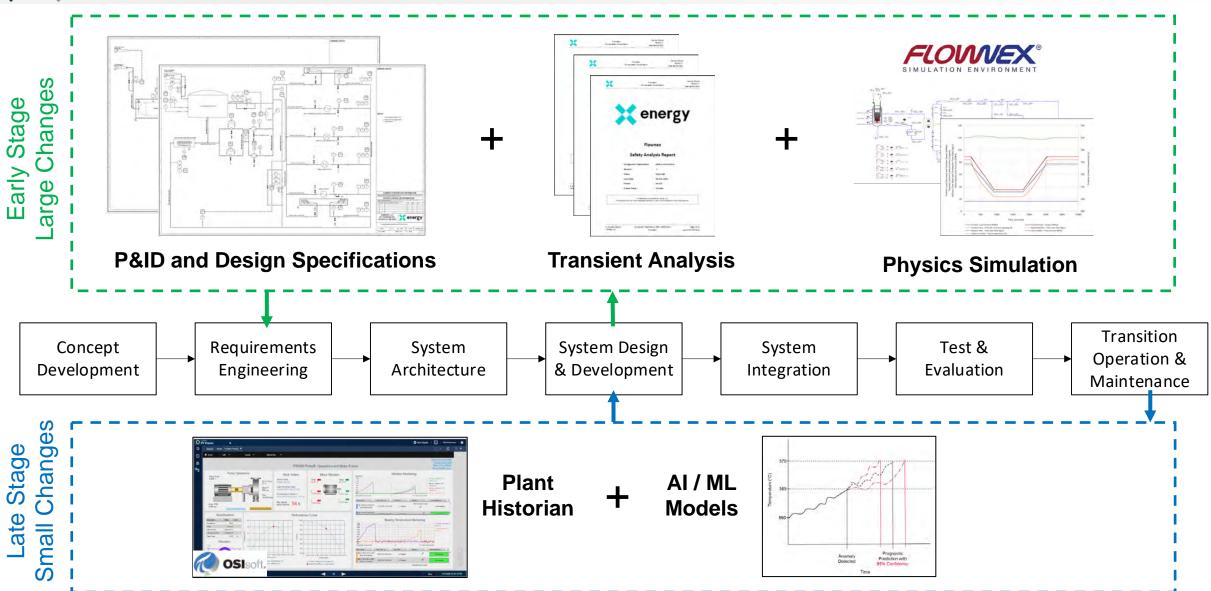






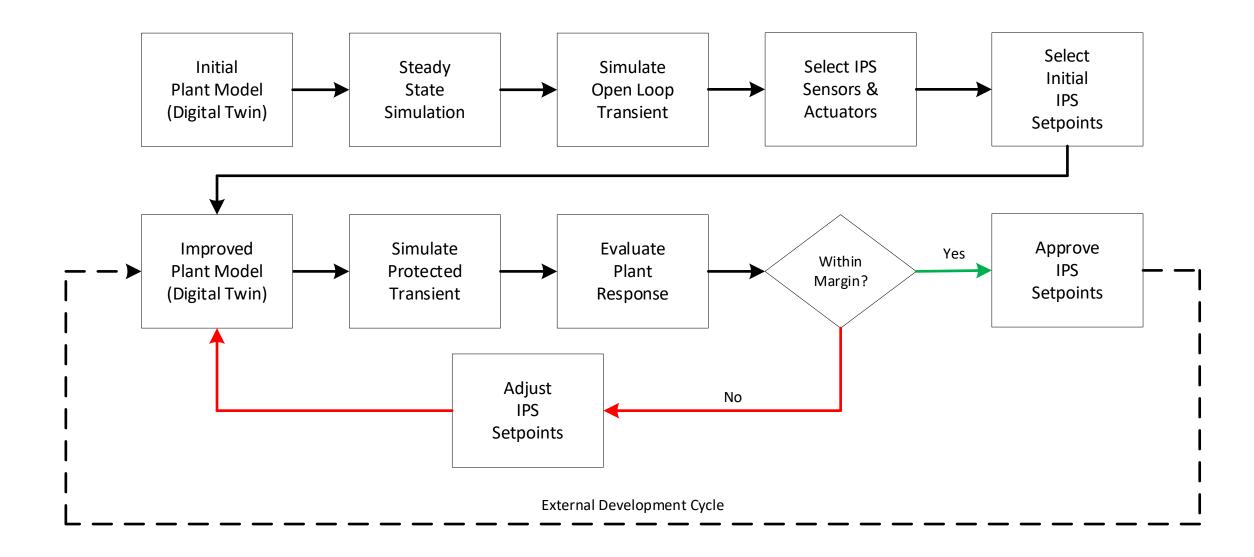


#### I&C Systems Engineering Process Flow + Digital Twin Toolsets



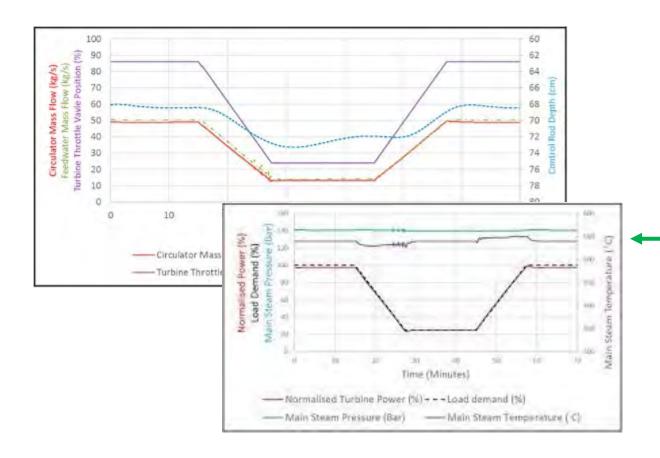


#### **DT Enabled Design by Analysis: IPS Trip Setpoints**





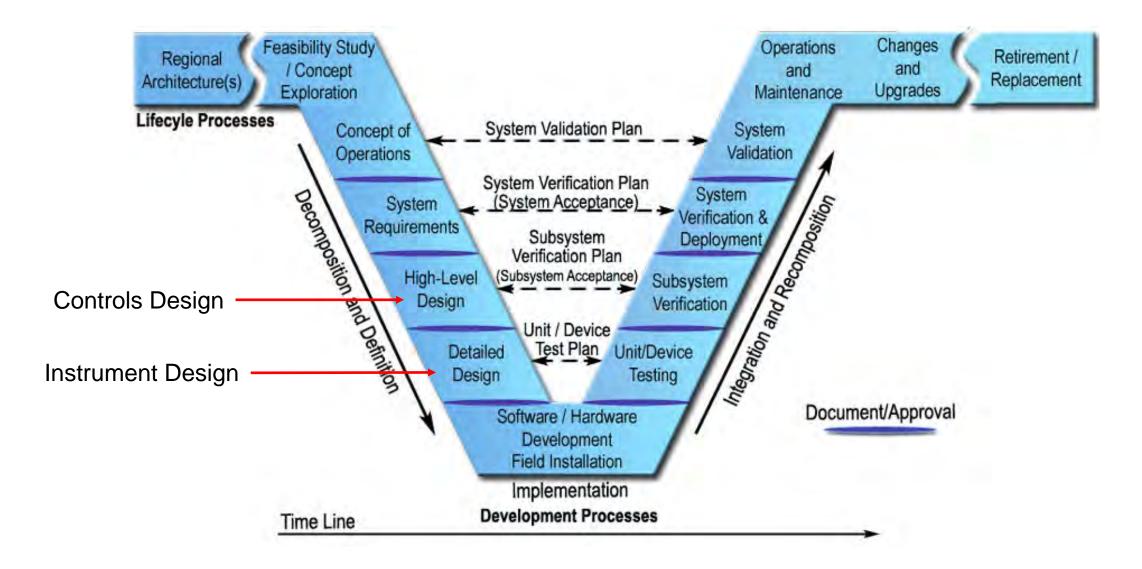
#### DT Enabled Design by Analysis: Control Trade Study



	Common		
Controlled Variable	Manipulated Variable	Secondary Variable	
Feedwater Valve dP	Feedwater Pump Speed	Feedwater Valve Flow	
Deaerator Level	CIFW Valve Position	CIFW Flow	
Deaerator Pressure	HPT Valve Position	HPT Valve Flow	
FWP Recirculation Flow	FWP Recirculation Valve Position	Flow/Speed Ratio	
Turbine Header Pressure	Turbine Bypass Valve	N/A	
	Option 1		
Controlled Variable	Manipulated Variable	Secondary Variable	
Main Steam Pressure	Feedwater Valve Position	Feedwater Flow	
Reactor Outlet Temperature	Control Rod Depth	Reactor Power	
Main Steam Temperature	Circulator Speed	Helium Mass Flow	
Turbine Power	Turbine Throttle Valve Position	Steam Flow	
	Option 2		
Controlled Variable	Manipulated Variable	Secondary Variable	
Main Steam Pressure	Circulator Speed	Helium Mass Flow	
Reactor Outlet Temperature	Control Rod Depth	Reactor Power	
Main Steam Temperature	Feedwater Valve Position	Feedwater Flow	
Turbine Power	Turbine Throttle Valve Position	Steam Flow	
	Option 3		
Controlled Variable	· · · · · · · · · · · · · · · · · · ·	Co condom: Vorighto	
	Manipulated Variable	Secondary Variable	
Main Steam Pressure	Feedwater Valve Position	Feedwater Flow	
Reactor Inlet Temperature	Circulator Speed	Helium Mass Flow	
Main Steam Temperature	Control Rod Depth	Reactor Power	
Turbine Power	Turbine Throttle Valve Position	Steam Flow	
	Option 4 (THTR)	<u> </u>	
<b>Controlled Variable</b>	Manipulated Variable	Secondary Variable	
Main Steam Pressure	Circulator Speed	Helium Mass Flow	
Reactor Inlet Temperature	Feedwater Valve Position	Feedwater Flow	
Main Steam Temperature	Control Rod Depth	Reactor Power	
Turbine Power	Turbine Throttle Valve Position	Steam Flow	



#### DT Enabled Design by Analysis: Xe-100 Next Steps





#### INNOVATING NUCLEAR TECHNOLOGY





Brent Shumaker and H.M. Hashemian AMS Corporation

#### Presented for:

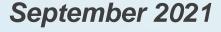
**2021 Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization** 













#### **All Plants Calibrate Transmitters Every Cycle**

#### **Procedure**

Prepare M&TE

Dress out for containment entry

Remove channel from service

Valve-off sensing lines

Inject test signal to transmitter

Calibrate (if needed)

Return everything to service

#### **Drawbacks**

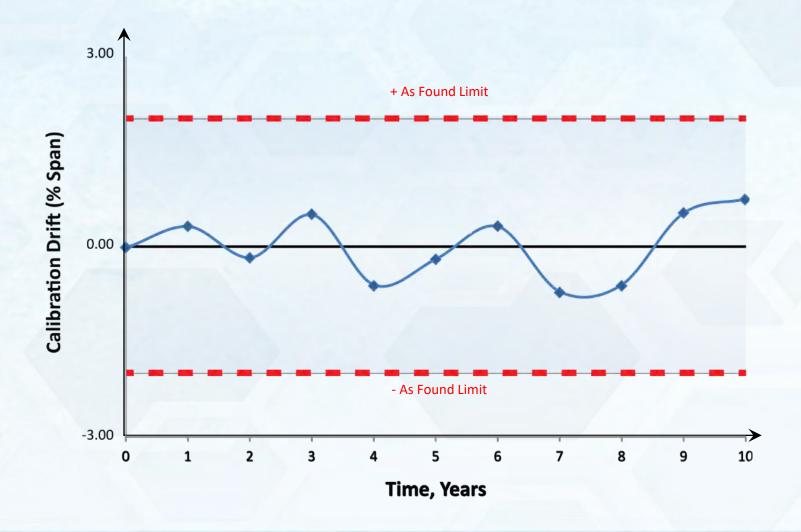
Radiation exposure
Potential to damage transmitters
Up to 5% experience maintenance-induced errors
Farley recently replaced numerous manifold valves due to wear and tear
Increased outage maintenance and critical path time







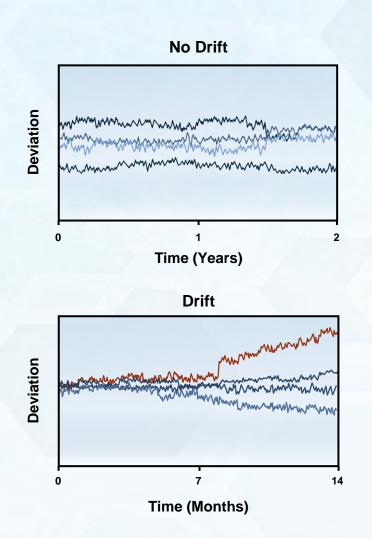
# Why Do We Say Transmitters Drift Very Little? (10-year calibration history of a typical nuclear grade transmitter)





# Online Monitoring (OLM) Identifies Drifting Transmitters (Actual PWR Plant Data)





#### **Traditional Calibration vs. OLM**

#### **Traditional Calibration**

Step 1. Determine if calibration is needed

Step 2. Calibrate if needed

#### OLM

Step 1. Determine if calibration is needed

Step 2. Calibrate if needed



# Typical OLM Results

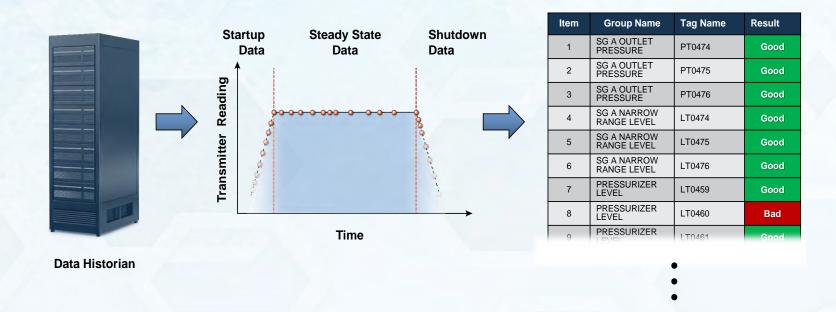
Item	Group Name	Tag Name	Result
1	SG C OUTLET PRESSURE	PT0494	Good
2	SG C OUTLET PRESSURE	PT0495	Good
3	SG C OUTLET PRESSURE	PT0496	Good
4	PRESSURIZER LEVEL	LT0459	Good
5	PRESSURIZER LEVEL	LT0460	Good
6	PRESSURIZER LEVEL	LT0461	Good
7	PRESSURIZER PRESSURE	PT0455	Bad
8	PRESSURIZER PRESSURE	PT0456	Good
9	PRESSURIZER PRESSURE	PT0457	Good
10	PRESSURIZER PRESSURE	PT0444A	Good
11	PRESSURIZER PRESSURE	PT0445A	Good
•	•	• \	•
•			

September 2021 **SLIDE 104 OF 14** www.ams-corp.com



# **OLM Process for Pressure Transmitter Calibration Extension**

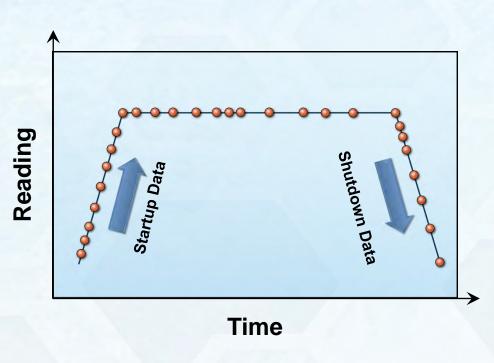
- Retrieve data from plant computer
- Analyze data to identify transmitters that have drifted out of tolerance
- Provide a list of transmitters to be calibrated

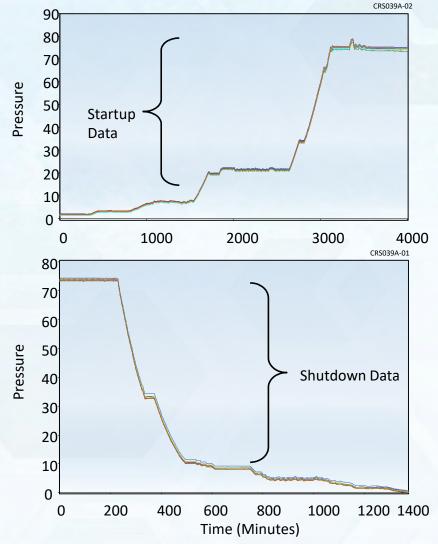


September 2021 www.ams-corp.com SLIDE 105 OF 14



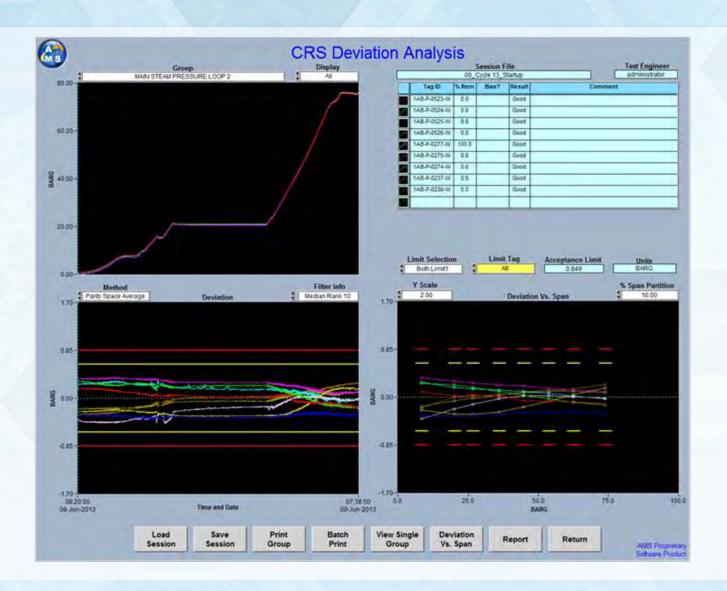
## M S Operating Range **OLM Checks Calibration Over Much of a Transmitter**







### **OLM Analysis at Startup**





# Commercial OLM Implementations in Nuclear Facilities

Sizewell B: 2005 - present

- Approved by UK's Nuclear Installations Inspectorate - 2005
- Calibration induced human error problems minimized

**Vogtle Units 1 and 2: 2018 - present** 

**Advanced Test Reactor: 2015 - present** 

Sizewell B



Vogtle

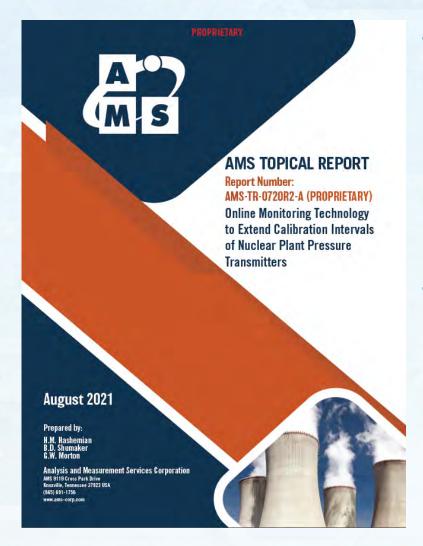


**Advanced Test Reactor** 





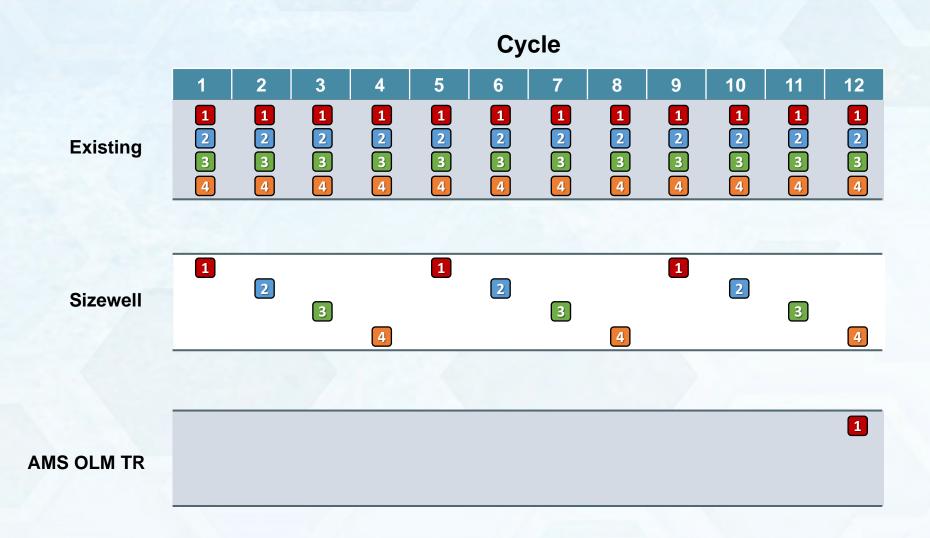
## **AMS OLM Topical Report**



- Step-by-Step OLM Implementation Methodology
  - Determine Calibration Intervals
  - Establish OLM Limits
  - Perform Drift Analysis
  - Perform Dynamic Failure Mode Assessment
- Example Changes to Existing Tech. Specs.
  - Description of OLM Program
  - Changes to SR Frequency Column
  - Changes to Bases Section of Each SR

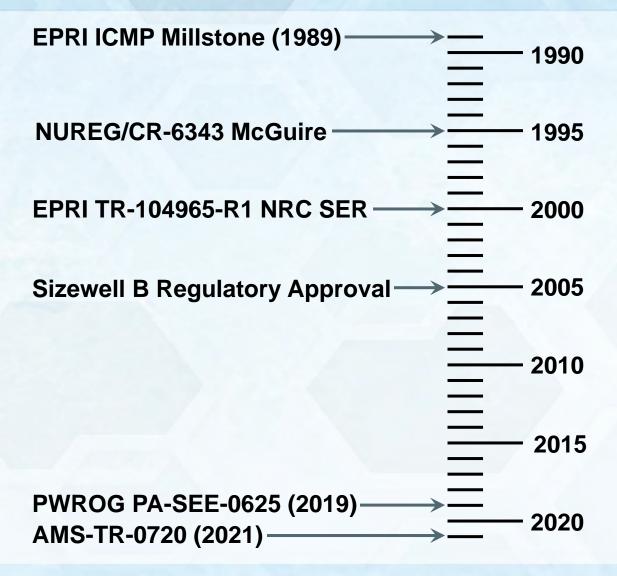


# **Example Calibration Schedules for a Group of 4 Redundant Transmitters**





# History of Transmitter Calibration Interval Extension







### **Thank You!**







Questions?











# Use Cases of Digital Twin Enabling Technologies in Nuclear Power Plants

**Technical Session – September 15, 2021** 

**Moderator: Gene Carpenter, DOE** 

Matthew Yarlett, Westinghouse Koushik A Manjunatha, INL Christophe Duquennoy, EDF Richard McGrath, EPRI Iikka Virkkunen, AALTO



Matthew Yarlett - PKMJ Technical Services, LLC

2021 Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization

September 15, 2021



In November 2018, the U.S Department of Energy selected PKMJ Technical Services, Idaho National Laboratory, and Public Services Enterprise & Group (PSEG) Nuclear, LLC for an Advanced Nuclear Technology Project.

https://www.energy.gov/ne/articles/us-advanced-nuclear-technology-projects-receive-18-million-us-department-energy

### **Project Objective**

Integrate advancements in online monitoring and data analytic techniques with advanced risk assessment methodologies

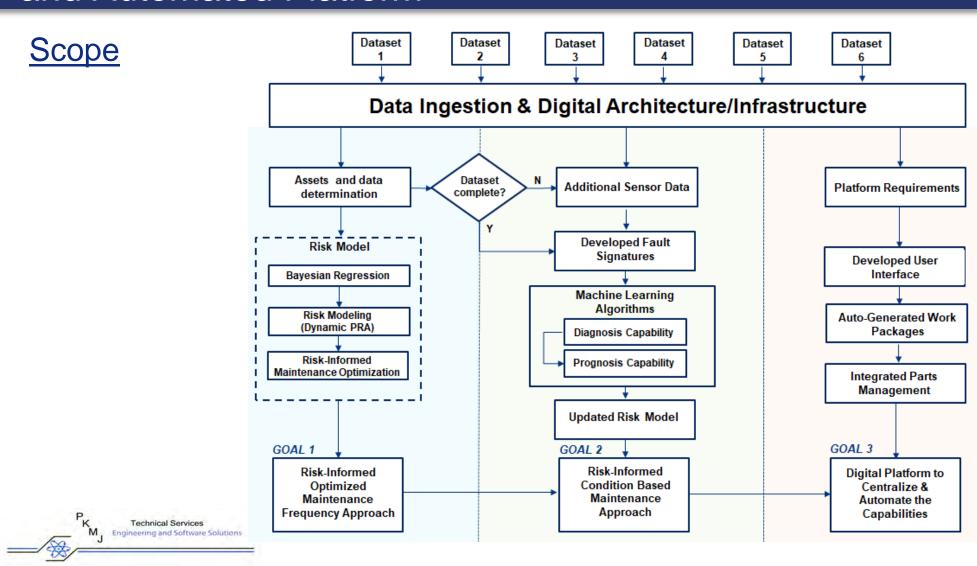
### Goals

- 1. Risk-informed approach to optimize equipment maintenance frequency
- 2. Risk-informed approach to condition-based maintenance
- 3. Develop and demonstrate a digital, automated platform

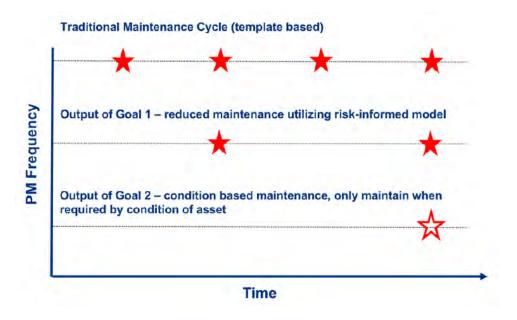
### **Duration**

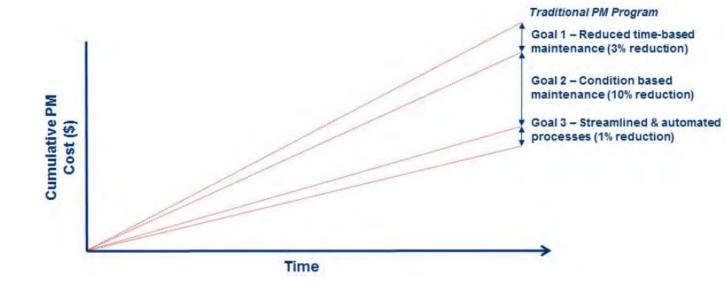
August 2019 - July 2021





### **Potential Benefits**







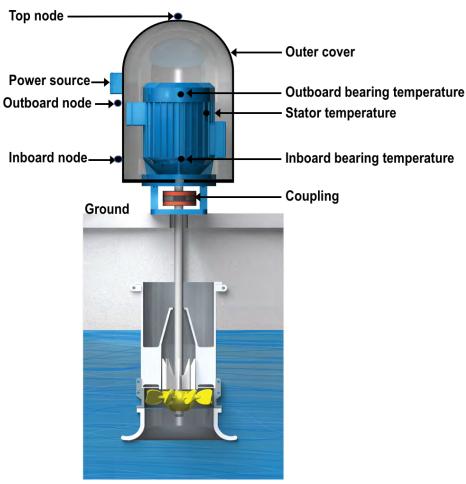
via Internet

# Integrated Risk-informed Condition-based Maintenance Capability and Automated Platform

### Accomplishments – Target Equipment Selected

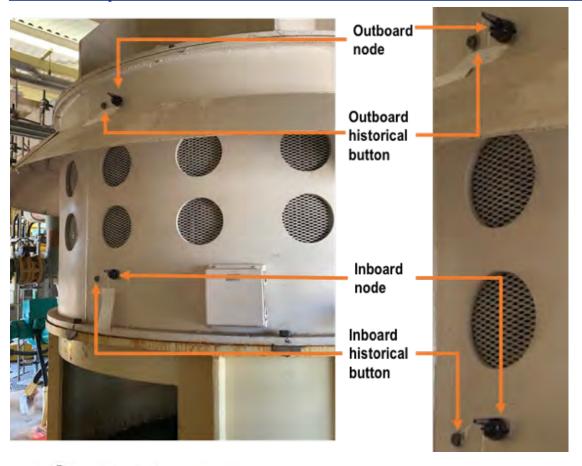


60 Vibration Sensor Nodes have been installed across Salem's 12 Circulating Water System pumps, motors and associated bypass valves. Each sensor node consists of two accelerometers sensitive to orthogonal in-plane motion and a temperature sensor.

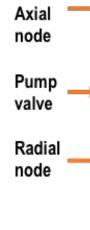




### Accomplishments – Wireless Vibration Sensor Installation









### <u>Accomplishments – Preventive Maintenance Optimization</u>

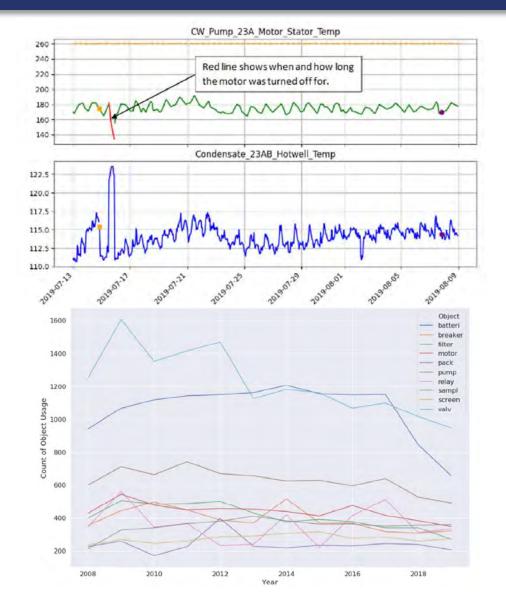
Equipment	Task Title	Current Frequency	Industry Average	Recommendation	Recommended Frequency
Pump	Refurbishment	6 years	14 years	Less Frequent	9 years
	External Visual Inspection	18/24 months	2.8 years	Кеер	18 months
Motor	Vibration Analysis	3 months	5.5 months	Less Frequent	6 months
	Oil Analysis	6 months	8 months	Keep	6 months
	Inspect/Electrical Testing	3 years	3 years	Keep	3 years
	Replace Motor	6 years	10.7 years	Less Frequent	9 years
Motor Cable	VLF TAN-Delta Testing	6 years	7 years	Keep	6 years
Protective Relays	Inspect/Calibrate	6 years	4 years	Keep	6 years
Pressure Switch	Calibration	4 years	4.2 years	Keep	4 years

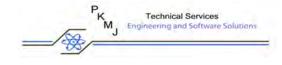
- Risk models developed by INL support the pump refurbishment and motor replacement frequency recommendations
- \$4.37M Net savings over next six years if implemented at the site



### <u>Accomplishments – Work Order Data Analysis</u>

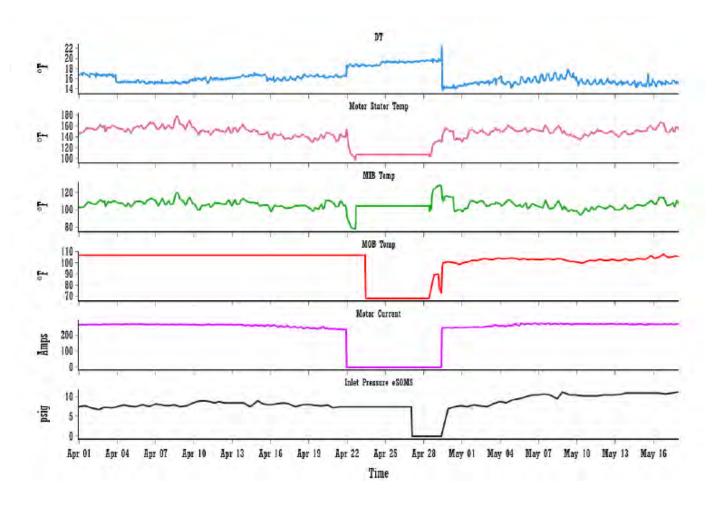
- WO Failure Classification was performed using Natural Language Processing (NLP) techniques to classify work orders associated with equipment degradation, failure, and other. NLP was also utilized to evaluate the primary object types (component types), conditions identified, and actions performed for work orders at Salem (see bottom figure).
- Created a plant process data and work order dashboard in PowerBI (see top figure). This was used for engineering review of events and work orders to better understand the impacts to plant process parameters when faults occurred.
- Developed WO work type classifier to determine when work requires part usage, improves health, includes inspection (check-up), identifies potential issues, restores condition to As Good As New, or others.
- Developed parts clustering technique to identify common parts used for work orders of a given type. This method allows for the identification of required and contingent parts, which can be used to enhance supply chain decision making based on future planned work.





# <u>Accomplishments – Identification of Fault Signatures</u>

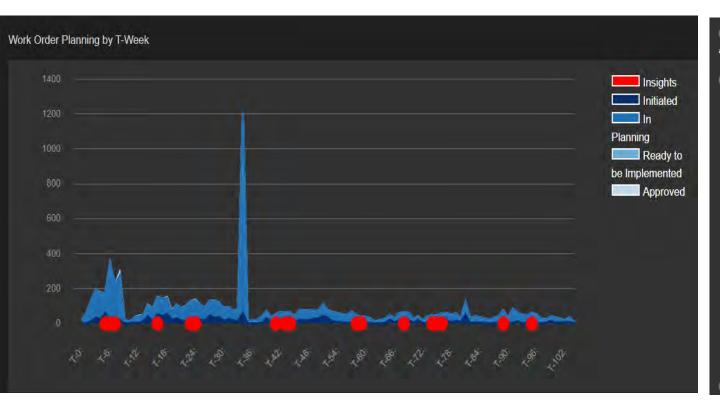
- PSEG Plant Process data was used to identify characteristics of specific faults that occurred with the Salem Circulating Water System equipment
- Time-domain and frequency-domain features were examined where possible to associate plant conditions with engineering characteristics
- Fault Signatures were used as the foundation for developing diagnostic models
- The development of fault signatures considered various scenarios based on the amount of indication available (i.e., without motor current, without vibration data, etc.) to provide insight into the flexibility of a fault signature solution

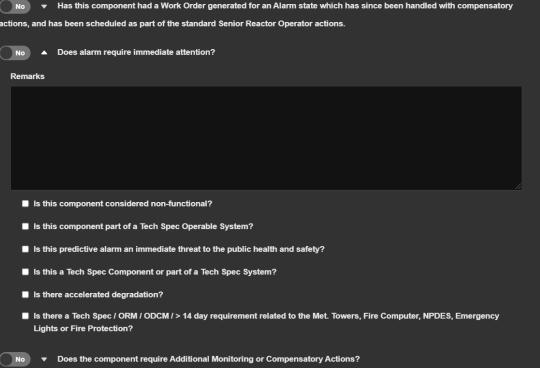




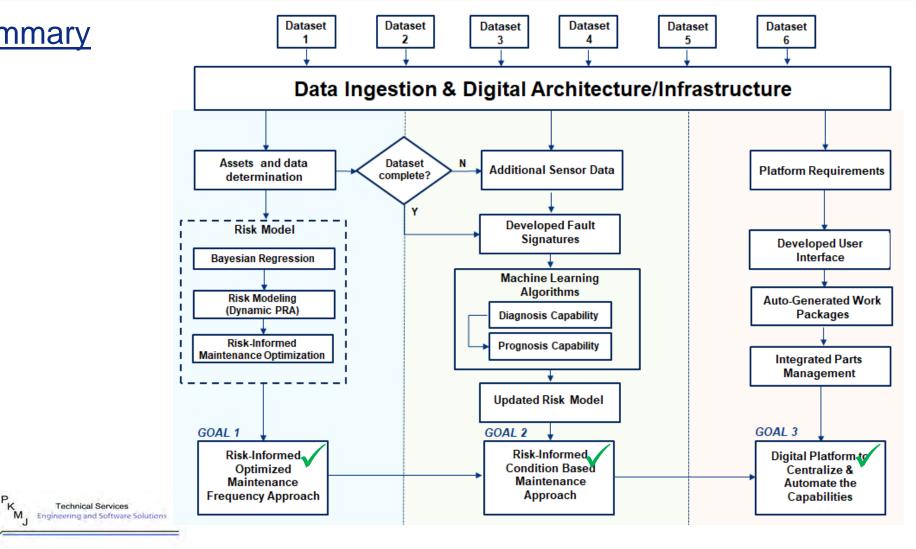
### Accomplishments - Development of an Automated Digital Platform

- PKMJ developed a cloud-hosted application using Microsoft Azure to enhance industry decision-making and data visualization
- Tools and services such as Azure Active Directory, Azure API Manager, Azure Function/Logic Apps, Azure DataLake Storage, DataFactory
  etc. were used to provide the structure, security, and data for the application
- The digital platform developed under the project supports maintenance strategy optimization, responding to fault signature alarms, and generation of automated work orders.









### **Future Development**

- PKMJ is in discussion with INL to continue the research, development, and commercialization of the digital platform solution. Areas of interest include:
  - Enhancements to Fault Signature Models
  - Integration of the Platform to Utility Systems
  - Improved Economic Modeling
- PKMJ is also working with its parent company, Westinghouse, to incorporate other Condition Monitoring solutions into the digital platform



### **QUESTIONS?**

### **Contact Information**

**Matthew Yarlett - PKMJ Technical Services** 

Email: Matthew.Yarlett@westinghouse.com

**Vivek Agarwal, PhD - Idaho National Laboratory** 

Email: Vivek.Agarwal@inl.gov

Harry Palas - Public Service Enterprise & Group (PSEG) Nuclear

Email: <u>Harry.Palas@pseg.com</u>



### References

- K. Technologies, "KCF Vibration Sensor Node SD-VSN-3," [Online]. Available: https://www.kcftech.com/smartdiagnostics/products/sensors.html. [Accessed May 2020].
- "KCF Wireless Interference and SmartDiagnostics®," [Online]. Available: https://www.kcftech.com/smartdiagnostics/resources/whitepapers/Application%20Note%20-%20Wireless%20Interference%20and%20SmartDiagnostics.pdf. [Accessed May 2020].
- E. Seneta, Non-negative matrices and Markov chains, 2nd ed., vol. XVI, (Originally published by Allen & Unwin Ltd., London 1973), 1981, p. 288.
- K. S. Trivedi, Probability and Statistics with Reliability, Queueing, and Computer Science Applications, New York: John Wiley & Sons, Inc., 2002.
- G. Bolch, S. Greiner, H. de Meer and K. S. Trivedi, Queueing Networks and Markov Chains, 2nd ed., John Wiley, 2006.
- R. Nelson, Probability, Stochastic Processes, and Queueing Theory, Springer-Verlag, 1995.
- "Serverless Web Application", Published by Microsoft [Online], https://docs.microsoft.com/en-us/azure/architecture/reference-architectures/serverless/web-app. [Accessed July 2021]



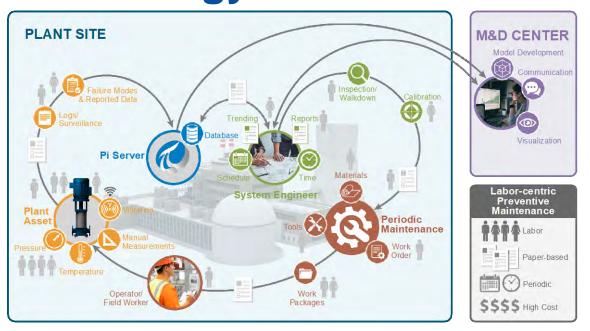
September 15, 2021

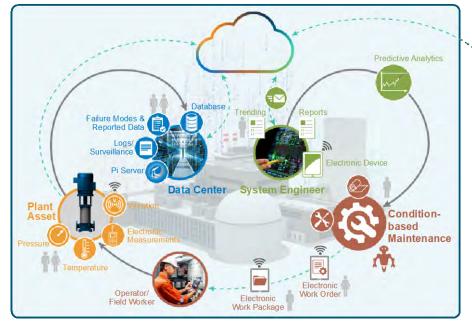
Koushik A. Manjunatha Staff Research Scientist, INL

## Artificial Intelligence (AI) Driven Scalable Condition based Predictive Maintenance Strategy



# Vision of Risk-informed Predictive Maintenance (PdM) Strategy

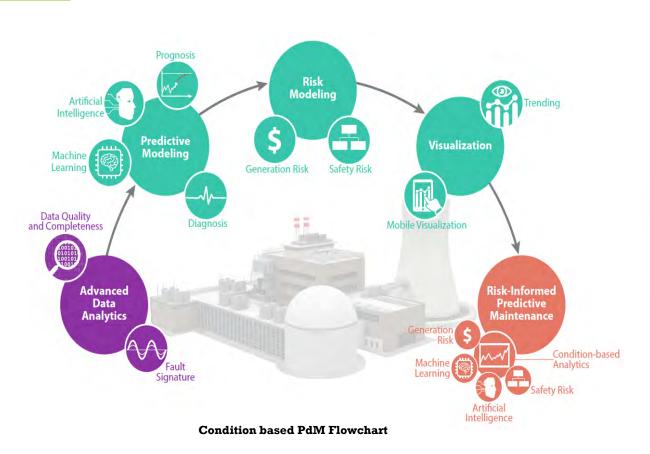


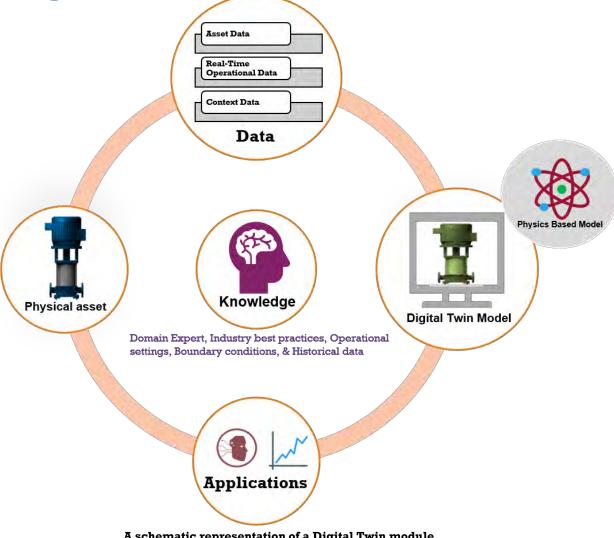






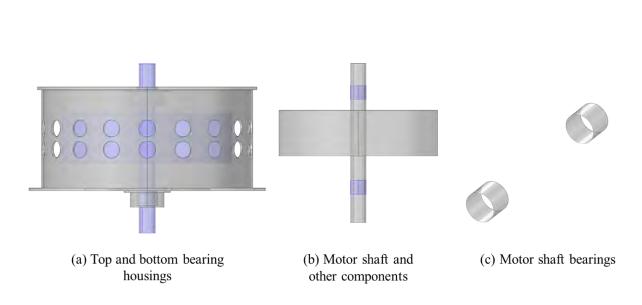
Data Analytics at Scale and Digital Twin

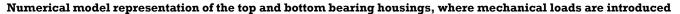


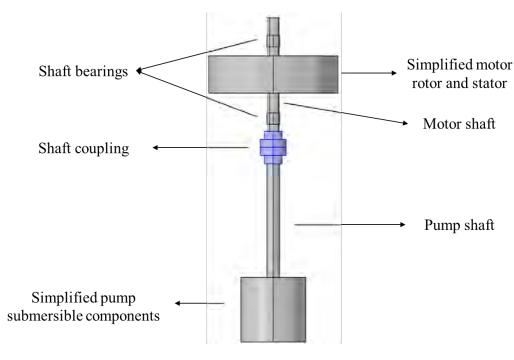


A schematic representation of a Digital Twin module

## **Physics based Modeling**







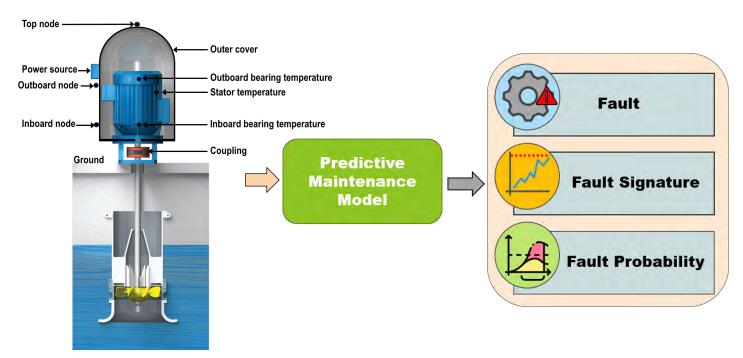
Simplified numerical model for motor, pump, and shaft coupling

#### Physics Driven model:

Requires: 1. domain knowledge (material, size etc.)

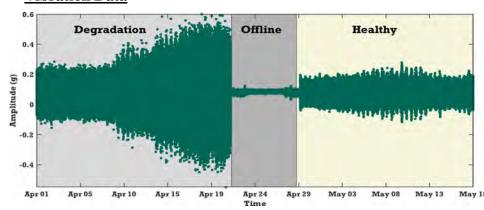
2. close interaction with operators

## **Condition based PdM Model and Outputs**

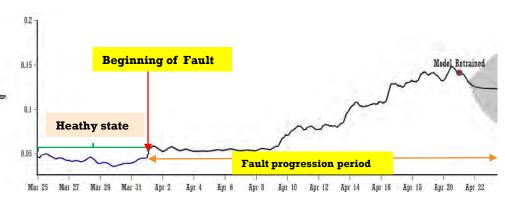


A schematic representation of a motor and pump with temperature measurement locations

#### **Vibration Data**



A representative figure of a particular fault captured in vibration signal



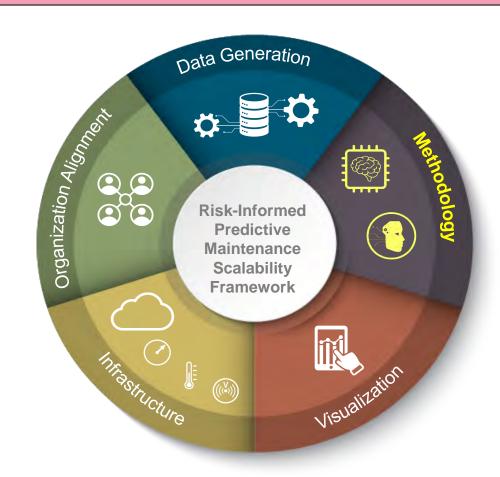
Forecasting dominant fault signature after prediction of a fault

## **Scalable Condition based PdM**

Scalability is defined as expanding capabilities of a target entity to meet current and future application-specific requirements

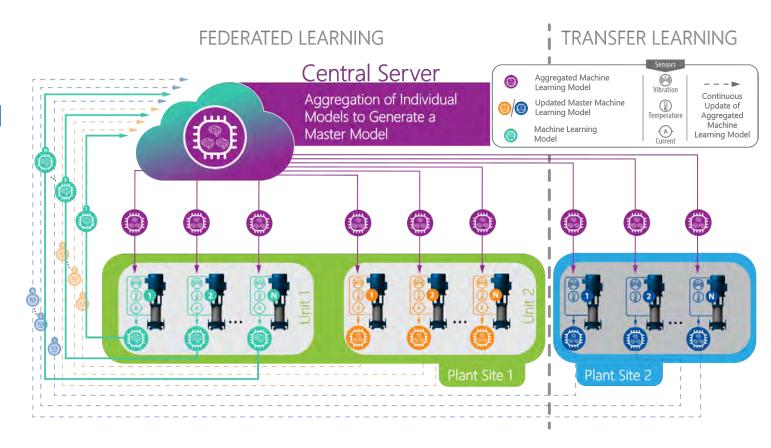
For NPPs, building a comprehensive AI model is challenging because

- Faults are rare events, and it is highly unlikely for all the faults to occur in each component;
- For a newly installed component/system or plant unit, it is infeasible to build AI models from scratch;
- Collecting data at a centralized location is limited by
  - High bandwidth costs
  - Real-time decision
  - Privacy, security, and commercial concerns



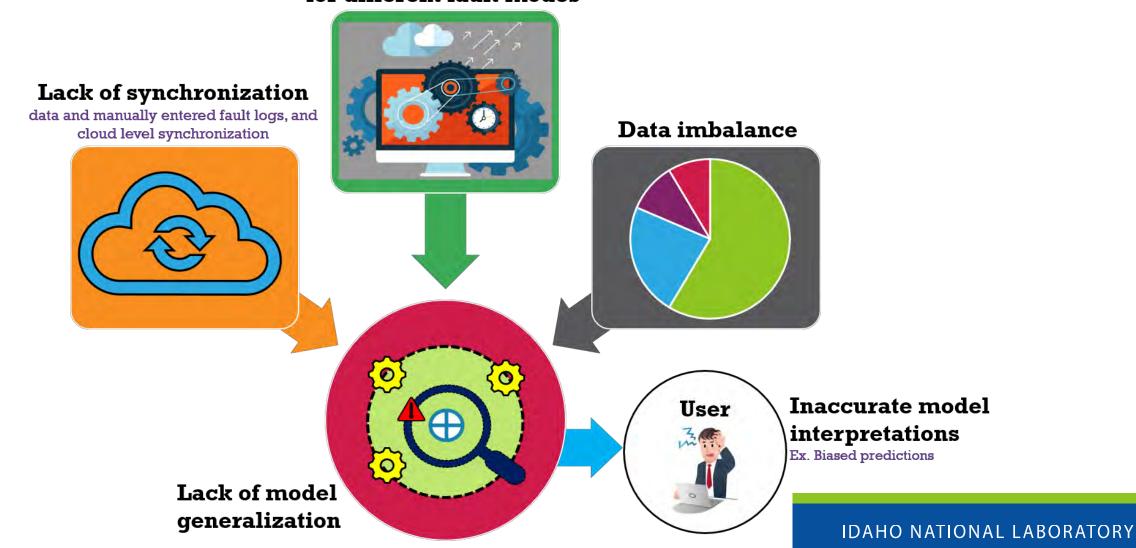
## **Federated Transfer Learning**

- Individual component-level model using component-specific available data sources
- Consolidating the knowledge gained from individual component models into a master model,
- Using the master model to make diagnostic and prognostic estimations of the entire system,
- Applying (i.e., transferring) the master model to similar plant systems, either at the same plant site or at different plants.



## Few Challenges to Consider

Lack of good quality data for different fault modes



### **Team**

- INL Team
- Public Service Enterprise Group, Nuclear LLC Team
- PKMJ Technical Services Team

Contact Information:
 Koushik A. Manjunatha
 koushik.manjunatha@inl.gov
 Idaho National Laboratory





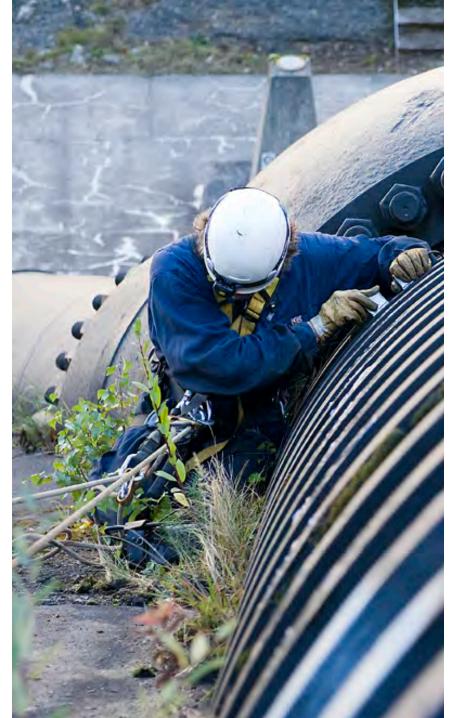
# THERMAL PERFORMANCE MANAGEMENT FOR NUCLEAR POWER PLANTS WITH DIGITAL TWINS

Use Cases based on EDF experience

09/15/21



Dr Christophe Duquennoy Nuclear Fleet Thermal Performance Expert <a href="mailto:christophe.duquennoy@edf.fr">christophe.duquennoy@edf.fr</a>



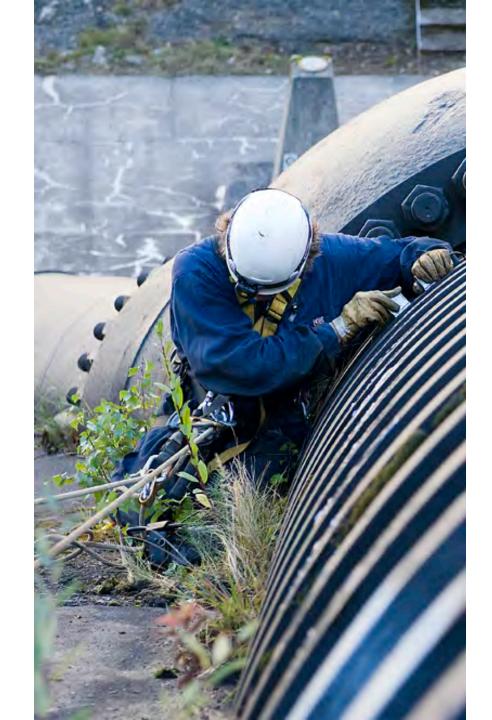


# THERMAL PERFORMANCE MANAGEMENT FOR NUCLEAR POWER PLANTS WITH DIGITAL TWINS

Use Cases based on EDF experience

09/15/21





## **AGENDA**

- 1. INTRODUCTION
- 2. DIGITAL TWINS FOR THERMAL PERFORMANCE
- 3. METROSCOPE SOFTWARE
- 4. USE CASES







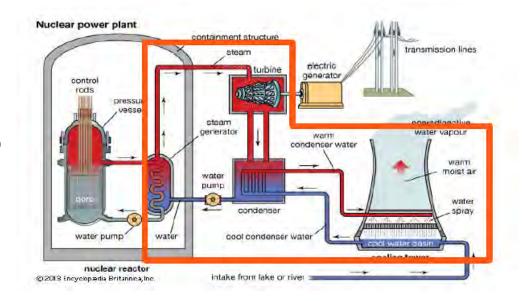




#### DIGITAL TWINS FOR THERMAL PERFORMANCE

# Thermodynamic Performance Testing and monitoring

- Thermodynamic measurement for NPP
- Performance monitoring, production and maintenance optimization

















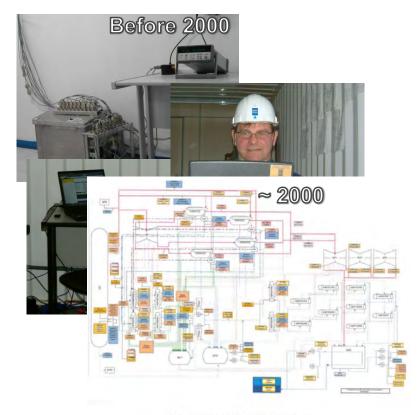


FIG. 12. Thermal model (with sensors)







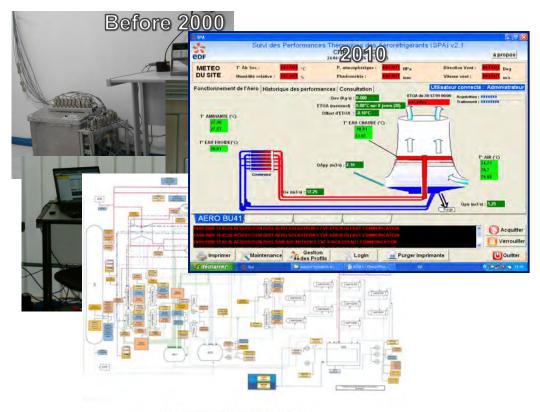
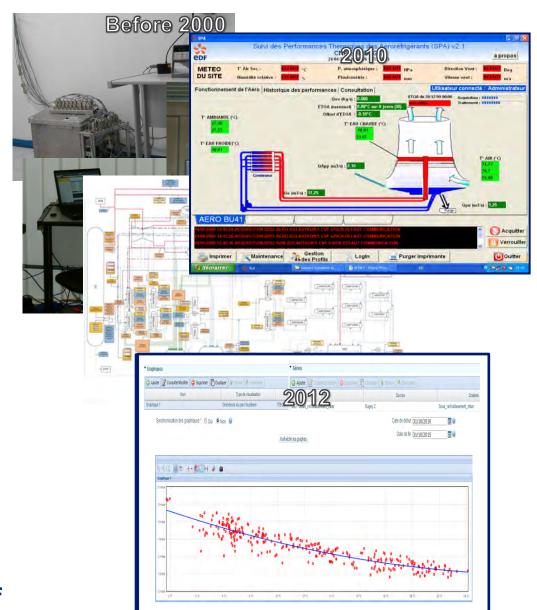


FIG. 12. Thermal model (with sensors)







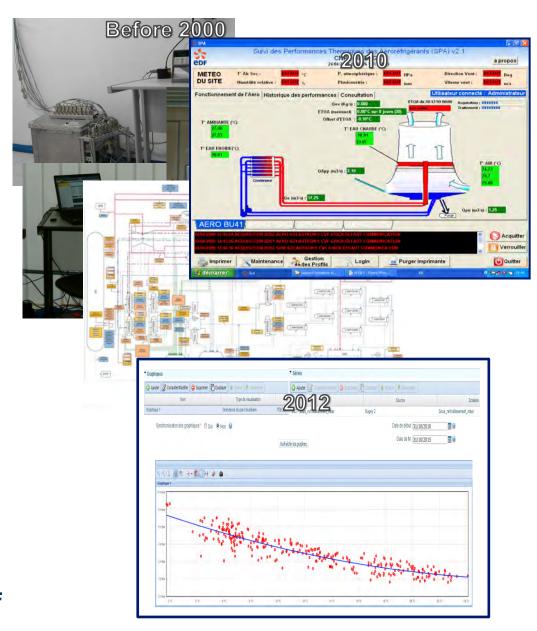








#### DIGITAL TWINS FOR THERMAL PERFORMANCE



#### 2014 Diagnostic methodologies

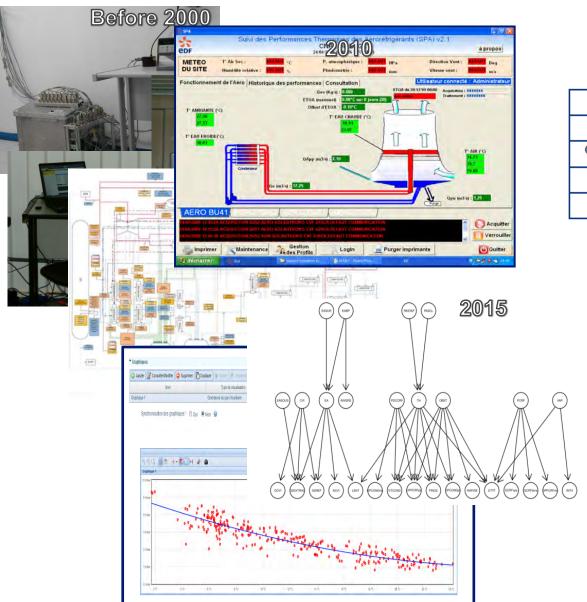
∆vide	
Ssref	
O <sub>2</sub> extrac	
ΔPpcrf	$\longrightarrow$
ΔTtf	<b>→</b>







#### DIGITAL TWINS FOR THERMAL PERFORMANCE



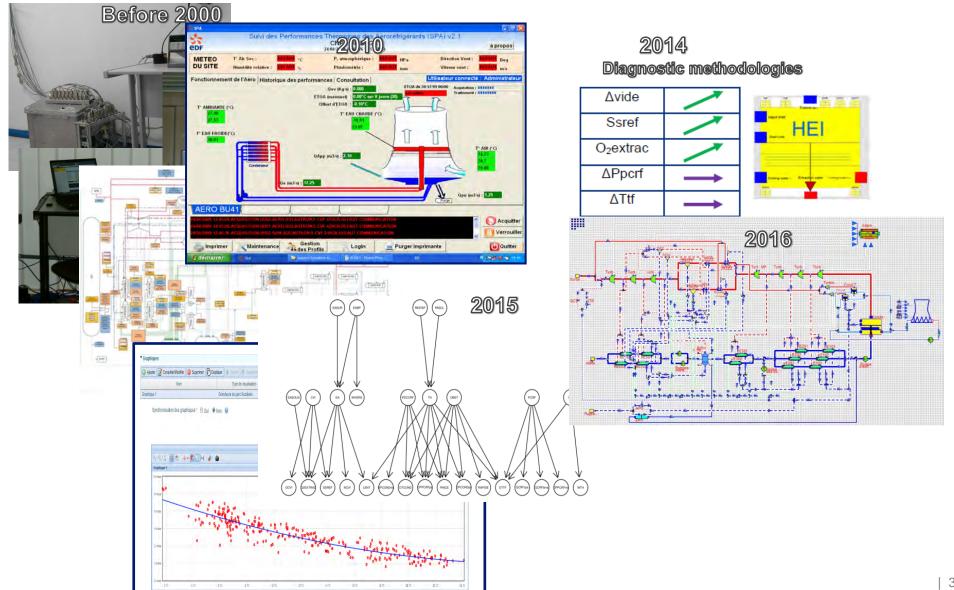
#### 2014 Diagnostic methodologies







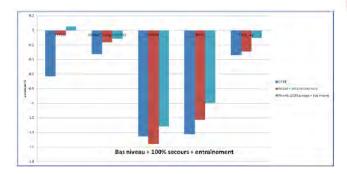


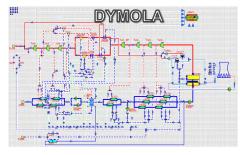


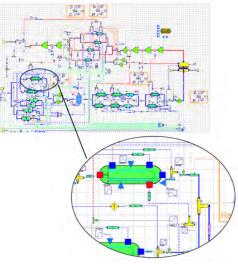


#### DIGITAL TWINS FOR THERMAL PERFORMANCE

- Saint Laurent NPP: Bypass leak on cooling tower vs Row Water leak in condenser
- □ We provided the maximum affordable power to allow to repair Row Water leak within operation. → Avoided Plant Shut down
- Cruas : Diagnostic for 14 Mwe lost in R501
   HP Heater
- Medialization of simultaneous defects: Low level in heater + condensates leak + vapor entrainment













#### DIGITAL TWINS FOR THERMAL PERFORMANCE

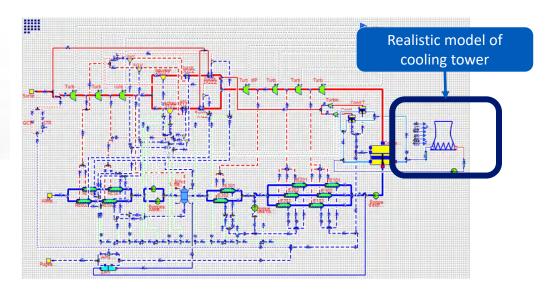


"We've got a drain piping under our water collectors which is about to break. Should we repair it before we turn on the plant?"

#### Nogent NPP draining pump rupture:

 Our analysis lead to maintenance and avoided ~600 k€ of loss.



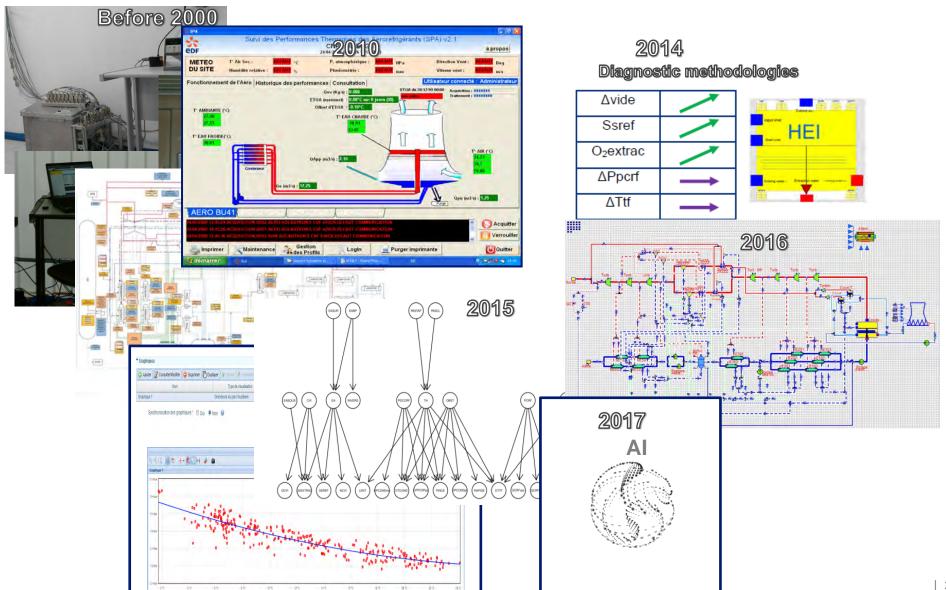








#### DIGITAL TWINS FOR THERMAL PERFORMANCE

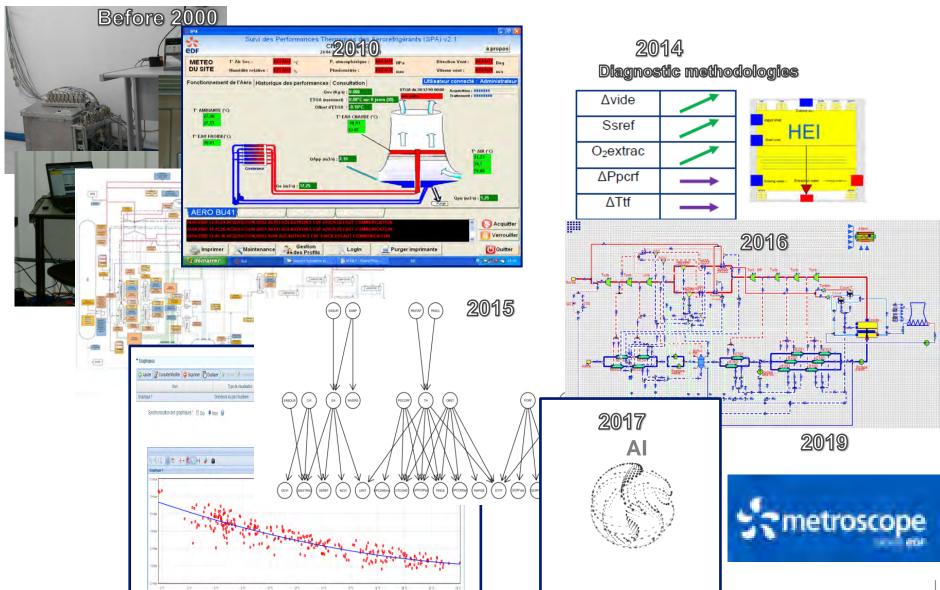








#### DIGITAL TWINS FOR THERMAL PERFORMANCE







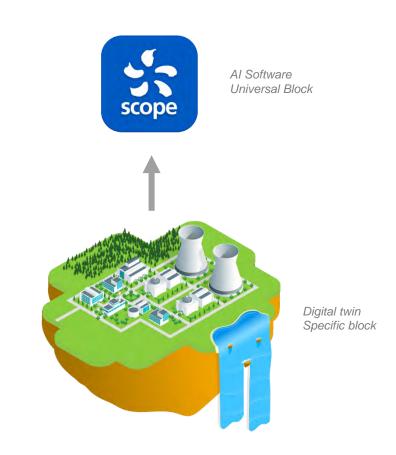


# **METROSCOPE SOFTWARE – GENERAL PRINCIPLE**

The software combines a Digital Twin of the process and AI to perform a diagnosis in operations

 $Digital\ twin + AI = Diagnosis$ 

The software uses existing information from sensors to analyze the plant performance and diagnose underlying causes impairing the expected behavior of the process







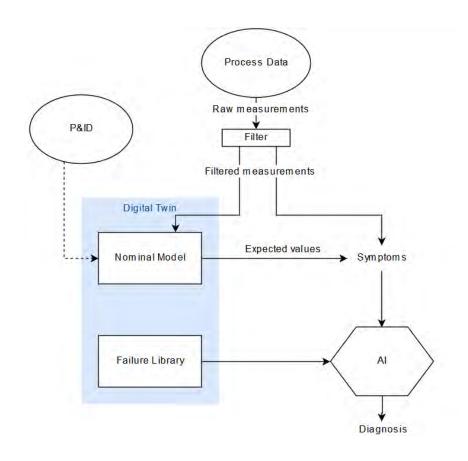
# **METROSCOPE SOFTWARE – GENERAL PRINCIPLE**

#### Three steps to perform a Diagnosis

- 1 Metrological analysis of the raw data
- 2 The symptoms are generated using the Digital Twin.
- 3 The diagnosis is produced thanks to AI

#### Two bricks to make a Digital Twin

- the nominal model represents the nominal behavior in operation of the process, calibrated on historical data
- the failure library is a mathematical description of the failure modes of the process and their impact on the measurements







# **METROSCOPE SOFTWARE - AI**

METROSCOPE AI belong to the domain of Symbolic AI (in opposition to Statistical AI)

It is meant to address Small Data problematics where decision relies on both Knowledge and Data.

METROSCOPE AI has been inspired by Medical Diagnosis

Most knowledge-based systems have two distinguishing features: a knowledge base (here the digital twin) and an inference engine (here the Bayesian network)

Awarded Best Innovation in Nuclear in 2019 by the Société Française de l'Enérgie Nucléaire

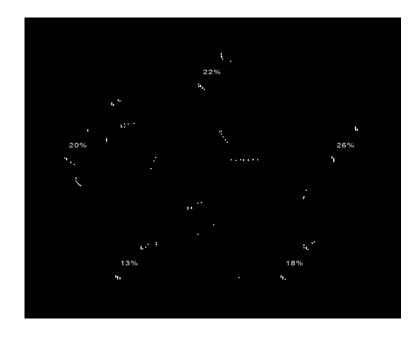


Illustration des chaînes de Markov







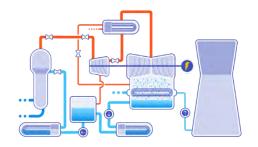
# **METROSCOPE SOFTWARE DIGITAL TWIN**

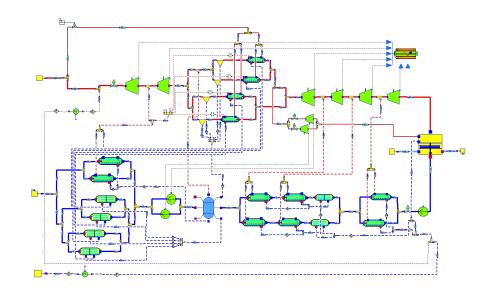
The Digital Twin (DT) is a numerical representation of the secondary side, able to simulate both nominal and impaired behaviors of the process.

DT is built from the P&ID of the plant and calibrated on historical data.

DT of a NPP is composed of around 10 000 equations and variable declarations.

Perimeter and Accuracy of the DT are meant to evolve overtime





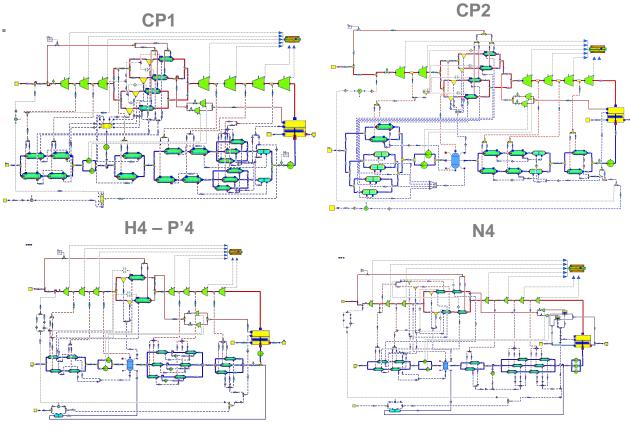
Jumeau numérique du circuit secondaire





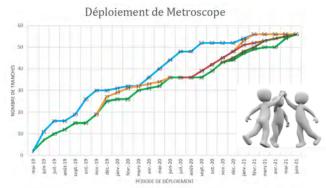
# **Sept**

# **METROSCOPE – DEPLOYEMENT ON EDF FLEET**



#### Fleet References:

- 58 condensers digital twins encapsulating the best ever seen performance based on historian data.
- 32 cooling towers digital twins tuned encapsulating the best ever seen performance based on historian data
- Métroscope compatible Digital Twins
- Production of all the reference models for French Nuclear Fleet's METROSCOPE: 56 units in 3 years
- Production of all defect libraries based on EDF operation feedback.
- In particular: Expertise on key components in terms of thermodynamic performance: Condensers, Cooling Towers, Heaters.









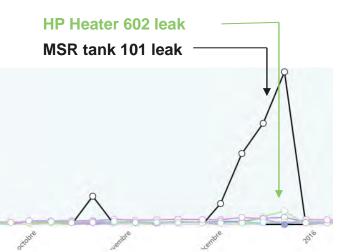
#### HIGH PRESSURE HEATER LEAK

Detection of simultaneous leaks on both MSR and HP Heaters :

- MSR condensate regulation fault was detected by the operator (since 2015 November 30th)
- Meanwhile heater's leak on condensates occured (2015 December 14th). It was **not detected by operators**.

 A lost of 2 MW between December 15th and 27th would have been avoided with METROSCOPE (not deployed at this time) Performance Gain

Avoidable loses 0,5 GWh





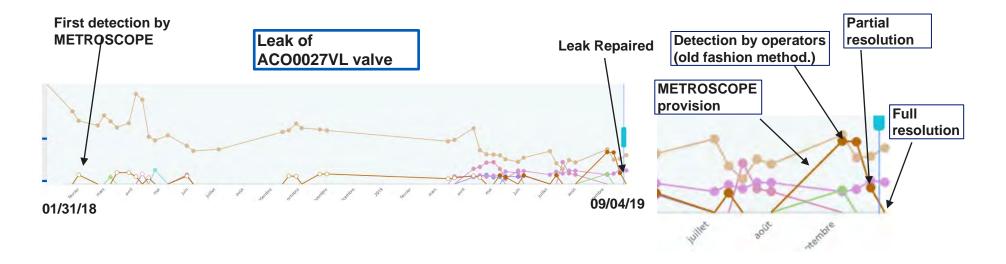


# CONDENSATE COLLECTOR TANK LEAK

- Early detection of small magnitude leak (~2% of the nominal flow)
- Reliable quantification of the impact (1.5 MWe)
- Diagnostic of multiple simultaneous defects

Performance

Avoidable loses ~ 8 GWh







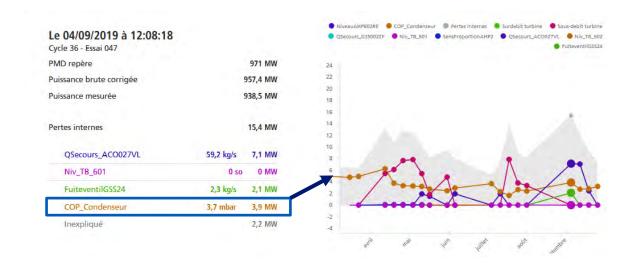
### **CONDENSER LOSES**

- Simultaneous Multiple Diagnostic
- Accurate Quantification of condenser losses due to updated reference Digital Twins

#### Performance

Avoidable losses

1 mbar = 1MWe





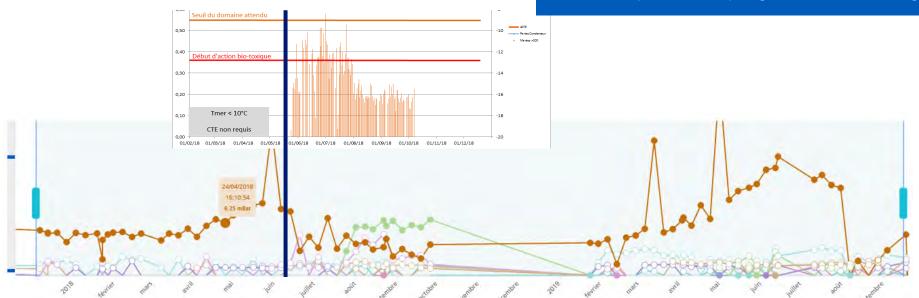


#### **CONDENSER LOSES**

- Real time reliable monitoring of condenser thermal performances
- Possibility to optimize biocidal injection during crises
- Accurate Quantification of condenser losses due to updated reference Digital Twins

#### Performance / Maintenance

Avoidable losses 1 mbar = 1MWe
Optimization of operating solutions:
Biocidal injection, Taproges CTA, reboiling.





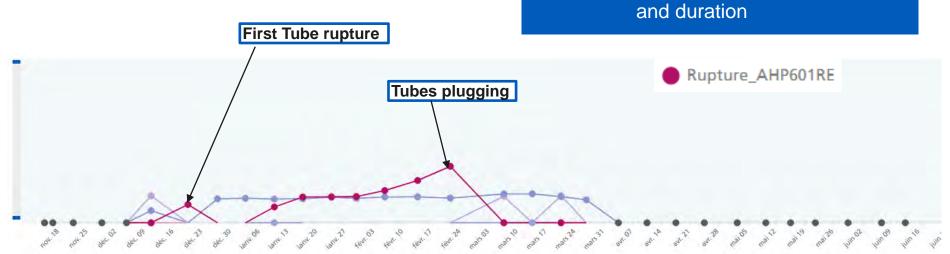


#### HEATERS TUBE RUPTURE

- Early detection (2 or 3 weeks)
- Real time monitoring of number broken tubes
- Maintenance schedule optimization
- Limit the duration of maintenance intervention (better prepared)
- Capitalization of feedback of evolution signatures.

Maintenance

Gain Optimization of maintenance scheduling







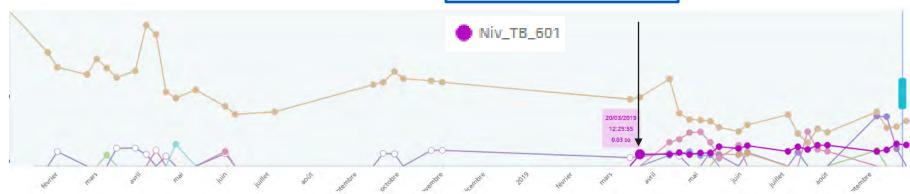
### LOW HP HEATERS LEVEL

- Situation before METROSCOPE not seen and solved by operators
- Detection of low level situations by METROSCOPE

Maintenance

Gain Plant adjustment

#### Low Heater Level detection





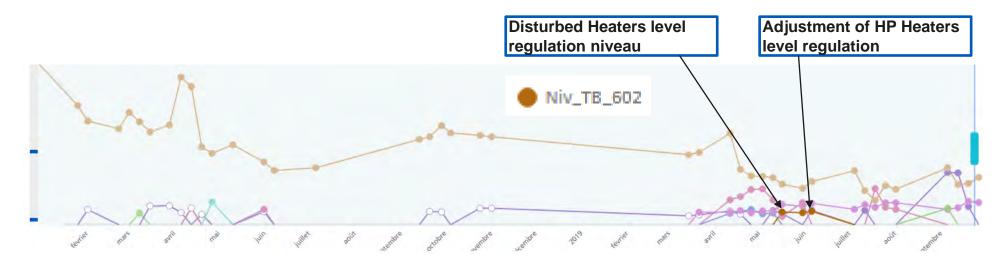


# HP HEATERS LEVEL REGULATION

- Real time detection of low level situations by METROSCOPE
- Optimization plant adjustment by operators

Maintenance

Gain Plant adjustment





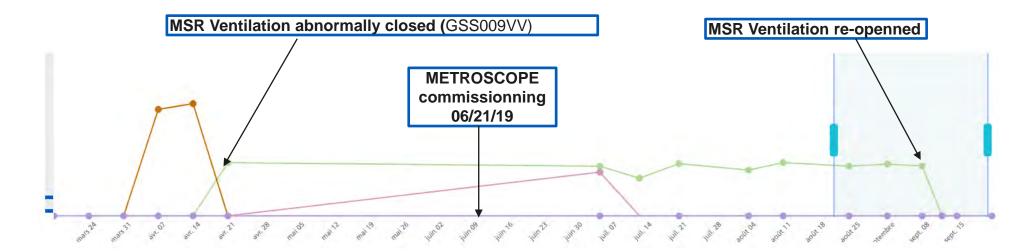


# MSR VENTILATION OBSTRUCTION

- Based on low signal detection (variation under 4% = 2 kg/s)
- · After operator intervention, the defect is solved

Maintenance

Gain: Incondensable gas management





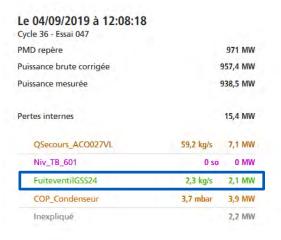


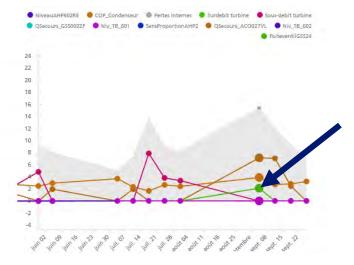
### MSR VENTILATION LEAK

- Simultaneous Multiple Diagnostic
- Almost MWe Losses are explained (~ 2 MW of residual losses)

#### Performance

Avoidable loses 0.4 GWh







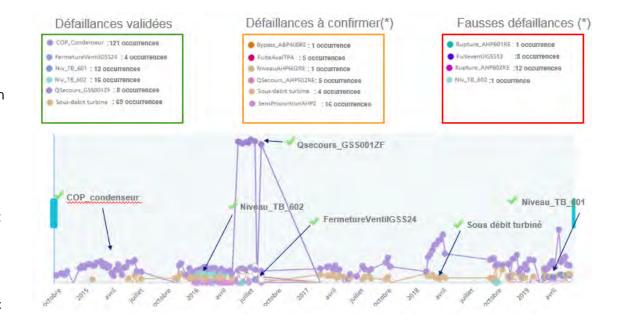


#### A large variety of situations are well catch

- Condenser Performance drift
- Steam flow decrease in the turbine
- Feed Water Heater unoptimal level regulation
- Feed water tank leak
- → METROSCOPE automated diagnosis strongly simplifies operators life
- → METROSCOPE is online, and provide a fleet wide supervision of the plants (enhance the production national reports for LTO)

#### Real life situations leads to

- Validation of DT (Reliable reference of best historical performance)
- Validation of Failure Libraries (Know ledge capitalization)





# **CONCLUSION**SYNTHETIS OF MAIN GAINS

#### **Performance Monitoring**

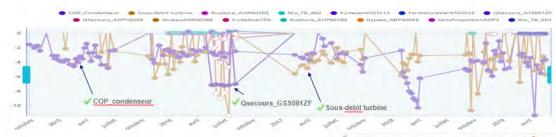
- Identification of major failures in terms of losses
- Smaller MW losses are catch
- Some failures impact maintenance only

With METROSCOPE unexplained MW losses are

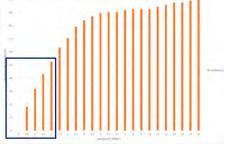
- Under 2 MW 53% of the time
- Under 3 Mw 70% of the time

→ Here METROSCOPE has been calibrated to detect just a little bit more than what's really happening

METROSCOPE interface is very useful to analyze unexpected behaviors of the tool

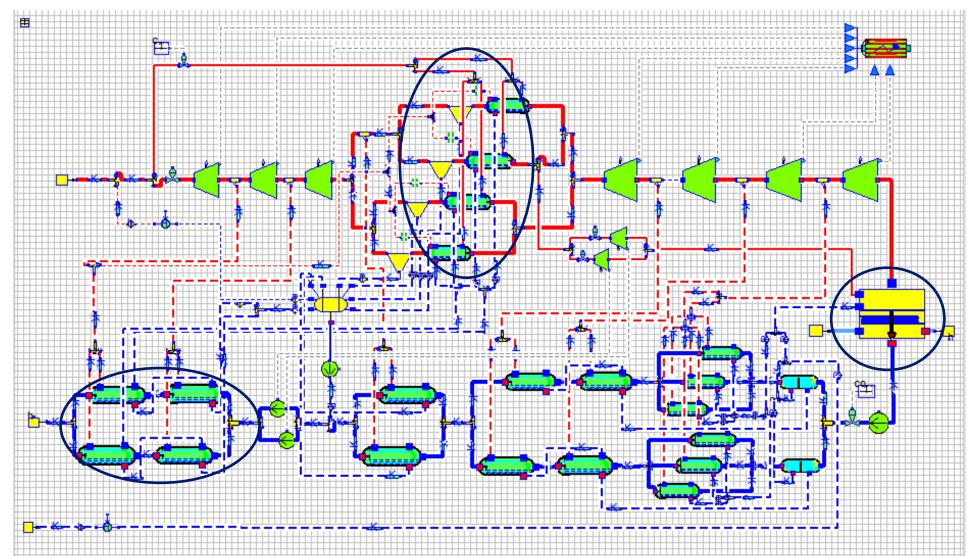


- 2 défaillances récurrentes avec pertes significatives (COP condenseur/ sous débit turbiné)
- 53% des essais avec inexpliqués<2MWe (70% <3MWe)





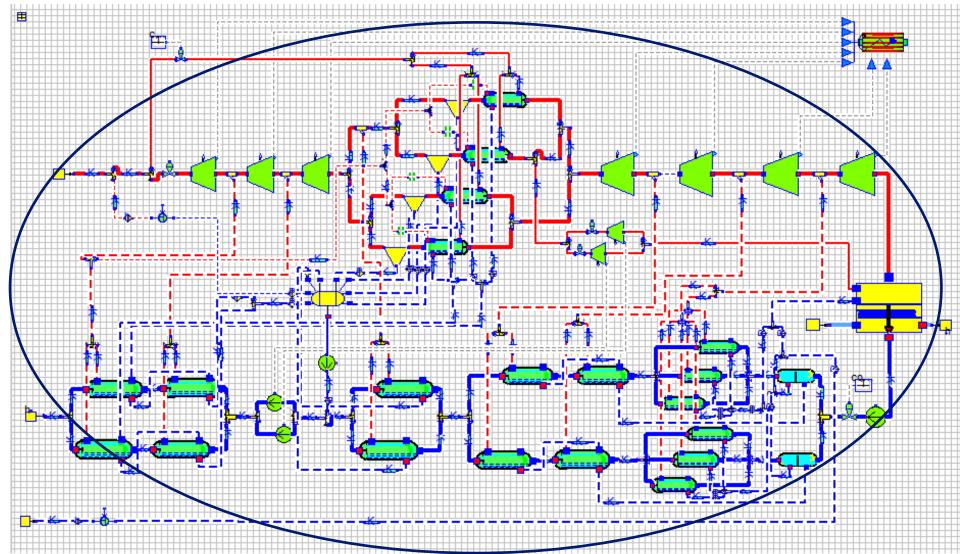
#### **EXTENSION OF AUTOMATIQUE DIAGNOSTIC**







#### **EXTENSION OF AUTOMATIQUE DIAGNOSTIC**





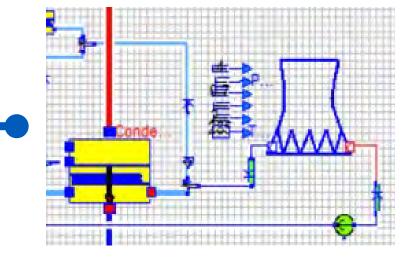




All diagnosis from physical Simulation



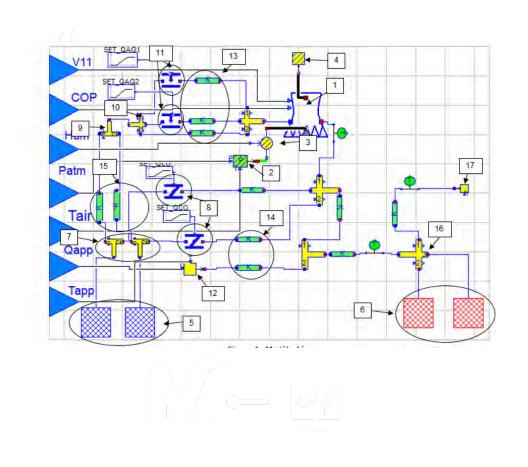
Numerical model plugged in the Metroscope for live diagnosis



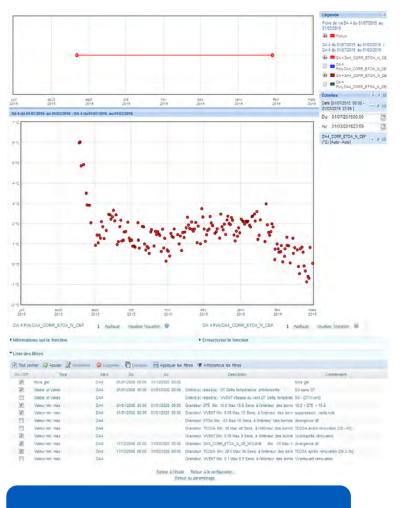




#### **TOMORROW – COOLING TOWER MONITORING WITH METROSCOPE**



**Develop Defect models** 



Use available data to validate





#### TOMORROW – COOLING TOWER MONITORING WITH METROSCOPE

#### **SYNTHETIC OVERVIEW OF SERVICES PROVIDED BY SPA**

SERVICE PROVIDED	ESTIMATION OF THE GAIN	LEVEL OF INVOLVEMENT	
REDUCTION OF PROCESSING TIME OF INCIDENTAL DEFECT	5 GWh/ major defect 20 GWh/other defect 4 major defects/year	trained operator + time required for level 1 support + functional SPA analysis	
OPTIMISATION OF MAINTENANCE	20 GWh/maintenance	competent support engineers + time required for level 2	
CHOICES	3 situations /years	support	
PRODUCTION FORECAST / OPTIMISATION CHOICES DURING CRISIS	Difficult to quantify	competent engineering unit+ Time required to go further in the analysis	
FEEDBACKS ON DESIGN CHOICES	Gain of 0,2°C ⇔ 160 GWh over 30 years	competent engineering unit+ + expertise on operation+ design engineering	





#### TOMORROW – COOLING TOWER MONITORING WITH METROSCOPE

#### SYNTHETIC OVERVIEW OF SERVICES PROVIDED BY SPA + METROSCOPE

SERVICE PROVIDED	ESTIMATION OF THE GAIN	LEVEL OF INVOLVEMENT	
REDUCTION OF PROCESSING TIME OF INCIDENTAL DEFECT	5 GWh/ major defect 20 GWh/other defect 4 major defects/year	trained operator + time required for level 1 support + functional SPA analysis	
OPTIMISATION OF MAINTENANCE CHOICES	20 GWh/maintenance 3 situations /years	+ time required for level 2 support	
PRODUCTION FORECAST / OPTIMISATION CHOICES DURING CRISIS	Difficult to quantify	Time required to go further in the analysis	
FEEDBACKS ON DESIGN CHOICES	Gain of 0,2°C ⇔ 160 GWh over 30 years	competent engineering unit+ + expertise on operation+ design engineering	





# THANK YOU!



# Any questions?



# APPENDIX









### **EDF-DTG** Our mission



#### 3 MAIN FIELDS:

#### **MEASUREMENT**

> Measurement engineering, metrology, data bases

#### **INSPECTION**

> Monitoring on behalf of operators

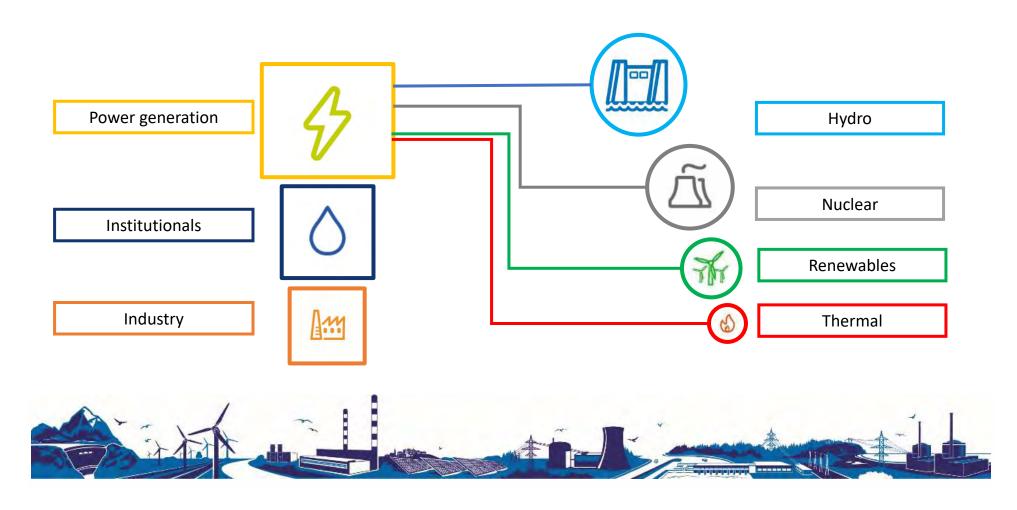
#### **EXPERTISE**

> Diagnosis, pronostics & consulting for operators and maintenance teams

- Monitoring, diagnosis and forecasts for EDF Group power plants (nuclear, hydro, fossil-fired and wind energy) to assist operators in making effective safety decisions and in managing performance
- Develop technical skills and expertise required to operate energy production facilities, to find solutions about energy development and environmental issues
- Integrate and make reliable new R&D monitoring solutions



# **EDF-DTG** Our customers







# **EDF-DTG** Our strength



**Accurate and independant** diagnosis & prognosis



**DATA** historian based on **50** years of experience



**Know-how to convert DATA into value** 



An unique expertise in O&M



**Innovative** solutions





# DESIGN & O&M Support Services



MECHANICAL INSPECTION AND DIAGNOSIS

- Inspection during production
- Assessment of key static components for power plants (penstock, secondary circuit...): non destructive testing, corrosion-erosion modeling, damage analysis and repair tracking



SYSTEMS AND VIBRATING ENGINEERING

- Monitoring mechanical behaviour of power plants, pipe systems and rotating machinery
- Acoustic testing of primary circuit for nuclear fleet
- Condition-based maintenance of nuclear valves



**ELECTRICAL EQUIPMENT** 

- Diagnosis of main electrical equipment needed for power plants operation, based on tests carried out on site and in the lab
- Diagnosis using infrared thermography (electrical trouble spots)





# **DESIGN & O&M Support Services**



#### SETTINGS, PROTECTION, OPTIMISATION PROCESSES, ANCILLARY SERVICES

- Testing the safety protection of power-generating facilities
- Testing the network reconstitution and voltage recovery following an incident
- System services monitoring and optimisation: contribution to system voltage and frequency stability



#### **ACOUSTICS**

- Noise control for all EDF Group facilities: acoustic impact studies and modeling, sound power measurements in situ, ...
- Optimising the soundproofing of installations and to ensure compliance with regulations



THERMODYNAMIC DIAGNOSIS
AND PROGNOSIS

- Measuring thermodynamic performances of thermal power plants (fossil-fired and nuclear)
- Monitoring energy performances, looking for productivity gains





# **DESIGN & O&M Support Services**



METROLOGY, IT & DATA technologies

- Development of appropriate IT solutions for the various business activities
- Data collection, management, processing and archiving
- Certified by Cofrac, the DTG laboratory is a reference for the calibration of temperature, pressure and humidity variables



MONITORING NUCLEAR FACILITIES

- Continuous monitoring of mechanical behaviour of equipment and civil engineering structures
- Containment tightness: monitoring and control during pressure tests



WATER RESOURCE FORECAST & MANAGEMENT

- Evaluation and management of the impacts of electricity generating activities on atmospheric, acoustic and aquatic environments
- Impact of aquatic environment on electricity generation



# Digital Twin of a Real-time Radiation Monitoring Network

# An EPRI NextGen RP Project

#### Rich McGrath

Principal Technical Leader rmcgrath@epri.com

2021 Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization

September 14-16, 2021





# **Discussion Topics**

- Background of Need for Digital Twins
- Results of Scoping Study of Available Technologies
- In Progress Project Tasks
- Future Work



# **Background of Need for Digital Twins**

- Utility RP Organizations need to seek additional measures for cost reduction while still maintaining excellent performance and safety while faced with the following challenges:
  - Shrinking contract RP technician resources,
  - Reductions in site RP staffing,
  - Cost efficiency,
  - Knowledge retention and transfer, and
  - Maintaining and/or enhancing worker safety.
- Currently, most radiation protection (RP) tasks for the collection of radiation field data are conducted manually at nuclear power plants. For example:
  - Most radiation field measurement and characterization activities (i.e., surveys) are conducted
  - Data from radiation surveys are manually entered into the plant database and subsequent data analyses to inform radiation protection, and/or source term management, and/or ALARA planning are conducted manually.
- The EPRI Digital Twin Project is evaluating:
  - Technologies that provide remote/automated radiological measurements that accurately reflect the radiological environment in the plant
  - If technologies are available, then creation of a Digital Twin of the radiological environ can be used to
    optimize work in the radiological areas of the plant.



## Digital Twin of a Real-time Radiation Monitoring Network

**Phase 1 - Demonstrate a Geophysical Application for Analyzing Radiological Survey Data** 

- Increased use of remote radiation monitoring technology is providing real-time radiological data.
- Measurement points are shown as single point sources with no values in between measurements.
- Geospatial technologies are available that can interpolate available remote detector readings to model the radiological conditions between measurement points.
- Applications could be applied in reverse to determine where remote monitoring devices should be placed to adequately monitor dose rates
- This EPRI project will in 2021:
  - Evaluate available software applications
  - Test geostatistical software tool(s) to see if tool accurately interpolates actual survey data

**Example of Actual Interpolation of Survey Data** 



**Survey Measurements** 

**Interpolation of Survey Data** 

# **Initial Tasks of EPRI Project**

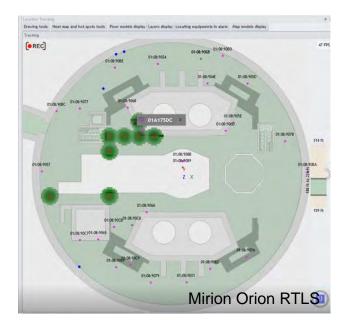
- EPRI Team has completed an initial scoping study to determine what software tools and associated technologies are currently available or planned that can create a digital twin of the radiation fields at a nuclear power plant.
- For each candidate software/technology supplier, the technology search attempted to determine the following:
  - The current state of development and/or planned future enhancements as it relates to this project.
  - Current state of deployment where it has been used or is being used at a nuclear power plant site
  - Willingness of the supplier to participate in a tabletop demonstration of the software tool

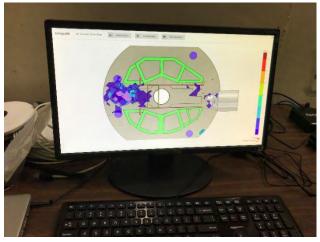


# **Examples of Supplier Capabilities for Tabletop Demos**



- EPRI Indoor Positioning System (IPS) system demonstrations showed functionality applicable to digital twin technology such as:
  - Track tags to ~1-2 meter accuracy
  - Collect dose rate data continuously
  - Generate live dose rate maps
- EPRI Demonstrations Performed at nuclear power plants:
  - Bluetooth Low Energy (BLE) based
     Quuppa
  - Ultra Wide Band (UWB) based Mirion Orion Real Time Locating System (RTLS)
- EPRI Technical Report for the IPS project to be published in 2021





DEI Telepath™ Live Map Display



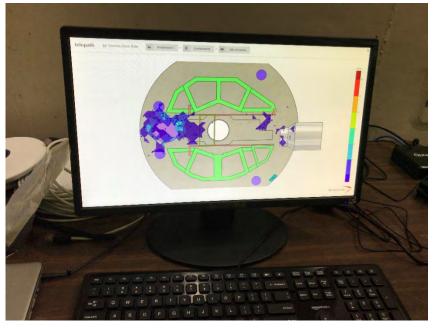
# **Quuppa Demonstration**

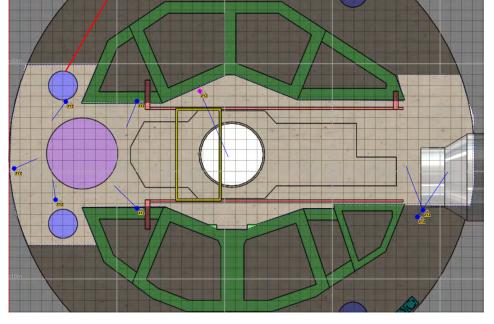
- Workers provided with positioning tags and teledosimeter
- Anchors with magnet mounts for receiving signals from dosimeters installed as shown in images
- Dose rate data from teledosimeter and positioning data from tags are merged together to create dose maps





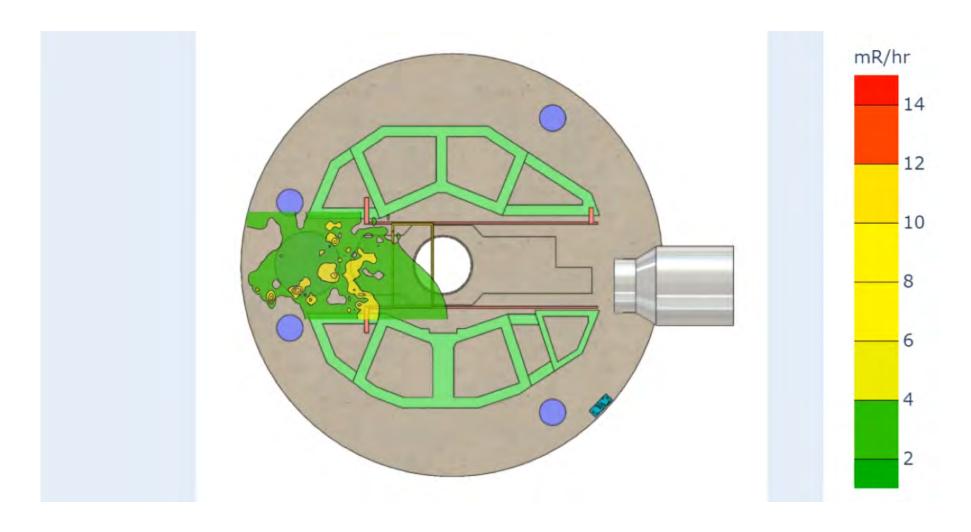








# **Quuppa Demonstration - Live Dose Rate Map**

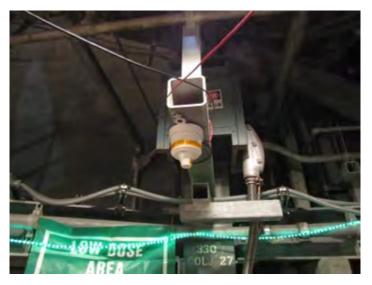


Rad Mapping from DEI Telepath™ Software



# **Anchor Placement**

 Similar approach but different anchor design for Mirion system







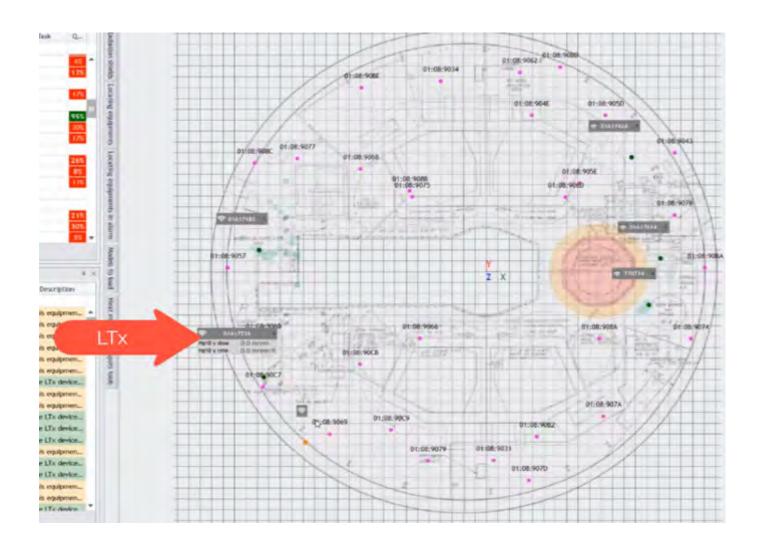




#### MIRION ORION™ RTLS DEMONSTRATION

# Accuracy

- Accuracy was observed to be 1-2 m
- For some isolated locations, the position was erratic due to structural interference
- Accuracy increases with number of anchors and transmit frequency
- Dose rate data displayed on the maps





# Other Technology Available

- Capabilities offered by other vendors:
  - Creation of 3D maps of the plant using LIDAR system incorporated into the portable radiation detector
    - Readings from detector automatically recorded onto the 3D map as the survey is conducted
- Upload new or existing 3D CAD generated maps of the plant into the software:
  - Radiation readings automatically loaded onto maps from:
    - Handheld survey meters as survey progresses
    - Electronic dosimetry as workers transit area
    - Fixed area monitors



# **EPRI Digital Twin Project Status – Phase 1**

### • Multiple vendors have:

- At least one technology that appears to be capable of supporting development of a digital twin of a radiation monitoring network, or have much of that development in place.
- Appear to have products that are fairly mature with variations on aspects of their product relative to each other (e.g., location tracking, radiation field heat map, use of fixed radiation monitors, etc.).
- Have products that were able to utilize 3D scanning technology that would subsequently support area mapping, heat map generation and location tracking.
- Expressed a willingness to participate in a tabletop demonstration
- Selection of supplier(s) for the Tabletop Demo this month
- Demos to be conducted in 2021



## Digital Twin of a Real-time Radiation Monitoring Network

**Phase 2 - Apply Machine Learning to Geophysical Radiological Survey Application** 

- Machine learning can be coupled with the geospatial algorithm to refine the radiation field estimates as measurements are updated
- The digital twin could be used to provide ongoing monitoring, trending, and alerts and allow for alternative maintenance and dose optimization scenarios to be investigated in cyberspace before the work is performed
- Simulation results could be visualized by the worker during job preparations, job briefings, and in the work environment using augmented reality techniques.
- Simulations would enable development of efficient maintenance practices to save time, reduce worker exposure, and reduce cost
- Could be used for event recreation and emergency plan exercises
- The proposed added scope for 2022-2023 includes:
  - Evaluate software tool(s) for use in machine learning of survey data
  - Test application to see if tool accurately predicts future radiation dose rate trends from survey data analyzed



Example of ALARA Planning Tool for Potential Further Development



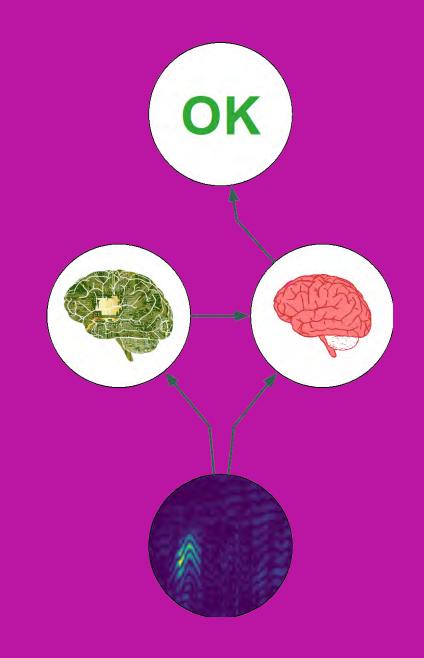


#### likka Virkkunen

Prof. (Adjunct) at Aalto University, Managing Director or Trueflaw, Ltd.

NDE4.0 and
Machine Learning
for In-service
Inspections





# NDE increases reliability

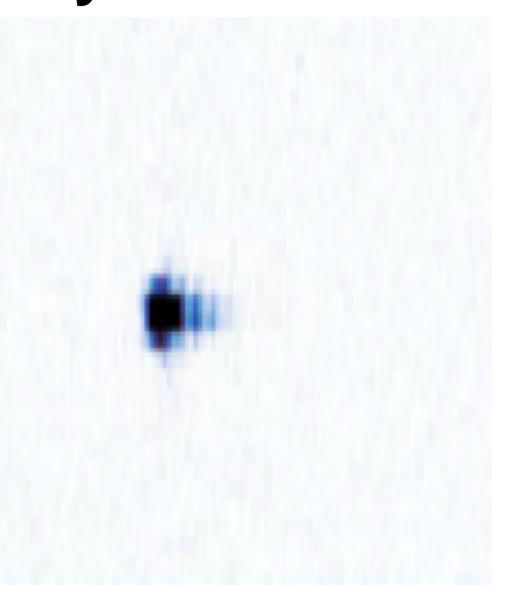
### Today's NDE3.0

- Qualified and reliable
- Rich digital data
- Important contribution to reliable and safe operation

#### **But**

- Time consuming
- Potential for human errors
- Limited information extracted
- More is less





# Machine learning enables human-level automated evaluation

Human-level performance in automated defect recognition

**Applicable to various inspections:** 

**Ultrasonics** 

Digital radiography

Already in field use in non-nuclear inspection



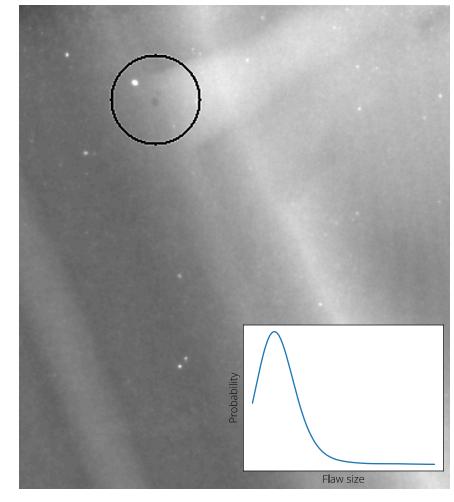


# Case: GKN digital X-ray of aerospace welds

High quality automated welds Small flaws, difficult to detect Current status:

- Human level detection
- Accurate sizing
- Criteria comparison

Additional data on small flaws



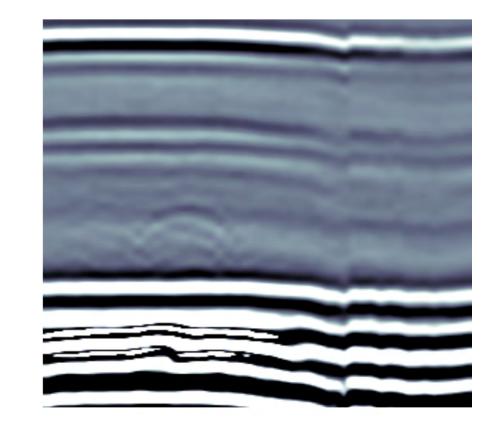


# Case: CRDM TOFD inspection

**EPRI – Trueflaw collaboration** 

Very difficult data
Limited real data available
Limited real flaws available
Human level performance

Next: field test planned





# NDE3.0 + ML = NDE3.5

- Automated analysis can manage large data
- Inspector work elevated to focus on the important parts
- More is more again

#### **Additional benefits**

- More sensitivity
- Predictive capability
- Added value from NDE



# NDE4.0 Integrated NDE data

- Connected systems can aggregate data
- Digital twin etc.
- Elevated to focus on the important parts for the system
- More is exponentially more due to network effects

#### **Additional benefits**

- Situation awareness
- Predictive capability
- Added value from NDE



# How do we get there, safely?

Compatibility with existing qualification

Compatibility with current practices

**Data security** 





# Qualification

#### 2020 ENIQ published a position paper:

• Qualifying (certain) ML systems filled within framework

#### **ENIQ RP13:**

Recommended practice for Qualification of Non-Destructive Testing Systems that Make Use of Machine Learning



# We now know how to qualify ML

# Repeatability of ML systems is good for qualification

#### Main changes are with test data

- ML systems may require more test data
- Combining open and blind trials may reduce test block needs

# Qualifications must be done using frozen software



#### **ENIQ RECOMMENDED PRACTICE**

**ENIQ Recommended Practice 13** 

Qualification of Non-Destructive Testing Systems that Make Use of Machine Learning

Issue 1

ENIQ Report No. 65

Technical Area 8

European Network for Inspection & Qualification



# Edge devices can secure data

- Stand-alone unit
- Easy to use
- Works with existing data files
- Multiple options for reports
- Can provide traditional reports
- Can integrate to digital twins





# Conclusions

For digital twins, NDE  $\Rightarrow$  4.0

#### The tools are ready:

- Machine learning
- Edge computing

#### The transition is clear

- Qualification RP13
- Adoption with stand-alone units













# Digital Twin Enabling Technologies in Advanced Reactor Applications

**Technical Session – September 15, 2021** 

**Moderator: Angela Buford, NRC** 

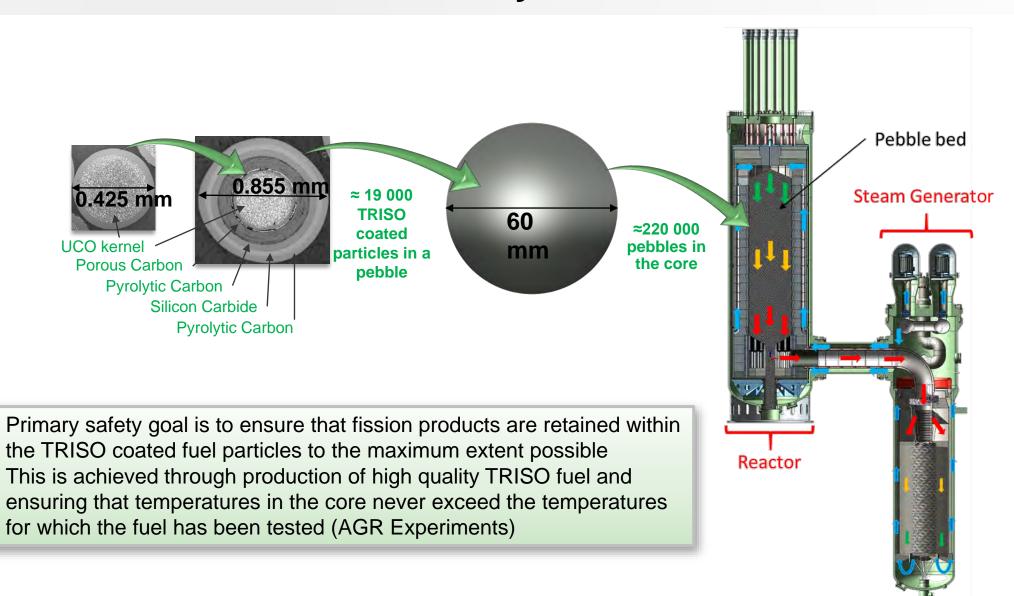
Ian Davis, X-energy
Nurali Virani, GE
Anthonie Cilliers, Kairos Power
Emilio Baglietto, MIT
Michael Buric, DOE
Bob Urberger & Roger Chin, Radiant







## **UCO TRISO Particle – Primary Fission Product Barrier**

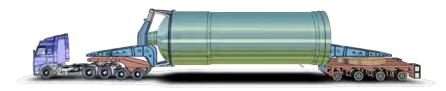


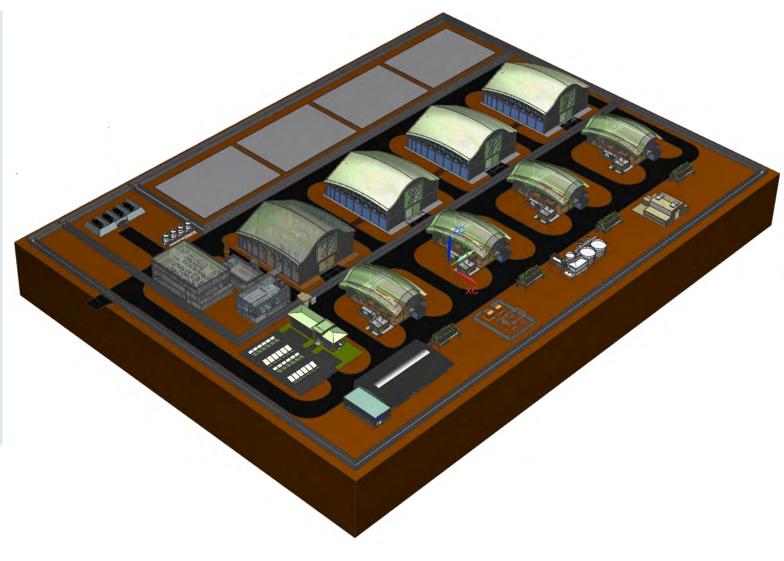


## **Background: Xe-100 Plant Overview**

Standard X-energy plant have 4 Reactors - 4 Turbines producing 320 MWe, attributes include:

- 200MWth/80MWe Per Module
- Process heat applications Proven intrinsically safe
- Meltdown proof
- Walk-away safe
- Modular construction
- Requires less time to construct (2.5-4 years)
- Road transportable for diverse
- geographic areas
  Uses factory-produced components
  Load-following to 40% power within 15 minutes
- Continuous fueling; resilient on-site fuel storage









## Xe-100 Digital Twin Tools





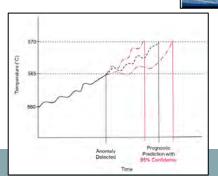
3D Models with AR / VR

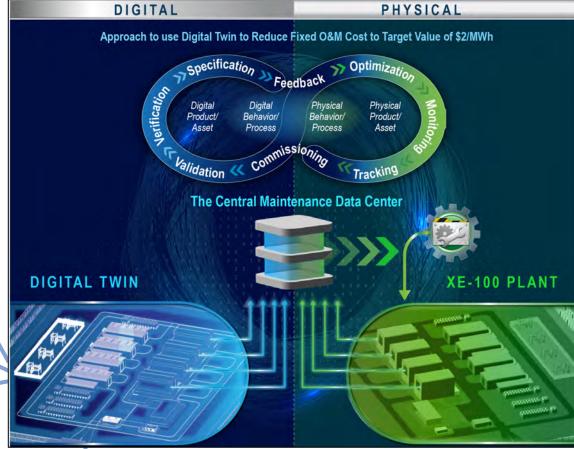


Operator Training
Simulator



Plant Historian



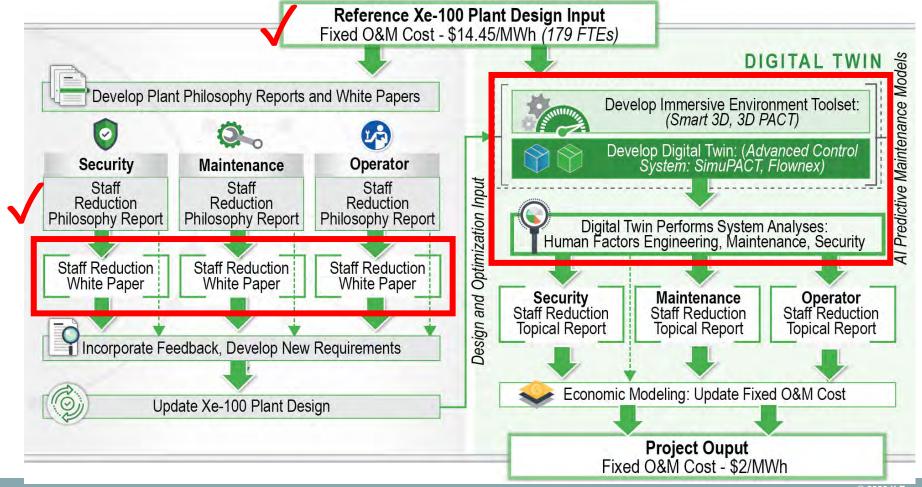


AI / ML Models



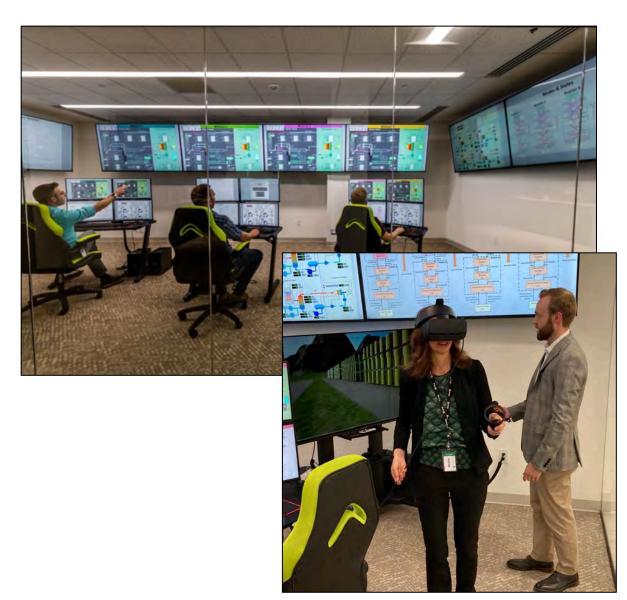
#### ARPA-E GEMINA Project Progress Summary

- Project Title: Advanced Operation & Maintenance Techniques Implemented in the Xe-100 Plant Digital Twin to Reduce Fixed O&M Cost
- \$7.5 Million award from DOE for Digital Twin (DT) and Central Maintenance Model (CMM) concepts





# Simulator and 3D Models









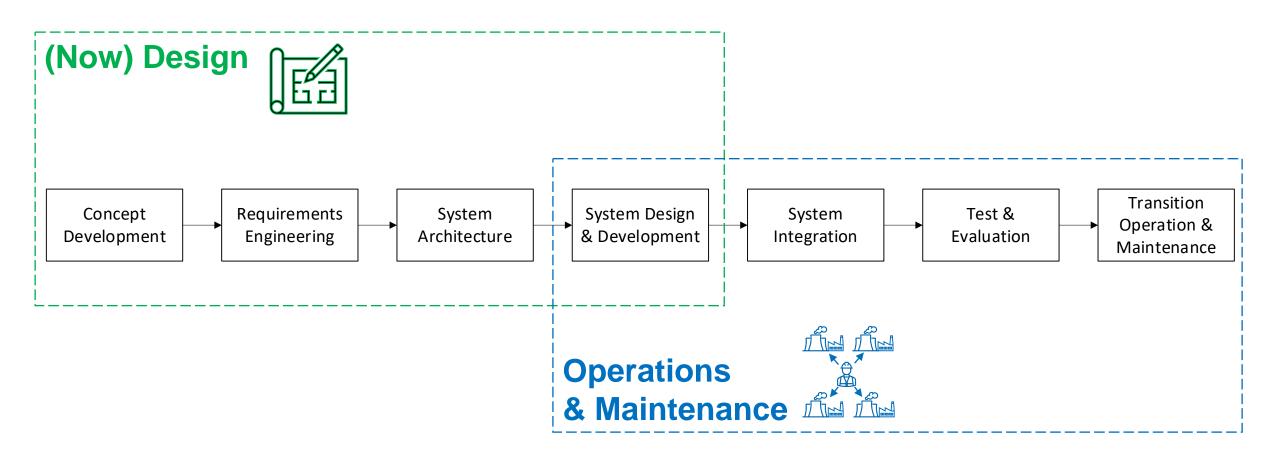


## Systems Engineering





## Systems Engineering with Xe-100 Digital Twin

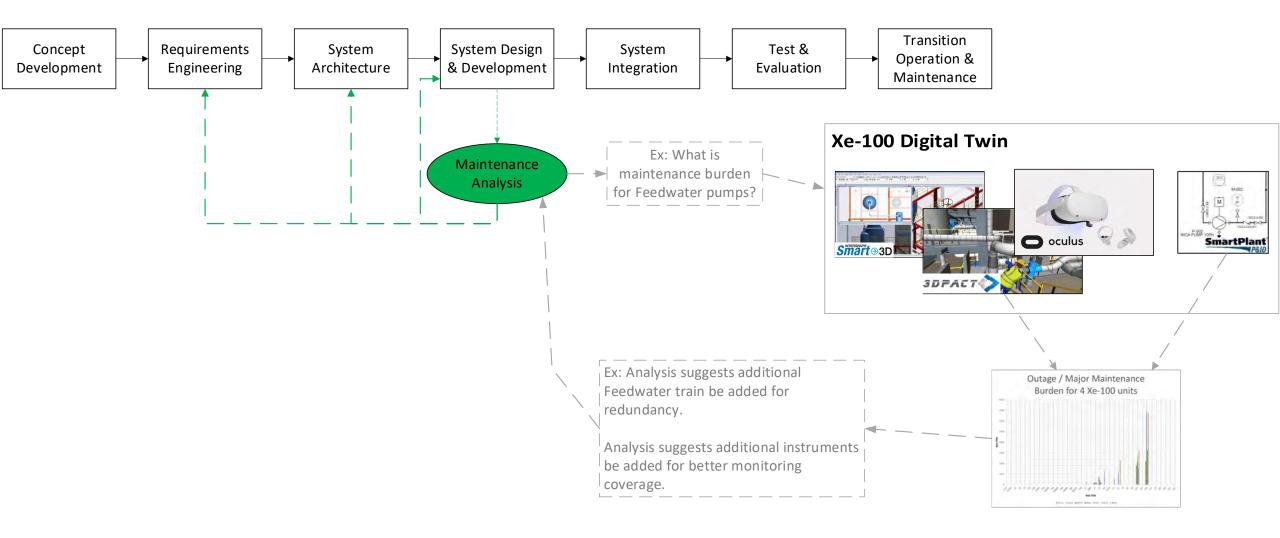






## Systems Engineering with Digital Twin: Maintenance

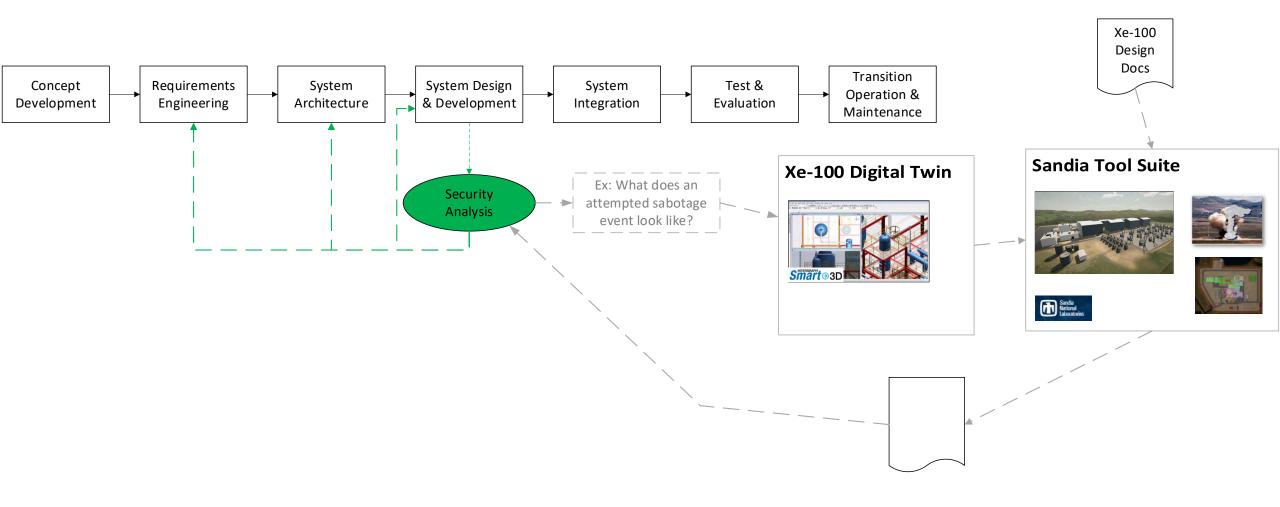






## Systems Engineering with Digital Twin: Security



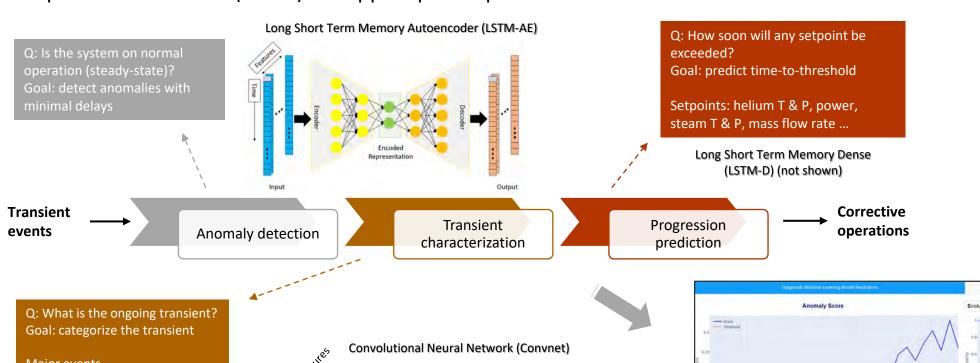




## Anomaly Detection with Machine Learning

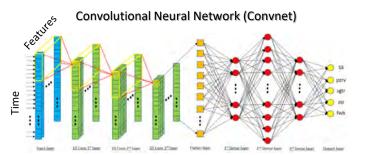


#### Deep Neural Networks (DNNs) to support plant operation

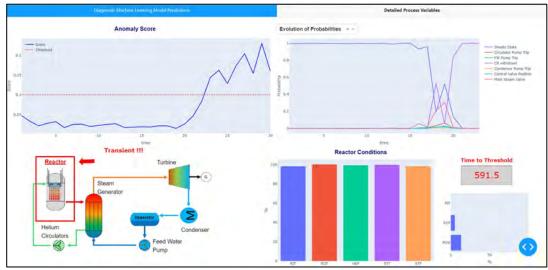


#### Major events

- Circulator pump trip
- Condenser pump trip
- Control valve position
- Clogging of MSV in turbine
- Control rod withdrawal
- Feedwater pump trip
- Unknown



#### **Human Machine Interface (HMI)**







## Wednesday, September 15<sup>th</sup> Digital Twin Enabling Technologies in Advanced Reactor Applications

# Humble AI for Reliable Machine Learning-based Health Twins

Dr. Nurali Virani

**Lead Scientist – Machine Learning, GE Research** 

December 7, 2021

## What are Machine Learning-based Health Twins?



Digital twins for fault detection, diagnosis, and health estimation of physical systems and components

#### Detection

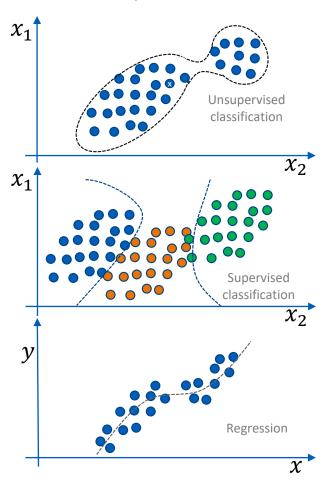
- Fault or anomaly detection with early warning is formulated as an unsupervised learning problem
- Needs data of nominal behavior from modeled system

#### Diagnosis

- Fault classification is formulated as a supervised learning problem
- Needs labeled data for multiple fault classes

#### Health estimation & Forecasting

- Health estimation and performance prediction of continuous-valued variable is formulated as supervised learning regression problem
- Needs input-output data pairs for training the model



Key gap in using health twins-based automation for critical industrial infrastructure is in characterization of reliability and trust as safety and performance are paramount



## Understanding prediction reliability can help improve outcomes and build trust in Al

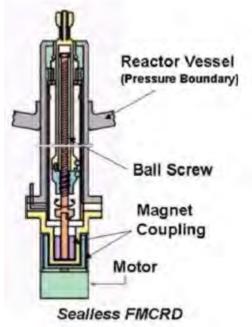
**Automated Fault Classification** 





- ❖ I know this is fault class 1, here is my evidence, please authorize to take appropriate actions.
- The *feedwater pump* might have fault 1 or 3, but not 2, send technicians who can address type 1 and 3 faults

Reduce O&M cost via automation, minimize repeated crew trips and avoidable shutdown events





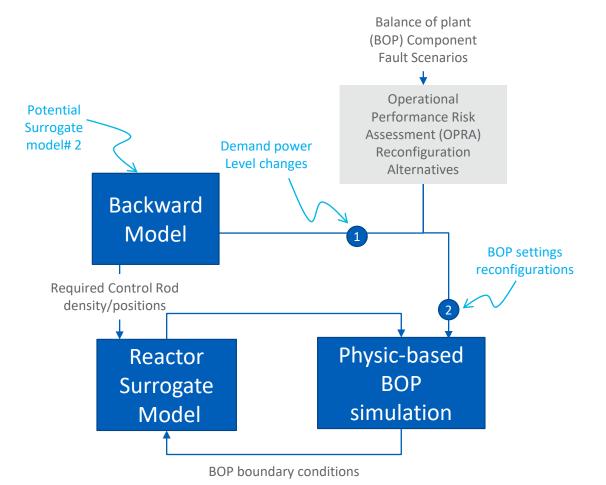
The *FMCRD* behavior is anomalous. I don't know if FMRCD has a known fault, partner can you investigate it and trigger safe mode operation.

Image source: FIG. 2.2 in https://aris.iaea.org/PDF/ABWR(Hitachi-GE) 2020.pdf

Better collaborative outcomes and ability to deal with novel scenarios

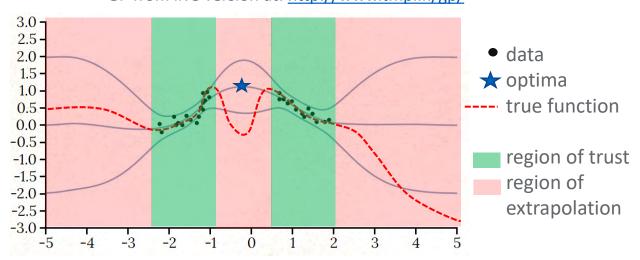


## Surrogate model-based optimization and control needs better assessment of prediction reliability



From ARPA-E GEMINA Predictive Maintenance with Digital Twin program (GE Research, ORNL, UTK, GE Hitachi, Exelon)

Script: <a href="https://github.com/to-mi/gp-demo-js">https://github.com/to-mi/gp-demo-js</a>
GP from live version at: <a href="http://www.tmpl.fi/gp/">https://github.com/to-mi/gp-demo-js</a>

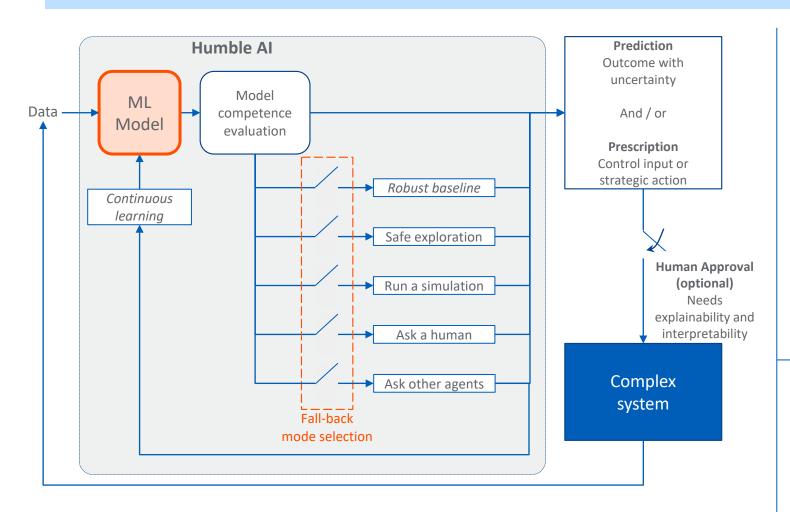


Lack of coverage: Complete variability that we expect during deployment is not available in training data

Can I trust the surrogate model prediction for the suggested BOP boundary conditions and control rod density values?

## Humble AI for Digital Twin

#### An AI that is aware of its own competence and improves its competence via learning



#### **Defining capabilities**

understand region of trust

quantify uncertainty

ask for help when incompetent

continual learning from 1 or more sources

- ✓ Humble AI will reduce Time to Value
- ✓ Humble AI will maintain safety



## Key Gap and Research Question

- Key Gap: Machine learning provides statistically impressive results which might be individually unreliable.
  - "My validation accuracy was high, so trust my belief"
  - "Soft-max value for predicted class is high, so trust my belief" (distance from hyperplane)
- In many situation, randomized inspection of samples is inadequate to verify reliable outcomes and complete inspection defeats the purpose of using AI

Can we characterize individual prediction reliability to understand the limitations of ML due to:

- Observability (or separability)?
- Brittle extrapolation?

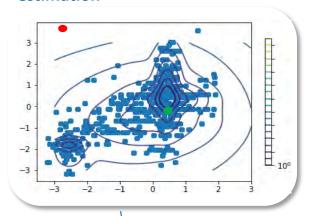


## Support types to create justification for model competence

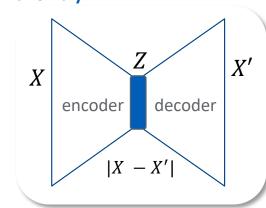
Using model input, model internal representation, and model outputs for support

#	Types of Support	Gua	Guard against	
1	Input anomaly	•	Input drift	
		•	Extrapolation	
2	Reconstruction error	•	Input drift	
_		•	Extrapolation	
3	Geometric neighborhood in	•	Extrapolation	
	embeddings	•	Ambiguity	
		•	Adversarial manipulations	
4	Output uncertainty anomaly	•	Input drift	
•		•	High process uncertainty	
5	Residual error drift	•	Concept drift	

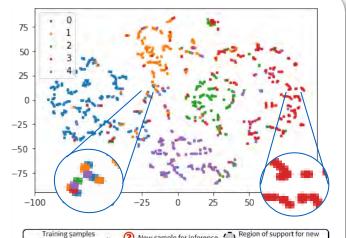
## (1) Input likelihood with density estimation



## (2) Reconstruction error-based anomaly



## (3) Geometric neighborhood in embeddings (latent spaces)

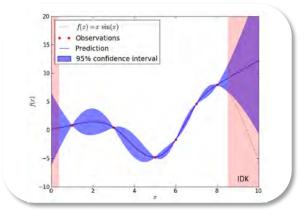




Neighborhood operator:

- (a) by number of neighbors
- (b) by radius (distance threshold)

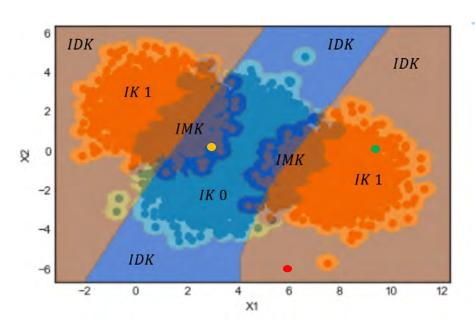
#### (4) Output uncertainty anomaly

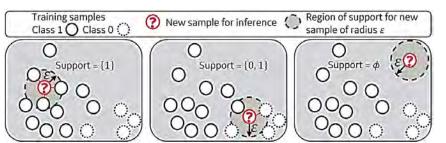


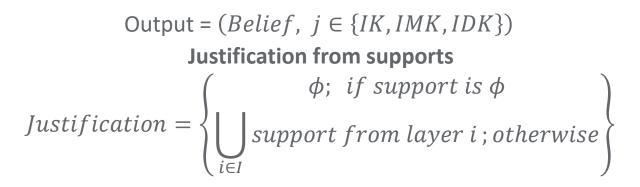
#### (5) Residual error drift



## Characterizing region of trust, overlap, and extrapolation for AI to get justification-based reliability







#### Justified belief is knowledge

 $Belief = Justification \rightarrow I \text{ know (Region of trust)}$ 

 $Belief \subset Justification \rightarrow I \text{ may know (Region of overlap/ambiguity)}$ 

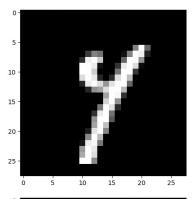
 $Belief \not\subset Justification \rightarrow I don't know (Region of extrapolation)$ 

1. Virani, N., Iyer, N. and Yang, Z. Justification-Based Reliability in Machine Learning. in AAAI 2020

2. Bhushan, C., Yang, Z., Virani, N. and Iyer, N., 2020. Variational encoder-based reliable classification. arXiv preprint arXiv:2002.08289. (IEEE ICIP 2020)



## Examples of evaluating competence using internal representations



Belief: 9 IMK [4, 9] Truth: 4

Dataset:

CIFAR (car and truck class)

Base model:

Residual Network

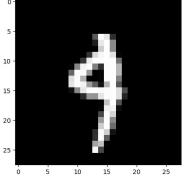
Layer:

Global average pooling

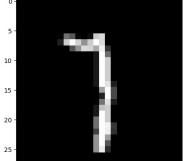
**Support:** 

 $\varepsilon$ -ball with

 $\ell_2$ -metric



Belief: 4 IMK [4, 9] Truth: 4



Belief: 1 IMK [1, 7] Truth: 7





Support from latent space as exemplars for interpretability



Visualization of Latent

Space - Car and Truck

Classes – IK, IMK, and

IDK

IK Car

IDK IMK

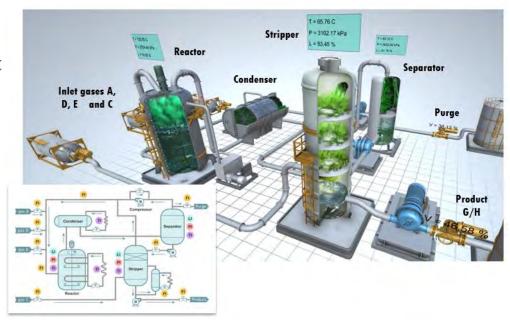
IK Truck

## Data to facilitate Health Twin development and validation



#### Tennessee Eastman Process (TEP) simulation data<sup>a</sup>

- Open-source benchmarking data of industrial chemical process for the purpose of developing, studying and evaluating process control and fault detection
- Process has 12 valves available for manipulation and 41 measurements available for monitoring or control.
- Time series sensor data, with faults injected using simulation
- Variables sampled every 3 minutes for a total of 25 hours (training data) and 48 hours (test data)
- Overall Data size: 15M rows x 52 variables x 1-label
- Labels: 20 fault-types + Normal



a:

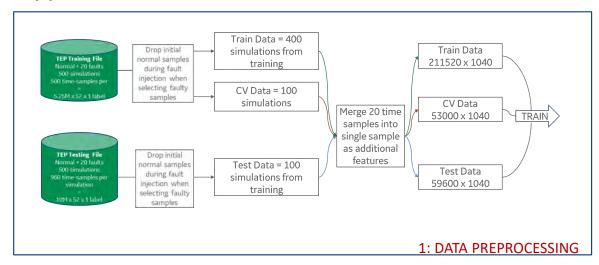
1> Downs, J.J. and Vogel, E.F., 1993. A plant-wide industrial process control problem. *Computers & chemical engineering*, *17*(3), pp.245-255

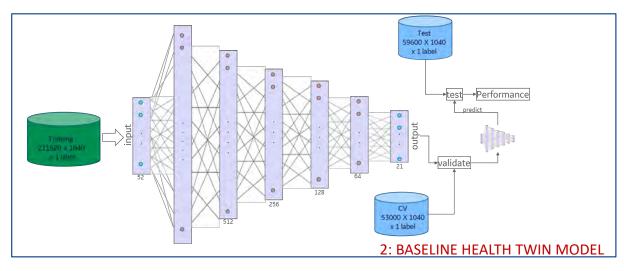
2> Rieth, Cory A.; Amsel, Ben D.; Tran, Randy; Cook, Maia B., 2017, "Additional Tennessee Eastman Process Simulation Data for Anomaly Detection Evaluation"

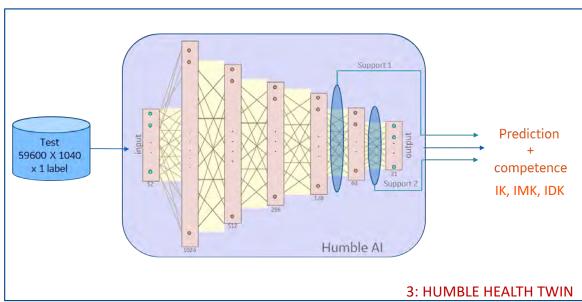


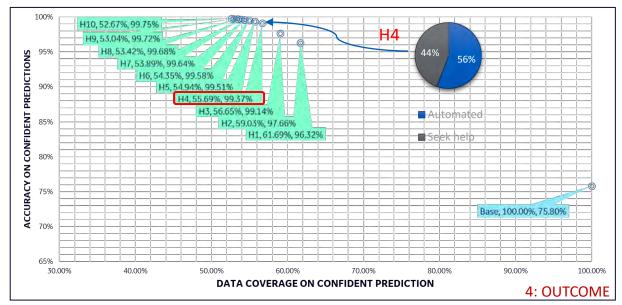
### Humble Health Twins for Fault Classification

#### Approach & Outcomes





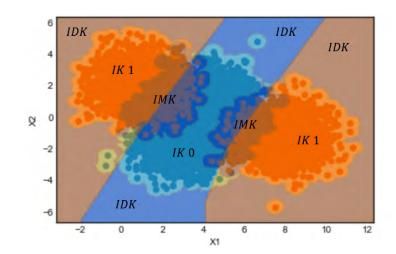


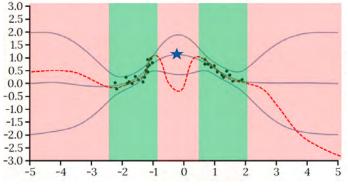




## Summary

- ✓ Humble AI is an AI that is aware of its own competence and improves its competence via learning
- ✓ Current focus of Humble AI is on ML-based digital twins (specifically health twins)
- ✓ Humble AI enables to characterize region of trust for health twins to get justification-based reliability where automation can be enabled
- ✓ Extend to use model input, model internal representation, and model outputs for support (beyond NN models)











## Building a world that works

Collaborate with us on research programs for a better tomorrow (nurali.virani@ge.com)



ENABLING TECHNOLOGIES FOR DIGITAL TWIN APPLICATIONS FOR THE KP-FHR

CHRIS PORESKY AND ANTHONIE CILLIERS

Kairos Power's mission is to enable the world's transition to clean energy, with the ultimate goal of dramatically improving people's quality of life while protecting the environment.

## Advanced reactors – challenges and opportunities

- Industry experience in plant simulators is entirely focused on light water reactors
- Challenging environments, more difficult component and structure qualification
- Limited historical data on performance and documentation of material and chemical properties
- Parallel engineering developing capabilities while developing application space simultaneously
- Advanced reactor priority phenomena may occur at different (i.e. slower) timescales
- Modern instrumentation and control may more easily lend itself to data communication
- Two-phase phenomena may be deprioritized for non-water coolants
- Passive safety may reduce the dependency on active control

## Overview of Kairos Power

#### KP-FHR Inherent Safety and Economic Potential are Unique:

#### Robust Inherent Safety

- Large *fuel temperature margins*
- fission products retained by fuel and primary coolant
- Low-pressure system
- Passive decay heat removal

#### Lower Capital Costs

- Reduced reliance on high-cost, nuclear-grade components and structures through FHR intrinsic safety and plant architecture
- Leverage conventional materials, existing industrial equipment, and conventional fabrication and construction methods

#### Improved Operating Economics

- High efficiency
- Flexible deployment of low-cost nuclear heat

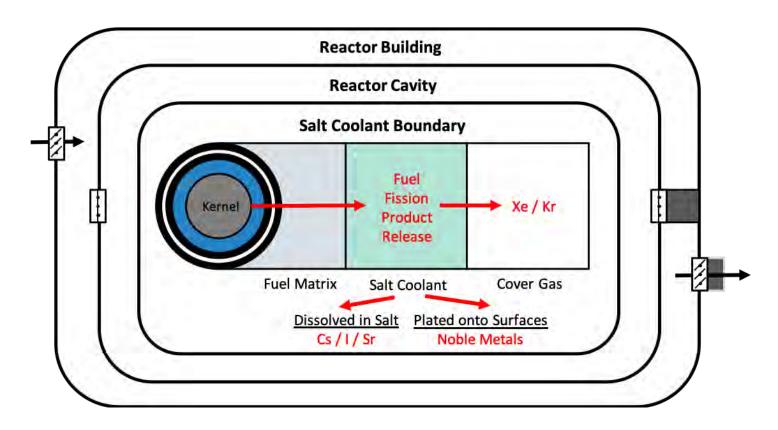


KP-FHR	HTGR	Fast Breeder Reactor	PWR SMR
Flibe	Helium	Sodium	Water
$280~\mathrm{MW}_\mathrm{th}$	$250~\mathrm{MW_{th}}$	$260~\mathrm{MW_{th}}$	$200~\mathrm{MW_{th}}$
120 MW <sub>e</sub>	100 MW <sub>e</sub>	100 MW <sub>e</sub>	60 MW <sub>e</sub>

<sup>\*</sup> Reactor vessels drawn to notional scale.

## KP-FHR Safety Case – Defense in Depth Barriers

• Radionuclide Retention Capabilities



## **KP-FHR I&C design**

## Robust Inherent Safety

- Low operating pressure
- Large fuel temperature margins
- Effective passive decay heat removal
- Uniquely large heat capacity
- Strong negative temperature coefficient
- Slow transient response



Safety Hazard Intervention and Event Limiting Defense (Shield)

KP-Sword: Active plant control

System with Operational Reliability and Diagnostics (Sword)

KP-Heart: Intelligent Health Monitoring

Health Evaluation and Analysis in Real-Time (Heart)

• KP-Sight: Semi-autonomous control room

Semi-autonomous Industrial Grade HMI Technology

KP-Bolt: Electrical supply

Basic Ohm Law Triangle (V = I.R)



 KP-Shield: Passive, robust, reliable safety shutdown system









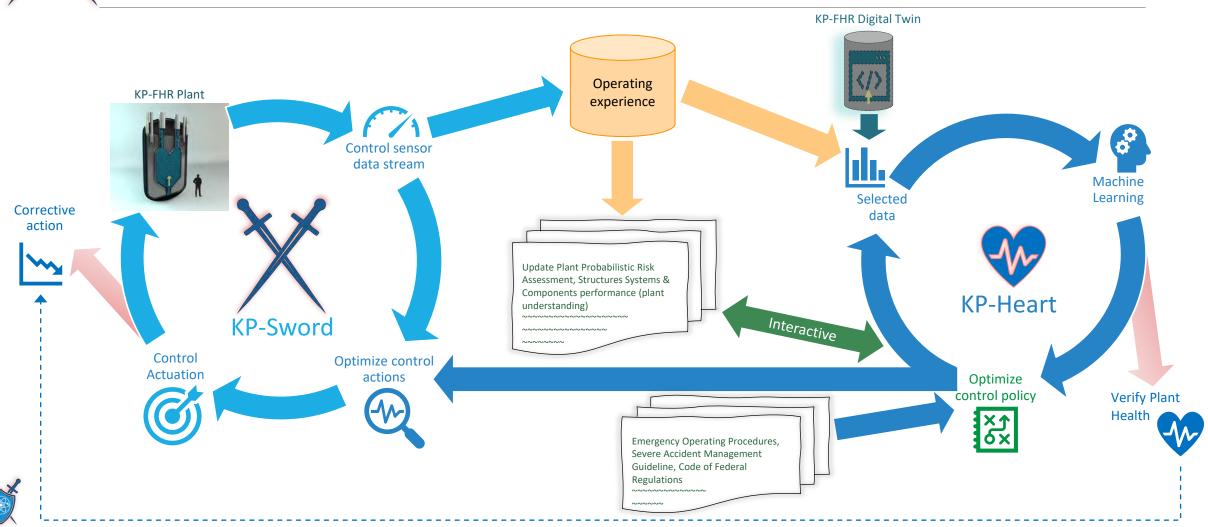




## Active plant Control & Health Monitoring: KP-Sword & KP-Heart 🕠



Normal Operations and Anticipated Operational Occurrences

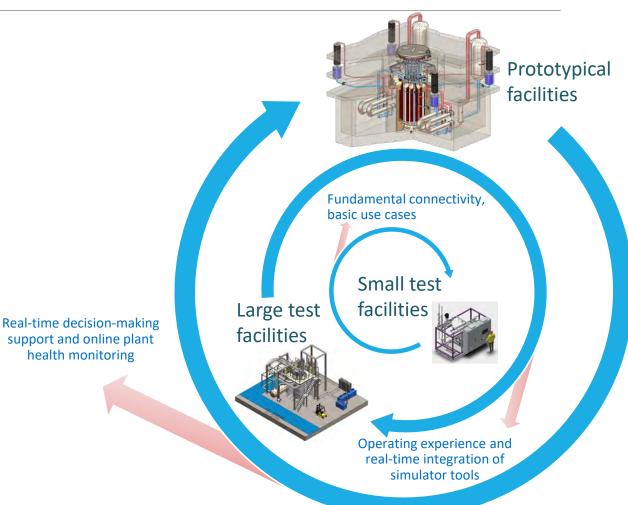


## KP-FHR digital twin working definition

- Collects, compiles, and contextualizes plant information from a variety of sources and makes it accessible to aid plant operation
  - Visualization
  - Database
  - Physics simulation
  - Real-time communications and feedback
  - Data analytics
    - Plant health monitoring
    - Operations optimization
    - Operator decision support

## KP-FHR digital twin development process

- Small test facilities
  - Demonstrate ability to connect facilities using different hardware and software
  - Validate simulation tools against representative physical systems
- Large test facilities
  - Deploy digital twins alongside tests to gather operating experience and train models
  - Integrate simulation into operation to improve understanding of system
- Prototypical facilities
  - Support decision-making in real time
  - Passively and continuously monitor plant health and flag unexpected behavior



## Desired capabilities and enabling technology

- Desired capabilities
  - Run continuously alongside plant operation
  - Provide lookahead and optimization capabilities to operators and support staff
  - Flag unexpected behavior and prompt important operations and maintenance activities
- Enabling technology
  - Communications infrastructure: talk to engineering tools, plant hardware, controls and simulation software
  - Physical processes: develop and specify all plant components and prototypical information
  - i: develop faster-than-real time physics simulation and data analysis capabilities

## A note on cybersecurity and plant design

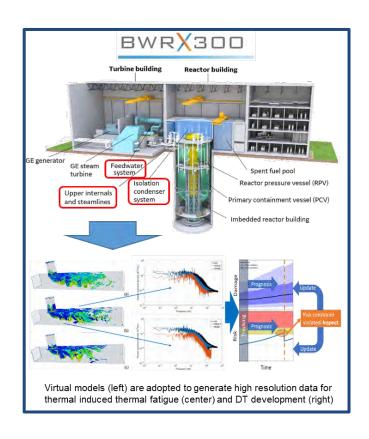
- If the plant design precludes consequence, the risk for all failures is lower
- If instrumentation and controls architecture precludes consequence, the reactor protection functions are not at risk
- If reactor protection functions are not at risk, the value of added capabilities becomes much more attractive despite potential risks
- Digital twins can **bolster** cybersecurity by improving the ability not only to detect but also to mitigate cyber threats
  - Mitigation of cyber threats and approach to cybersecurity should be baked into plant design the same as it is for mitigation of insider threat, access control, intellectual property management, etc.

#### NSE

Nuclear Science & Engineering at MIT

science: systems: society





## HIGH FIDELITY DIGITAL TWINS FOR BWRX-300 CRITICAL SYSTEMS

Focus on "enabling" technology

#### **Emilio Baglietto**

Massachusetts Institute of Technology
Department of Nuclear Science and Engineering

## **High Fidelity Digital Twins for BWRX-300**









GE Global Research



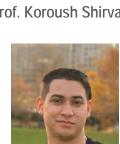
Prof. Emilio Baglietto (PI)



Prof. Koroush Shirvan (co-Pl)



Yu-Jou Wang (PhD)



Brandon Aranda (MS)



Dr. Panos Tsilifis (co-Pl)



Genghis Khan (co-Pl)



### **HITACHI**



Dr. Christer Dahlgren(co-



Douglas McDonald

**David Hinds** 

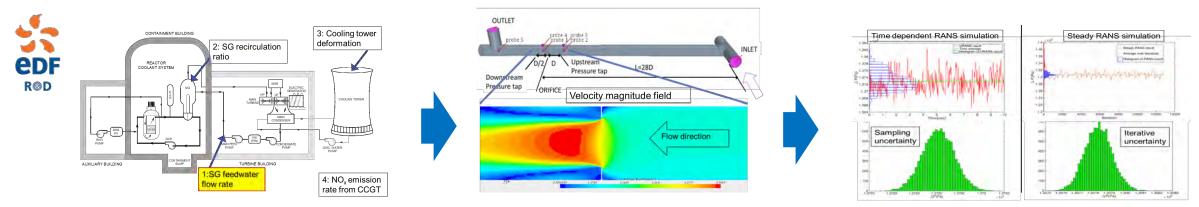


Charles Heck

## The Backbone

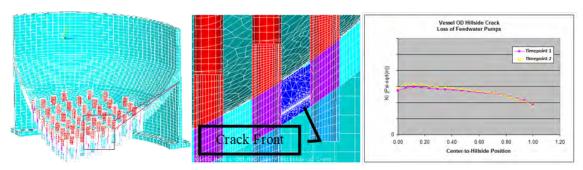


MIT: e.g. demonstrated reduction of operating uncertainty through high fidelity simulations

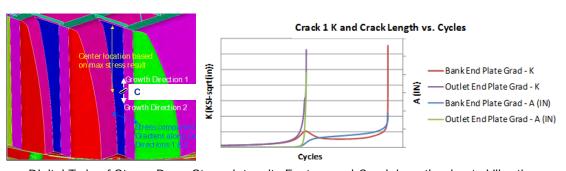


U.Otgonbaatar, E.Baglietto, Y.Caffari, N.E.Todreas and G.Lenci - A METHODOLOGY FOR CHARACTERIZING REPRESENTATIVENESS UNCERTAINTY IN PERFORMANCE INDICATOR MEASUREMENTS OF POWER GENERATING SYSTEMS - JVVUQ

• **GE**: e.g. digital twin deployed in Nuclear and Aerospace Industries



Digital Twin of BWR Stress Intensity Factor after simulated loss of feedwater pumps

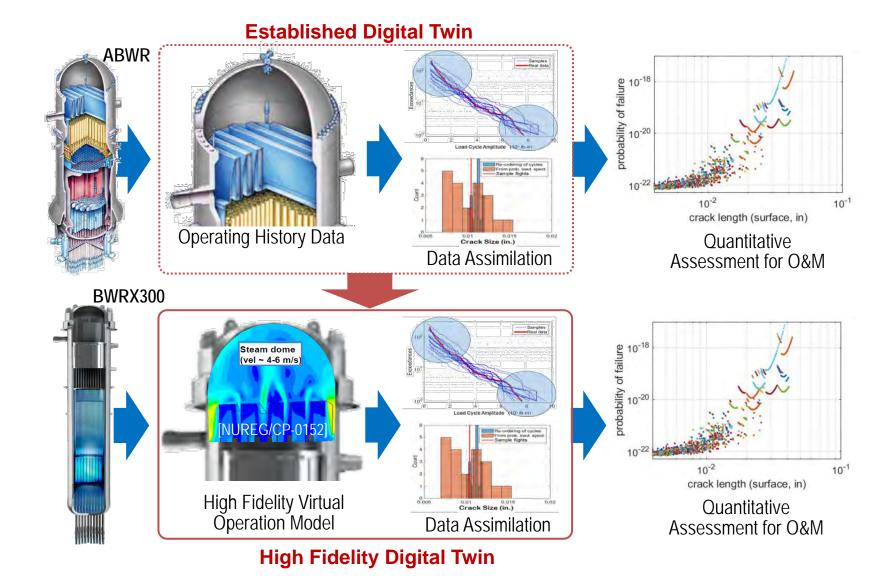


Digital Twin of Steam Dryer Stress Intensity Factors and Crack Lengths due to Vibration

## Why high-fidelity and why now ...



- Advancement and demonstration of highfidelity simulations based maintenance approaches and model based fault system detection techniques.
- Address mechanical and thermal fatigue failure modes which drive O&M activities of BWRX-300



## Why high-fidelity and why now ...



- Advancement and demonstration of highfidelity simulations based maintenance approaches and model based fault system detection techniques.
- Address mechanical and thermal fatigue failure modes which drive O&M activities of BWRX-300, and are extendable to all advanced reactors (ARs) where a flowing fluid is present.



## ... why now ...



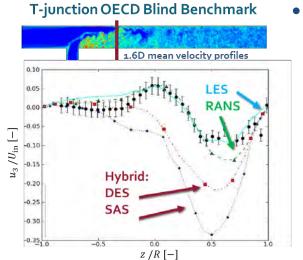
- Computational Fluid Dynamics is an old tool, it has been leveraged by various industries for many years, and it has provided great support to design and safety of nuclear reactors as a complement to experimental and experience based approaches
- CFD has a much greater potential: "to provide high-fidelity data to support efficient operation" of NPPs... bootstrapping the lack of operational data
- The incomplete maturity of the simulation methods (and teams) and the excessive computational costs has hindered this last major jump.

 Leveraging the last 10 years of DOE and industry sponsored effort we are in position to demonstrate this jump

## ... why now ...



- Computational Fluid Dynamics is an old tool, it has been leveraged by various industries for many year, and it has provided great support to design and safety of nuclear reactors as a complement to experimental and experience based approaches
- But CFD has a much greater potential: "to provide high-fidelity data to support efficient operation" of NPPs... bootstrapping the lack of operational data
- The incomplete maturity of the simulation methods (and teams) and the excessive computational costs has hindered this last major jump.



- T-junction blind international benchmark 2013 had good and bad news
  - LES works very well, but its too slow to drive Operational DTs
  - Many acceleration ideas proposed for external aero, some successful, but not for Nuclear Applications

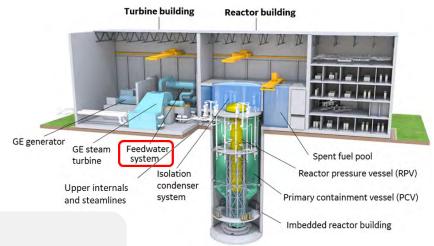
#### For Nuclear Applications we need robust reliability:

- We are looking for local hybridization in presence of turbulent structures
- Independence from grid resolution
- Independence from time stepping and spatial interpolation methods
- Leveraging the last 10 years of DOE and industry sponsored effort we are in position to deliver this capability

## Thermal striping driven fatigue prediction

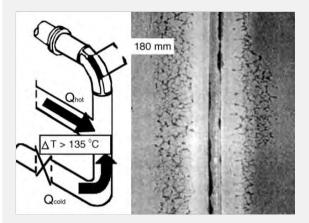


 Thermal striping fatigue has largest applicability to all AR concepts, beyond BWRX-300 – FIRST OBJECTIVE



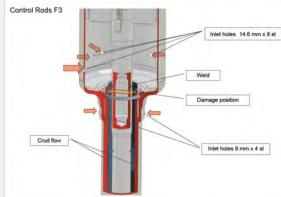
## A few notorious examples

#### Civaux (PWR)



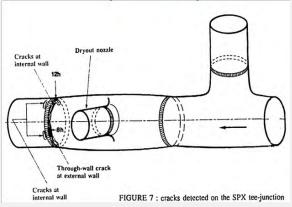
Civaux, EDF, crack in an elbow weld [K.-J. Metzner, U. Wilke /Nuclear Eng. and Design 235 (2005) 473–484 4811

#### Swedish BWRs CR Stems



Fatigue Inside Control Rod Guide
Tubes [Eric Lillberg, ICONE21-16632]

## Superfenix (LMFBR)

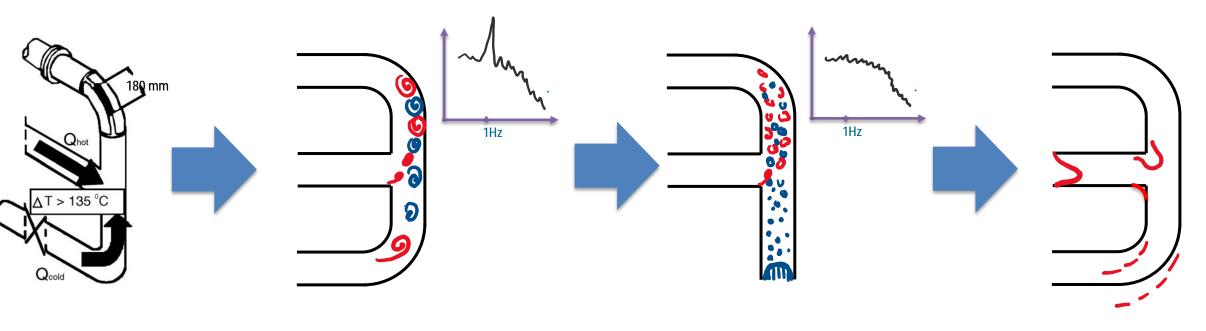


Auxiliary pipe in secondary circuit [O. Gelineau, M. Sperandio/IAEA-IWGFR/90 (1994)]

## The challenge in a (hand drawn) nutshell



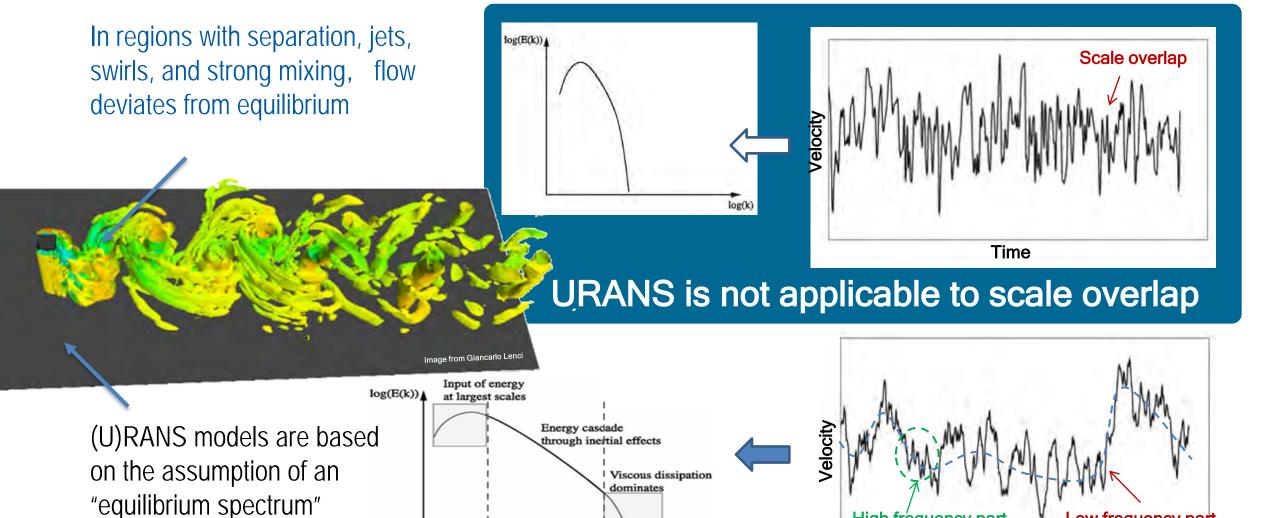
We can leverage the Civaux failure for discussion



- The turbulent structures generated at the 90° elbow interact at the T and lead to low frequency (1-3 Hz) large temperature oscillation that lead to accelerated fatigue failure
- The same T connection without the elbow does not suffer of these oscillations (but small flow / geometry variations could lead to the opposite results)
- Phenomenon is driven by formation and interaction of large turbulent structures and is strongly non-linear, not prone to "lumping" and generalization

## The STRUCT idea (Lenci and Baglietto, 2016)





log(k)

143

Low frequency part

Inertial Range

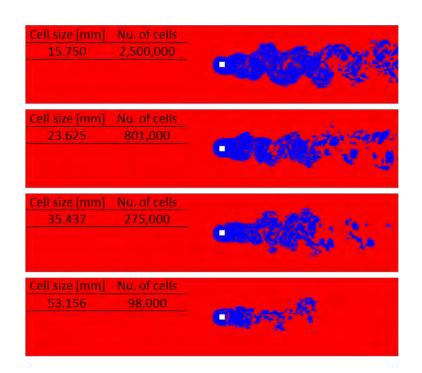
Time

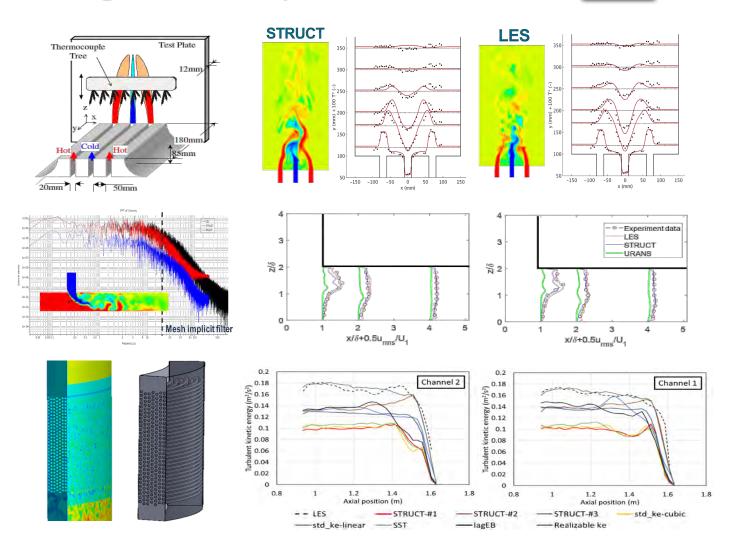
High frequency part

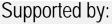
## The STRUCT idea (Lenci and Baglietto, 2016)

NSE
Nuclear Science
and Engineering

- STRUCT has demonstrated consistent grid convergence and accelerations between 50-100x on a number of validations
- Thanks to DOE and Industrial sponsorship













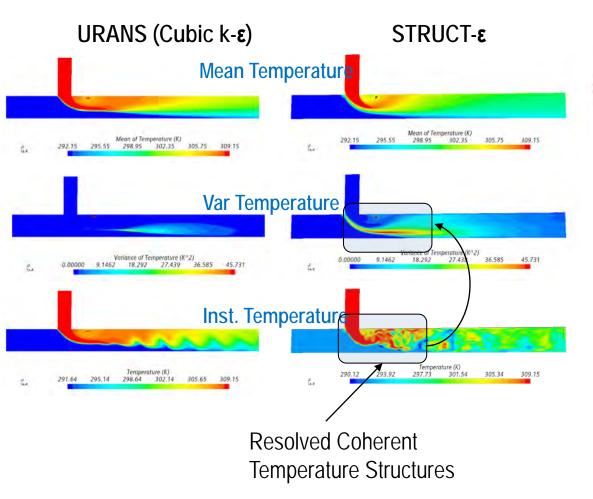


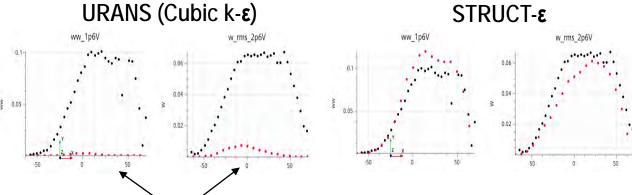


#### Recap: coherent flow structures drive T variation



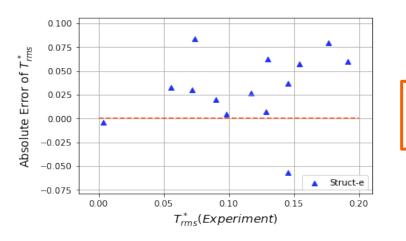
Maturity of models and application experience, at reduced computational cost





URANS Cannot predict Velocity and Temperature Fluctuations responsible for fatigue

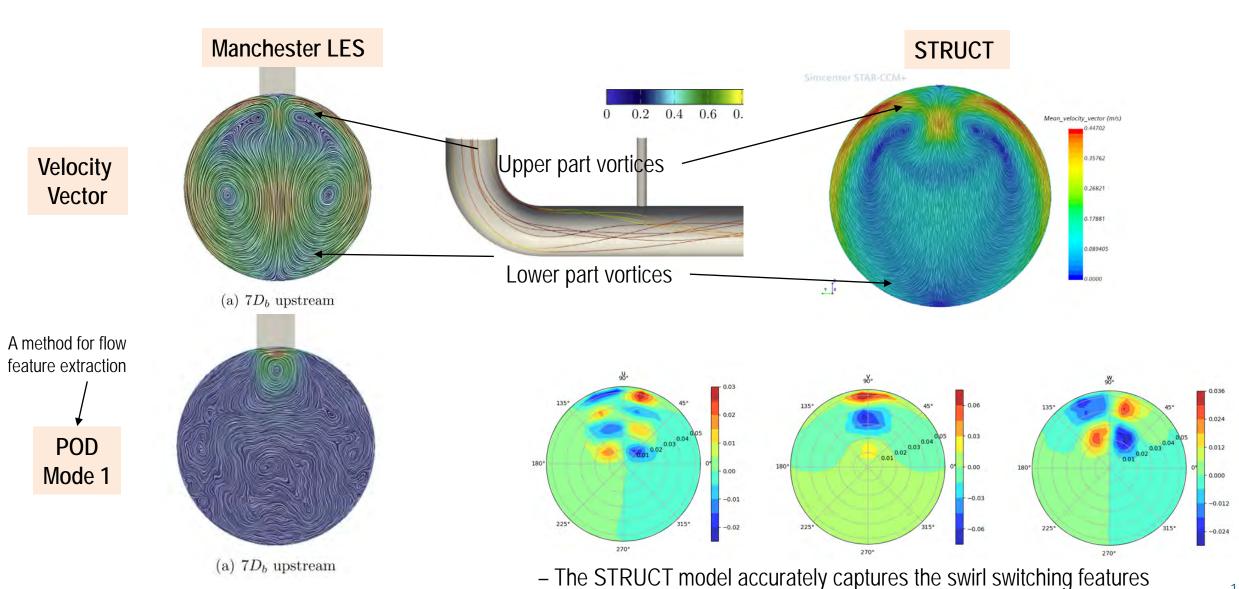
STRUCT predicts both velocity and Temperature fluctuations with high accuracy



Errors on T<sub>rms</sub> order of 1%

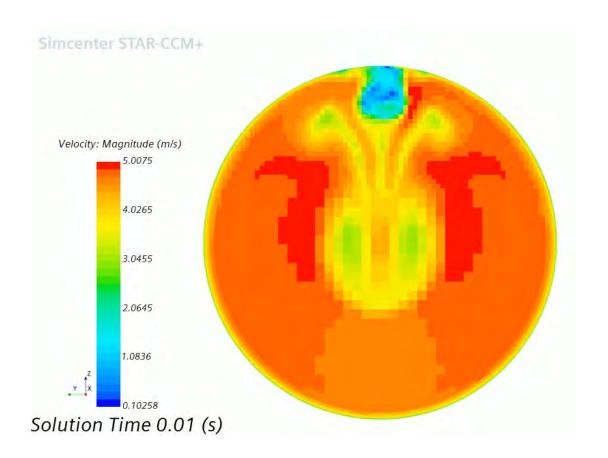
#### Application on feedwater system recap



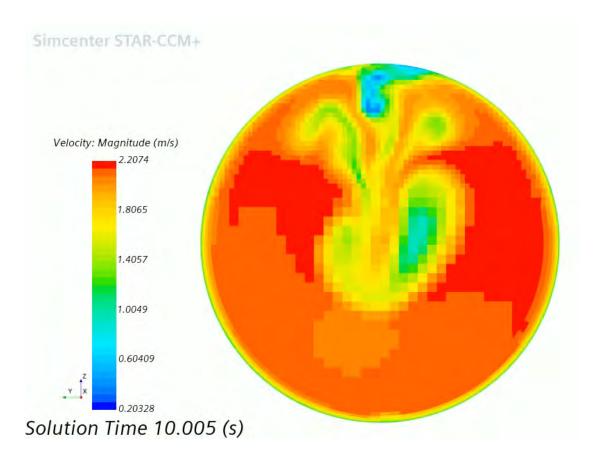


# Swirl-Switching example Fluid (~3.5 inch downstream)

• 100%



• 50%



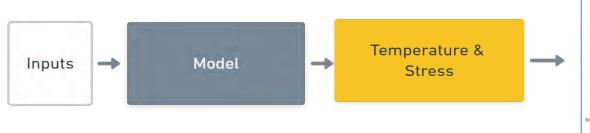
#### High-fidelity simulation based DT generation

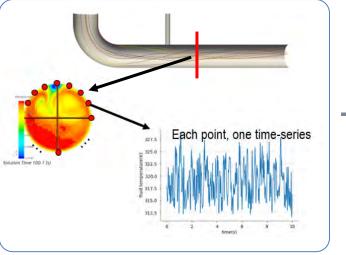


Damage or Fatigue

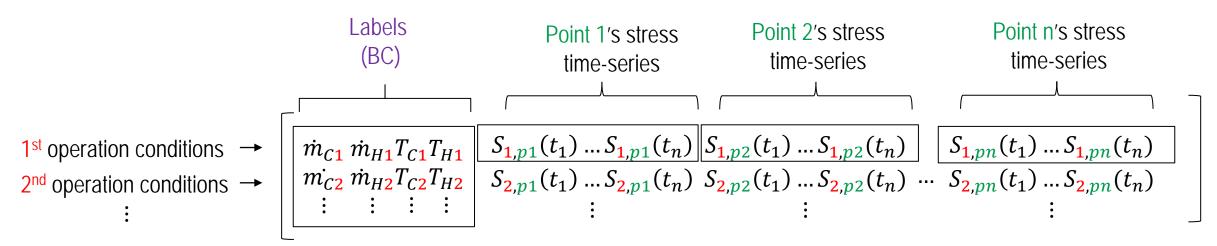
lifetime







The model outputs will look like this:

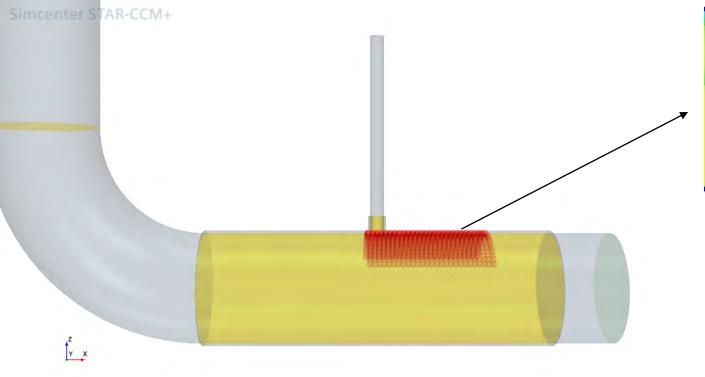


- The high-fidelity data provide a uniquely rich database for DT generation
- Capability of including sensitivity to many parameter variations

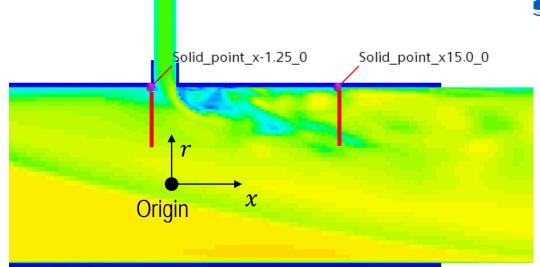
### **High fidelity Data Collection**





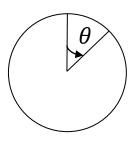


	Spec.	O.D.	Wall thickness	I.D
Main	316L 40S	16"	0.375"	15.25"
Branch	316L 80S	2 ½"	0.276"	1.948"



#### Probes:

$$x = -1.25 in \sim 15 in$$
  
 $\theta = -60 \sim 60 deg$ 



- 675 points at the inner wall (r = 7.625 in)
- 675 points at the outer wall (r = 8 in)

#### **Digital Twin Prototype**

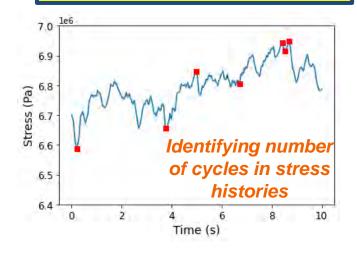
#### Feedwater subsystem

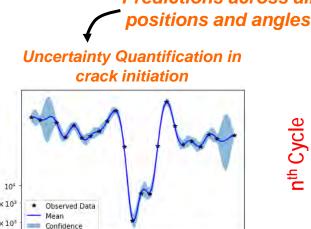
Analysis of number of cycles to damage initiation  $(N_i)$ using simulated data at each power level:

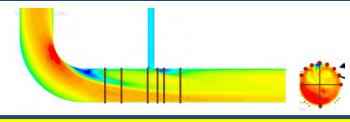
- Selection of power level (user input)
- Running Rainflow Counting on the available simulated stress histories
- Calculating N<sub>i</sub> values
- Modeling  $N_i$  across all angles and positions in the data for each cycle using Gaussian Process models

log(Ni)

#### **Rainflow Counting Algorithm**



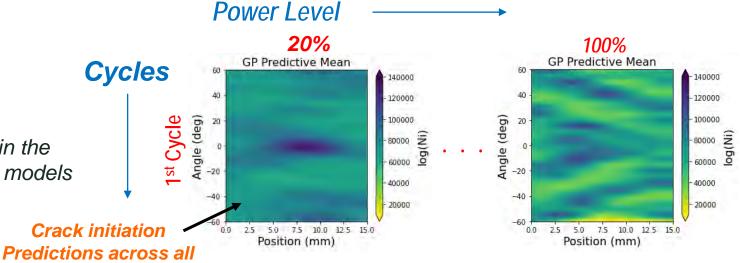


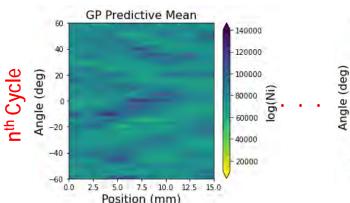


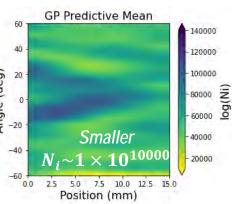




#### **Uncertainty Quantification in Crack Initiation**









# Molten salt-loop development acceleration with distributed single-crystal harsh-environment optical fiber-sensors

Presented by: Michael Buric, Staff Scientist NETL

2019-2022 ARPA-e TINA, 2021 update



#### Crystal fiber distributed sensing

#### **Project Objectives**

- Introducing fully-distributed sensing to Molten-Salt Reactors
- Growing new cladded single-crystal optical fibers for molten-salt environments
- Gathering thousands of data-points to map reactor coolant-path temperatures or other parameters
- Mapping in-core temperature distributions
- Next-gen sensing replaces single-point sensors like thermocouples
- Providing data to guide reactor design and improvement through thermal efficiency
- Support LFMSR Licensing Basis





#### Crystal fiber distributed sensing

Team

- National Energy Technology Lab (fiber growth, sensor design, interrogator design)
  - Michael Buric (PI, fiber optics and systems)
  - Guensik Lim (LHPG)
  - Juddha Thapa (LHPG, materials)
  - Jeff Wuenschell (DTS, testing)
- Idaho National Lab (reactor expertise, system implementation and testing)
  - Pattrick Calderoni (in-pile instrumentation director, co-PI)
  - Joshua Daw (nuclear instrumentation)
  - Ruchi Gakhar (nuclear materials)
- MIT (material compatibility, efficacy simulations)
  - David Carpenter (Irradiation Engineering Director)
  - Koroush Shirvan (reactor design and simulation, co-PI)
  - Tony Zheng, Yeongshin Jeong



Idaho National Laboratory





# GOAL: MSR Thermal Response Analysis for sensor specification and placement

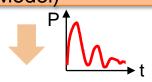
 Providing information on the transient responses under possible transients of molten salt reactor for monitoring system with fiber optic sensor

# **Neutronics** (SERPENT)



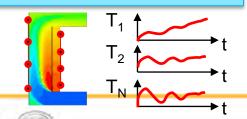
- ✓ Generation of neutronics parameters
- ✓ 0-D/3-D, Steady state
  - > Neutronics: static fuel at constant temperature

# Coupled neutronics/T-H (Simplified MSR Kinetics Model)



- ✓ Transient response of state variables (power, salt and structure temperature, reactivity, etc.)
- ✓ Lumped model, transient
  - ➤ Neutronics: effects of precursor transport due to flow T/H: energy balance only, constant flow rate

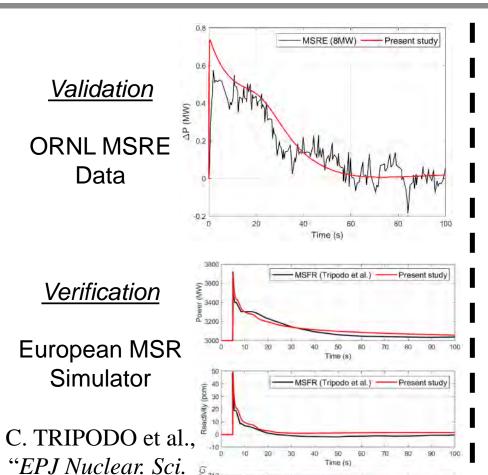
# Thermal-Hydraulics (Star-CCM+)



- ✓ Thermal response of MSR in time and space
- ✓ 2-D (or 3-D), transient
  - Neutronics: power changes imported from simulator
  - ➤ T/H: momentum/energy balance incl. convective and radiative heat transfer of salt



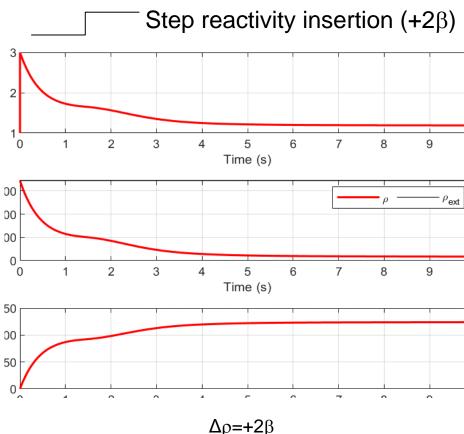
#### Simple System Transient Model Completed



704

P 702

MCFR Application (Based on Available Literature on TerraPower MCFR Design)



Transient response of power, reactivity, and fuel salt temperature changes (1)



*Technol.*, **5**, 13

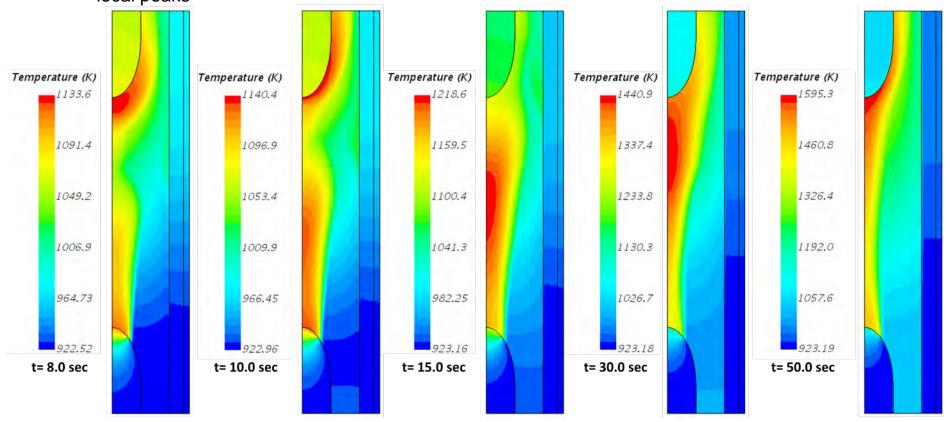
(2019).



#### **3D Accident Analysis Completed**

Demonstrates the Techno-Economic Value of Continuous Temperature Monitoring

 Temperature distribution in the salt, reflector and vessel walls experience meaningful gradients and local peaks

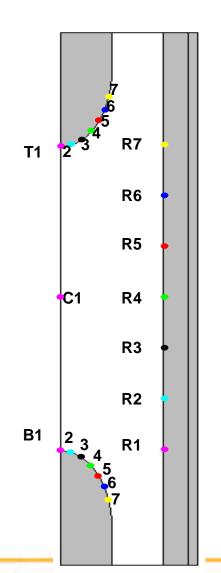


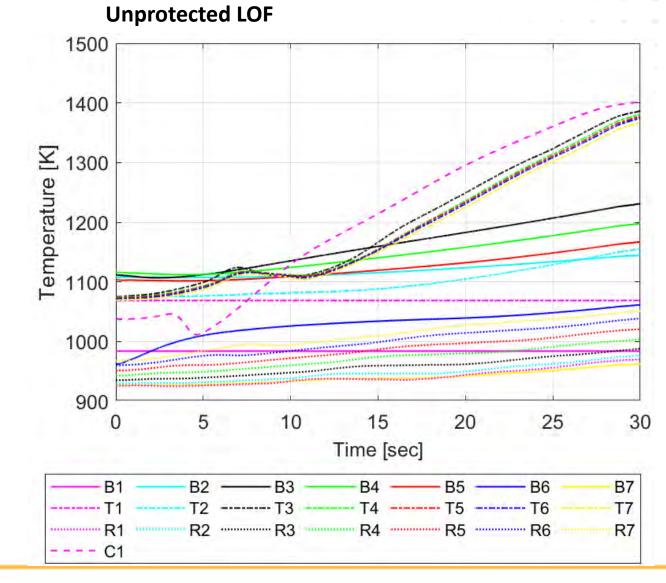
Unprotected LOF (Decrease of fuel salt flow rate to 80 % exponentially with time constant of 5 sec)





#### **3D Accident Analysis**





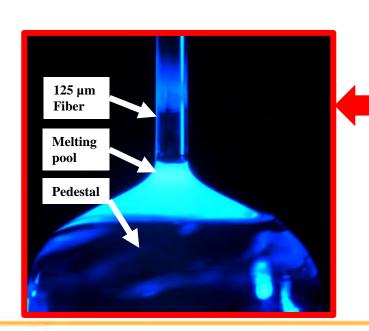


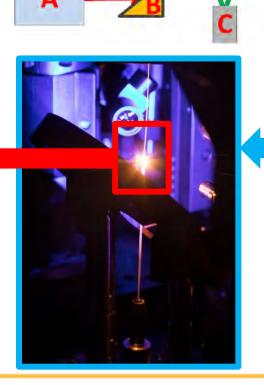


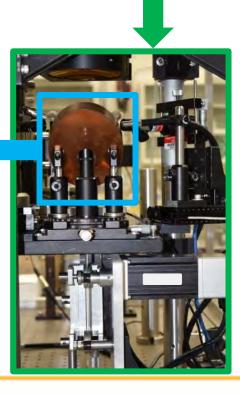
#### Crystal fiber distributed sensing - LHPG

#### **How LHPG works:**

- CO2 laser melts oxide feedstock
- Seed crystal lowered into melt
- Controlled motion of seed and feedstock upward
- Fiber is grown from the melt



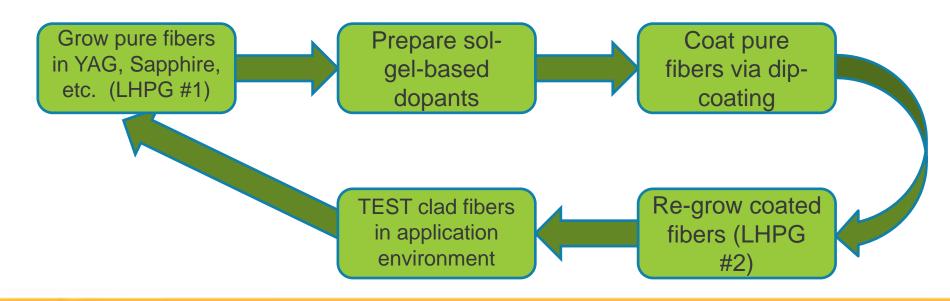






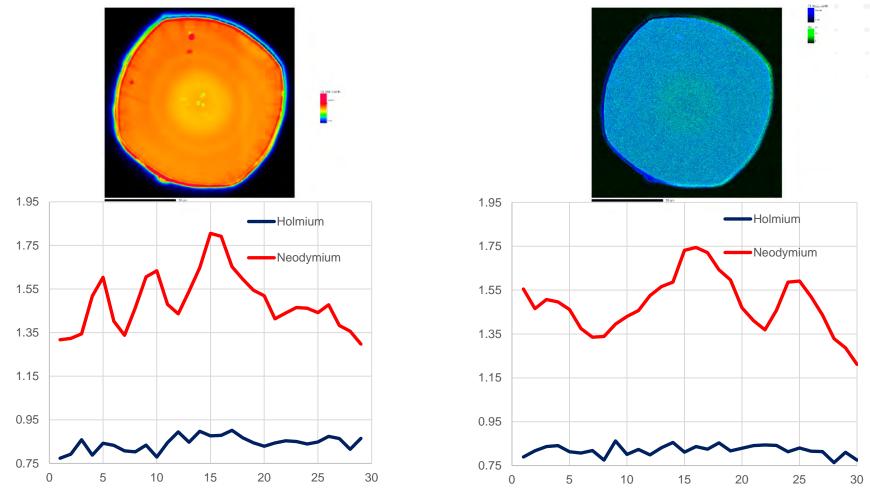
#### Crystal fiber distributed sensing

- Grow cladded fibers with 2-stage LHPG
  - Sapphire or YAG
  - Sol-gel (or other) dopant additions
- Evaluate materials compatibility in fluoride and chloride salts (bench tests)
- Evaluate radiation durability (gamma source, research reactor)





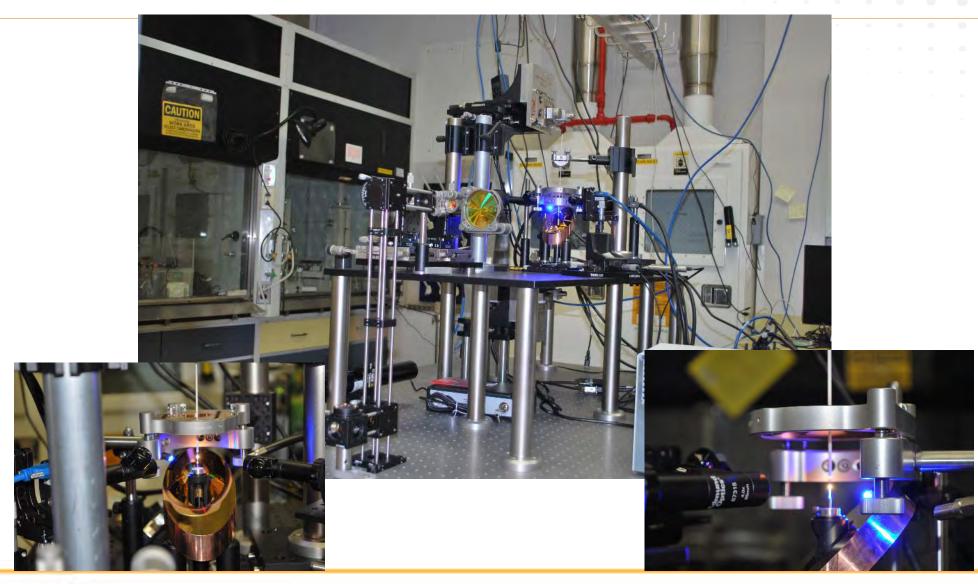
#### Crystal fiber distributed sensing - Claddings



Automatic Dopant Segregation through LHPG: Top left: Visible light guiding in GRIN YAG fiber, Top right: EMPA map of Nd concentration in a GRIN YAG fiber, Bottom plots: Co-doped Nd and Ho: YAG fiber dopant concentrations in X (left) and Y (right)



## **Regrowth LHPG System**



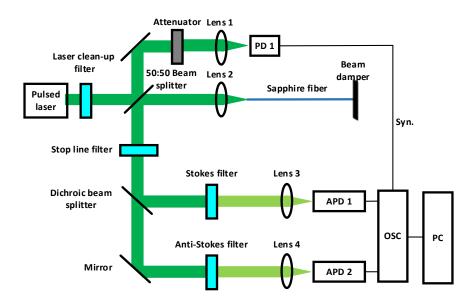


#### Distributed Sensing – Raman OTDR

-25

- OTDR with Single Crystal fiber
- Useful for high-rad/ high-T
- ~5cm, ~1C resolution

Raman OTDR, Liu et al Opt. Lett., 2016



Sapphire fiber attenuation at Virginia 532nm, measured by a Raman OTDR system

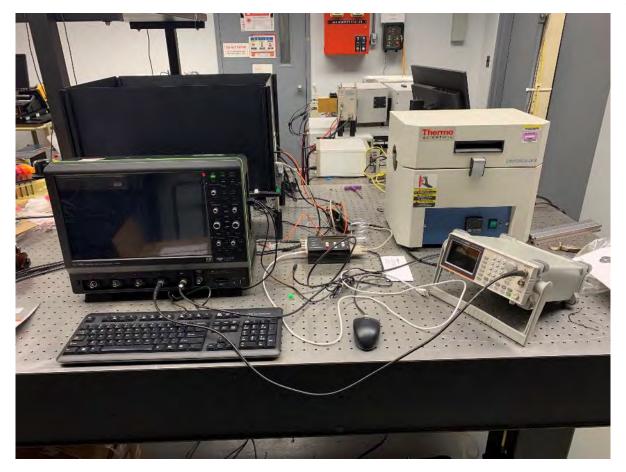
Other of the system o

Distance (m)

2.5



#### Raman DTS design and operations

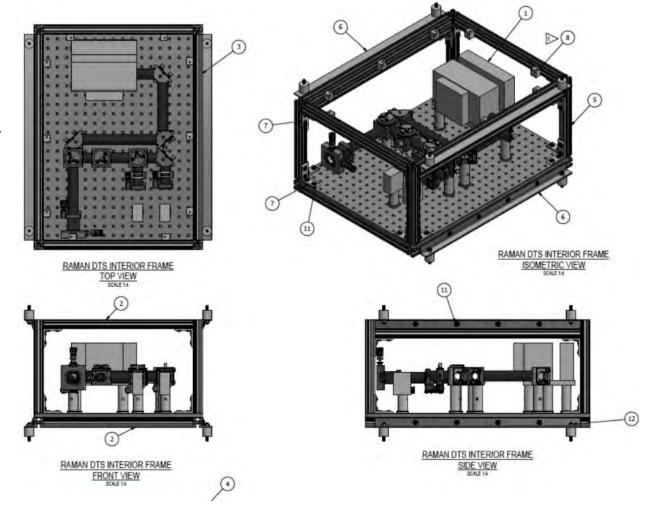


Current view of Raman DTS in the lab with external test systems



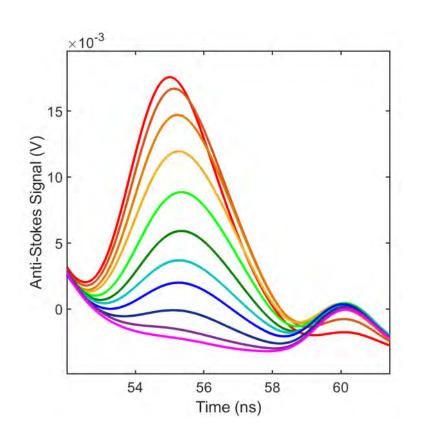
#### DTS design for portability / finished product prototype

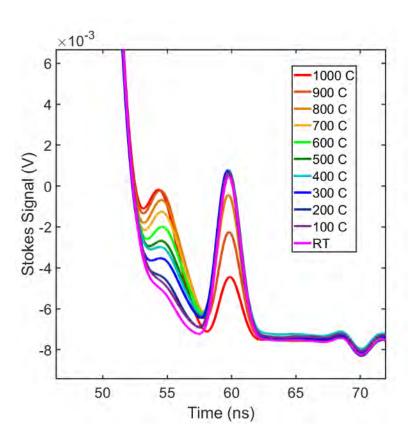
- Flight case design
- Laser safety electrical interlocks
- Software for lead-in fiber
- Field tests in July at INL
- (even more important) field test in October at MITRR





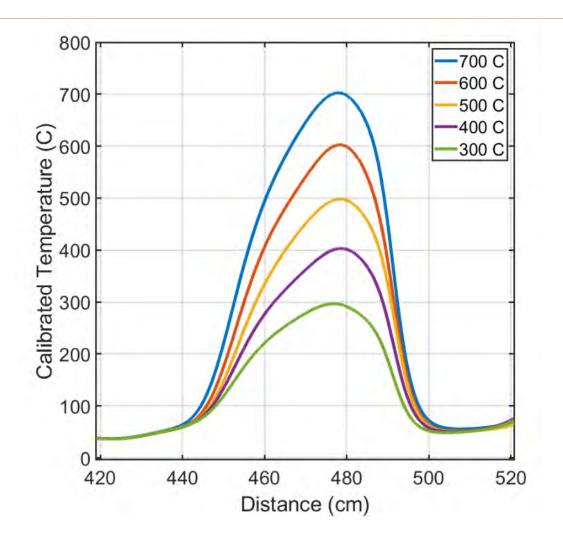
#### Furnace calibration with 50' standoff







#### New data processing for long-standoff



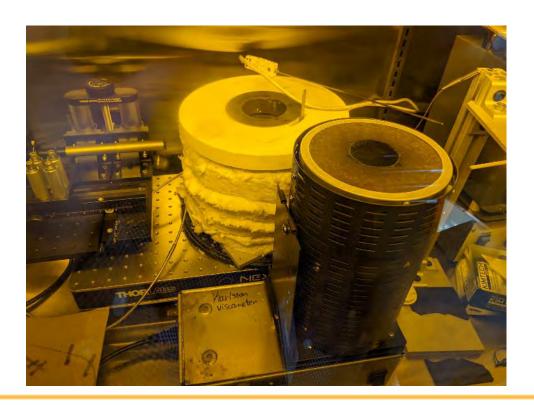
- With 50-ft. lead fiber
- Apply small delay in Stokes vs. Anti-Stokes based on reflection peak delay (one light-ns is 300mm)
- Calibrating only on range where temperature-based losses are minimal (300-700 C).
- Lower loss fiber leads to higher Temp capability
- Smaller fiber leads to higher temp capability



#### **INL Molten salt field tests**

- ► Testing up to ~750C
- Molten Chlorides
- Assortment of crystal fiber materials and protection layers







#### MITR test planning

- Encapsulation tubing (pressure boundary) for fiber
- "dummy" fuel element insertion
- ► DTS standoff at ~60 feet
- October 11<sup>th</sup> installation

Dry irradiation vehicle Dummy fuel element Gas/instrumentation tubes



#### **Conclusions**

- Distributed sensing is coming to numerous industries
- Single-crystal Optical fiber technology can extend into nuclear harsh-environments
- Raman DTS is a good distributed platform for SC-fiber
- Temperature mapping needed for LFMSR transient response
- Amazing new levels of visibility and automation are here!





# Digital Twins to Production Reactors

#### **The Simulation Continuum**

Bob Urberger, Roger Chin

09/15/2021



# Who Are We? | Radiant

- Radiant was founded by former SpaceX engineers in 2019 to make nuclear power portable.
- Currently designing and building Kaleidos, a 1 MW electrical reactor that fits in an ISO shipping container.
- We believe our experience from the aerospace and software industries will allow us to bring a safe, reliable product to market quickly.



# Digital Twins as a Tool

Digital Twin

#### Regulator

Parameters and Documentation Source

Verification and Validation

High Fidelity
Modeling & Simulation

#### Developer

Control Systems/Realtime Modeling & Simulation

Anomaly Response Modeling

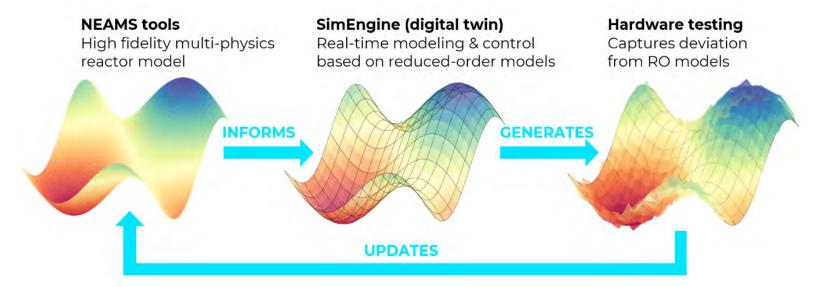
Performance Assessment

Real-Time
Modeling & Simulation

Machine Learning
Applications

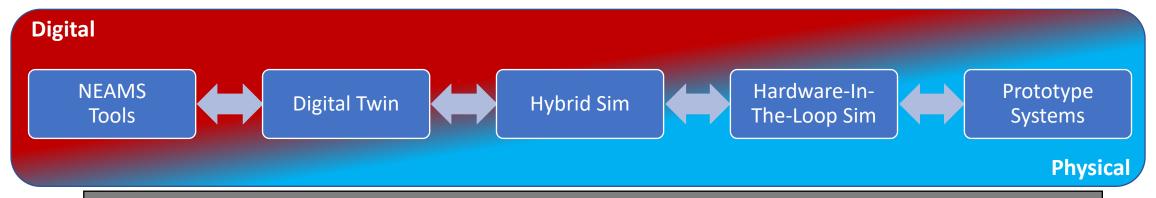
# Why Do We Simulate?

- Inform Design Aspects
- Test Operational Procedures
- Iterate Quickly and Inexpensively
  - Real systems are always correct and used for validation
  - Simulation purpose must be explicitly defined and designed for

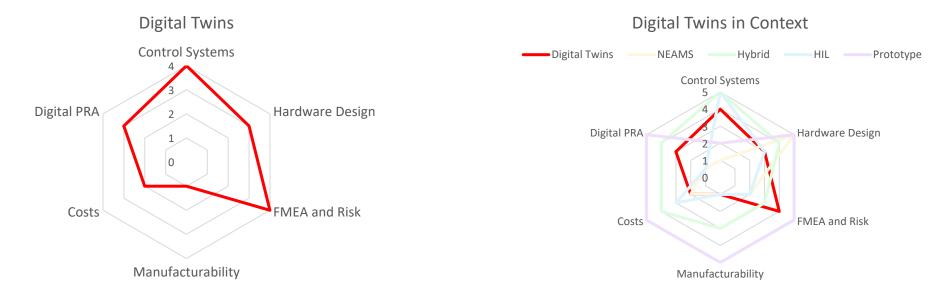


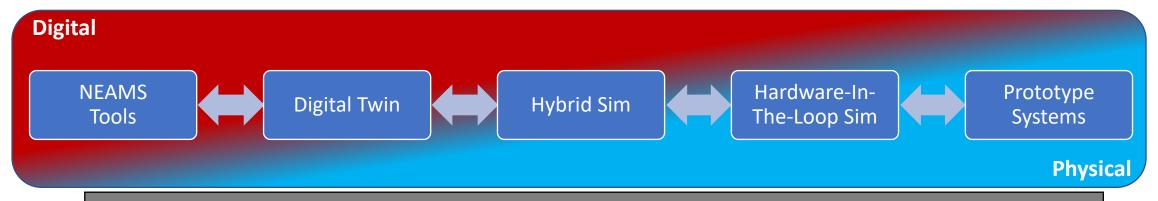
#### **The Radiant Simulation Continuum**

- Our digital twin is low fidelity. How do we bridge the gaps?
- Exploit the strengths of other simulation technologies.
- The scope of each simulation must be carefully defined.
  - What does this sim model?
  - How accurate is it?
  - What can it NOT model?



# **Digital Twins Strengths**





Digital Twins are complementary to existing technologies for a development cycle.

#### **The Radiant Simulation Continuum**

- SimEngine is Radiant's in-house simulation tool.
  - It can run in real time.
  - It can run with any mix of real and simulated hardware
  - Each model can be anchored to either a higher fidelity digital model or a higher fidelity physical model
    - Real hardware is used as the source of truth when possible.
    - NEAMS tool solutions used as source of truth elsewhere.



## **Summary**

- Reality is the ultimate truth. Digital twins have limitations.
- Simulations must have a defined scope.
  - What can/can't they predict? How accurate are they?
- Simulations must be anchored.
  - How do we know the simulation is correct? We need to know the source of truth and the simulation's relationship to it.
- A single high-fidelity digital sim isn't practical for Radiant's use. Instead, we cover a larger simulation space with multiple narrowly focused simulations.



# Thank you for your attention











#### **Steps Toward Regulatory Realization of Digital Twins**

**Closing Panel Session – September 16, 2021** 

**Moderator: Eric Benner, NRC** 

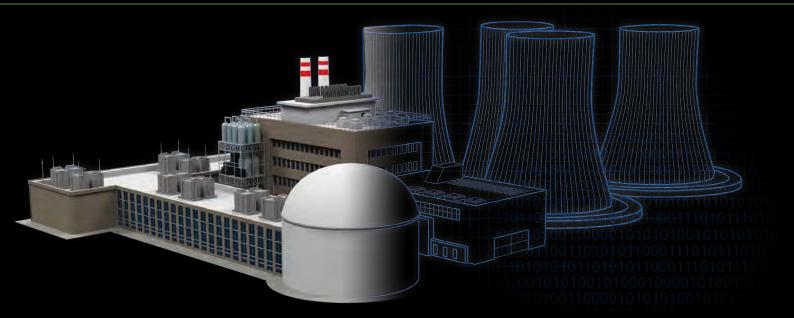
Jeremy Bowen, NRC
Tom Braudt & Steve Vaughn, X-energy
Paul Keutelian, Radiant
James Slider, NEI
Anthonie Cilliers, Kairos Power
Brian Golchert, Westinghouse







# Digital Twins Regulatory Viability Jeremy Bowen



The 2021 Workshop on Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization September 14 – 16, 2021



## Digital Twins Project Plan



TECHNICAL PREPAREDNESS



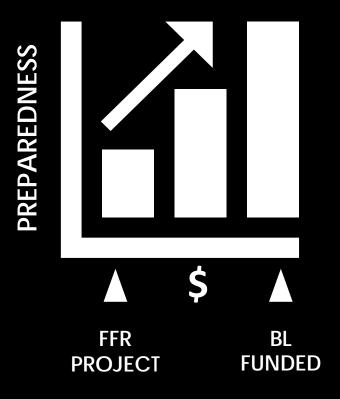
REGULATORY READINESS



ASSESSMENT OF STANDARDS



COMMUNICATION & KNOWLEDGE MANAGEMENT





## Takeaways Thus Far



The State of Technology of Application of Digital Twins



Public Workshop #1



Assessment of the state of DT technology in nuclear reactor applications



Widespread interest with >400 participants across the globe



Proven benefits



Technique is increasing and developing rapidly



Development of a common understanding



Need for community of practice to collaborate

Access Link: ML21160A074

Access Link: ML21083A132



## Active & Future Tasks



Technical Challenges and Gaps for Digital Twins in Using Data Analytics, ML/AI and Multi-Physics Models



Technical Preparedness – Safeguards and Security in Digital Twins



Regulatory Readiness Levels and Gaps Pertaining to Digital Twin Technologies for Nuclear Reactor Applications



Online Monitoring for Enhanced Diagnostics and Prognostics





### Xe-100 Digital Twin Tools





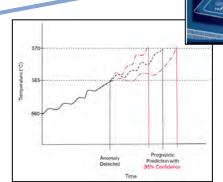
3D Models with AR / VR



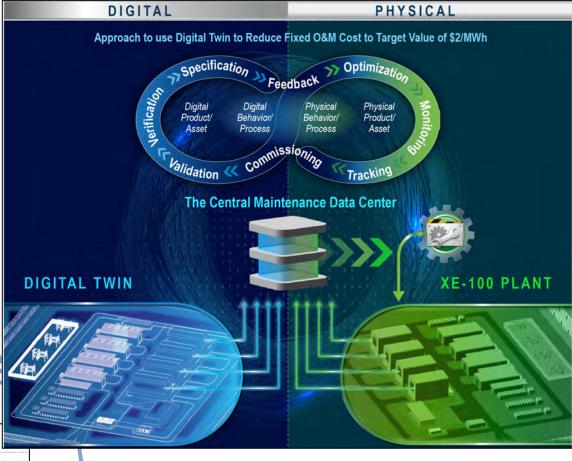
Operator Training Simulator



Plant Historian

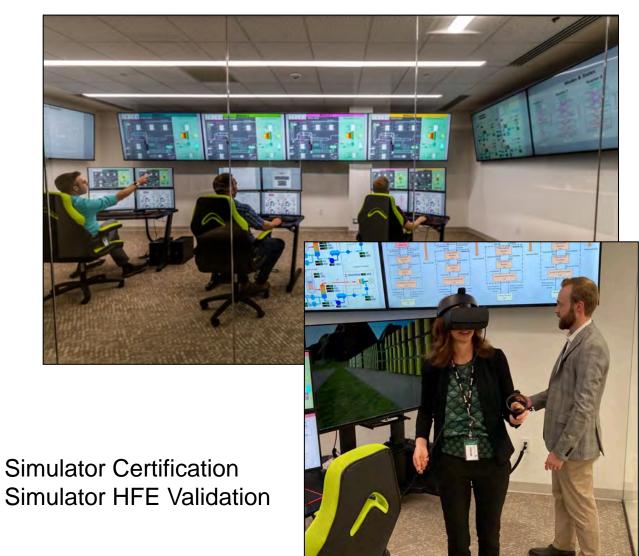


AI / ML Models





### **Hi-Fidelity Simulator and 3D Models**













### Regulatory Intersections with the Xe-100 Digital Twin

## Digital Twin Primary Purpose is Plant Economics – Not Regulated Some Intersectionality with:

- Xe-100: extensive use of digital control and monitoring systems
- Maintenance Rule predictive maintenance
- Simulator fidelity and configuration management
- Human Factors Engineering predictive trending and intelligent alarms reduce operator workload
- NLO/Maintenance/I&C/RP Training multi-role approach
- ALARA dose reduction
- Security analytics and what-if scenarios
- Fire Detection intelligent dispatch and response
- Support load-follow, industrial heat, hydrogen production

## Strong NRC Staff Interest in Understanding Xe-100 Digital Twin Capabilities and Usage

- Control Room staffing methodology
- Operator Training methodology and workload





### **Digital Twin Regulator Pathway – Future Considerations**

### Use of AI / ML for Control Functions

- Ensuring regulatory framework positioned to support
- Leverage lessons-learned from safety-related digital retrofits (cyber)
- NUREG/CR-7273 starting point

Dynamic PRA Usage - initiating event prediction via Al Pre-emptive utilization of "Remaining Useful Life" (RUL) for equipment Sensor Drift Detection

Cyber Security Requirements for Control Functions – DoD Lessons Learned

**Encourage Research at NR/ INL/ORNL to Support Control Functions** 

Making 'digital twins' to monitor and control tomorrow's reactor designs - INL





# Using Digital Twins to Support Regulations

Paul Keutelian

September 16, 2021





## Vision | Radiant



Make Nuclear Portable

Maximize speed of iteration to get to a buildable product

Provide streamlined regulatory analysis



**Regulatory Intents** 

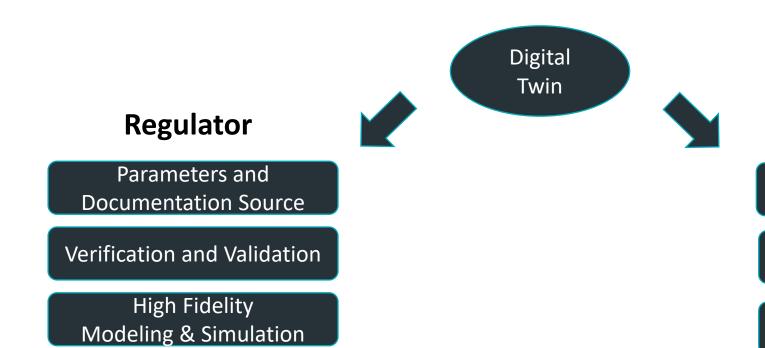
## **Novel Challenges**

**Protect Personnel Protect Environment Digital Twins Protect Hardware Prove the Above** 

Digital Twins are a powerful potential to answer these core questions, the challenge comes from understanding where is it appropriate to use this tool, and how do we work with regulators to build confidence?



## **Digital Twins as a Tool**



#### Developer

Control Systems/Realtime Modeling & Simulation

Anomaly Response Modeling

Performance Assessment

Real-Time
Modeling & Simulation

Machine Learning
Applications

Digital Twins are a common tool between regulators and developers to ensure common sources of information for aspects relevant to them.



## Contributions

- Internal analysis documentation standardization
- Visualizations for eased introduction to analysis
- Automation of explicit analysis flows
- Common interface with Regulators



## **JAMES SLIDER**

TECHNICAL ADVISOR



### **ANTHONIE CILLIERS**

SENIOR MANAGER
INSTRUMENTATION, CONTROLS AND ELECTRICAL



### **BRIAN GOLCHERT**

PRINCIPAL ENGINEER











## **Enabling Technologies for Digital Twin Applications for Advanced Reactors and Plant Modernization**

September 14-16, 2021

Thank you for your participation in this workshop. The proceedings for the workshop will be publicly available in the next few months. Please provide any comments or feedback to one of these workshop sponsors.

Jesse Carlson, NRC - jesse.carlson@nrc.gov

Vaibhav Yadav, INL - vaibhav.yadav@inl.gov

Jenifer Shafer, ARPA-E - jenifer.shafer@hq.doe.gov

Hasan Charkas, EPRI - <a href="https://hcharkas@epri.com">hcharkas@epri.com</a>

Prashant Jain, ORNL - jainpk@ornl.gov

