



Defense Nuclear Nonproliferation Research & Development
Nonproliferation Enabling Capabilities
Remote Detection
DAM 2021 Meeting

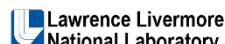
**Persistent DyNAMICS: Remote Sensing Based
on Domain-Informed Analytics**

Thomas J. Kulp and Siddharth Manay

Chris Burt, Jereme Haack, Romie Morales, Norma Pawley, Paul Pope,
Bob Priest, Mark W. Smith, Tom Vestrand

Persistent-DyNAMICS:

Persistent Dynamic Nuclear Activity Monitoring via
Intelligent, Coordinated Sensing



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SAND2021-????



6 Labs: LANL, LLNL, ORNL, PNNL, SNL, STL

Sponsor: Office of Proliferation Detection



Venture Management Team:

Venture Manager: Norma Pawley, LANL

Science Integration Lead: Tom Kulp, SNL

Experimental Integration Lead: Bob Priest, LLNL

Mission Assurance Lead: Paul Pope, LANL



Data Management Team:

LANL – Paul Pope, Phillip Romero

SNL – Rose Borden, Craig Ulmer

ORNL – Janice Greenwod



Testbed Team: Jeff Johnson, ORNL (Lead)

ORNL – Chris Young, Russ Henderson, Michael Willis, Riley Hunley, Andrew Duncan



Interpretation:

Siddharth Manay, LLNL (Lead)

LLNL - Brenda Ng, Goran Konjevod, Lance Bentley-Tammaro, Phan Nguyen

PNNL – Romarie Morales Rosado, Jereme Haack, Lee Burke, James Bradford, Paul Keller, Dave Engel, Jackson Chin



Prediction: Tom Kulp, SNL (Lead)

SNL – Sam Eaton, Randy Brost, Benjamin Cipiti, Philip Honnold, Tom Reichardt, Jason Reinhardt, David Farley



Coordination: Tom Vestrand, LANL (Lead)

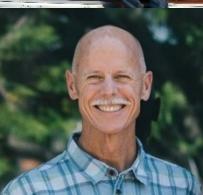
LANL – Przemek Wozniak, Sean Brennan, Troy McVay, David Palmer, Lucas Parker



Collections: Bob Priest, LLNL (Lead)

SNL – Mark Smith, Stephen McConnell, James Ramsey, Jon Slater, John VanderLaan, David Yocky, Holly Eagleston

LLNL – Aaron Wegner, Randy Sanchez, Josh Dunn, Garth Pratt, Steve Shiromizu, Jim Curry



STL – Chris Burt, Kevin Lee, Carmelo Gonzalez, Ljuboslav Boskic, Ian Bortins, John DiBenedetto



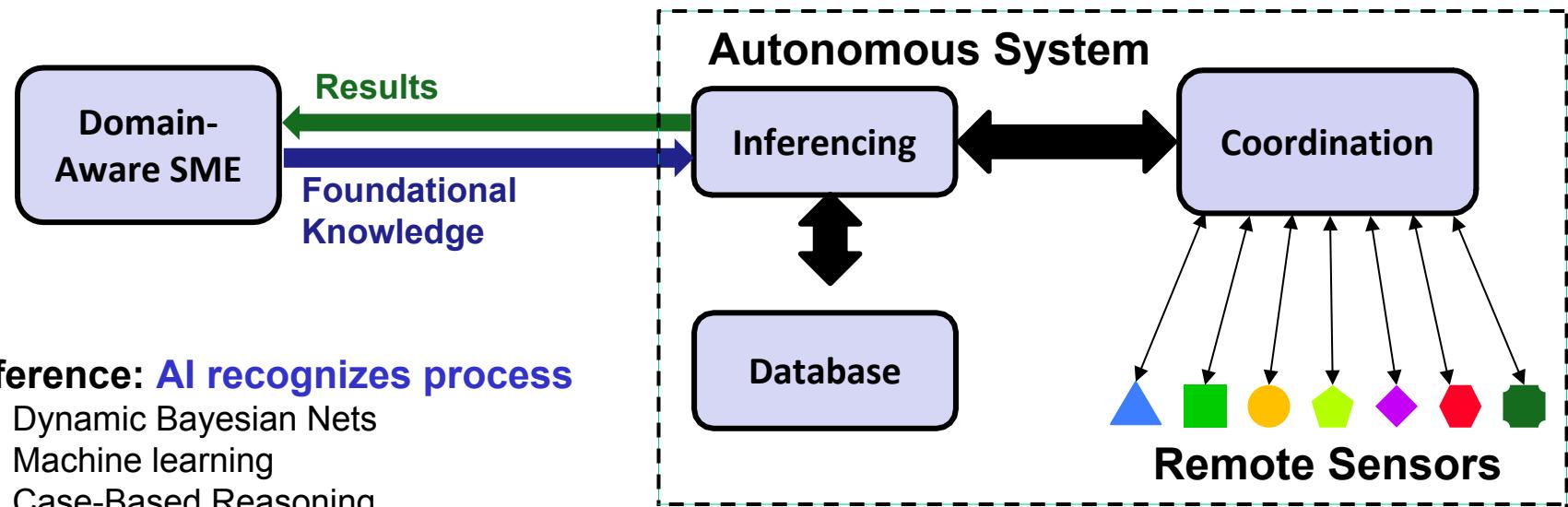
Integration: Michele Decroix, LANL (Lead)

Kriste Henson, PNNL



We are developing a system to characterize processes

Here, we will discuss how this system uses domain-informed analytics



Inference: AI recognizes process

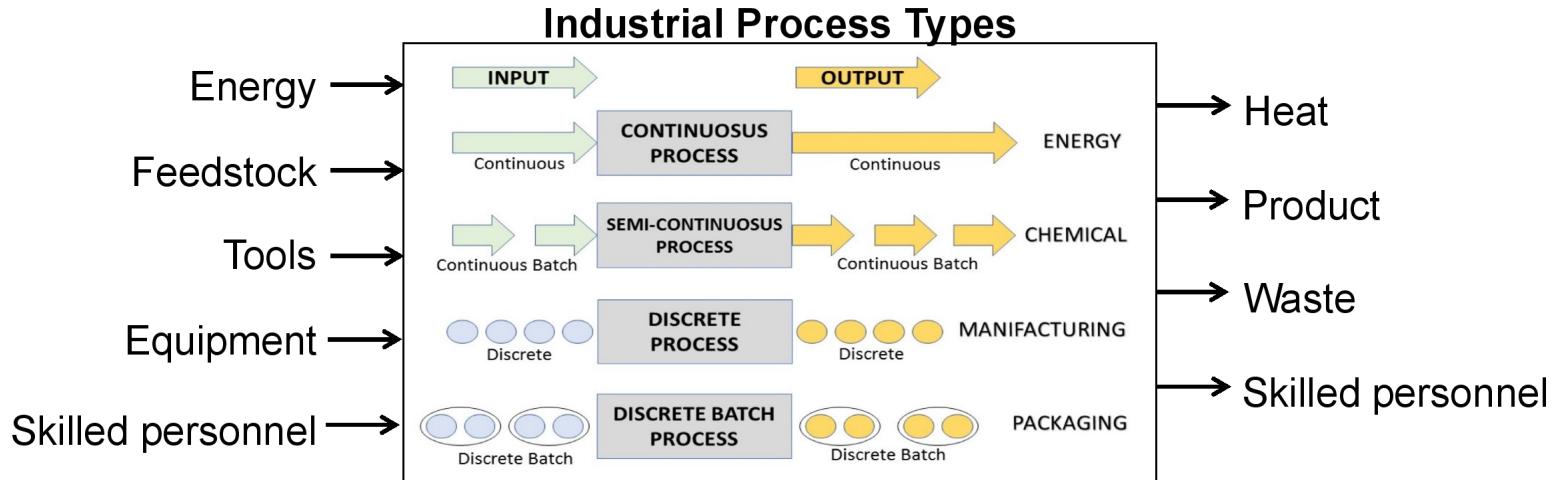
- Dynamic Bayesian Nets
- Machine learning
- Case-Based Reasoning

Foundational Knowledge: Domain Knowledge codified for AI

Coordination: Measurements scheduled using domain “script”



Industrial processes follow a domain-informed “script”



- Activities are both technical and “pattern of life”
- Each is driven/influenced by science and “engineering best practice”

With domain knowledge:

↓ Need for data

↑ Explainability

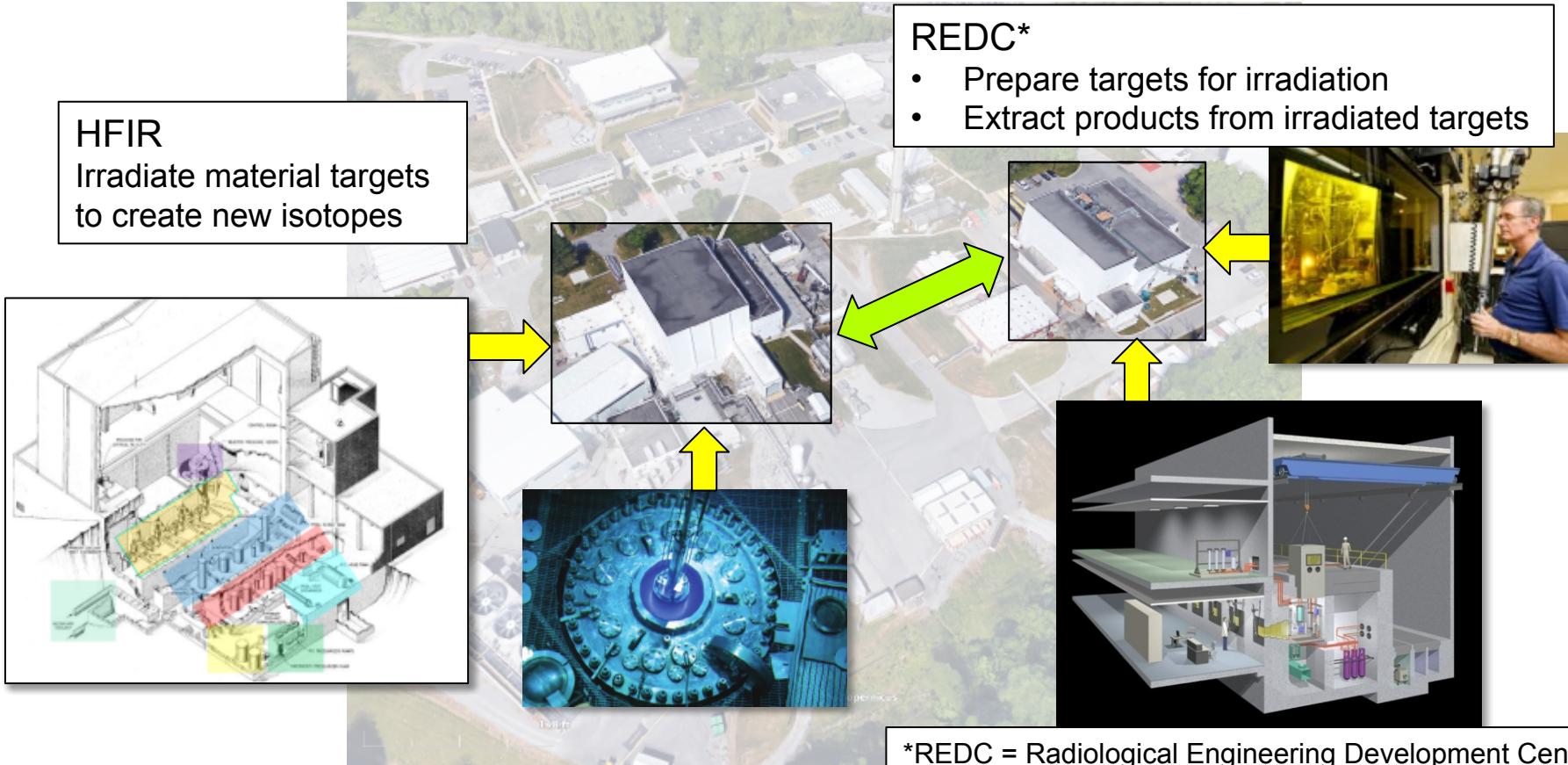
↑ Transferability

HFIR is the current Persistent DyNAMICS Testbed



ORNL High Flux Isotope Reactor (HFIR) Complex

There are two connected facilities at the Complex



Process observables being measured

Key Hardware

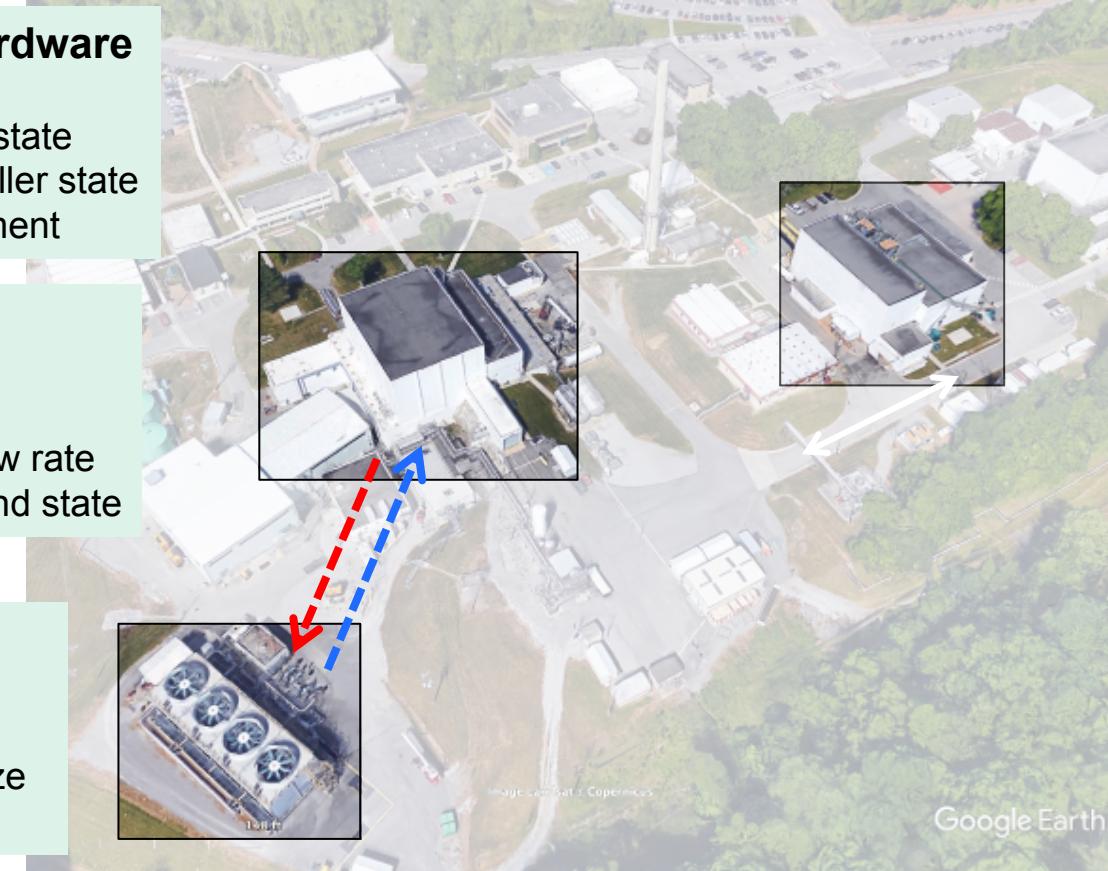
- State
- Motor state
- Controller state
- Movement

Water flow

- Pipe state
- Pump state
- Water state and flow rate
- Valve movement and state

Heat Disposal

- Tower state
- Fan state
- Plume presence / size
- Temperature



Process observables being measured

Key Hardware

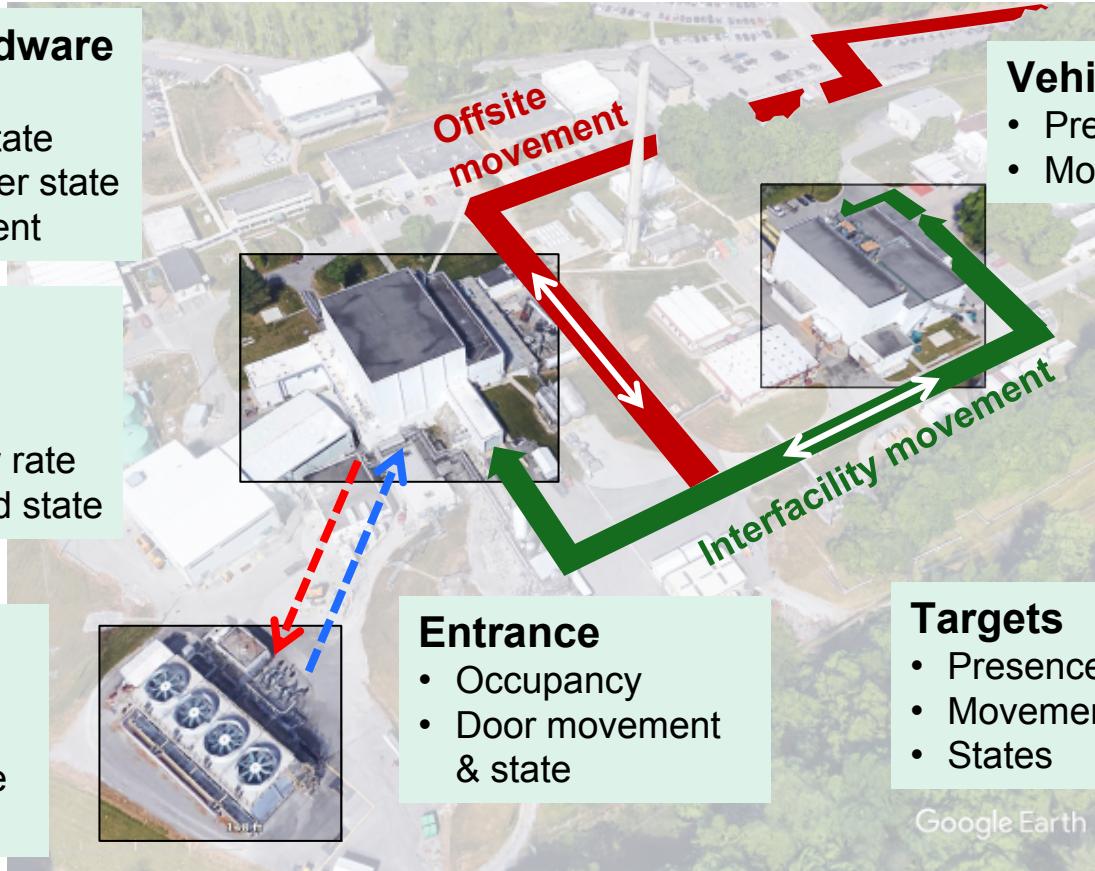
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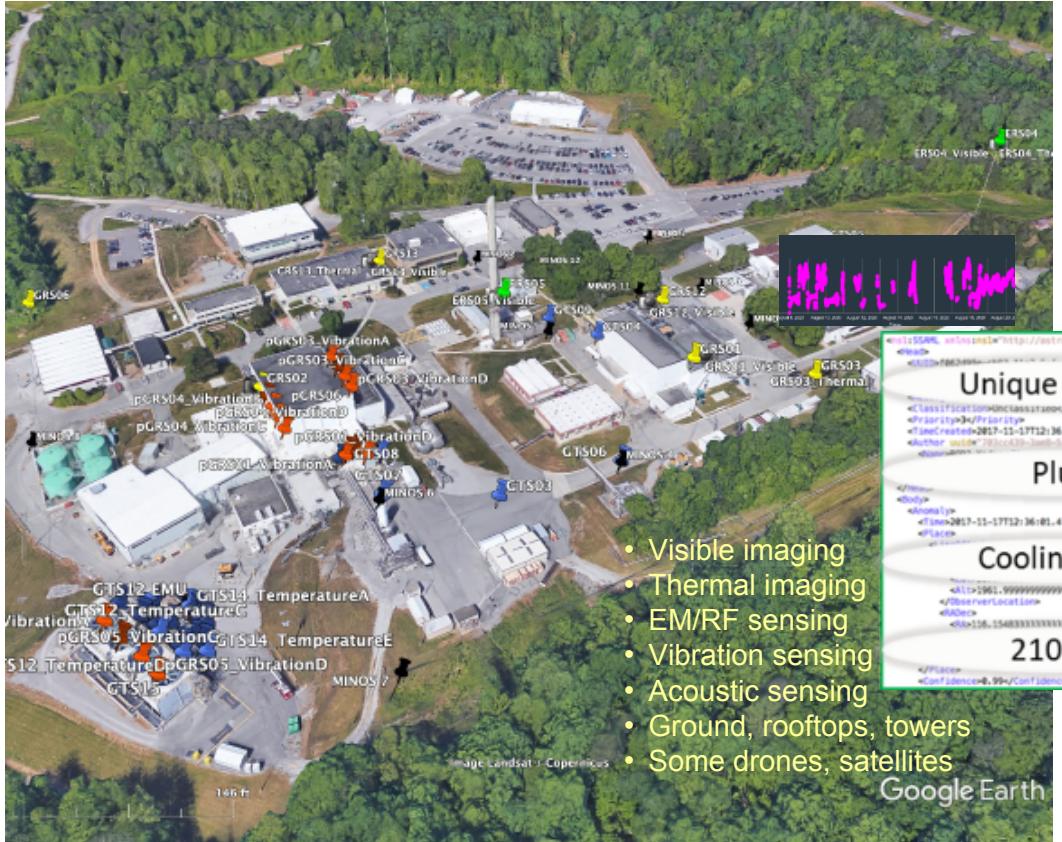
Vehicles & Containers

- Presence (at locations)
- Movement

Targets

- Presence (at locations)
- Movement
- States

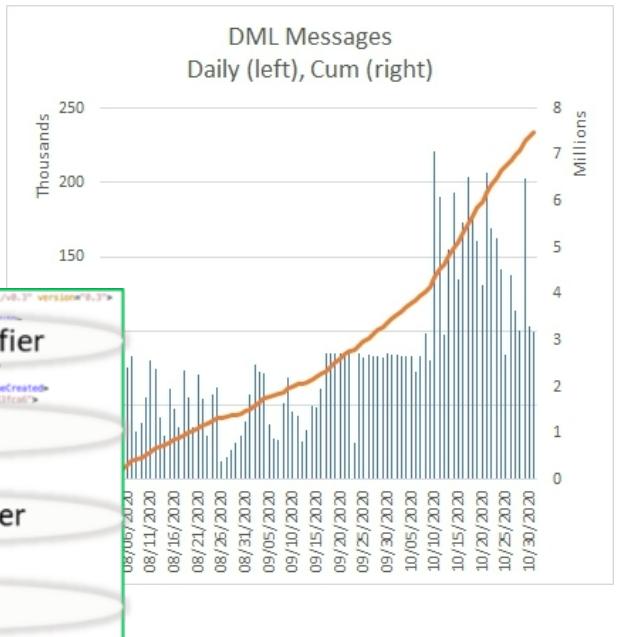
PD sensors have been collecting since 15 Jul 2020



- Visible imaging
- Thermal imaging
- EM/RF sensing
- Vibration sensing
- Acoustic sensing
- Ground, rooftops, towers
- Some drones, satellites

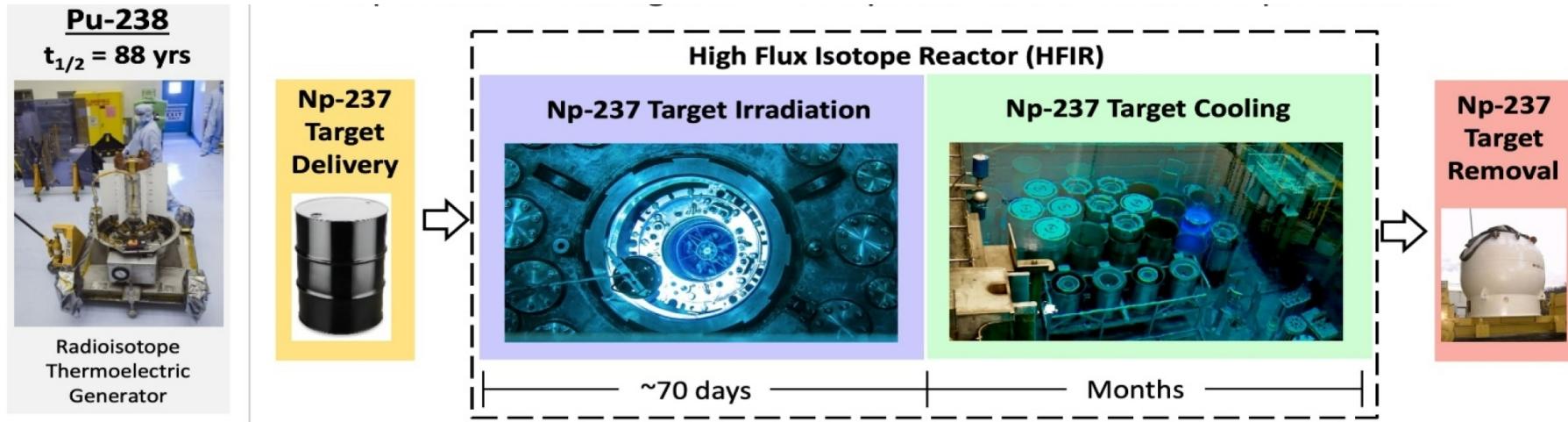
Google Earth

Edge Sensors emit Sensed
Information text messages 24/7



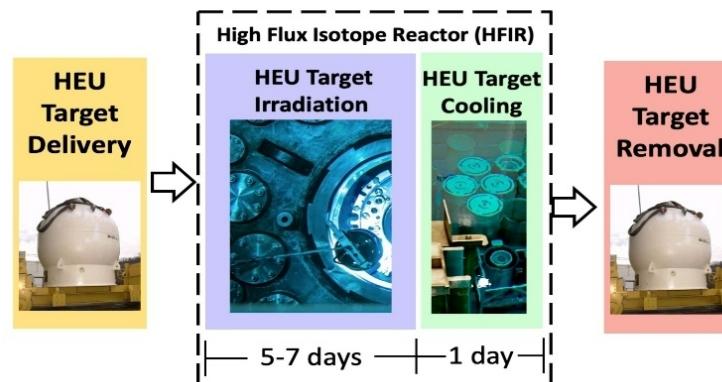
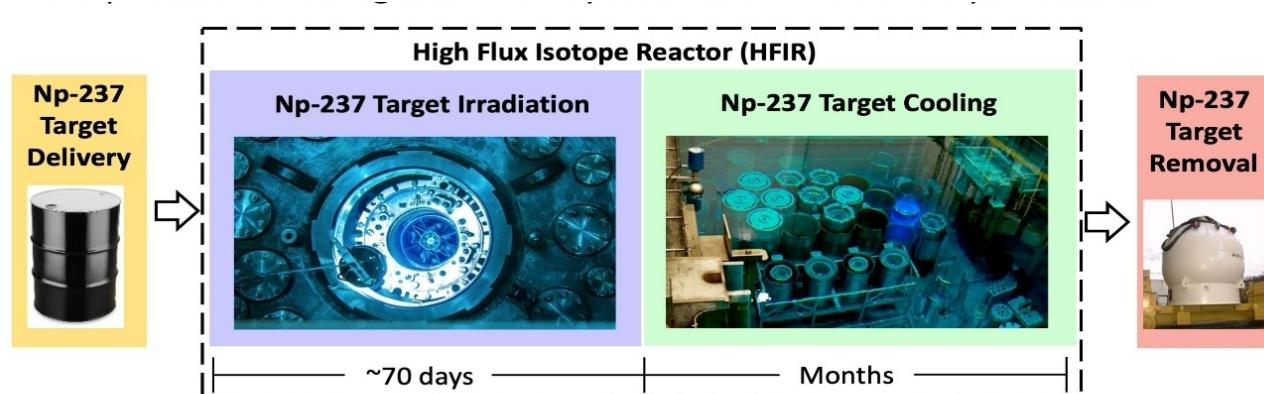
Use Cases provide a context for test and evaluation

- Use Case 1 (UC1) hypothesis: *The target facility is performing activities consistent with the production of a short-lived medical isotope.*



Use Cases provide a context for test and evaluation

Physics constrains tempo of activity sequences and ensures transferability



Transportation conditions and irradiation/cooling cadences are dictated by isotope half-lives → represent significant differences between production processes.

Encoding Domain Knowledge for UC1

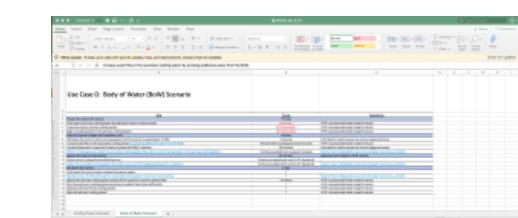
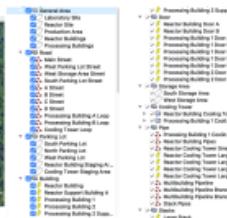
Initial SME elicitation



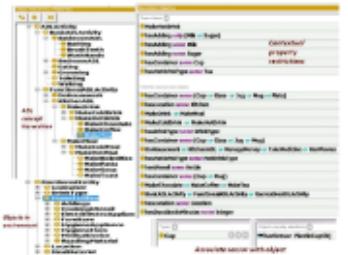
Primer (text)



Geospatial labeling



First-cut sequence “script”



General Domain Ontology



Functional Tools

- **Sequence Model** – Informs Dynamic Bayesian Networks and Case-Based Reasoning
- **High-Level Process Model** – Generative simulator of state vectors for Machine Learning
- **Lexicon** – Defines “language” for agent-based communication

Maximize transferability →

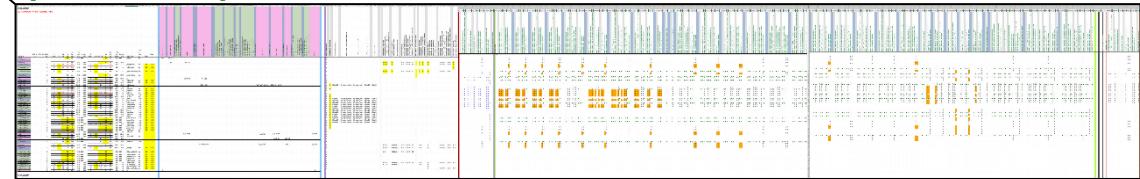
- Focus on science-constrained activities
- Operate “abstractly” (avoid designing for HFIR)

Functional tool examples

Sequence Steps

4_reactorOperation
4.2_reactorOperationPrepare
4.2.1_reactorOperationPrepareLoad
4.2.2_reactorOperationPrepareSealRectorVessel
4.2.3_reactorOperationPreparePressurizePrimary
4.2.4_reactorOperationPreparePreOpChecks
4.3_reactorOperationStartup
4.3.1_reactorOperationStartupWdrawRods
4.3.2_reactorOperationStartupIncrSecondaryFlow
4.3.3_reactorOperationStartupIncrPowerTo85MW
4.3.4_reactorOperationStartupIncrPowerCoolTower
4.4_reactorOperationRun
4.4.1_reactorOperationRunCoolTower
4.5_reactorOperationShutdown
4.5.1_reactorOperationShutdownInsertRodsReducePower

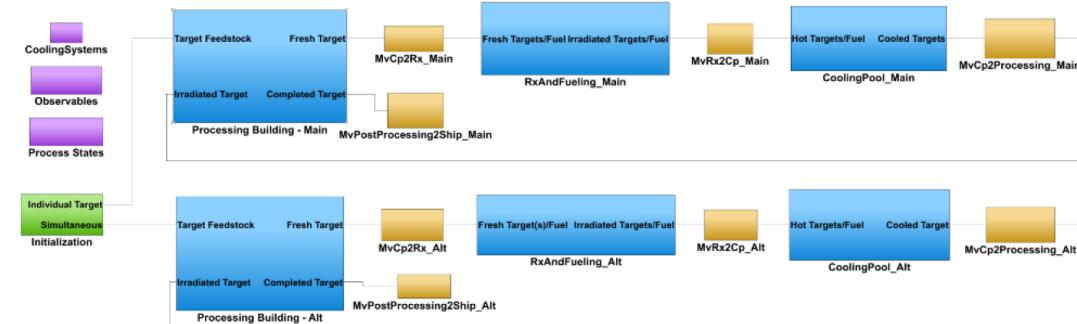
Sequence Model (for DBN)



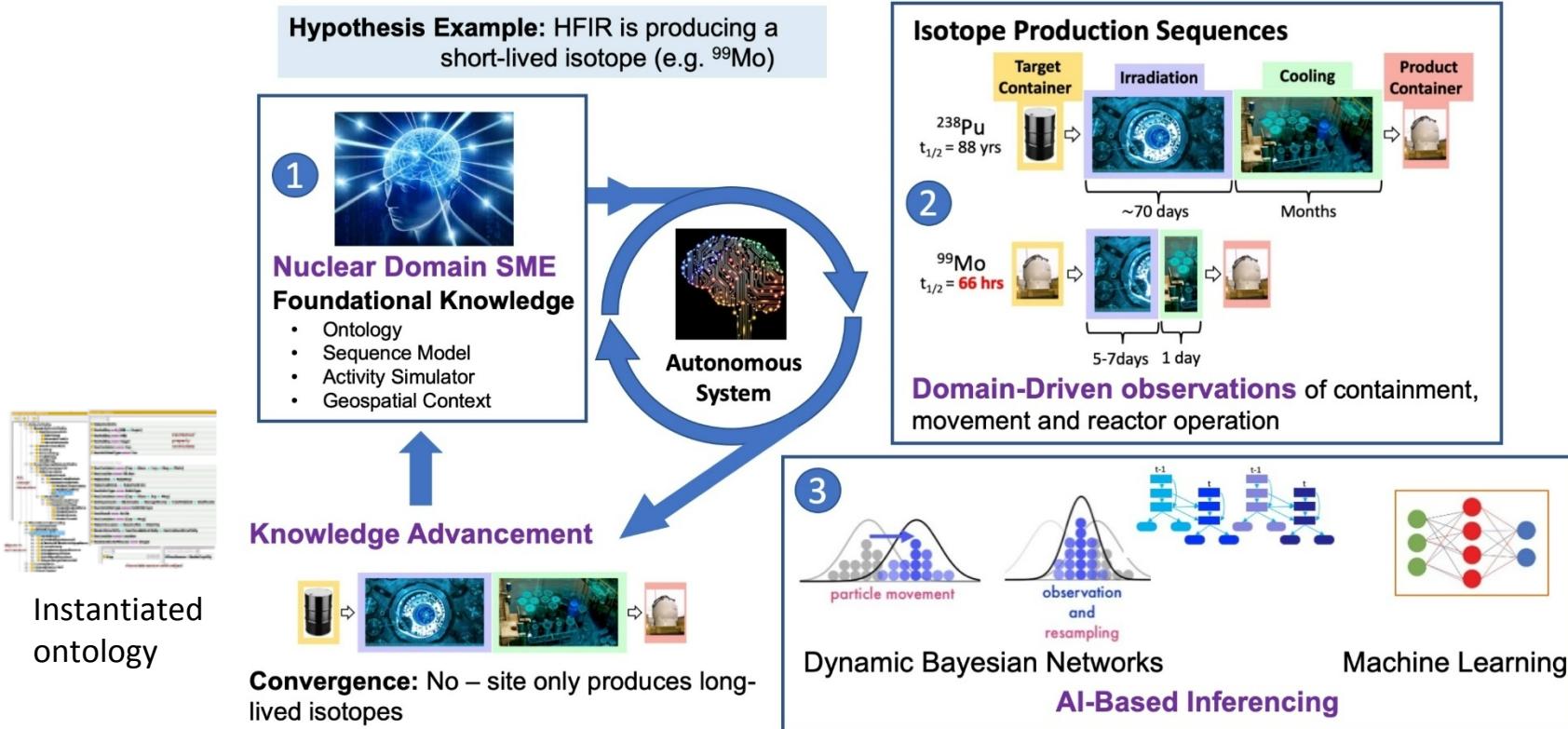
Sensed Information (SI) Messages

State Transition
Matrix

Simulink-Based HLPM Simulator (for ML training)



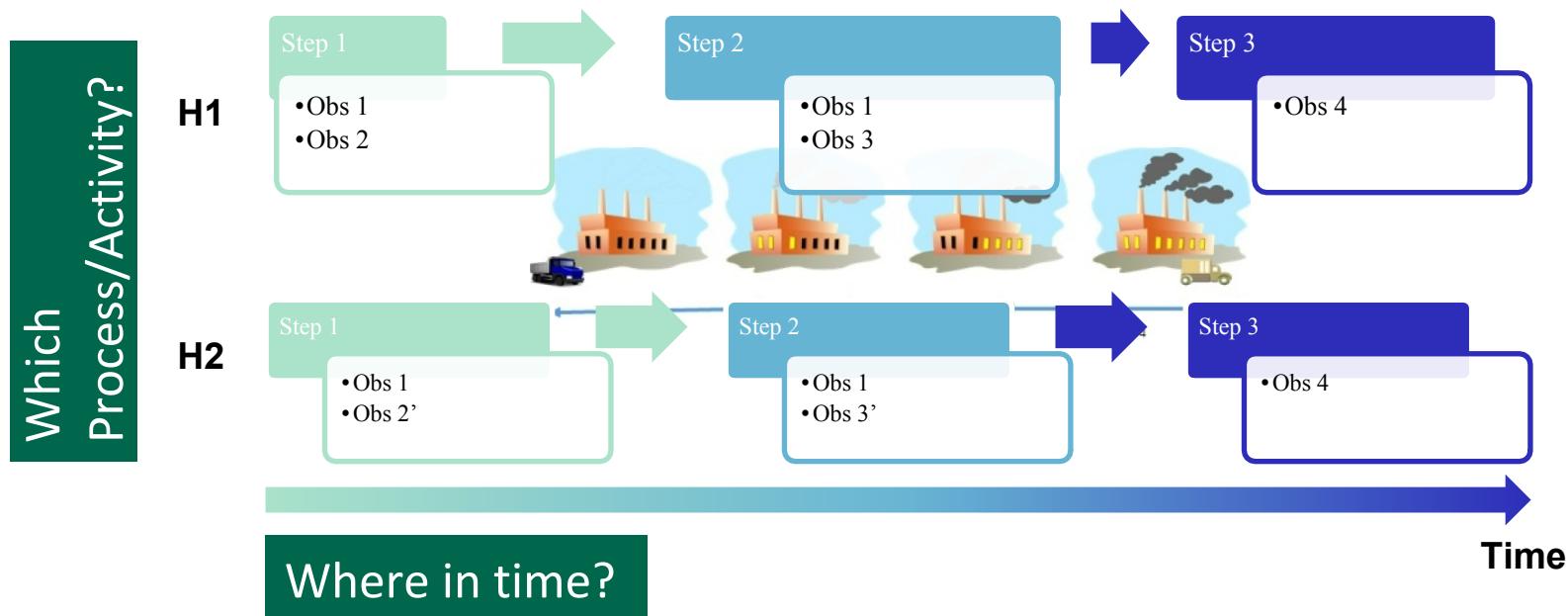
The system is designed for *updating* Knowledge



As Knowledge advances, a Site-Specific Ontology is developed

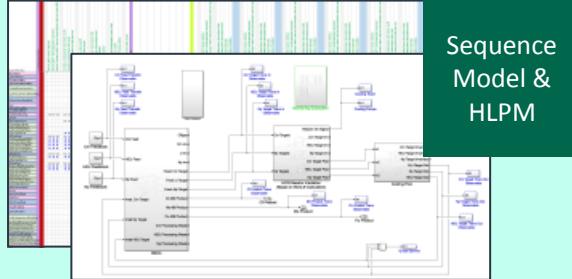
System purpose: characterize industrial activities

- Current focus on two questions about an industrial site:
 - What process/activity are they doing?
 - Where are they in the process?



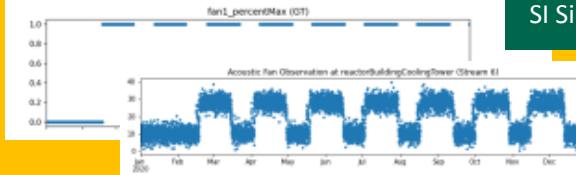
Embedding Foundational Knowledge into Inference Algorithms

Foundational Knowledge



Sensor & exploitation
Design info, performance
characteristics,
deployment...

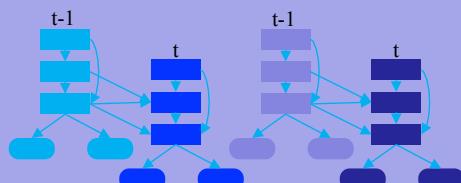
Data-Driven ML: Use simulator



Train
XGBoost



Design DBN



Knowledge-Driven
DBN: Design based
on FK

Domain Aware Inference Methods at work

- Preliminary result:
 - DBN and ML Inference algorithms tested on simplified model (Nov 2019).
 - ML Inference algorithms trained and tested on simulated SI messages. ➔
- Ongoing work and next steps
 - Testing and debugging inference methods on real HFIR data.
 - Knowledge iteration cycle with SMEs.



Conclusion

- Persistent DyNAMICS is developing an autonomous *system* to characterize industrial processes
- The system uses Domain-Informed AI to characterize the “script” of industrial processes that combine technical actions and “patterns of life”
- This serves to:
 - Bridge the gap between sparse data and automated inferences.
 - Maximize transferability
 - Maximize explainability
- Subject Matter Expertise is the key enabler!