



Sandia  
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# A New Approach to Insider Threat Detection & Mitigation for High Consequence Facilities & Critical Infrastructure

*Artificial Neural Networks & Risk Significance*



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Introduction to the Role of Artificial Intelligence in Strengthening the Security of Nuclear  
Facilities | February 6-8, 2024 | Vienna, Austria



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U.S. Department of Energy's National  
Nuclear Security Administration under  
contract DE-NA0003525.

**SAND2023-14192C**

# Roadmap



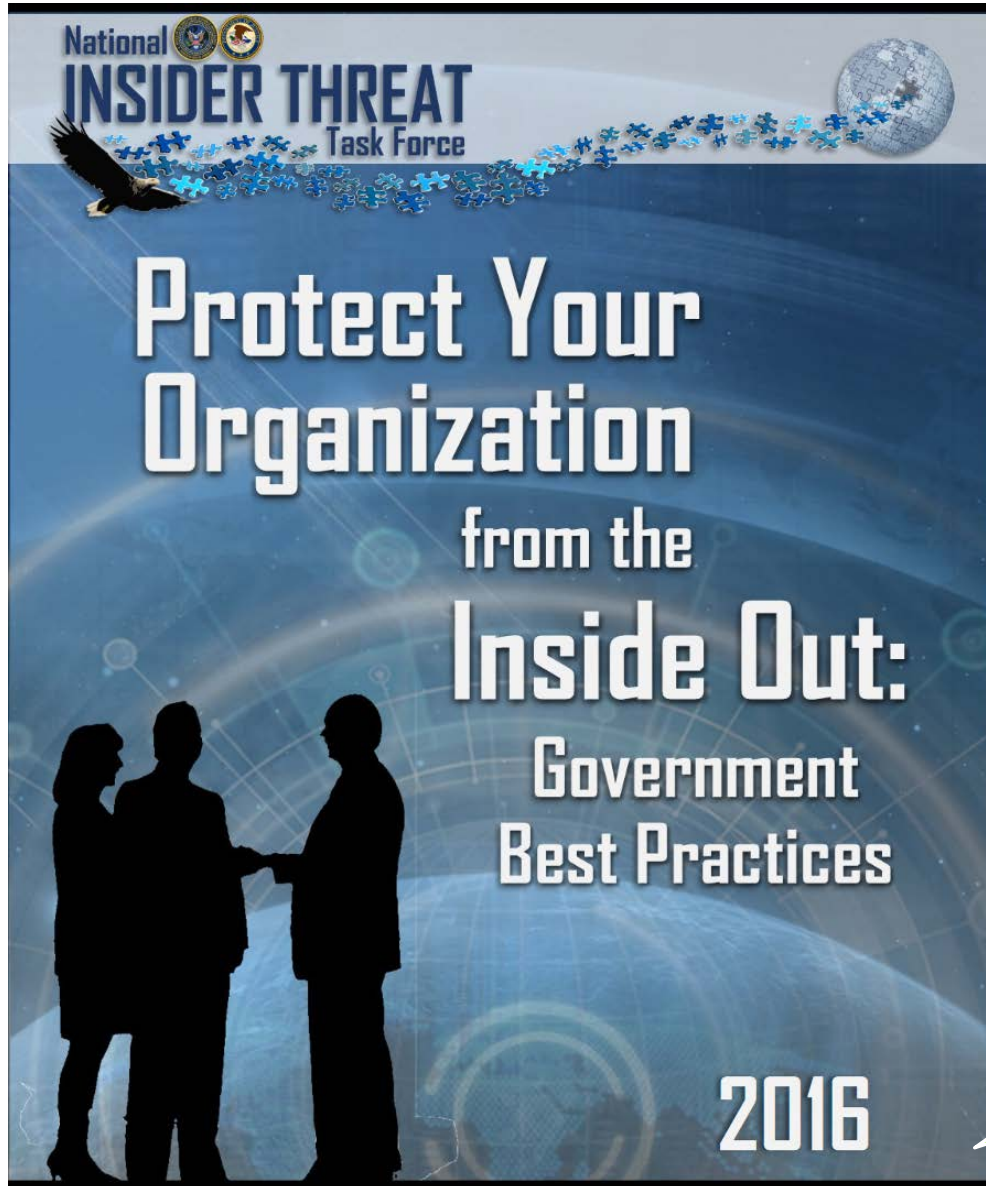
Introduction

Situating a New Approach to ITDM

Methods & Data Collection

Demonstrating a New Approach to ITDM

Conclusions, Insights & Implications



“the **risk** [that] an insider will use their **authorized access**, wittingly or unwittingly, **to do harm** to their organization. This can include theft of proprietary information and technology; damage to company facilities, systems or equipment; actual or threatened harm to employees; or other actions **that would prevent the company** from carrying out its **normal business practices.**”

# Introduction



## Insider threat definitions:

- NRC** → “Once an individual has been granted unescorted access to protected and vital areas ... preventing an adverse event becomes dependent on detecting ... and/or denying ... the **opportunity to commit** the act”
- IAEA** → “an individual with authorized access to [nuclear material,] associated facilities or associated activities or to sensitive information or sensitive information assets, who **could commit, or facilitate the commission** of criminal or intentional unauthorized acts ... [with] an adverse impact on nuclear security”
- DHS/CISA** → “is the **potential** for an insider to use their authorized access or special understanding of an organization to harm that organization”



# Introduction



## Insider threat definitions:

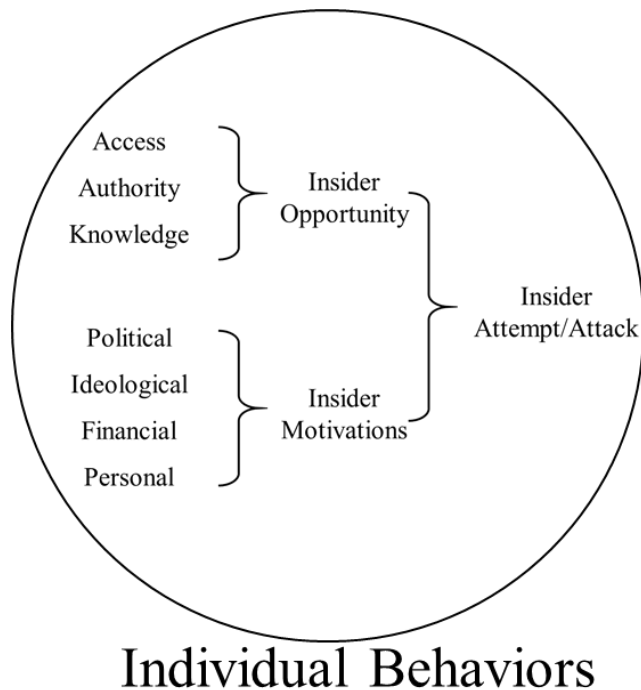
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“opportunity” or “could” or “potential” → risk significance

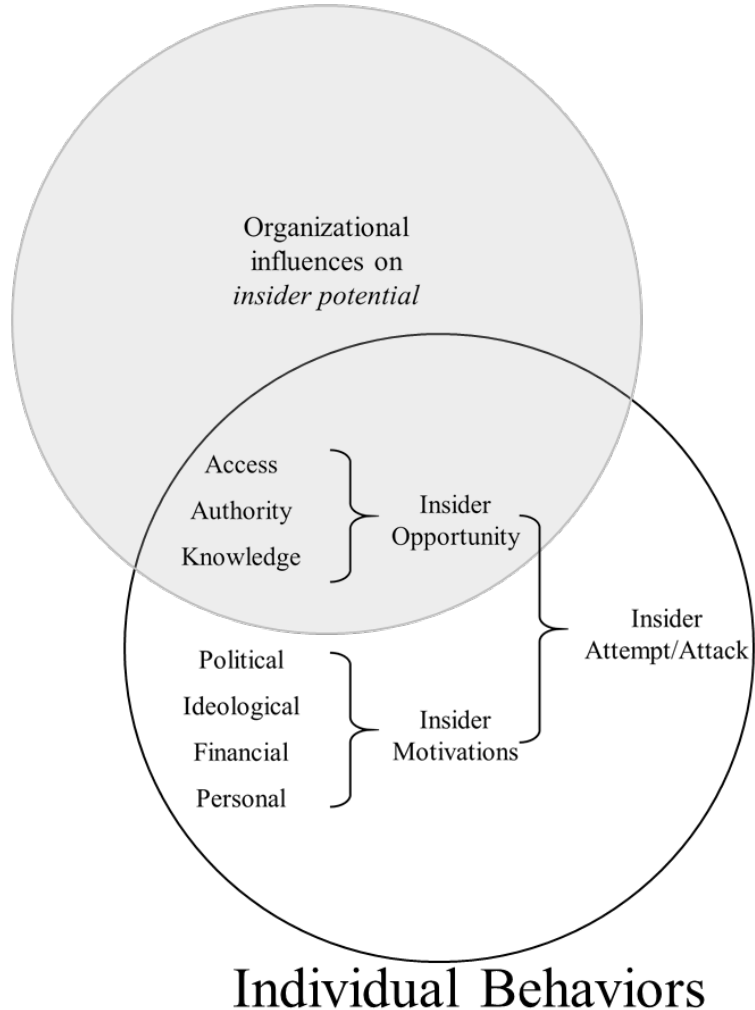


## **Traditional approaches** to Insider Threat Detection & Mitigation (ITDM)

- Focus on individual characteristics
  - Difficult to identify, almost impossible to measure/quantify
- Based on “prevention” and “protection” concepts
  - Best practices, for example
- Struggle to anticipate growing “insider threat potential”
  - Underlying “reactionary” paradigm



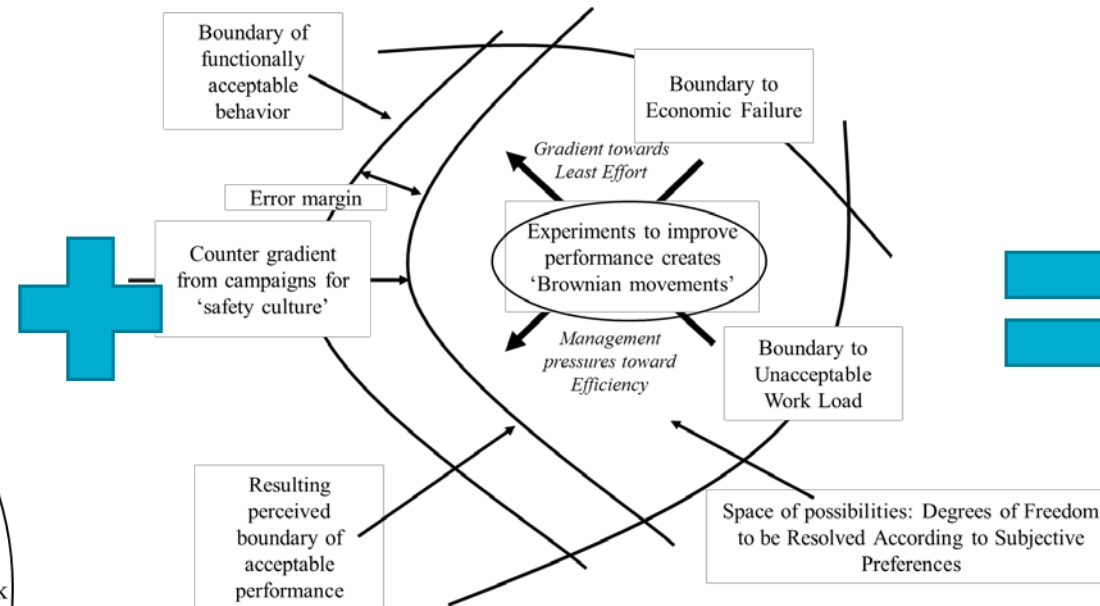
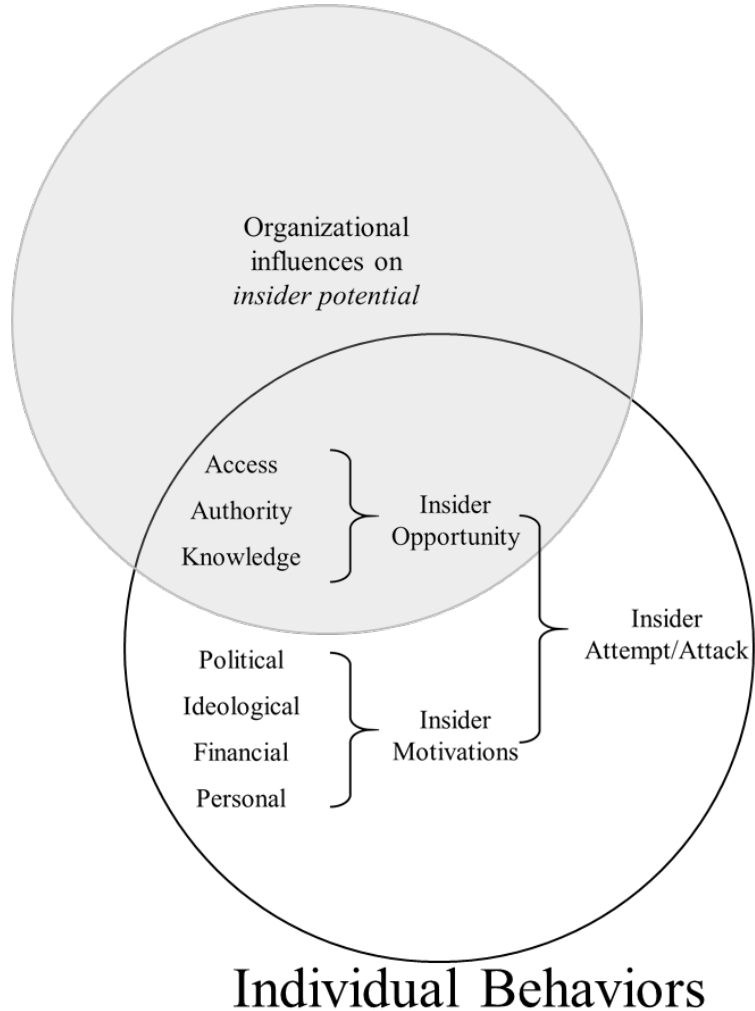
## Collective Behaviors



A ***new approach*** for potential improvement, based on several observations:

- People working in nuclear facilities settle into “operational rhythms”
- These rhythms can be described with data/signals already being collected at nuclear facilities
- Recast “preventive” & “protective” approaches as boundaries on these rhythms

## Collective Behaviors



A new approach :

- "workplace rhythms"
- data/signals already being collected
- Recast approaches as boundaries on these rhythms



# Situating: Artificial Neural Networks

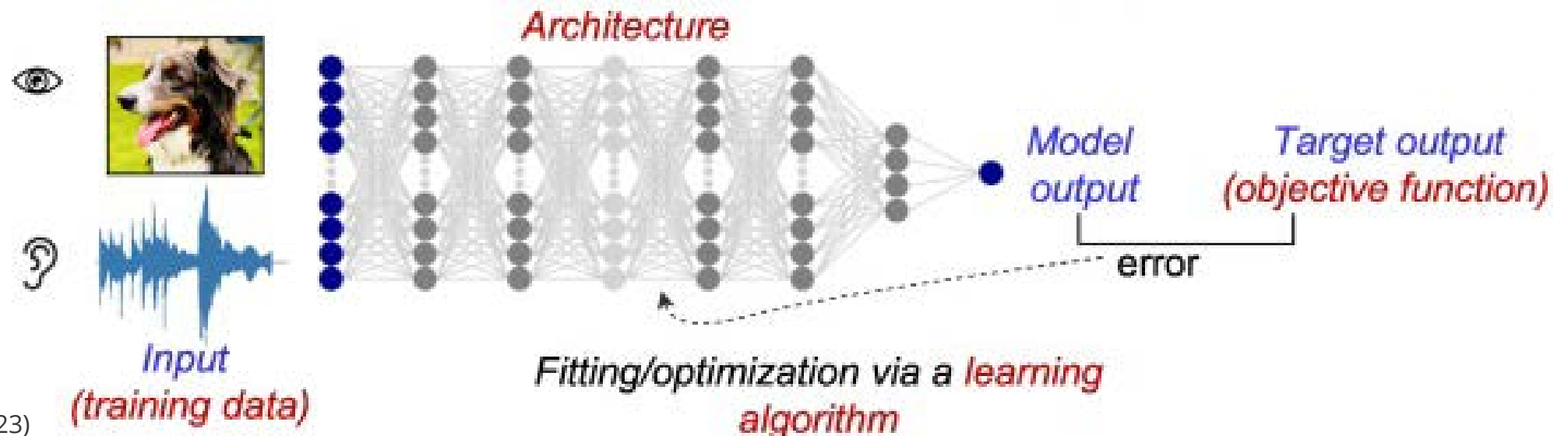


Assumption:

- Insider threat **attempts** represent a deviation from these “operational rhythms”

Conclusion:

- Humans are **creatures of habit & unpredictable** – can deviation from normal rhythms ID insiders?
- Anomaly detection **may** identify the **potential** for an insider opportunity to manifest into action
- Artificial neural networks** (ANNs) can be trained to ID patterns/deviations in operational rhythms

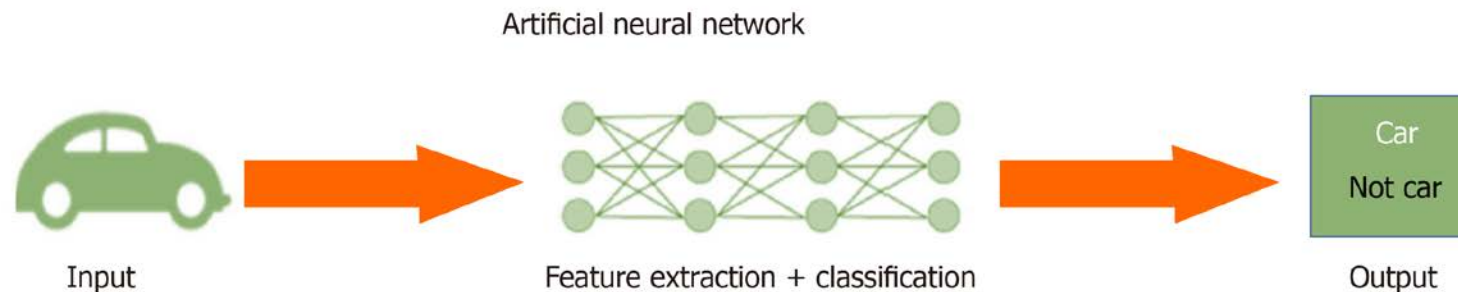
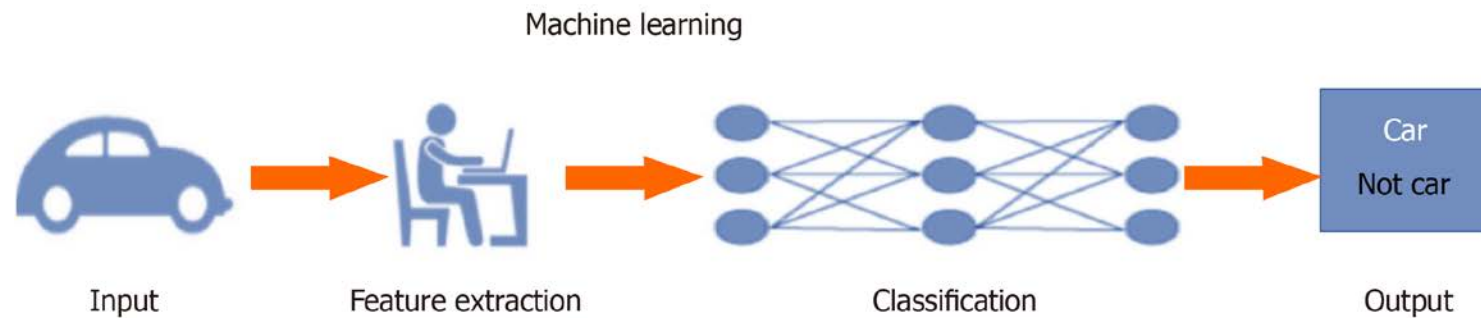


# Situating: Artificial Neural Networks



Hypothesis: ANNs can evaluate facility data signals to support ITDM

- Unusual access times as monitored by access control points like badge readers
- Attempts to access physical areas beyond current access level as monitored by access control points
- Increased or routine alarms from personnel radiation portal monitors



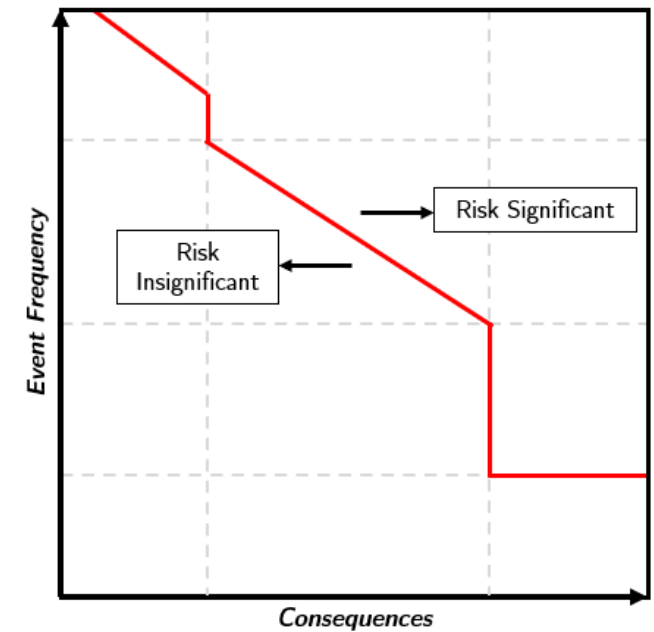
# Situating: Risk Significance

Borrowing the concept of **risk significance** from nuclear safety:

- Risk significance → does an accident sequence exceed a predetermined risk limit?
  - $f$  (event frequency, consequences)
- If yes, then those accidents are risk significant

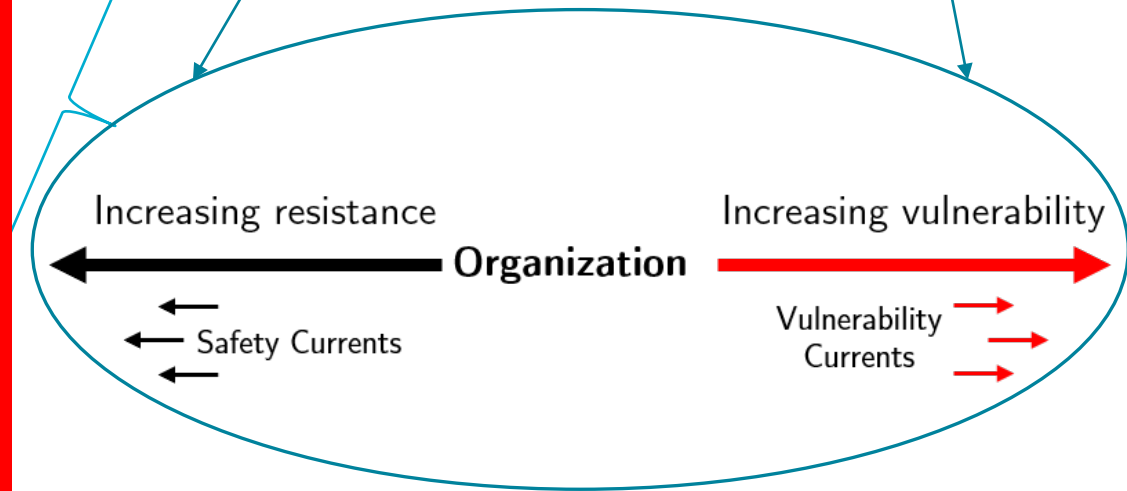
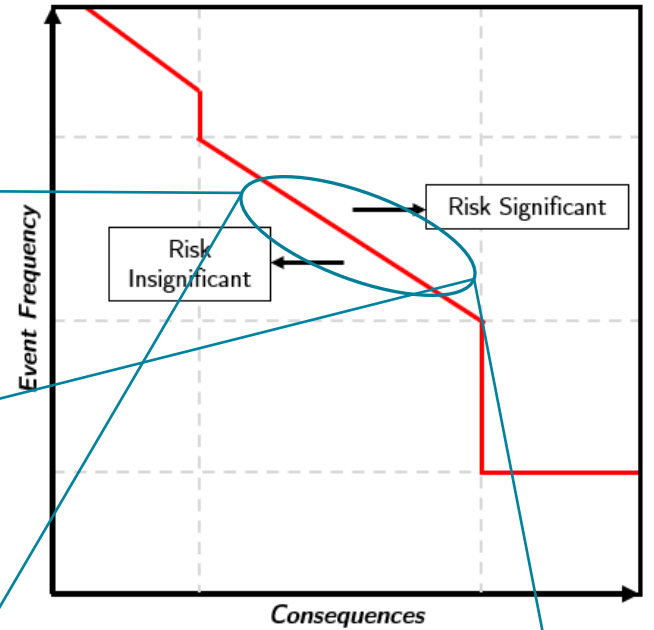
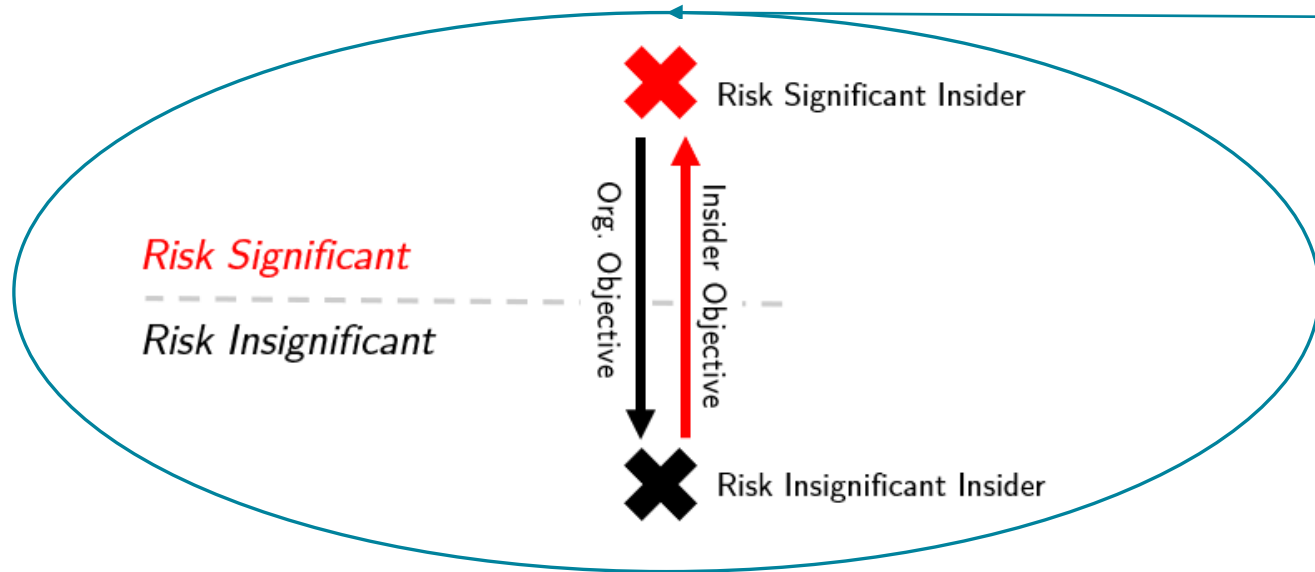
Therefore, **risk significance for an insider** considers:

- Best described as a time-variant continuous variable
- Related to the ability to successfully execute an act
- Both individual & facility characteristics
  - Ex: Individuals conduct business according to the access & authority (sometimes knowledge) bestowed by the facility



**Workplace  
rhythms**

# Situating: Risk Significance



**A risk significant insider:**

- has capabilities that exceed the ability of security measures for detection, including
- **Type I Non-Detection** = the lack of detection **before** an insider
- **Type II Non-Detection** = the lack of detection **after** an insider act



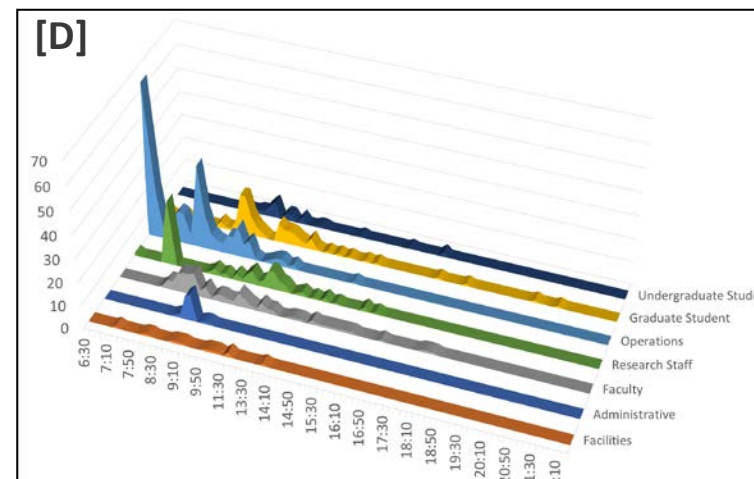
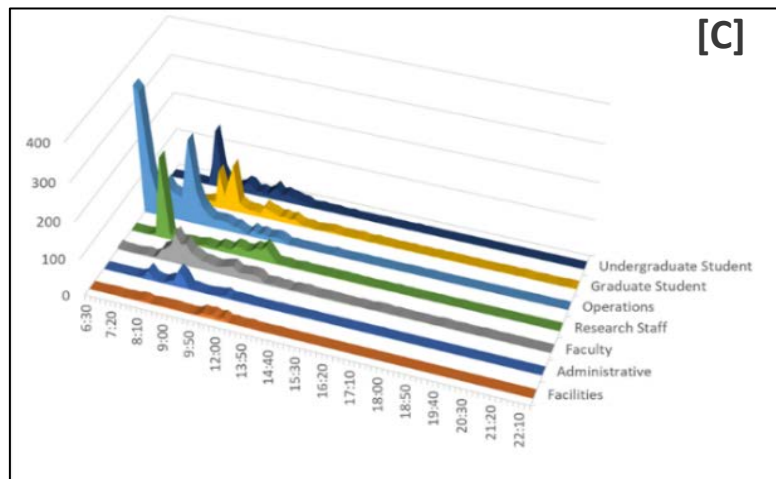
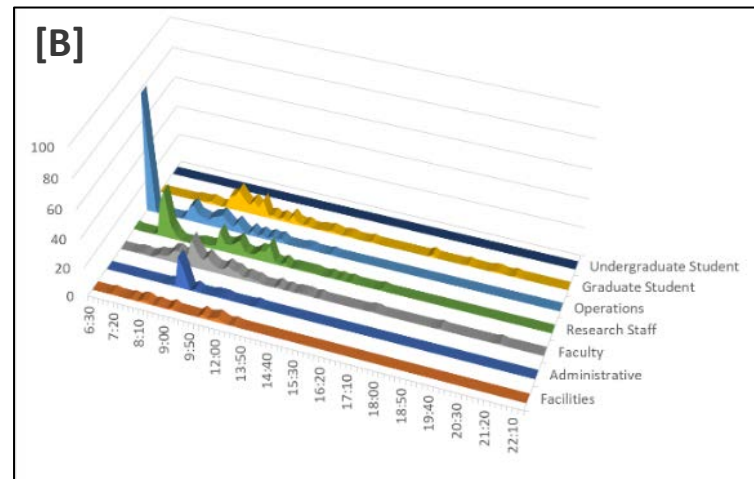
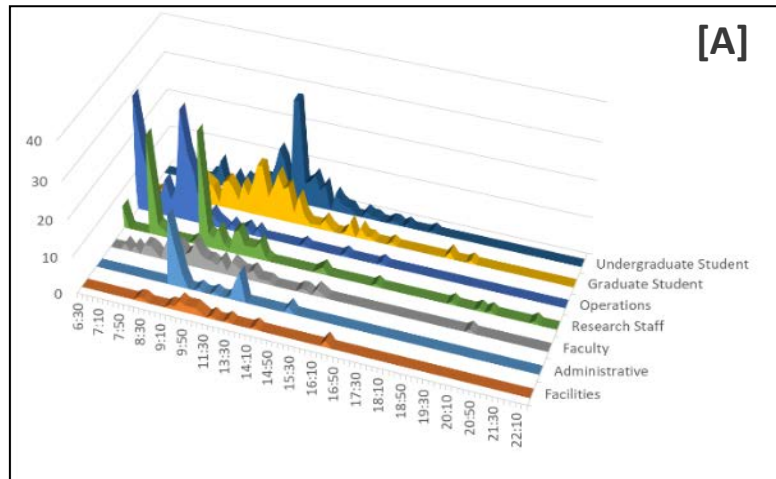
	Description	Implication
Single access point (SAP)	All access control data was organized by sensor location in the facility, date and time of allowed access, and then by identity used for access	Allowed for observation of patterns of accesses in time including bounds for when particular accesses are expected to occur for all individuals as well as for specific individuals
Time-sequenced, multiple access points (TS/MAP)	All access control data was organized by identity used for access, by date and time of allowed access, and then by location in the facility	Allowed for observation of patterns of access by individuals including bounds for when particular individuals would be expected to complete a sequence of access to different locations
Time of access by personnel type	All access control data was organized by access point, date and time of allowed access and then by grouping the identity used for access into a personnel type	Allowed for observation of pattern differences between personnel groups: Facilities, Administrative, Faculty, Research Staff, Operations, Graduate Student, Undergraduate Student



Type	Sensor Type	Data Type	Representative Activity
Access Control	<ul style="list-style-type: none"> <li>Badge reader                             <ul style="list-style-type: none"> <li>ORG B entry</li> <li>Security control panel</li> <li>Limited area</li> <li>Reactor control room</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Badge readers:                             <ul style="list-style-type: none"> <li># authorized attempts</li> <li># unauthorized attempts (false negative + false positives)</li> <li>Time of access attempts</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Personnel arrival to facility</li> <li>Researchers approaching the reactor</li> <li>Reactor operator arriving for shift</li> </ul>
	<ul style="list-style-type: none"> <li>Balanced magnetic switch                             <ul style="list-style-type: none"> <li>Limited area</li> <li>Security control panel</li> <li>Reactor control room</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Balanced magnetic switches:                             <ul style="list-style-type: none"> <li># times switch opened</li> <li>Time at which switch opens</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Researchers approaching the reactor</li> <li>Maintenance of security control panel</li> <li>Reactor operator arriving for shift</li> </ul>
Intrusion Detection	<ul style="list-style-type: none"> <li>Area motion sensor                             <ul style="list-style-type: none"> <li>Reactor bay</li> <li>Fuel storage surveillance</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Area motion sensors:                             <ul style="list-style-type: none"> <li># times change in physical phenomena registered</li> <li>Time at which change in physical phenomena registered</li> </ul> </li> </ul>	<ul style="list-style-type: none"> <li>Custodial services around the reactor</li> <li>Transfer of fresh/used fuel into/out of ORG B</li> </ul>

Data Characteristic	Data Set I	Data Set II	Data Set III	Data Set IV
<b>ANN Solution</b>	Tool 1	Tool 1	Tool 1	Tool 2
<b>Date range</b>	10/12/2019 to 03/14/2020	03/15/2020 to 09/25/2020	09/26/2020 to 03/31/2022	03/15/2023 to 09/15/2023
<b>Access control data points</b>	13,653	18,986	74,922	27,653
<b>Intrusion detection data points</b>	694	923	4211	1102
<b>Categories for organizing data points<sup>a</sup></b>	SAP TSMAP	SAP TSMAP	SAP TSMAP	SAP TSMAP

# Demonstrating a New Approach: SAP Frequency



- Somewhat surprising level of regularity
- Time bounds → baseline patterns for ANN
- Key Results:
  - collected data signals can reflect patterns and rhythms in behaviors
  - common patterns and rhythms can form profiles associated with particular personnel categories
  - such personnel category profiles can be used as a baseline of expected behaviors

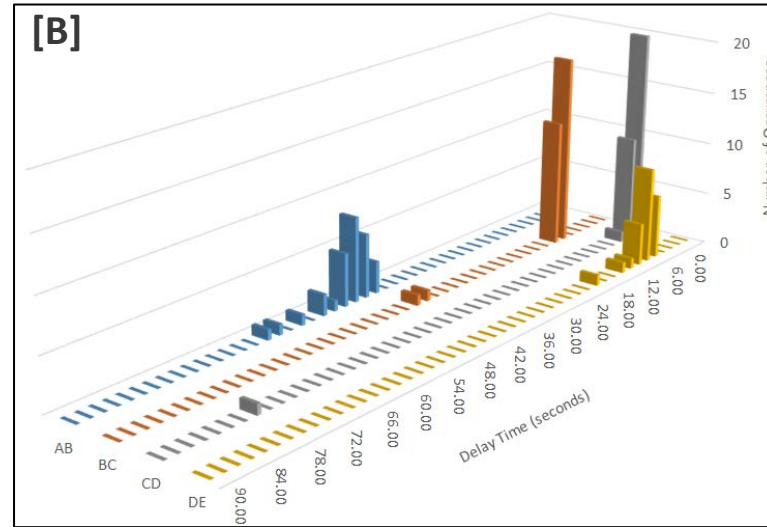
# Demonstrating a New Approach: TSMAP Frequency



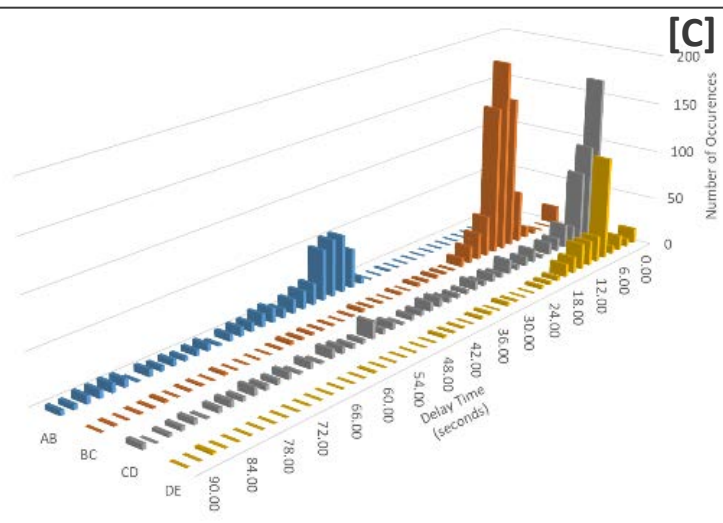
These results were not captured for Data Set I

[A]

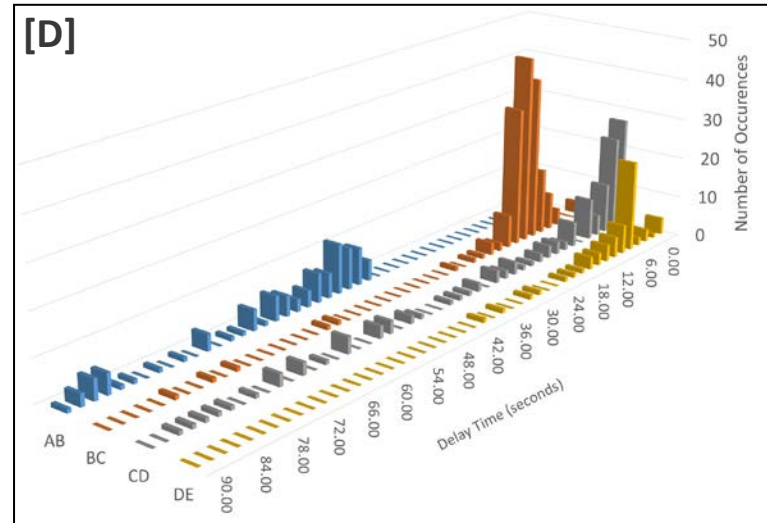
[B]



[C]



[D]



- Pathway-based patterns
  - $A \rightarrow B \rightarrow C \rightarrow D \rightarrow E$
- Higher fidelity + more nuanced description of patterns
  - Ex: "this individual is expected to take 42-66 seconds to move from access point A to access point B" (Data Set II)
- Key Results:
  - Higher validity & structure for anomaly detection
  - Captures dynamism of workplace rhythms

# Demonstrating a New Approach: Experimental Results

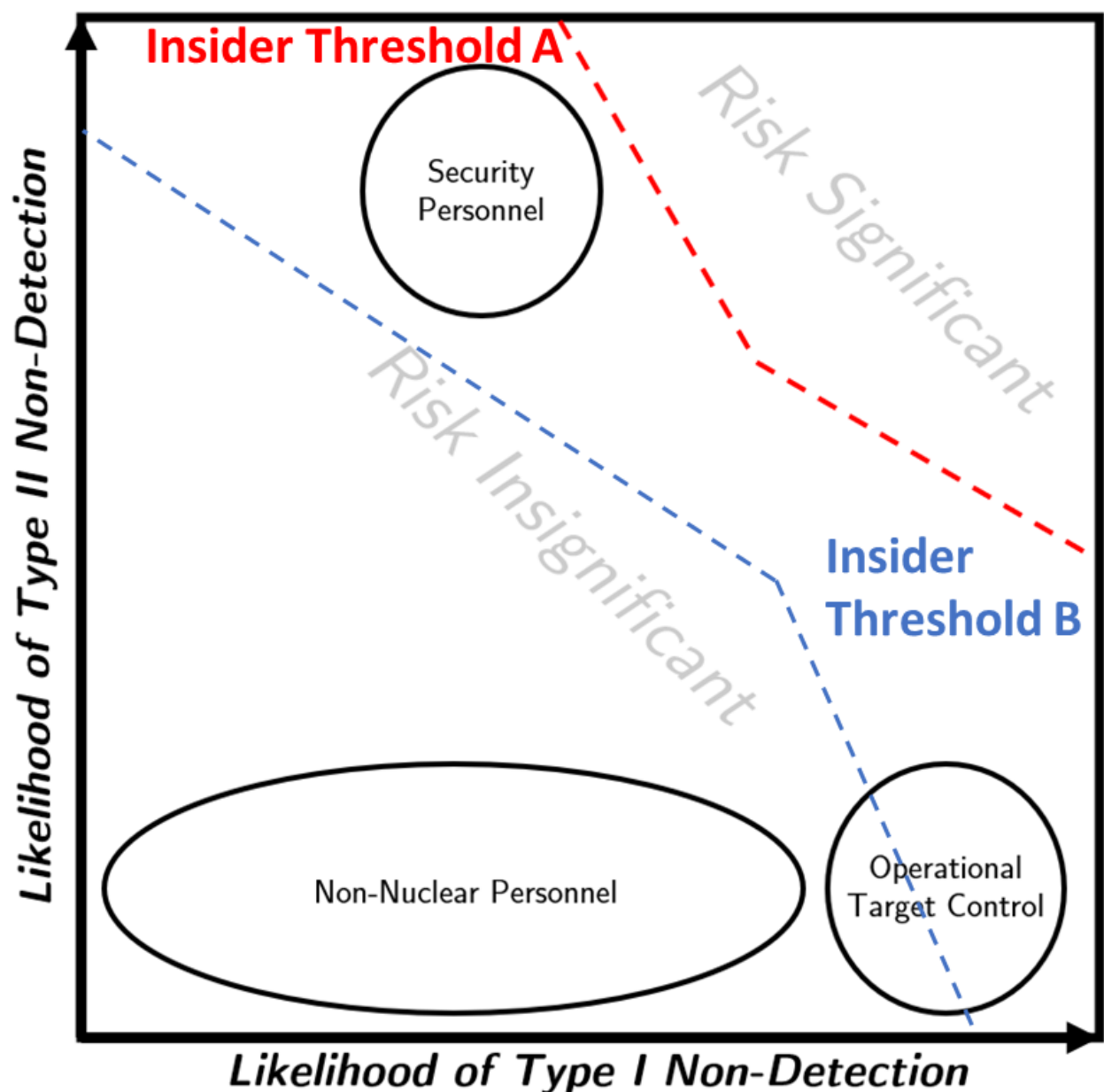


Scenario Name (#)	Test Description (Scenario # & Pathway Name)	Data Set I Results*	Data Set II Results	Data Set III Results	Data Set IV Results
Security Closet Access (1)	Unauthorized Access Attempt (1A)	Detected & Denied in <b>ALL</b> Cases [SAP]	Detected & Denied in <b>ALL</b> Cases [SAP]	Detected & Denied in <b>ALL</b> Cases [SAP]	Detected & Denied in <b>ALL</b> Cases [SAP]
	Authorized Access Credentials Used by Unauthorized Individual Who Entered Building Using Their Own Credentials (1B)	Detected & Denied in <b>MOST</b> Cases [SAP; TSMAP]	Detected & Denied in <b>MOST</b> Cases [SAP; TSMAP]	Detected & Denied in <b>MOST</b> Cases [SAP; TSMAP]	Detected & Denied in <b>NO</b> Cases [SAP; TSMAP]
	Authorized Access Credentials Used by Unauthorized Individual Who Entered Building Using Authorized Individual's Credentials (1C)	Detected & Denies in <b>NO</b> Cases [TSMAP]	Detected & Denies in <b>NO</b> Cases [TSMAP]	Detected & Denies in <b>MOST</b> Cases [TSMAP]	Detected & Denied in <b>MOST</b> Cases [SAP; TSMAP]
Reactor Bay Access (2)	Unauthorized Access to Reactor Bay (2A)	Detected & Denied in <b>ALL</b> Cases [TSMAP]	Detected & Denied in <b>ALL</b> Cases [TSMAP]	Detected & Denied in <b>ALL</b> Cases [TSMAP]	Detected & Denied in <b>ALL</b> Cases [TSMAP]
	Early Detection by Motion Sensor (2B)	Not Tested	Detected in <b>MOST</b> Cases	Detected in <b>MOST</b> Cases	Detected & Denied in <b>NO</b> Cases [SAP; TSMAP]
Fuel Storage Surveillance (3)	Insider Surveillance (3A)	<i>Difficult to Detect Without Additional Sensing Input [TSMAP]</i>	<i>Difficult to Detect Without Additional Sensing Input [TSMAP]</i>	<i>Difficult to Detect Without Additional Sensing Input [TSMAP]</i>	Detected & Denied in <b>NO</b> Cases [SAP; TSMAP]
	Insider Alarm Testing (3B)	Not Tested	<i>Difficult to Detect Without Additional Sensing Input [TSMAP]</i>	<i>Difficult to Detect Without Additional Sensing Input [TSMAP]</i>	Detected & Denied in <b>NO</b> Cases [SAP; TSMAP]

- Point 1
- Point 2



# Demonstrating a New Approach: Risk Significance



- SAP-based or TSMAP profiles → scaffold for functionally unacceptable behaviors or thresholds
  - Or, frame for risk significant insider potential as quantified deviation from expected behaviors
- Benefits:
  - Thresholds derived from ANN-identified patterns
  - Multiple thresholds on same framework (red & blue lines)
  - Clear mapping of different personal categories
  - Provides opportunity for *anticipatory* ITDM



## ***Positive results from ongoing data collection & early experiments***

Empirical support for theoretical & technical approach to ITDM based on “workplace rhythms”



## ***Shift toward “insider potential” a new, useful framing***

Encourages use of facility & system-related data streams; aligns with “workplace rhythms” interpretation



## ***Incorporating risk significance = a data-driven approach***

Supports quantitative descriptions of insider potential *not* heavily biased with individual psychometric indicators



## ***Incorporating risk significance = inclusive of data already being collected***

Leverages wealth of data (e.g., quality assurance) + mitigates common challenges to efficacy of behavioral reporting systems



## ***Incorporating risk significance = streamlines anomaly detection***

Helps prioritize deviations in workplace rhythms, with opportunity to anticipate/categorize future deviations in workplace