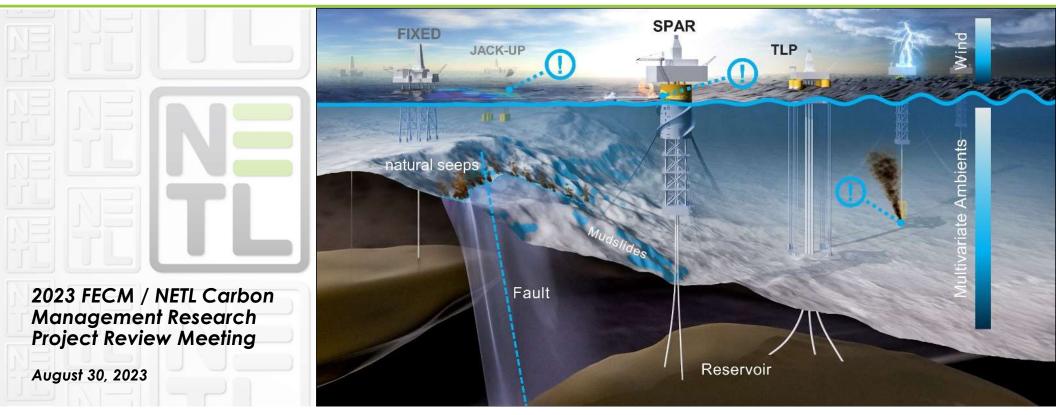
## Assessing Offshore Infrastructure Reuse Potential for CS



### An Award-Winning AI Model to Forecast Resiliency





### Disclaimer



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## Offshore Energy Operations: Potential Risks & Impacts

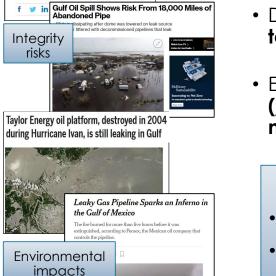
#### Solution:

- Applied NETL big data, big data computing, and advanced modeling capabilities to evaluate existing energy infrastructure
- Developed multiple machine learning (ML) models & spatiotemporal analytics to understand existing infrastructure integrity
- Enable model-driven insights & visualization with AIIM (<u>Advanced Infrastructure Integrity Modeling</u>) published method-turned-tool (alpha)

#### Values Delivered

- Evaluates the current state of offshore infrastructure for reuse potential
- Inform lifespan extension, remediation, and safe-use/reuse strategies
- Support environmental & operational risk prevention

AllM Technical Report (2021): <u>osti.gov/servlets/purl/1780656/</u>



Overview

Natural hazards

Offshore drilling regulator

of Mexico

warns of bolt failures in Gulf

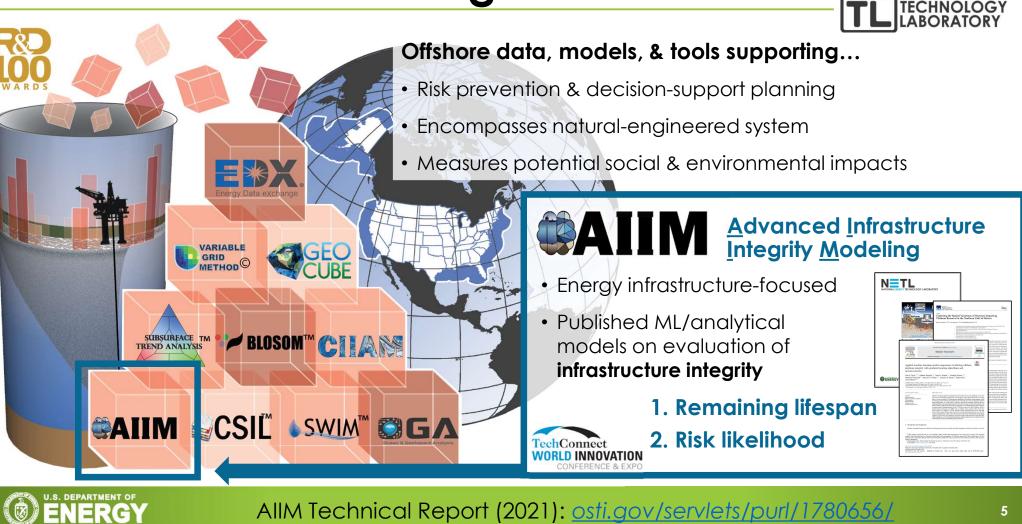
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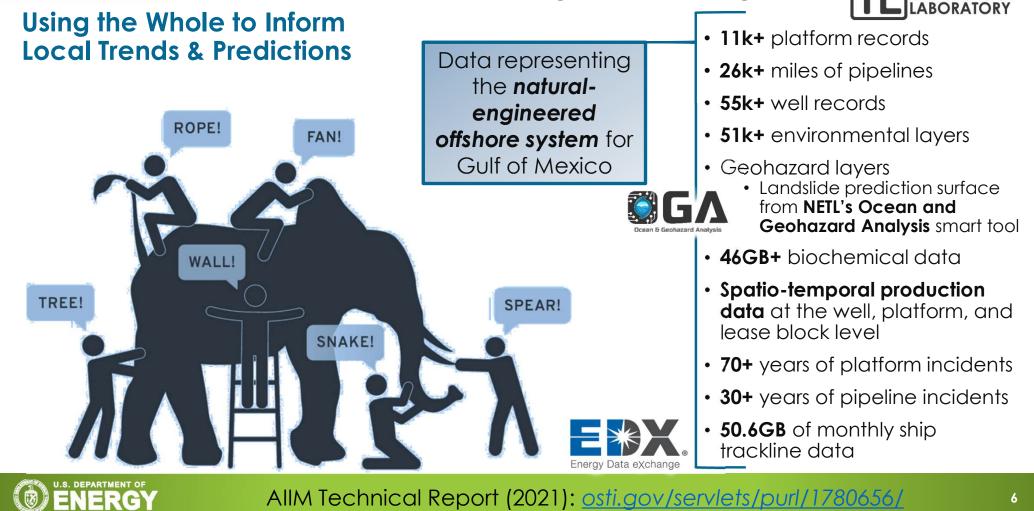
## **Offshore Risk Modeling Suite**



AllM Technical Report (2021): osti.gov/servlets/purl/1780656/

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## The **AIIM** Approach:

### Evaluating Multi-ML Model & Advanced Analytical Insights

#### Machine Learning Models

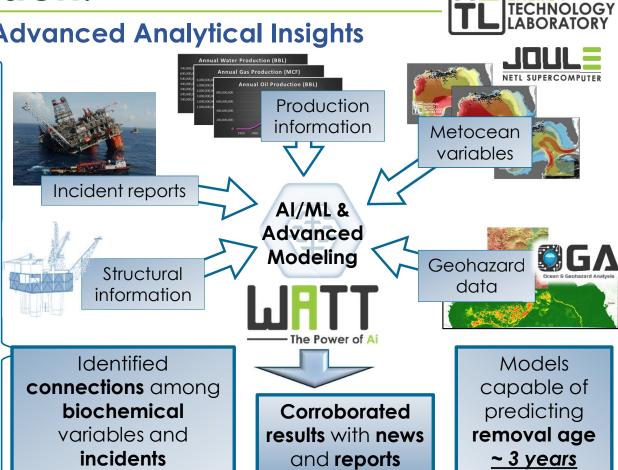
(Dyer et al. 2022)

- Gradient Boosted Decision Trees (GBDT)(2 models)
- Artificial Neural Network (ANN) (2 models)
- Bayesian Network

#### **Advanced Analytics**

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- Geographically Weighted Regression (Nelson et al. 2021)
- Causality/Time Series Analytics



Applied machine learning model comparison: Predicting offshore platform integrity with gradient boosting algorithms and neural networks - ScienceDirect

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## **AIIM**: Research Products & Insights

### **Evaluating Platform Infrastructure**

 Integrated 70+ years of data Ages of Existing Baton Rouge Platforms as of 11k+ structure records November, 2022 NET NATIONAL ENERGY TECHNOLOG<sup>\*</sup> LABORATORY • 1,700+ features Updated incidents & metocean statistics Federal Waters Sources: Bureau of Safety and Updated ML models capable Environmental Enforcement DW\* UDW\*\* and Livet al. 2018 of predicting removal age \*Deepwater (1,000ft+) Age Distribution within  $\sim$ 3 years \*\*Ultra-deepwater (5,000ft+) 17 70 - 80 ENX NETL's Energy Data e • Key stressors included: 60 - 70210 50 - 60404 Storm occurrences e GOM Federal Waters P Comprehensive 40 - 50479 (wind gust, minimum **Platform Dataset** 30 - 40834 central pressure) 20 - 30902 to be updated 10 - 20759 Corrosive factors 1 - 10225 (Romeo, 2021) 0 - 1(biochemical variables) Energy Data exchange U.S. DEPARTMENT OF

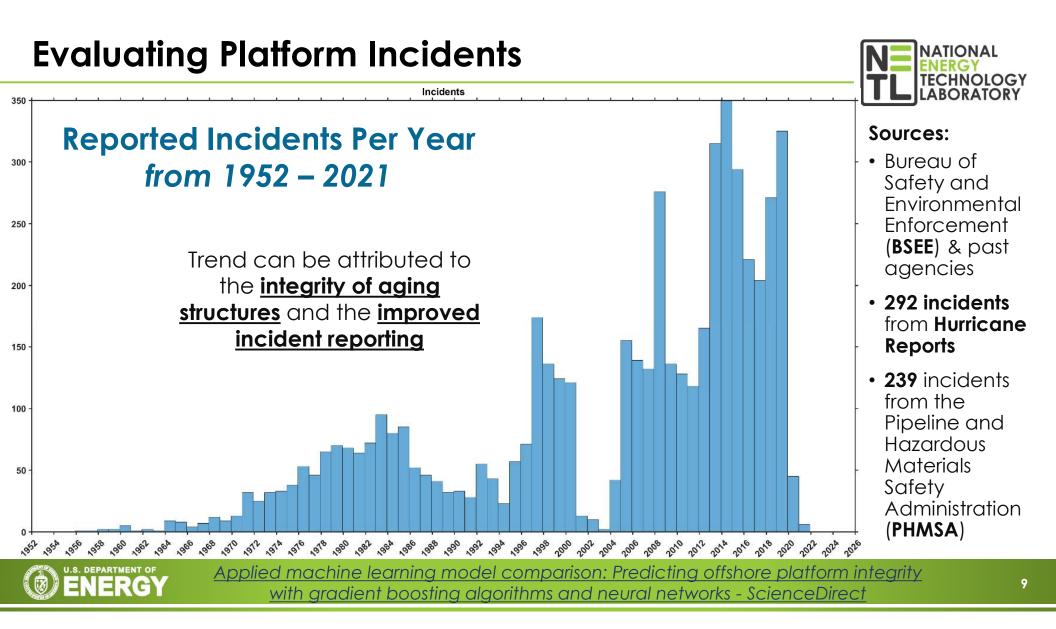
Applied machine learning model comparison: Predicting offshore platform integrity with aradient boosting algorithms and neural networks - ScienceDirect

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	Model	Roof Mean Square Error (RMSE)
	ANN	5.3 years
Ð	GBDT (XGBoost)	3.4 years
	GBDT (CatBoost)	3.1 years

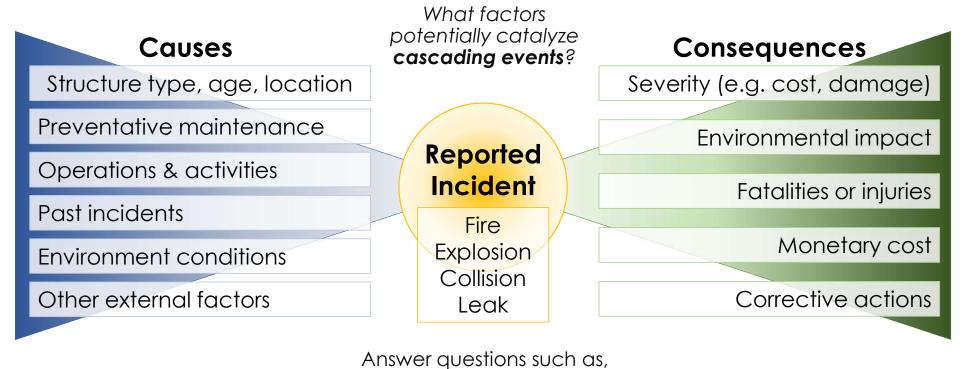
Number of well slots Water depth

**Reported** incidents



## **Expanding Analytics on Reported Incidents**

#### **Examining Risk Cause and Effect**



- At what age are platforms most prone to incidents?
- Where are infrastructure more vulnerable to incident?



<u>https://edx.netl.doe.gov/offshore/portfolio-items/assessing-</u> current-and-future-infrastructure-hazards/



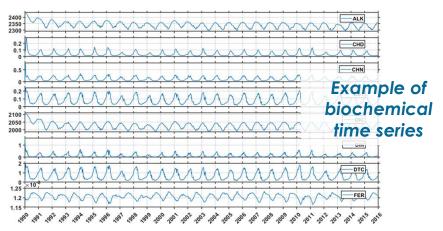
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## **Time Series Analytics**



- Aggregated incidents and biochemical (Yool et al., 2013) time series to estimate **transfer** of information
  - 26 years of biochemical data (1990-2015)
- Causal relationship was not found
  - Biochemical variables  $\rightarrow$  incident time series
- Identified a causal relationship
  - Biochemical variables → rate of change of incident time series



#### Biochemical properties cause incidents over **two main periods**: **10 years** and **20 years**

Model	Variables
Medusa/Orca (U.K. NOC)	Alkaline, diatom chlorophyll, nondiatom chlorophyll, detritus, inorganic carbon, inorganic nitrogen, detritic carbon, dissolved iron, dissolved oxygen, biologenic silicon, silicon, diatom phytoplankton, non-diatom phytoplankton, silicate, meso zooplankton, micro zooplankton
	https://edx.netl.doe.gov/offshore/portfolio-items/assessing- current-and-future-infrastructure-hazards/

## **AIIM**: Research Products & Insights

## **Evaluating Pipeline Infrastructure**

**Root Mean** • 26k+ miles of pipelines Preliminary model runs capable Model **Square Error** of predicting abandonment age (RMSE) • 100k+ locations (every 1km + end points) within  $\sim$ 3 years 400+ features ANN 3.3 years • Key stressors included: Added seafloor metocean data **Biochemical variables** (floor current, bottom pressure) GBDT 3.1 years Installation date/proxy date (XGBoost) **Pipelines by Status** Cathodic code Source: Bureau of Safety 0.7 years\*\* GBDT and Environmental<sup>o<sup>1</sup></sup> Status code Enforcement (CatBoost) 0.99 R<sup>2</sup> Facility operator Status Miles Pipeline Database Abandoned or 23.838 slated for Removed Fall 2023 Release! Database 19,731 Active Publication is in Cancelled, Proposed 3,914 Abandon or Remove preparation! DW\* UDW Federal Waters Proposed 109 (Pfander et al. in prep) Energy Data exchange https://edx.netl.doe.gov/offshore/portfolio-items/assessing-



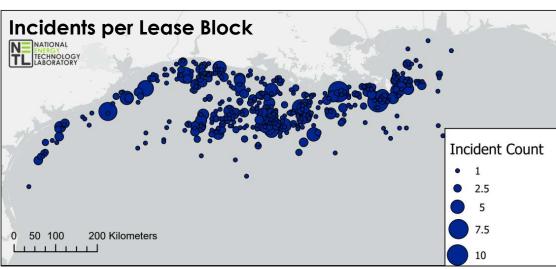
current-and-future-infrastructure-hazards/

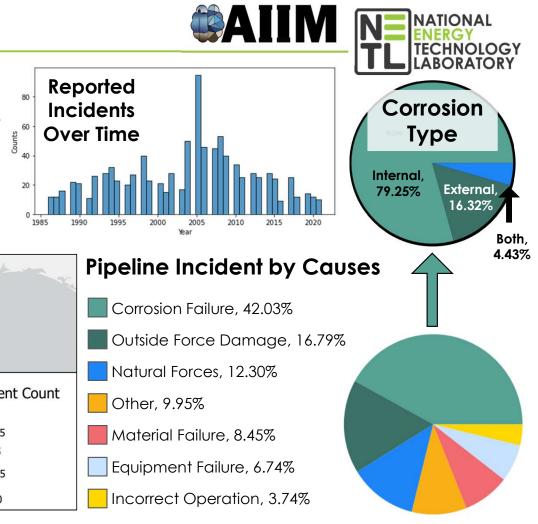


## **Pipeline Incident Processing**

#### **Understanding 30+ Years of Incidents**

- Compiled, cleaned, and mapped **970 incidents** (Pipeline and Hazardous Materials Safety Administration)
- 30+ years of incidents (1986 2021)
- Mapped more than 80% to lease blocks
- Calculated impact-based severity





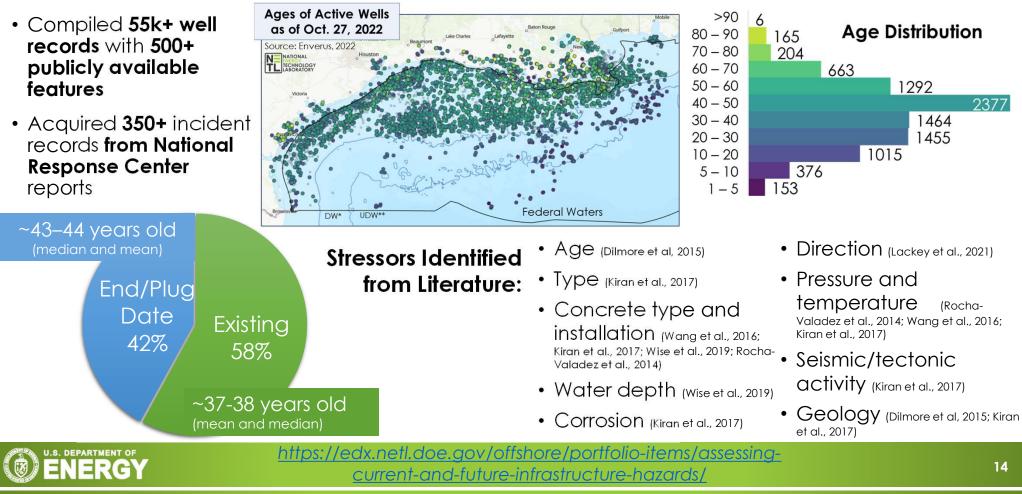


<u>https://edx.netl.doe.gov/offshore/portfolio-items/assessing-</u> <u>current-and-future-infrastructure-hazards/</u>

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## **Integrating Well Infrastructure**

#### Utilizing Past Research to Inform New Insights



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## **AIIM** Dashboard (alpha)

## **Delivering Data- and Model-Driven Insights**

- Developed as an ESRI Dashboard to support data visualization & interrogation
- Contains **pipeline data** and model results
- Adding in updated platform data and model results
- Enable spatially querying by **areas of interest** (lease blocks, protraction areas, etc.)

Offshore Advanced Infrastructure Integrity Modeling (AIIM) Dashboard Active Lease Blocks Lease Effective Date: 10/31/2018 Lease Status: Within initial term of 5, 8, or 10 year contract Area Code & Lease Block: AC24 Lease Status: Held by production of a minera 47.6k 47.6k 22.8k Area Code & Lease Block: AC240 se Status: Within intial term of 5, 8, or 10 year miles miles miles Area Code & Lease Block: AC245 ease Status: Within intial term of 5, 8, or Lease Effective Date: 11/30/1988 Lease Status: Included in approved unit agreemen Shallow (< 1.000 ft OCS Lease Block Incident Information eep Water (1 000-5 000 ft) Area Code and Lease Block: BM5 Total Incidents: 8 Material Failure Incide Ultra Deep Water (> 5,000 ft Geohazard Incidents: 2 Incorrect Operation Incidents: Area Code and Lease Block: El266 Total Incidents: 7 Material Failure Incidents: Geohazard Incidents: 0 Incorrect Operation Incidents: ( Geohazard Influenced Equipment Failure 61 Area Code and Lease Block: EI30 Total Incidents: 6 Material Failure Incidents: 6 ard Incidents Area Code and Lease Block: SS15 200 m Total Incidents: ONANP, Esri, HERE, Garmin, FAO, NOAA, USGS, EPA | Bure GOM Overview Pipelines

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https://edx.netl.doe.gov/offshore/portfolio-items/assessingcurrent-and-future-infrastructure-hazards/



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