



# Open Power AI Consortium Member Representative Committee Meeting



September 2025

# Agenda

Thursday, August 21, 2025		
Time (EST)	Topic	Lead
11:00 am	Welcome & Key Updates <ul style="list-style-type: none"><li>• Meeting Overview</li><li>• Work Groups</li><li>• What's Coming?</li></ul>	Jeremy Renshaw, EPRI
11:10 am	Storm Response & Aurora Weather Model	Pat Lo, Microsoft
11:30 am	Active Edge Network Management & Storm ERT Prediction	Rajagopal Iyengar & Jason Duncan-Wilson, Oracle
11:50 am	Roundtable Discussion	All
12:00 pm	Adjourn	All

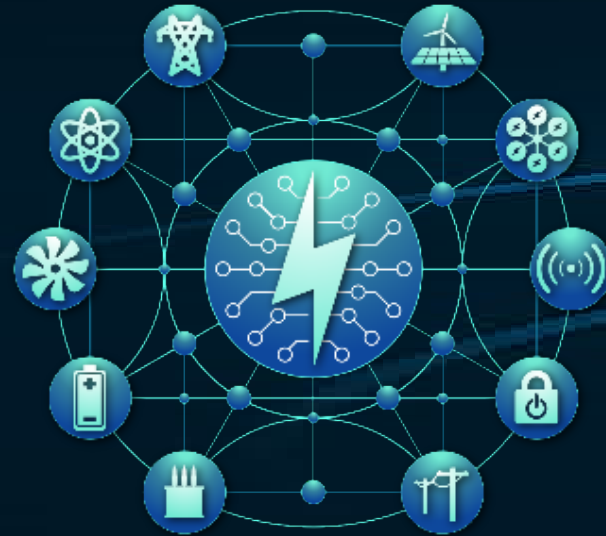
# Open Power AI Consortium

Creating an ecosystem for stakeholders to identify, develop, validate, and deploy AI solutions to transform the electric sector

Develop AI Models,  
Data & Roadmaps

Create an AI  
Sandbox

Implementation and  
Lessons Learned



**OPEN POWER  
AI CONSORTIUM**

**>200** Members

**>100** MOU/Requests  
In-Process

**>300** Use cases  
identified

# OPAI Participants

## AI and Energy Technology Partners



## Academia & Other Strategic Partners



## Energy Partners





# OPAI Work Status - Initial Priorities for the Three Work Groups



## Electric DSM WG

**Lead: Ben Sooter**



## AI Use Case WG

**Lead: Adrian Kelly**



## Deployment WG

**Lead: Jason Hollern**

# What Do We Need from You?

- Fill out the MRC Survey
  - <https://www.surveymonkey.com/r/participantsurvey>
  - Assign staff to Work Groups (if not done)
- Share public examples of relevant use cases
  - Proposed
  - Ongoing
  - Completed
- Reach out to [jrenshaw@epri.com](mailto:jrenshaw@epri.com) to speak in future meetings



**Participant  
Survey QR Code**



# Data Readiness White Paper

How can we properly leverage data to support AI initiatives?



EPRI



2025 White Paper

AI Readiness in Utilities: Turning Data into Strategic Advantage

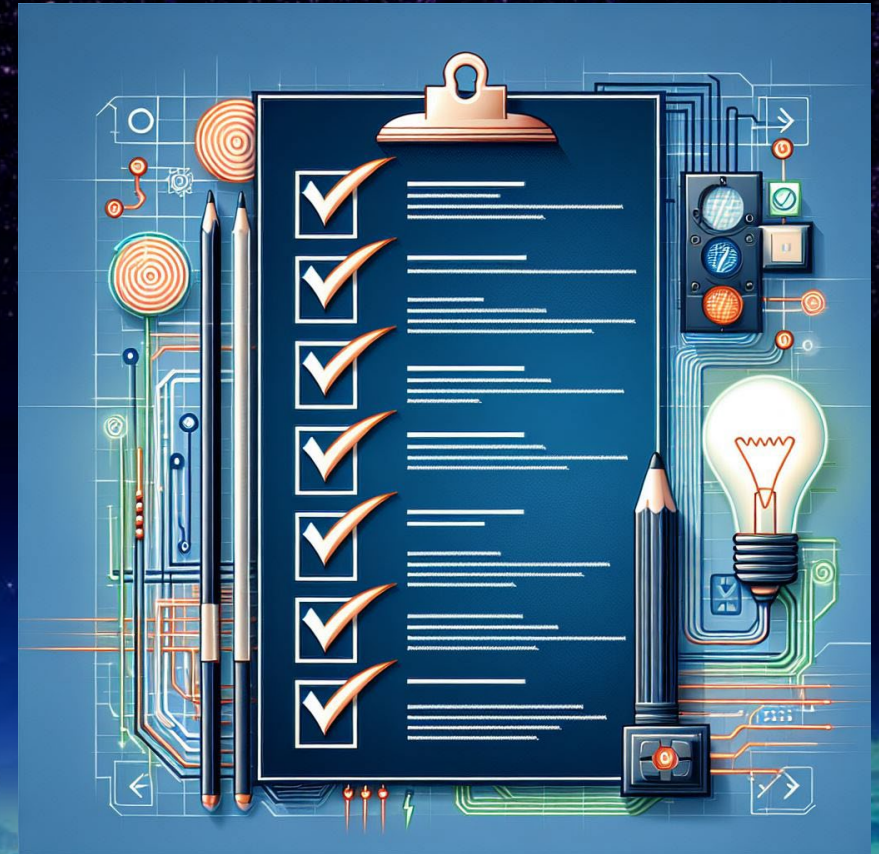


The first in a series of guidance documents to accelerate AI implementation



# What's Coming?

- Work Group meetings starting (October)
- Domain-specific model and benchmarking results (October)
- Use Case database published (October)
- More opportunities to engage
- New meeting invite!





# Storm Response & Aurora Weather Model (Microsoft)



## Active Edge Network Management & Storm ERT Prediction (Oracle)

The Oracle logo, consisting of the word "ORACLE" in a red, bold, sans-serif font.



# Storm Response

AI for Power & Utilities

Leslie Connolly

AI Lead for Storm Response - Energy

Pat Lo

Americas Executive Leader – Power & Utilities

# Solve for X | Utility Challenges Today

1. Unlocking New Capacity | Rapid Planning and Building to connect more customers
2. Extreme Weather and Reliability | Predict, Prevent, and Restore Power
3. Data as an Asset | Extracting more value out of existing software and systems to unlock new businesses and save opex \$

# Storm Response

## Predict

AI for Extreme Weather Forecasting

AI for Field Worker Dispatch

## Prevent

AI for Structure Damage Assessment

AI for Post Event Analysis

## Restore

Data & AI for Outage Reporting

Agents for Storm Response Process

Microsoft Planetary Computer Pro



# Introducing Aurora: The first large-scale foundation model of the atmosphere

Aurora is a 1.3 billion parameter foundation model for high-resolution forecasting of weather and atmospheric processes.

Aurora leverages the strengths of the foundation modelling approach to produce operational forecasts for a wide variety of atmospheric prediction problems, including those with limited training data, and extreme events. In under a minute, Aurora produces 5-day global air pollution predictions and 10-day high-resolution weather forecasts that outperform state-of-the-art classical simulation tools and the best specialized deep learning models.

Aurora's effectiveness lies in its training on more than a million hours of diverse weather and climate simulations, which enables it to develop a comprehensive understanding of atmospheric dynamics. This allows the model to excel at a wide range of prediction tasks, even in data-sparse regions or extreme weather scenarios.



# AI For Good: Microsoft Damage Assessment and Storm Recovery Tool

The frequency of natural disasters is increasing globally, affecting 350 million people annually and causing billions in damage. Providing timely humanitarian interventions like shelters, medical aid, and food is challenging.

Microsoft's solution uses high-resolution satellite imagery and a convolutional neural network model to localize buildings and assess damage levels, categorized into four levels based on the xView2 dataset.

The primary value lies in inference speed, with the solution operating three times faster than the fastest xView2 challenge-winning solution. The model achieves a pixel-wise F1 score of 0.74 for building localization and 0.6 for damage classification. Additionally, a web-based visualizer displays before and after imagery with damage predictions. This study, conducted with a humanitarian organization, aims to deploy and assess the model and visualizer for disaster response efforts.

Visualizer for Hurricane Milton in  
(2024)

Restore | AI for Reliability

## Hurricane Milton Response



Juan M. Lavista Ferres • 2nd  
CVP and Chief Data Scientist at Microsoft  
3w •

+ Follow

Our thoughts are with those affected by the recent storms that impacted South  
Brandenton, Sarasota and surrounding areas.

We ran our damage assessment AI models on images provided by NGS and have  
mapped out the affected buildings. If your organization would benefit from access  
to the underlying data in this report, please reach out to me. We are committed to  
sharing this information to support ongoing recovery efforts.

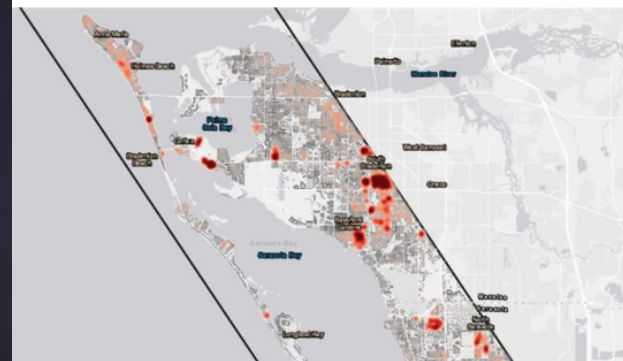
From the 650 sq km of imagery analyzed, we found:

24,486 (17.3%) of buildings are directly affected  
73,537 (51.9%) of buildings are unaffected  
68,095 (48.1%) of buildings have damage within 20m of their surrounding

While the data provides a valuable first look, it should serve as a preliminary guide  
and will require on-the-ground verification for a complete understanding. Any  
building classified as >0% damaged likely has nearby debris and should be further  
inspected.

The AI for Good Lab remains dedicated to helping impacted communities recover  
and stands ready to assist with data and insights that can accelerate relief efforts.

Link to visualizer:  
<https://lnkd.in/gf-hSnt5>





# Microsoft Planetary Computer Pro



## Power & Utility Use Cases

### Compliance & Reporting

Environmental Monitoring

Land Use Tracking

### Risk & Forecasting

Risk Assessment & Modelling

Supply Chain Optimization

Weather & Climate Analysis

Scenario Modelling

### Operations & Optimization

Asset Monitoring

Site Intelligence

High Resolution Mapping

Area of Interest Modelling



# Southern Company leverages Microsoft & Databricks for Outage & Storm Modelling



**RAMP**, which stands for Reliability Analytics Metrics and Performance, is an Azure-based reliability application that provides a comprehensive view of the power grid's performance, including reported values, customer experience values, and device failures. The application helps identify areas of improvement and provides insights into the root causes of reliability issues.

**SPEAR**, which stands for Storm Planning, ETR and Reporting, is a forecasting application on cloud that uses data from weather vendors and internal systems to predict the impact of severe weather events on the power grid. The application provides a detailed forecast of the number of incidents, resources needed, and estimated time of restoration, allowing the company's storm center to make more informed decisions and allocate resources more effectively.



## Forecasted Impact:

- 5% targeted reduction (3500 outages) to save **\$17.5M**
- 99.7% (3600x) efficiency gain historical customer outage history recovery
- Potential savings of **\$2.8M** per 10-day storm event



# Disruption Prepare

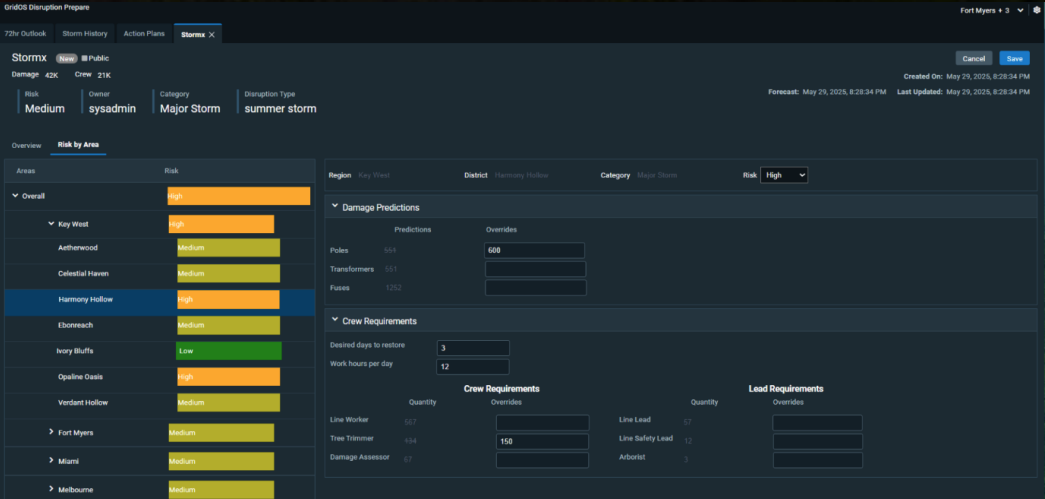
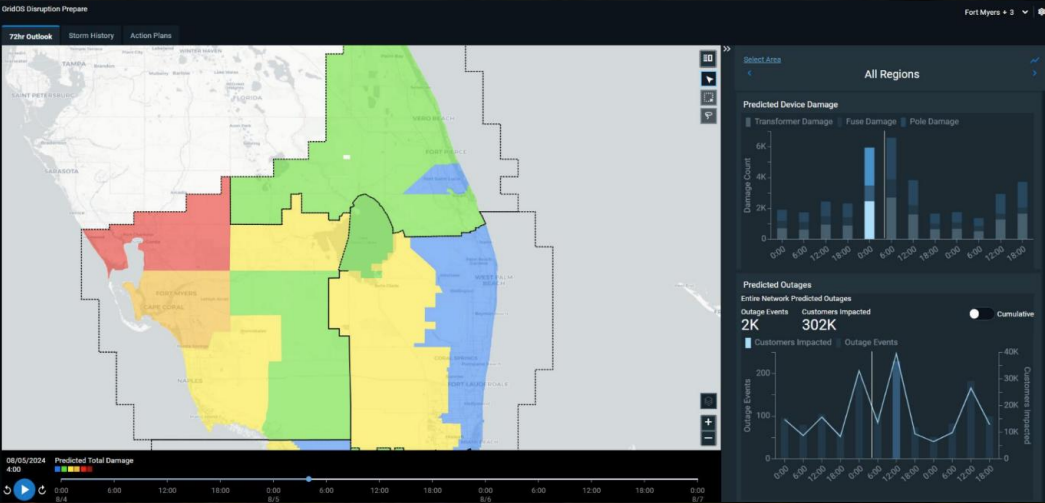
Together, GE Vernova and Microsoft are equipping operators with tools to reduce risks and recover quickly from large-scale disruptions such as wildfires and storms. Using Vision AI, GIS, LIDAR, drone, and satellite data, the companies are equipping operators with a single data plane to enable resilient operations.



GE VERNOVA



Microsoft





# Storm Response Agents

Microsoft AI for Grid Resiliency



Microsoft



# AI For Storm

## Weather Prediction

- Better Predict extreme weather events with generative models
- Forecast Accuracy/Confidence Agent
- Forecast
- Impact and timing chat bot.

## Storm Response

- Outage Detection Agent
- Intelligent Dispatch optimizer
- Grid Reconfiguration Advisor
- Excluded event predictor

## Storm Response Planning

- Storm Impact Prediction
- Storm material planning
- Crew resource forecasting
- Grid resilience planning
- Vegetation management
- Proactive reliability action agent

## Storm Assessment

- Remote sensing storm damage assessment.
- Exclusions
- Generate compliance-ready financial reports



## Storm Restoration

- Equitable Store restoration agent
- Restoration validation
- Lessons learned compilation
- ETR Optimization

## Customer Outage Support

- Customer Communication Bot
- Call center agent assist

## Storm Recovery

- Storm exclusion agent.
- Data review and correction
- Storm reporting agent for regulators and leadership

## Response Coordination

- Mutual aid mobile chat bot
- Invoice review and approval agent
- Storm data correction/reconciliation
- Storm Material reconciliation

# StormGrid Collaborative



Microsoft is forming a first-of-its-kind industry consortium to co-develop AI-powered storm response tools—predictive outage models, intelligent crew dispatch, and resilient infrastructure strategies.

Shaped by utility leaders, this is your chance to turn weather into a solvable equation and lead the future of operational resilience

Microsoft Research

Microsoft Planetary

Microsoft Energy

[lconnolly@microsoft.com](mailto:lconnolly@microsoft.com)

[patlo@microsoft.com](mailto:patlo@microsoft.com)



# Asset Health Failure AI Models



Microsoft and EPRI are entering into a partnership to leverage EPRI's vast data sets and build first of its kind asset health failure models.

These models will be available exclusive to EPRI members and be enabled on a customer's Azure tenant.

## Transformer Health through Dissolved Gas Analysis

Microsoft and EPRI collaborated on accelerating the creation of AI Asset Health Failure models leveraging the technology used traditionally for diagnosis of health issues from blood samples. The Dissolved Gas and Oil Analysis model (DGA) will allow utilities to feed their own transformer DGA data into the EPRI and Microsoft AI accelerator to diagnosis the true health condition of these transformers, leveraging the knowledge, expertise, and data from EPRI's existing Transformer Asset Health Database

## Outcomes

- **100%** improvement in predicting at risk transformers
- **23%** improvement in false positive rate



# AI Solutions for the Modern Electric Grid

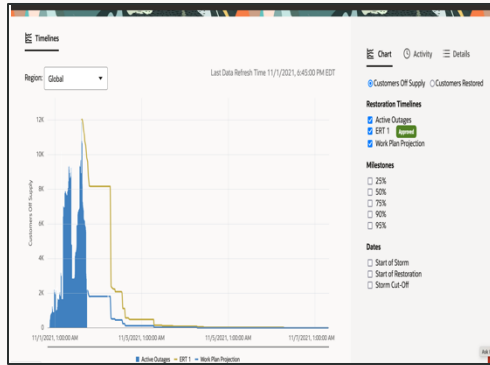
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Raj Iyengar (Oracle Utilities Data Science Lead) Jason  
Duncan-Wilson (Industry Data Platform Lead)



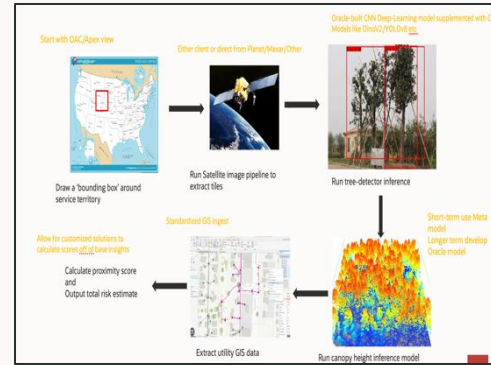
# Operate – Service Interruptions

Reliability and safety are critical. Extreme weather and other events can disrupt lives and services. Utilities need tools to predict, prevent, and protect against interruptions.



## Predict....

Using AI, historic weather and outage data, utilities can now model future events and prepare their storm response.



## Prevent....

AI analysis of satellite, drone and ground-based footage with GIS, spatial and network models: Improve vegetation management.



## Protect....

Respond to power line incursions in milliseconds. Combining the sensor technology + AI enabling time critical decision making at the edge.

# Predict

## Storm Estimated Restoration Time (ERT)



# Storm ERT

*A solution designed to help plan and manage storm restoration more effectively*



**Pre-Storm**

*Scope & crew*



**Post-Storm**

*Approval plan*



**Evaluate**

*Adjust & update*

✓ Single source of truth

✓ Integrated tools

✓ Progress updates

✓ AI/ML-based estimates

✓ Regulatory audits

Collaboration partner:



# Example Workflow: Pre-Storm Scope and Actions

## 1: Landing page: inputs here

Create Storm Model

### Storm Model Details

Name

1/29 Overhead Event

Start Date

1/29/2025, 2:00:00 PM EST

Cut-Off Date

1/30/2025, 11:00:00 PM EST

End of Restoration

2/1/2025, 7:00:00 PM EST

Operating Company

Cecony

☐

Exclude Single Customer Outages

Name

Pre-Storm ERT

Crew Schedules								
Zone	Estimated Customers	Actual Customers	Estimated Outages	Actual Outages	90% Restoration Date	1/29/2025 7:00am Crews	Efficiency	1/29/2025 7:00pm Crews
Global	14,082	15,454	474	1,119	1/31/2025, 7:01:54 AM EST	107	5	0
Con Edison	14,082	15,454	474	1,119	1/31/2025, 7:01:54 AM EST	107	5	0

## 2: Similar storm selector

Create Storm Model

### Similar Storm Events

The new Storm Model will be based on one or more ERTs. Choose up to six Storm Events and Approved and Archived Models to use for the first ERT. Results are ordered by similarity scores which indicate how much their weather matches that for the current storm. Some Storms may be selected by default.

Selected Storm Events

10/16/19 Northern Region 16-OCT-19 | Weight: 5 | 5/15/18 Northern Region 15-MAY-18 | Weight: 5

Q

State Historic | Duration Hours 24 - 72 | Season (3)

Type

Level

Customers 500 - 1,500

Outages 50 - 150

More Filters

19 items

	State	Name	Season	Storm Start	Storm Cut-Off	Duration Hours	Type	Level
<input type="checkbox"/>	Historic	4/03/24 Northern Region 04/04/24	Spring	4/3/2024	4/6/2024	57	Plus 10 2 0 Hrs	MAJOR_STORM
<input checked="" type="checkbox"/>	Historic	10/16/19 Northern Region 16-OCT-19	Fall	10/16/2019	10/18/2019	34	Plus 0 0 0 Hrs	
<input type="checkbox"/>	Historic	4/15/18 Northern Region 15-APR-18	Spring	4/15/2018	4/17/2018	29	Plus 0 0 0 Hrs	
<input type="checkbox"/>	Historic	8/21/21 Con Edison 21-AUG-21	Summer	8/22/2021	8/24/2021	50	Plus 0 0 0 Hrs	
<input checked="" type="checkbox"/>	Historic	5/15/18 Northern Region 15-MAY-18	Spring	5/15/2018	5/17/2018	37	Plus 0 0 0 Hrs	

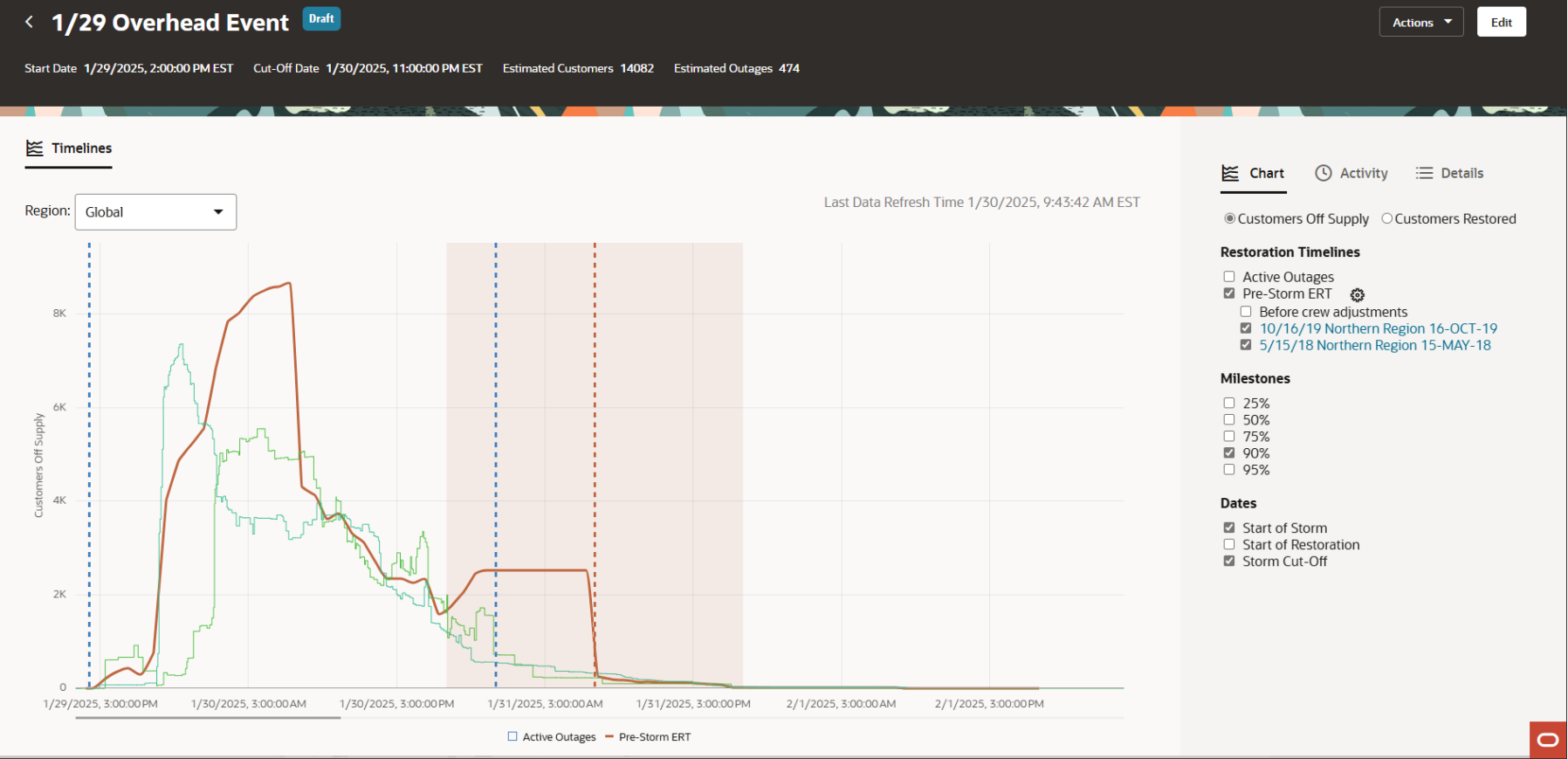
Cancel Continue

## 3: Crewing Adjustments

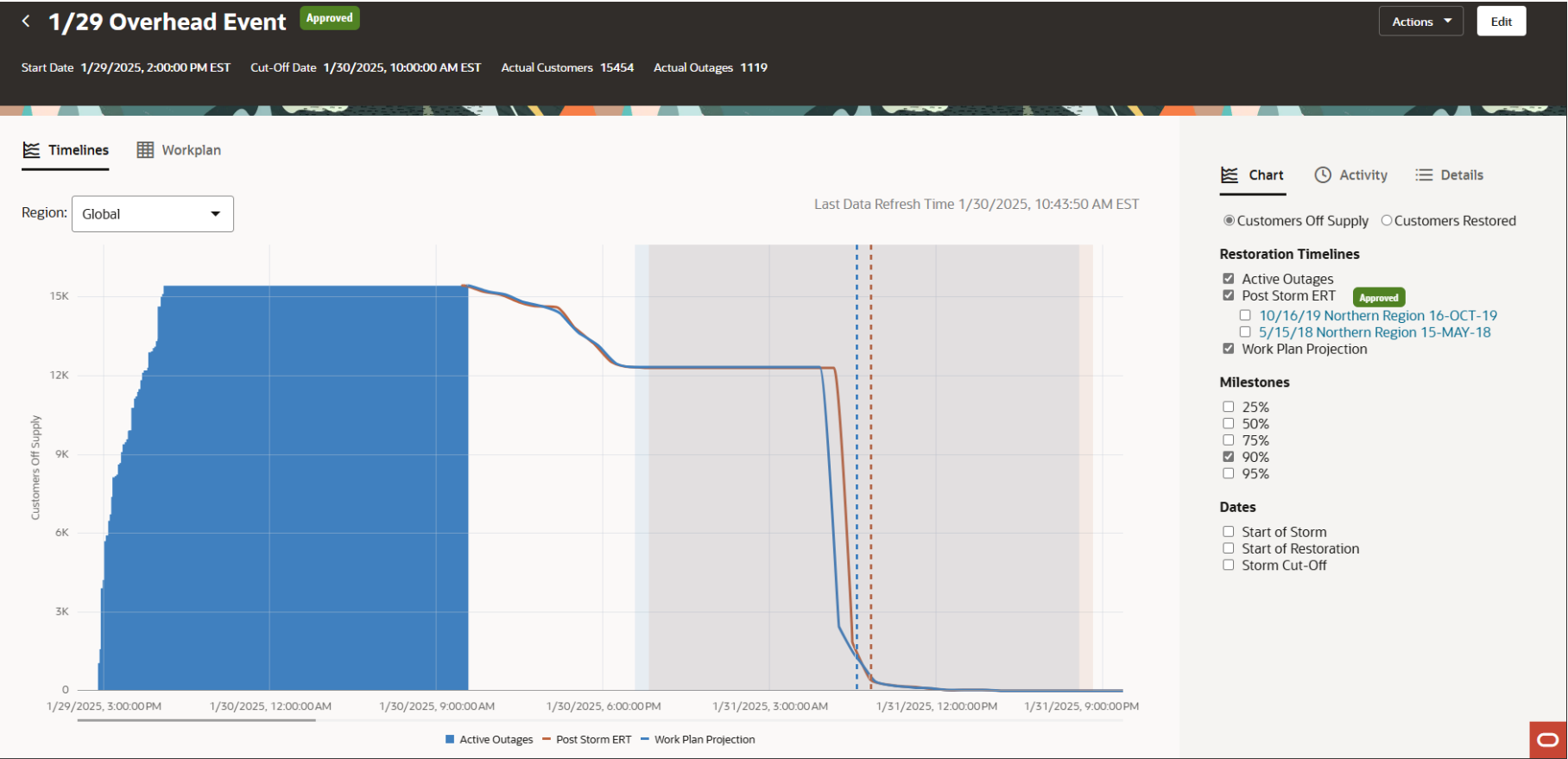




# Example Workflow: Pre-Storm Prediction



# Example Workflow: Adjustments During Restoration

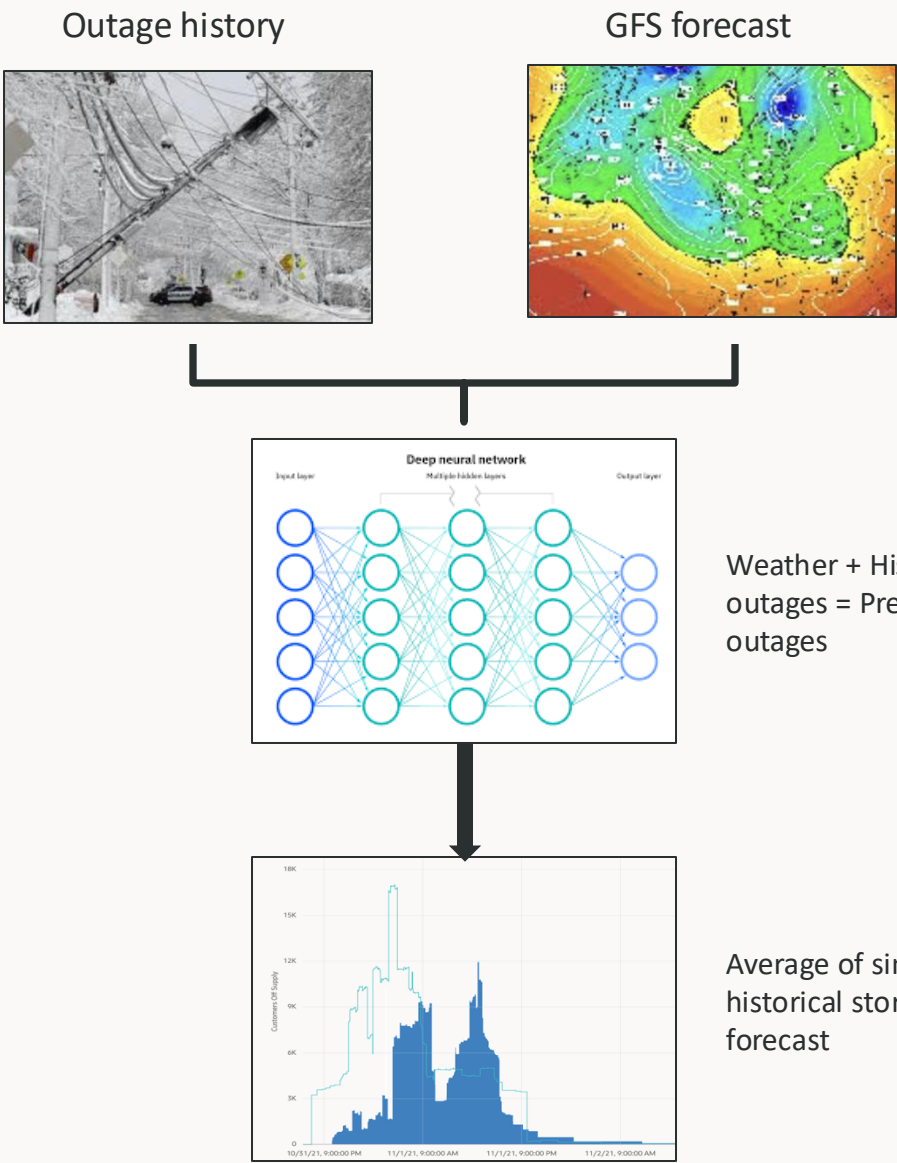




# Modeling Storm Similarity

- *Automates* similar storm selection
- “Select the top ‘n’ most-similar storms to the forecast period”

Storm name	Duration	“Distance”	Outages	Customers Count
Snow Event d011622	64	0	86	12,767
Storm Event d032921	67	39.01	176	16,210
Storm Event d04212021	55	43.49	116	8,568
Zeta d103020	55	43.65	131	16,570
Wind Storm d121121	61	44.00	96	3,298



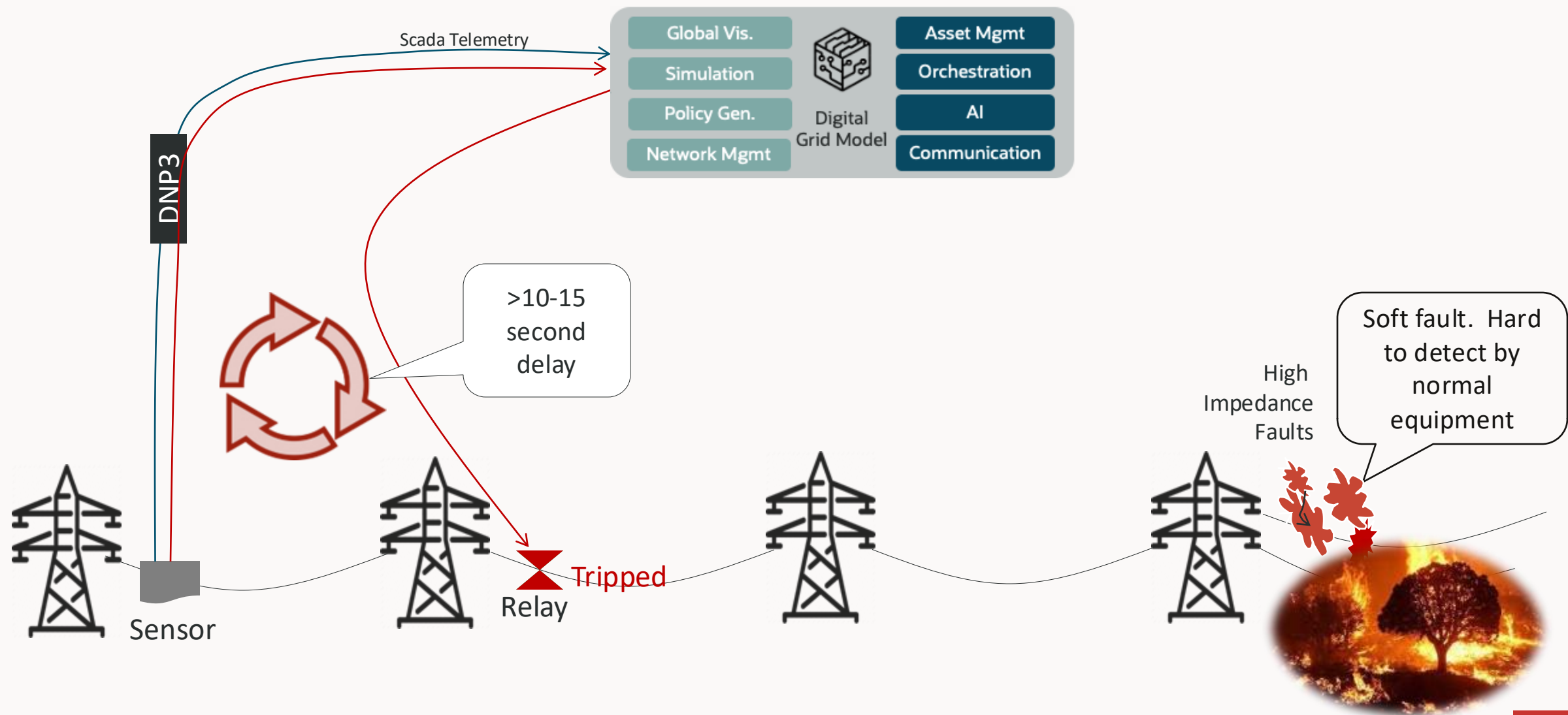
# Protect

## Active Edge Network Management



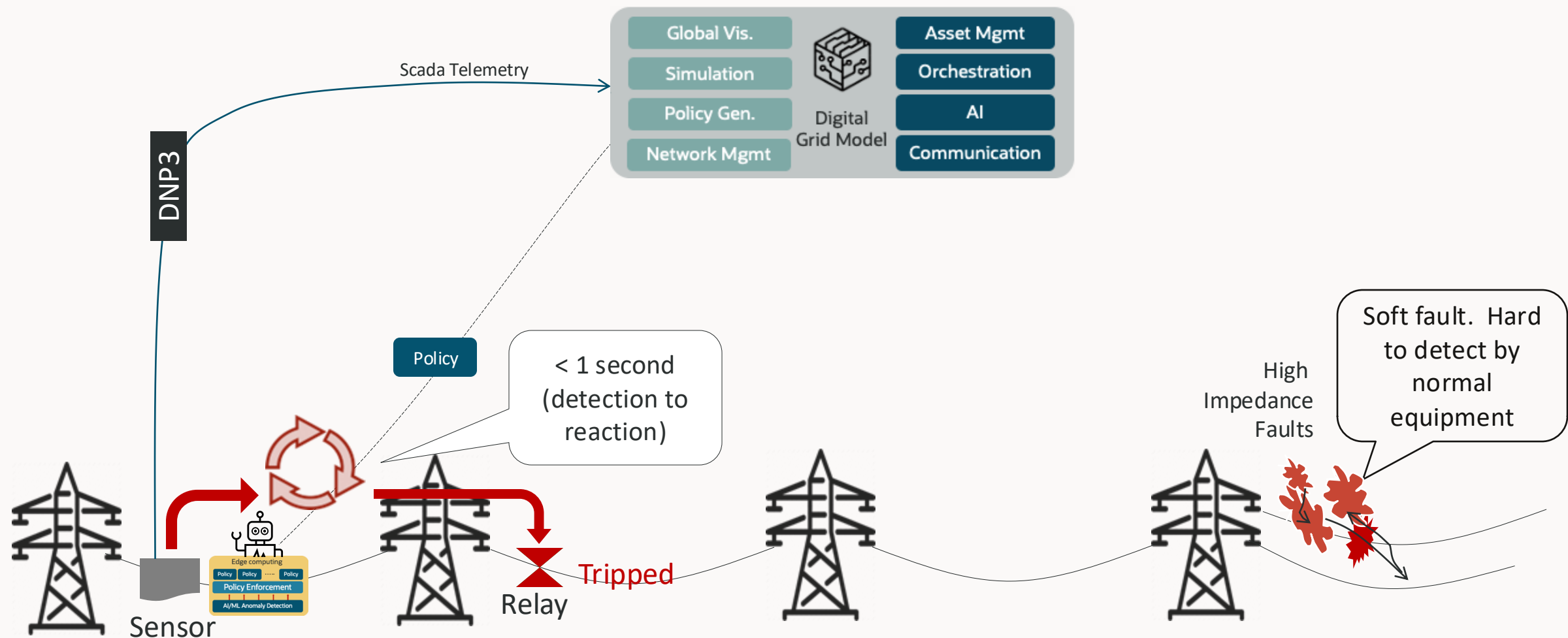
# Existing grid protection mechanism

High impedance fault & wildfire mitigation



# Existing grid protection with AENM

High impedance fault & wildfire mitigation





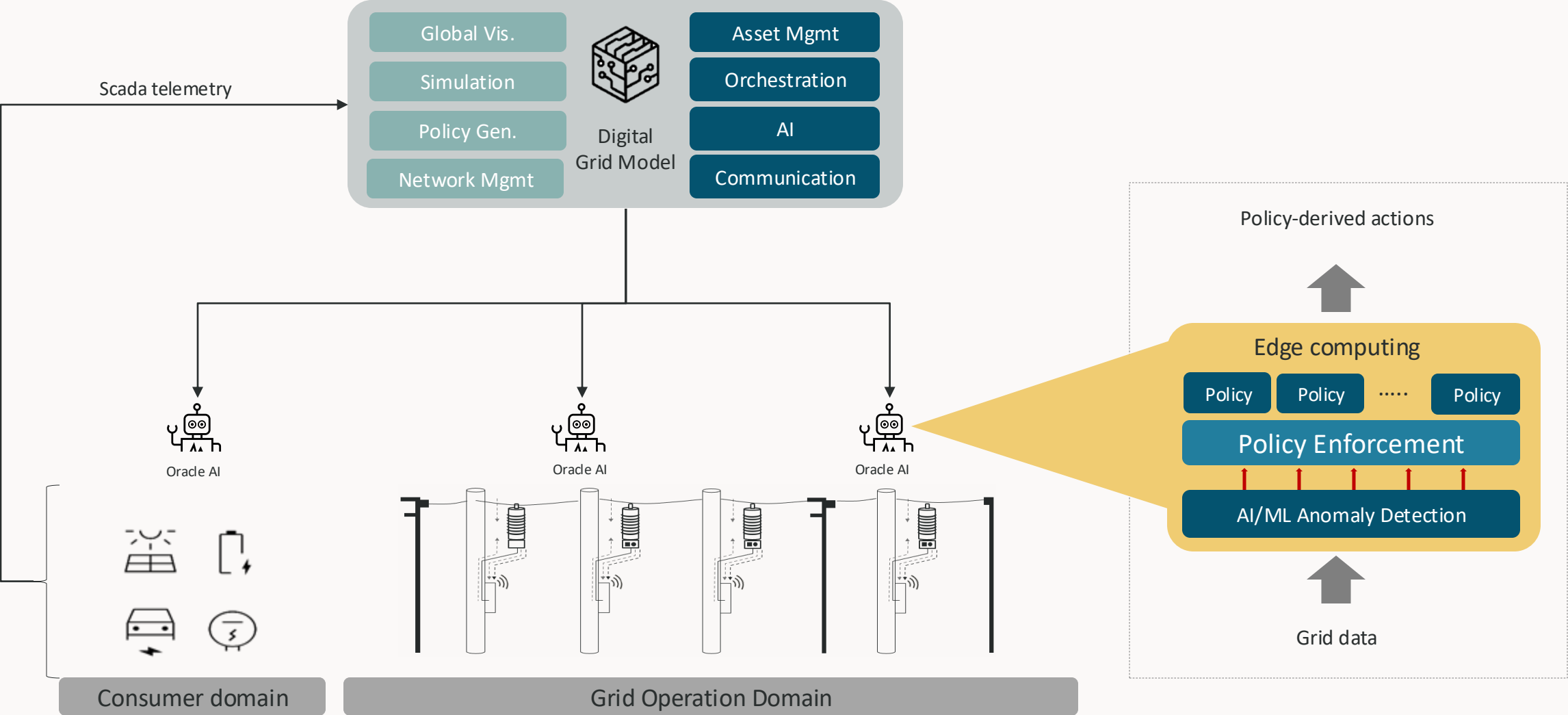
# High Impedance Fault (HIF) ML Model

- Developed a classifier to detect high impedance faults (HIFs) in power distribution networks, to prevent fire and ensure grid safety
- Data sources
  - Department of Energy’s Grid Events Signal Library (GESL)
  - Micatu sensor data (lab-generated)
- Datasets: Labeled events from DoE GESL covering a range of fault types
- The initial model is built with a binary classification model to detect whether an event is normal (label 0) or fault (label 1)
- 80% of data used for training, 20% used for testing

Event Type	Number of Samples	Label
Tripped	27	Fault
Tree Contact	23	Fault
Arcing	2	Fault
Capacitor Switching	5	Non-Fault
Motor Starting	8	Non-Fault
Load Changing	18	Non-Fault
Normal	35	Non-Fault



# X-Domain Agentic AI for Utilities





## Lab PoC with partners and customers



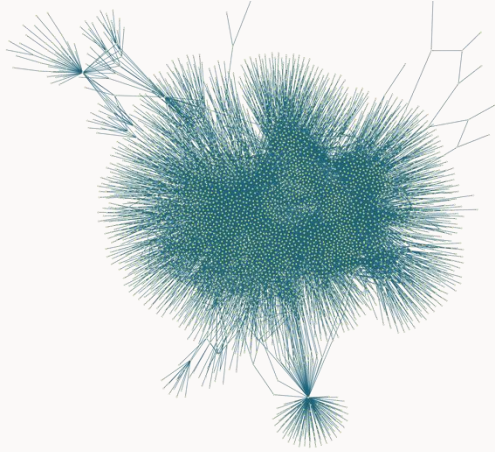
### AENM innovation paves the way for additional use cases

- ✓ Over/under frequency management
- ✓ Feeder load balancing
- ✓ Autonomous microgrid islanding
- ✓ Distributed Volt/VAr optimization
- ✓ Loss of central plants and their spinning inertia needed for grid system stability
- ✓ And many more

# Scale

## Oracle Utilities Data Exchange

# Industry Depth, Public Cloud Scale



## Ontology

connected industry  
intelligence with  
shared meaning



## Industry Data

data from utilities,  
emerging innovators, &  
broader industry data (e.g.  
EIA, NOAA, DOE)

ORACLE CLOUD  
Infrastructure



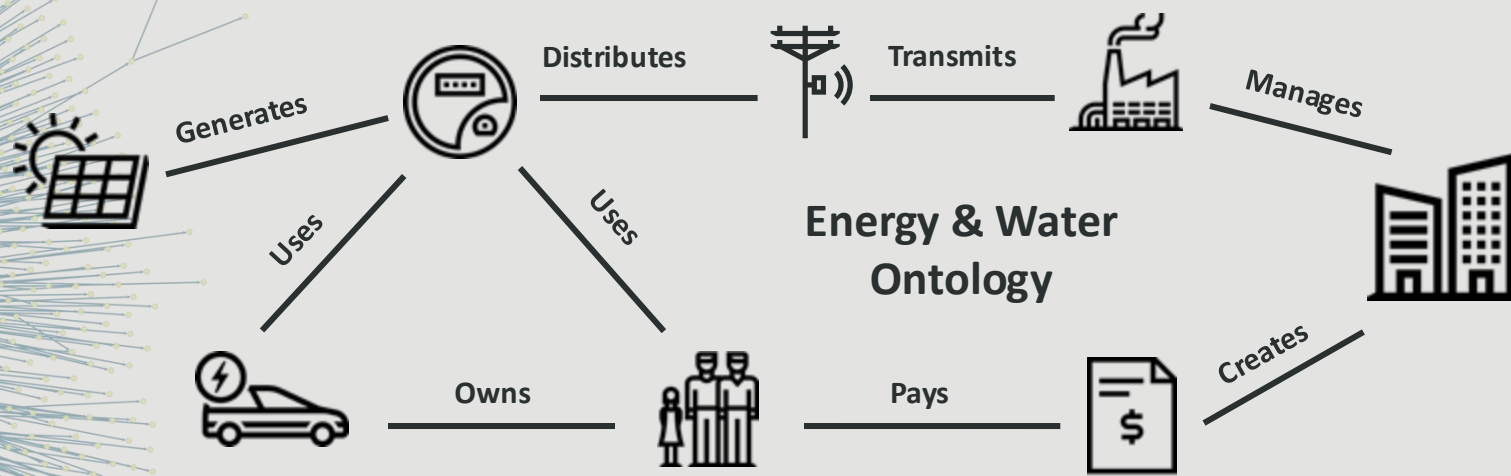
## Public Cloud

Enables velocity,  
scale, & secure  
collaboration



# From raw data to actionable data

## ORACLE Energy & Water Data Exchange



### Enterprise data

PER\_T : ID A 98234    PER\_C : F\_NAME = "Bob"  
ACCT : ID = YT0293    PER\_C : L\_NAME = "Jones"  
BILL\_P : CURR\_AMT = 360.93

### To language data for Gen AI

Bob Jones is a [person](#)  
and has a [utility bill](#)  
with an [amount due](#) of \$360.93  
with a [total account balance](#) of \$360.93.

# Industry Data

## Data from companies

### from utility companies...

Customer, Grid, DER, CX, etc

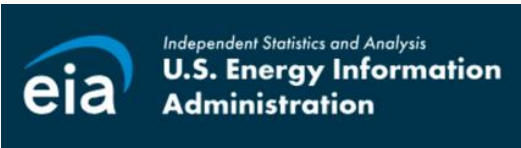


### from emerging innovators...

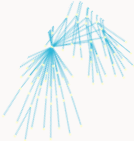
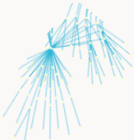
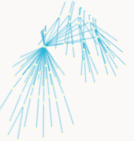
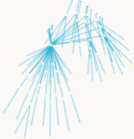
Customer, Grid, DER, CX, etc



## Data from industry organizations



# Panoramic View of Anything for AI

	Employee AI Conversations	Utility Customer AI Conversations	Smarter bulk insights
	Data prepped by Data Exchange		
	More Intelligent Interactions with Gen AI		
 <p><b>Customer E2E</b> all data connected to a customer</p>	<p><i>“Why is this customer’s bill so high? What changed?”</i></p>	<p><i>“Why is my bill higher this month than it was before?”</i></p>	<p>Calculate an overall sentiment for this customer</p>
 <p><b>Grid Circuit E2E</b> all data connected to a grid circuit</p>	<p><i>“Which customers on this grid circuit have smart thermostats?”</i></p>	<p><i>“Why am I experiencing an outage right now?”</i></p>	<p>Propose options for circuit management based on customer attributes and risks</p>
 <p><b>Asset E2E</b> all data connected to an asset</p>	<p><i>“Propose a more optimal maintenance schedule for this asset based on history”</i></p>	<p>n/a</p>	<p>Calculate asset risk based on all data and shifting conditions</p>
 <p><b>Transformer E2E</b> all data connected to a transformer</p>	<p><i>“Is this transformer at risk from rising EV adoption?”</i></p>	<p>n/a</p>	<p>Calculate transformer load scores that factor in behavior</p>





ORACLE

# Open Discussion