

**Electricity  
Transmission**

# Framework for Risk Analysis and Modelling of Events

**(FRAME)**

## Presenters

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**nationalgrid**



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# FRAME

- Electricity demand is rising, networks are more interconnected, and extreme weather is increasing.
- Ofgem's RIIO-3 framework pushes for proactive, data-driven asset risk management.
- NGET identified the need for a unified system to track events, asset health, and substation resilience.
- Current tools are fragmented across multiple data sources, limiting visibility of true network risk.
- This fragmentation slows decision-making during storms or major national events.
- Result: higher chance of missed risks, inefficient maintenance, and avoidable operational issues.

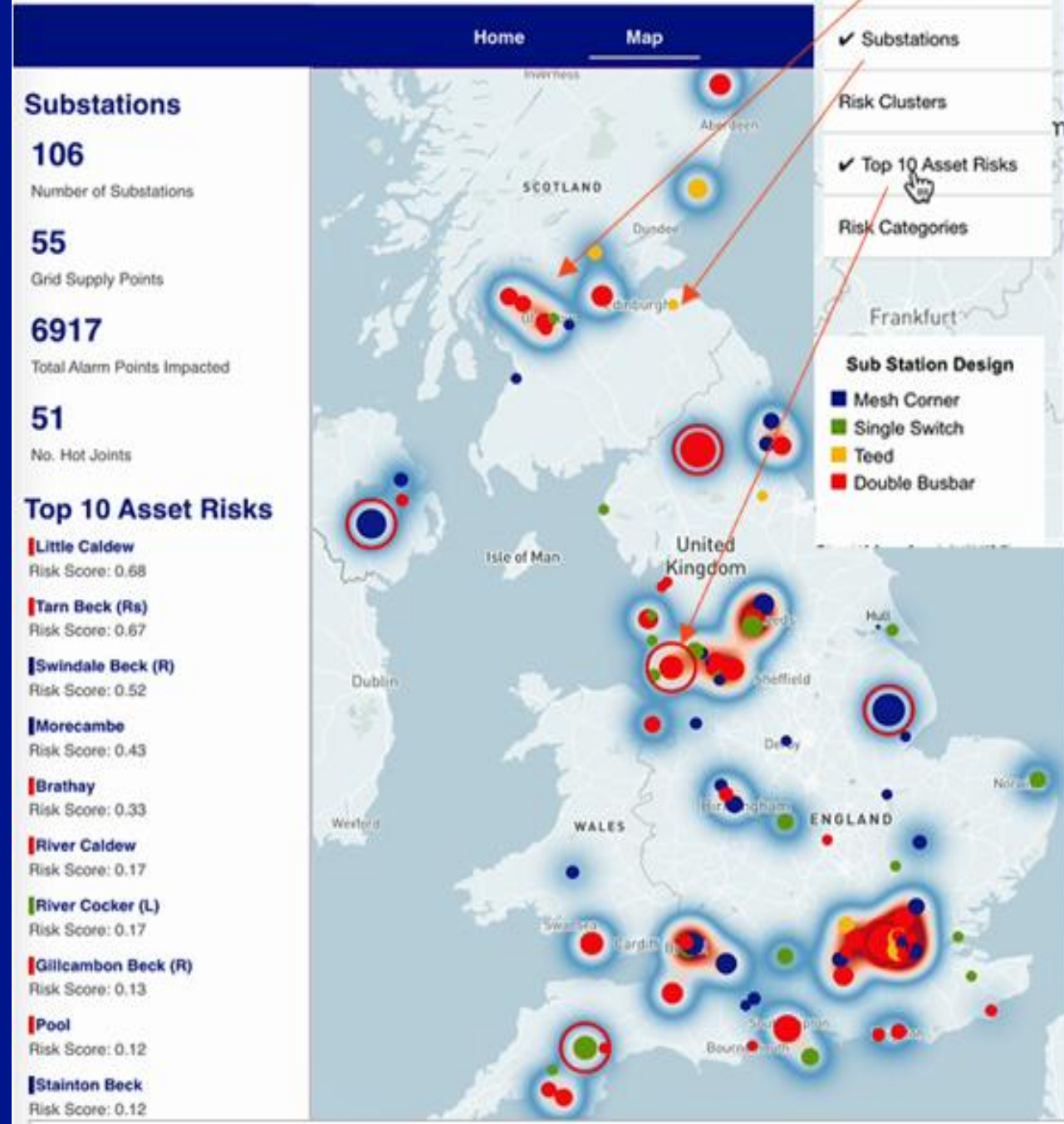
Fund : Network Innovation Allowance (NIA)

Duration : 12 Months

## Partners



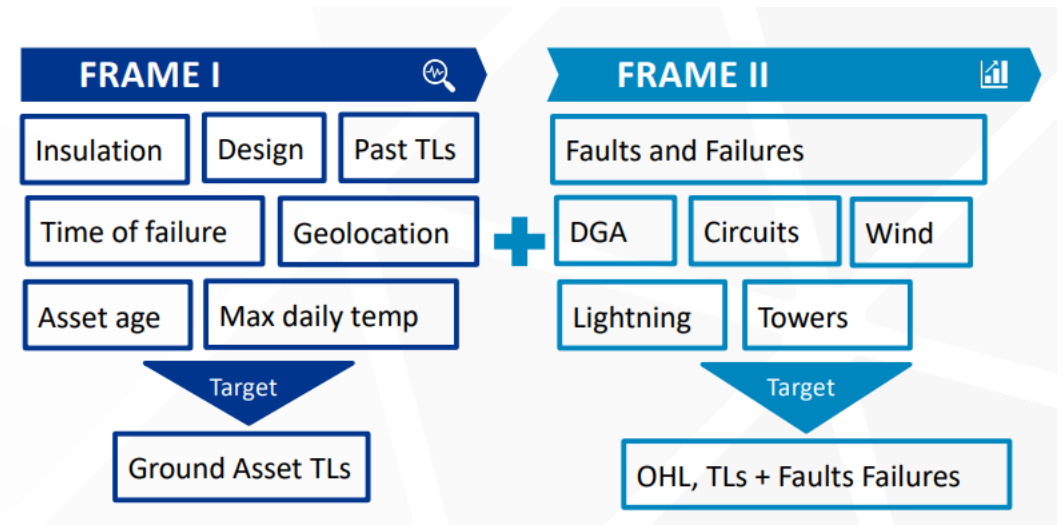
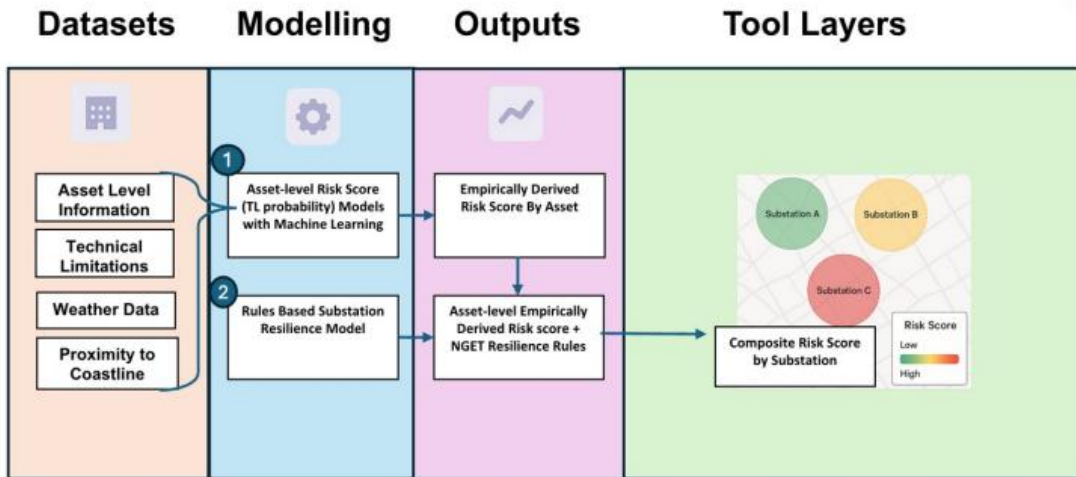
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# Summary

A mathematical risk modelling and visualisation tool (Proof of Concept) which is capable of interpreting and assessing multiple operational data sources to identify intrinsic risks/issues to operability of the transmission network based on proven systems engineering techniques.

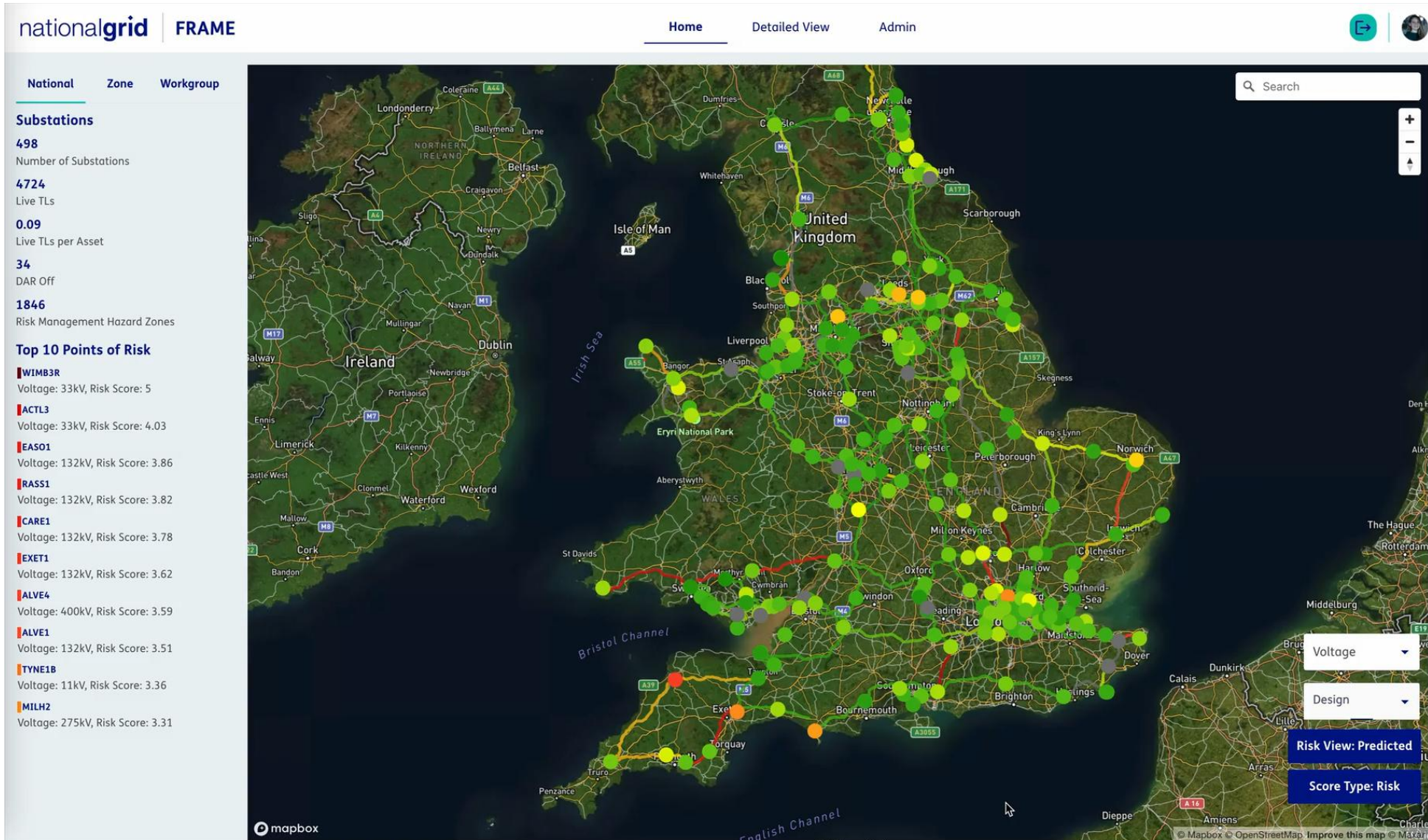
- Geospatial **Resilience Risk Score**.
- **AI** - identify trends and patterns.
- Different **zone/site/national** level holistic view.
- **Quick** identification of a worry list
- **Improved** decision-making to support CR.







# Benefits

<b>Network Resilience (Resilience and Policy Team) and TNCC (Control Room)</b>	<b>AO (Planning &amp; Performance managers, lead ACEs, operational managers):</b>	<b>NOI and AO management</b>
<b>Unified Map View &amp; Real-Time Risk Visualization</b>		
<b>Trend Identification</b>	<b>Issue Identification</b>	<b>Comprehensive Data Source</b>
<b>Resilience Risk Scoring</b>	<b>Justification for Investment</b>	<b>Proactive Storm Response</b>
<b>Collaboration</b>	<b>Optimized Resource Allocation / planned work</b>	<b>Cost Reduction</b>

# Demonstration

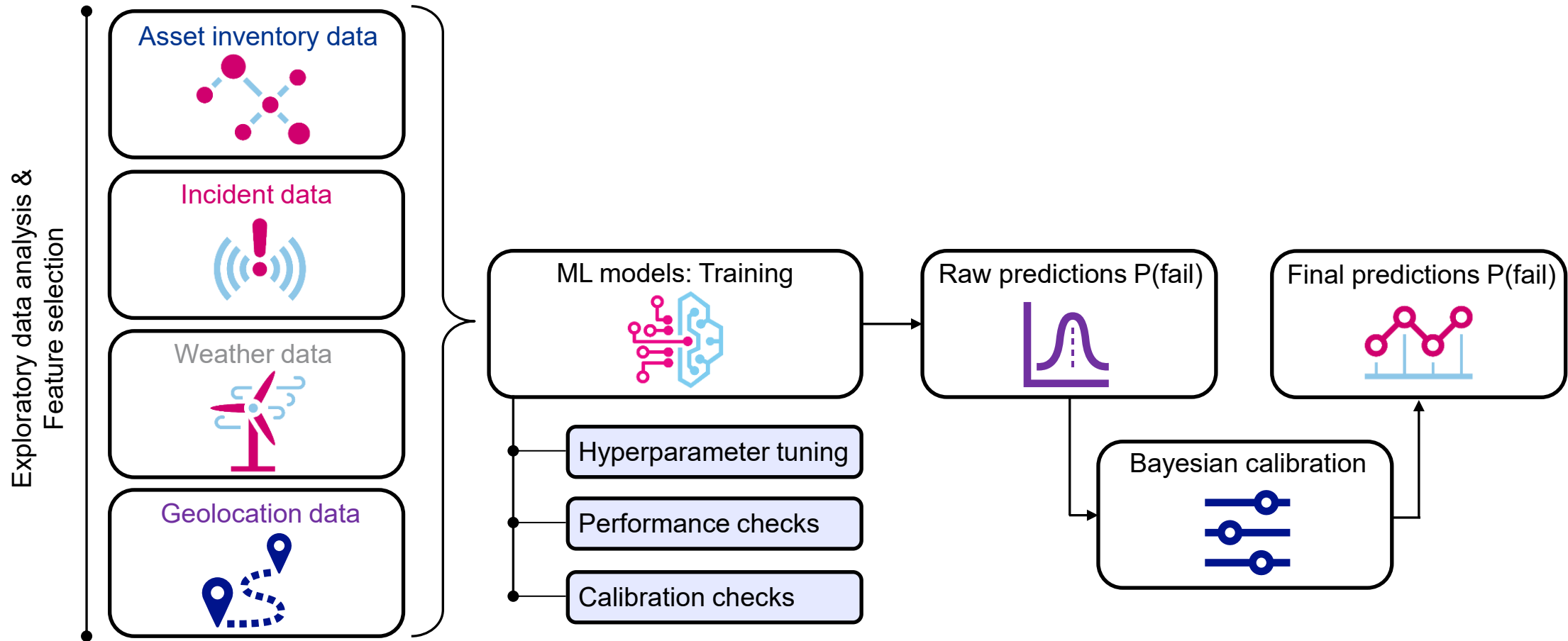


# Methodology – Datasets and features explored

Data Type	Ground assets	OHL assets
<b>Asset inventory data</b> 	<ul style="list-style-type: none"><li>• Asset type</li><li>• Age</li><li>• Manufacturer</li><li>• Insulation, design</li></ul>	<ul style="list-style-type: none"><li>• Conductor type</li><li>• Age</li><li>• Length</li><li>• Line greasing, operating voltage</li></ul>
<b>Incident data</b> 	<ul style="list-style-type: none"><li>• Technical limitations, seasonality</li><li>• Faults &amp; Failures, seasonality</li><li>• Frequency of historical incidents</li><li>• Maintenance events</li></ul>	<ul style="list-style-type: none"><li>• Faults &amp; Failures, seasonality</li></ul>
<b>Weather data</b> 	<ul style="list-style-type: none"><li>• Wind speeds</li><li>• Rainfall, Snowfall</li><li>• Temperature</li><li>• Lightning (CAPE, K-index)</li></ul>	
<b>Geolocation data</b> 	<ul style="list-style-type: none"><li>• Latitude, Longitude</li><li>• Proximity to industrial sites</li><li>• Level of emissions</li><li>• Distance to coastline</li></ul>	

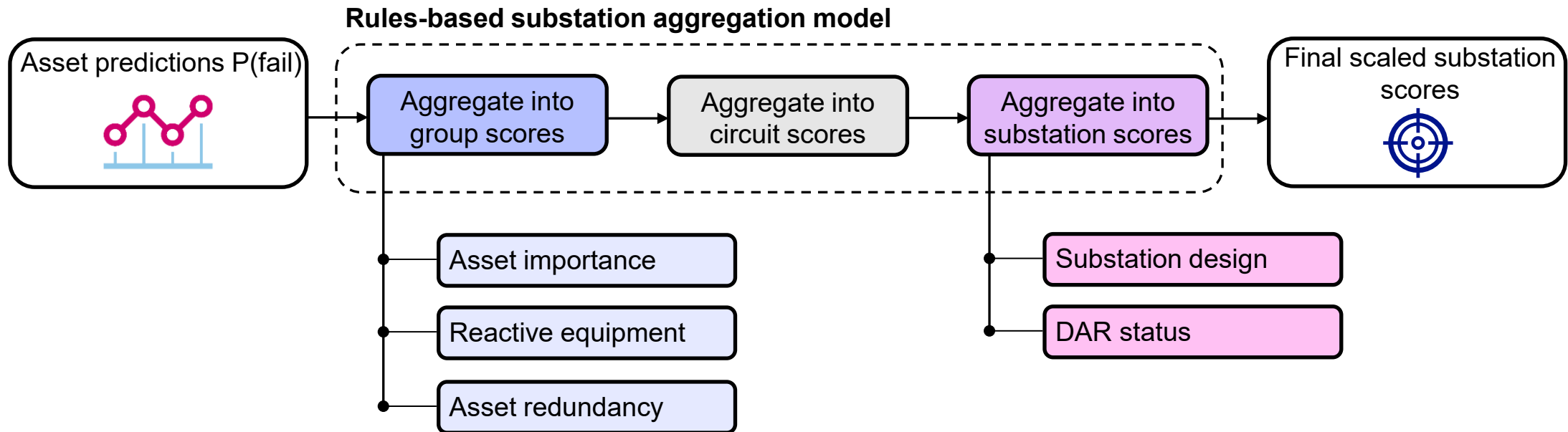
# Methodology – Asset risk modelling

**Asset  $P(\text{fail})$ :** ML models (supervised learning) are trained to predict probability of TLs/Faults & Failures across assets. Ground assets and OHL are modelled separately, using distinct binary classifier models. Predictions are generated at monthly granularity for ground assets and at weekly granularity for OHL (more exposure to weather hazards).



# Methodology – Substation risk modelling

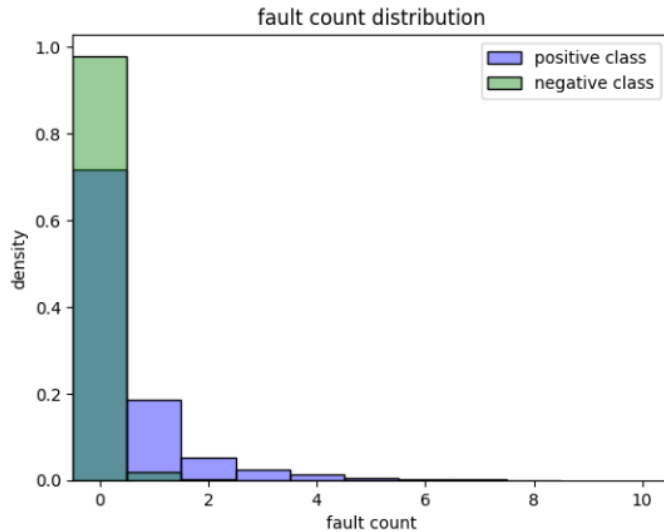
**Substation risk score:** Ground assets P(fail) predictions from ML models are aggregated into substation risk scores using a rules-based model, incorporating SME knowledge. The final scaled substation risk scores are bounded between 1-5 (1 corresponding to low risk, and 5 to high risk).



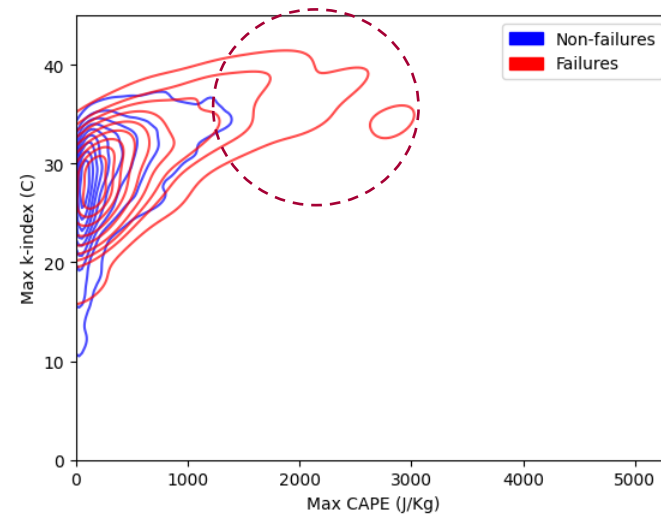
# Methodology – Key observations

## Notable risk drivers:

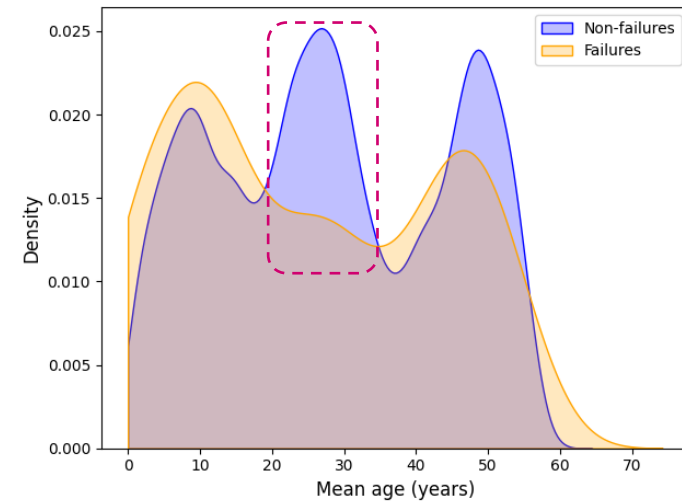
- Frequency of past incidents (historic TLs or faults/failures) and asset age for ground assets.
- Lightning risk (CAPE, K-index), wind gust speed, precipitation, conductor length and age for OHL.



**Fig 1** - Positive (fault) and negative (non-fault) histograms vs past number of faults



**Fig 2** – Bivariate distributions of fault and non-fault OHL vs lightning metrics (CAPE and K-index)



**Fig 3** – Distributions of failure and non-failure of OHL vs conductor age (in years)

## Modelling challenges/considerations:

- Extreme class imbalance impacts model training. Re-balancing improves this but impacts precision and increases false-positive rates. Additional Bayesian calibration adjusts predicted P(fail)s to match operational conditions.
- Inclusion of additional asset health features, increased TLs/faults & failures data timespans and maintenance records could improve model performance.

# Recap

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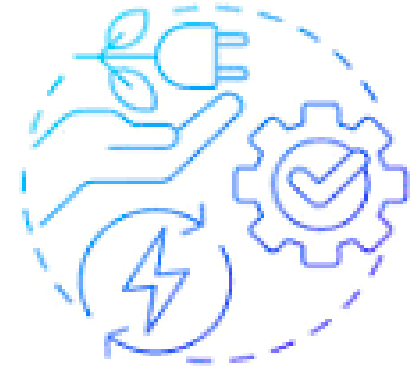
## FRAME

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## Innovative Solution

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## Benefits



RESILIENCE  
BENEFITS

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