



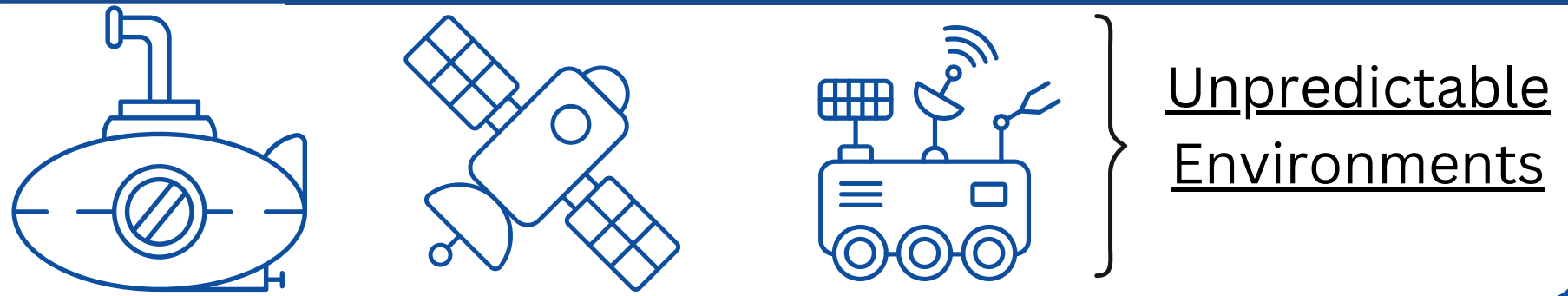
Reliability Estimation of Adaptive Distributed Embedded Systems Using GATs

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A system with these qualities can be defined as an ADES:

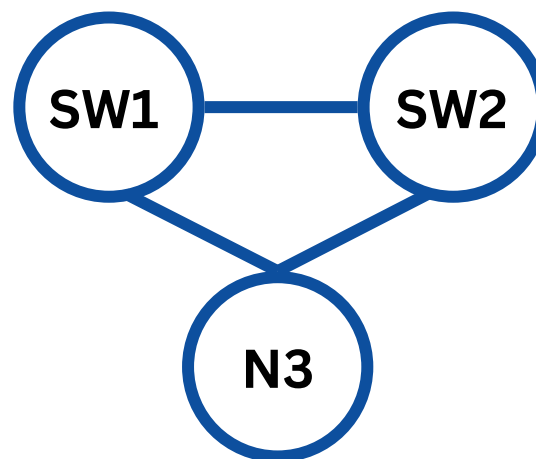
Distributed	Embedded	Adaptive
Made up of multiple components that communicate and share information.	Integrated into a larger system to perform specific functions.	Able to adjust and reorganize itself by managing its components.

Use Cases for ADES



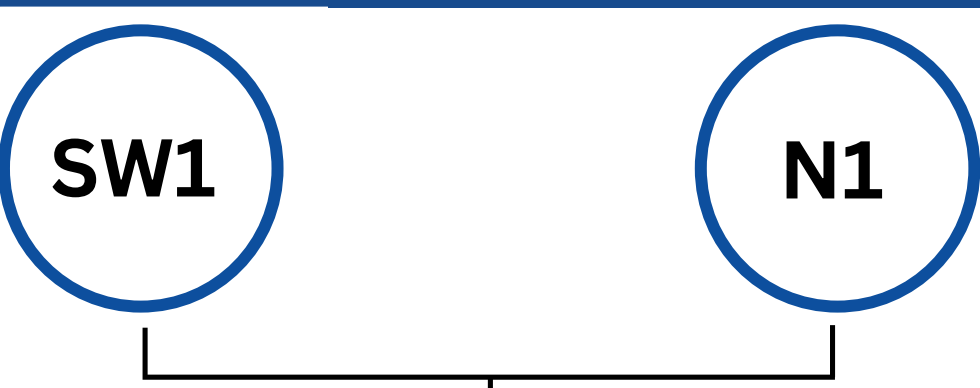
Dataset

The dataset comprises an initial fully connected graph (3SW/5N) and **all possible degraded configurations**. Every data entry has an adjacency matrix, timestamp, and reliability value. There are **86,760 entries** in the range of 0 to 8500 hours.



Degraded Graph at 1000 hours

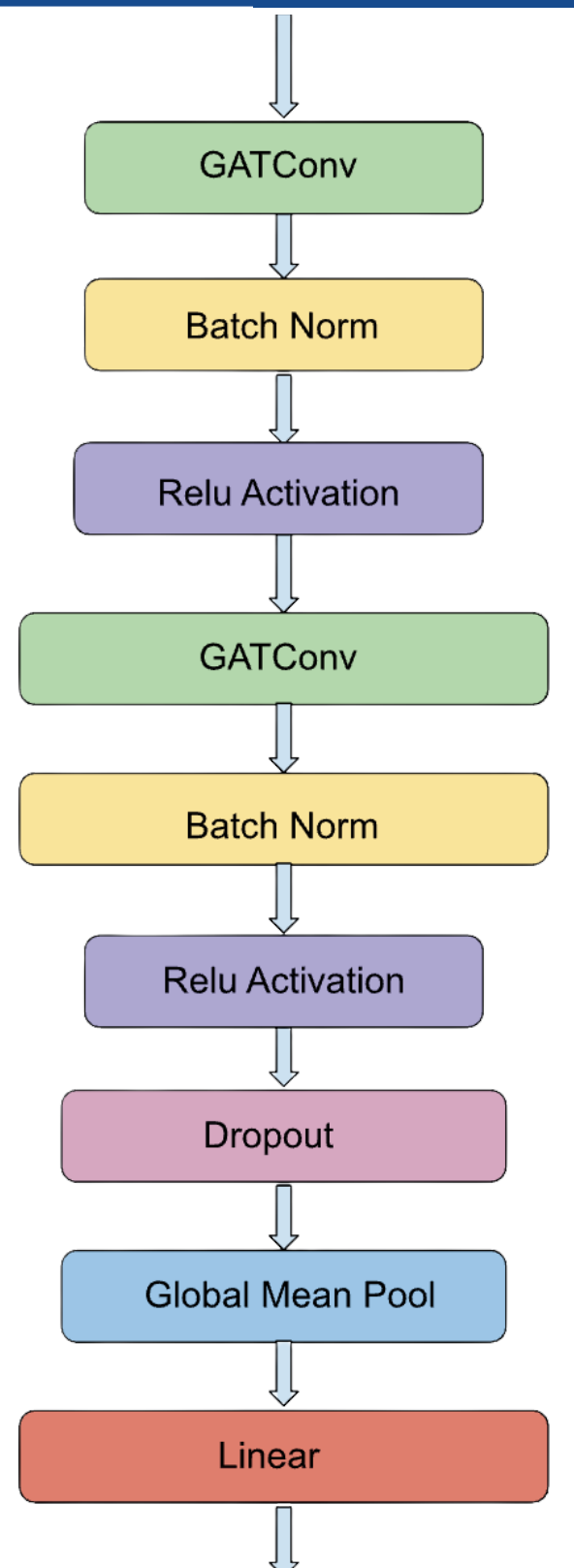
Vertex Features



- Centrality Measures
- K-core number
- Degree
- Mean Failure rate of neighboring vertices
- Count per vertex type

The following values were added to each vertex of the graph. They were selected based on their high correlation to the target variable out of 20 features.

Single GAT Model



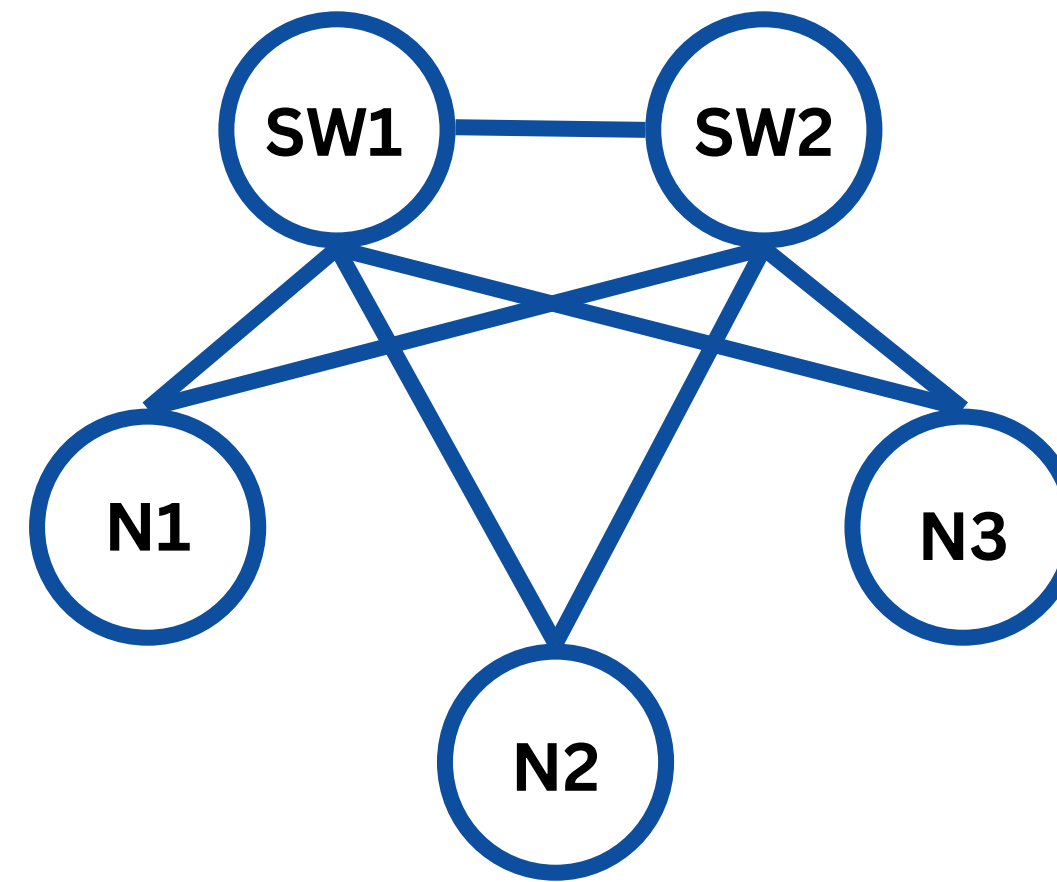
How are ADES represented in our project?

As fully connected graphs

Switches (SW)
Coordinate node communication.

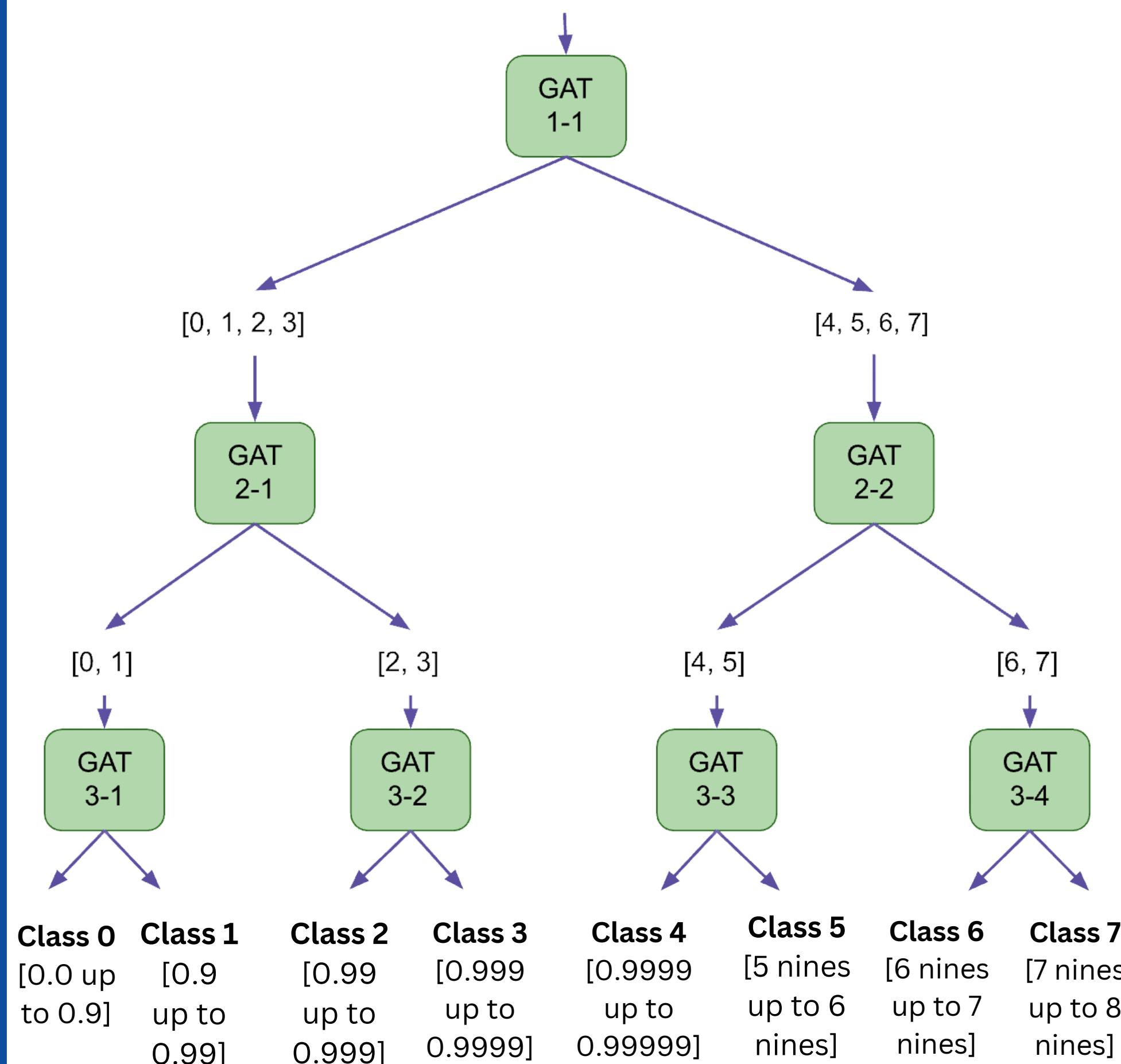
Nodes (N)
Execute designated functions and communicate output to switches.

Links
Physical connections between components.



Model Architecture

Each sample was assigned a class based on the number of 9's in its reliability value. Seven binary classification Graph Attention Networks (GAT) were arranged in a binary tree composition. The ensemble allows the model to narrow down the reliability class, with each level of the tree handling a more refined classification task.



Problems with ADES

Any component may fail unexpectedly. While the system's adaptive nature **allows reconfigurations** to address this, it introduces a new problem: **each reconfiguration requires a reliability re-estimation.**

The time required for this makes them **impractical for critical scenarios**, as a new configuration cannot be applied **without ensuring it meets the system's reliability requirements.**

Project Motivation

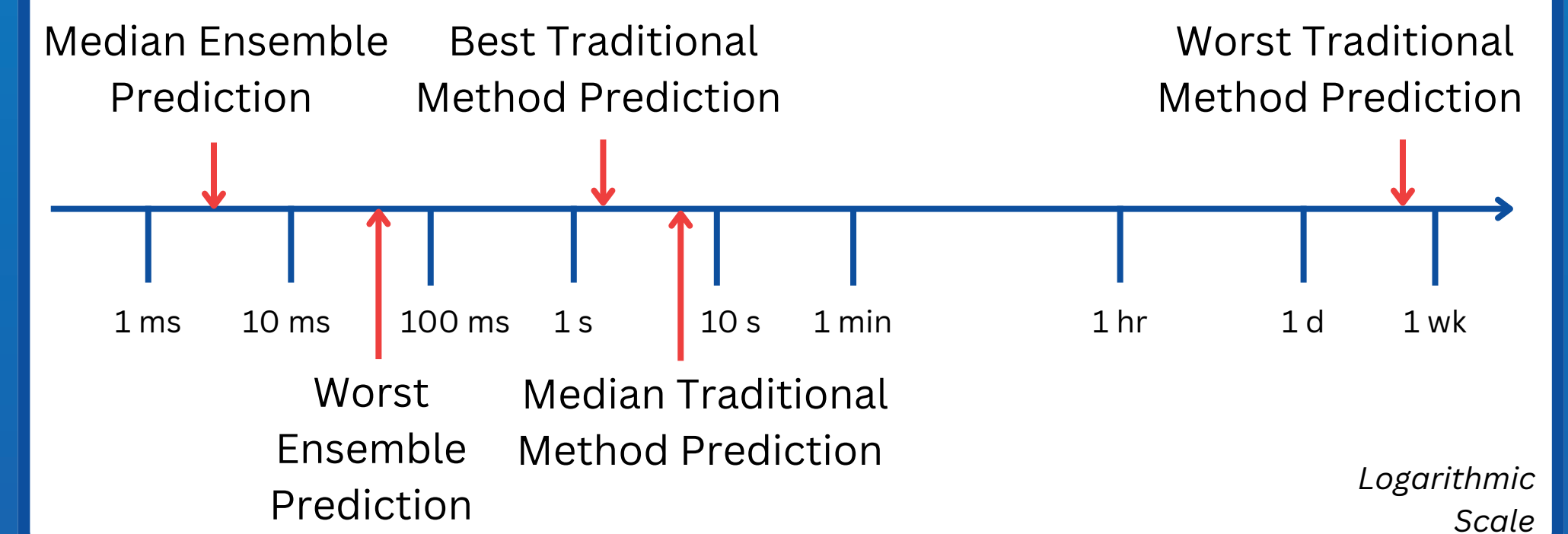
Explore machine learning solutions to estimate the reliability of ADES faster than traditional methods.

Model Evaluation

True label \ Predicted label	0	1	2	3	4	5	6	7
0	273	5	0	0	0	0	0	0
1	57	1586	91	0	0	0	0	0
2	0	357	3056	204	0	0	0	0
3	0	14	515	3436	402	49	5	0
4	0	0	14	231	1773	149	19	1
5	0	0	1	124	244	1665	232	0
6	0	0	3	2	32	224	1619	83
7	0	0	0	0	7	4	73	815

The matrix on the left shows the model got **82%** of samples correct, with only **1.8%** more than one class off.

The plot below compares the inference times of the ensemble and traditional methods. Our model is **several orders of magnitude faster.**



Conclusion

The inference time of our model was, on average, **three orders of magnitude quicker** than conventional methods. The reliability estimation based on a graph and a timestamp achieved an **accuracy of 82%**. This supports the notion that Graph Attention Networks (GATs) can be **effectively utilized in this area of research.**