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Centro Nazionale di Ricerca in HPC,
Big Data and Quantum Computing



Centro Nazionale di Ricerca in HPC,
Big Data and Quantum Computing

Predictive Maintenance

R.Cornali for INFN-CNAF, INAF, UniBO, ENI

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The background is a deep blue gradient. On the left side, there are numerous bright blue light trails and particles that appear to be moving towards the center, creating a sense of depth and motion. The trails are composed of many small, bright blue dots connected by thin lines, resembling a data stream or a network of connections. The overall effect is futuristic and technological.

Predictive maintenance

What is predictive maintenance?

- **Equipment maintenance** that leverages data analytics and condition monitoring.
- Traditional maintenance strategies rely on scheduled inspections and reactive interventions.
- Predict when maintenance is needed to schedule interventions at the most convenient times.
- Advancements in technologies such as Internet of Things (IoT) and machine learning allowed this innovation.

What is the goal of predictive maintenance?

How?

- Data are collected from the various sensors installed on the machinery.
- Data are continuously analyzed to **identify patterns and anomalies**.

The main objectives are:

- Optimize the performance of equipments.
- Minimize the unplanned downtime and maintenance costs.

Research Framework

Innovation Grant: *"Harnessing the Power of Artificial Intelligence for Predictive Maintenance of Industrial Plants"*

Goals:

- Demonstrate the usability of AI-based techniques for predictive maintenance
- Modelling interdependencies in complex industrial apparatuses

Partners:

- ENI, UNIBO, INAF, INFN



Case studies

- **Case study 1:** We analyze signals from industrial rotating machines. Our aim is to predict **anomalous instances**.
- **Case study 2:** We estimate the output of a liquefaction unit given a subset of the operational parameters. Then, we try to **exploit data to improve the production process**.

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Case study 1: INAF

Hybrid clustering method

- Data are pre-processed and clustered through a multistep pipeline.
- The cluster labels are treated as features, and used to augment the data. This process seems promising in improving the robustness of regression and classification methods.
- Different classification models are tested to classify the data between "standard condition" and "anomaly".



Results

- Adding the clustering features improved accuracy on average by 4.87%.
- Pre-clustering reduced the training time of the classifiers, but the reduction wasn't statistically significant.
- Overall, data pre-clustering improved the classification results.

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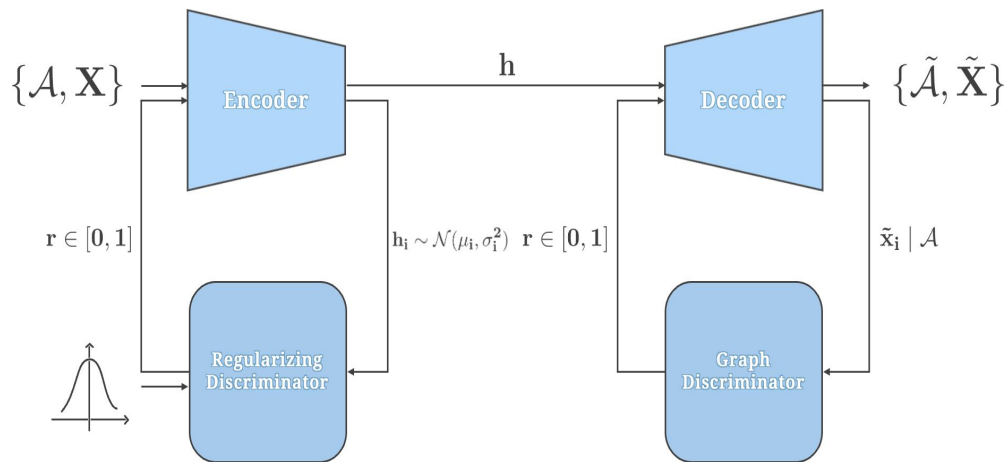
Case study 1: INFN-CNAF

Graph Neural Networks and GANs

- The method is based on the “**reconstruction error** framework” for anomaly detection.
- Time and spatial relationships in time series are modeled with a **static graph**. This choice can improve interpretability and model effectiveness.
- **Generative Adversarial Networks (GANs)** are employed to obtain robust, regularized reconstructions.

Proposed model

- The model has three core elements: a graph autoencoder and two critics.
- The graph encoder maps each node to a latent representation, the decoder reconstructs the nodes and edges from the latent representations.
- One discriminator regularizes the latent space.
- One discriminator judges the truthfulness of the reconstructed graphs.

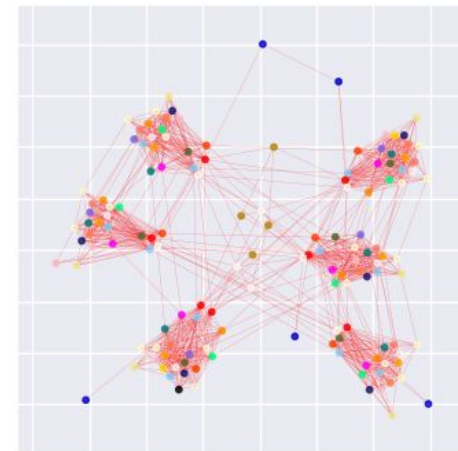
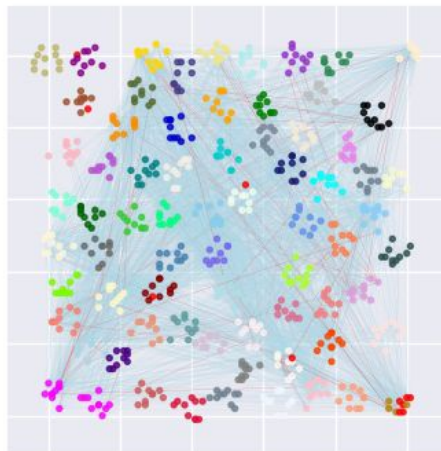


Reconstruction errors and critic scores are used as anomaly measures.

Open questions

The model has promising results, there are some open issues:

- Implement a general, effective, and efficient graph structure.
- Find an adequate thresholding and aggregation rule for the anomaly scores.



A graph representing a small time-interval, and a subgraph of anomalous connections.

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Case study 2: INFN-CNAF

Open problems of case study 2

Challenges:

- Exploit data from a LNG liquefaction plant to improve the choices of operational parameters.
- Reduce the risks associated to the use of a machine learning algorithm in a control setting.

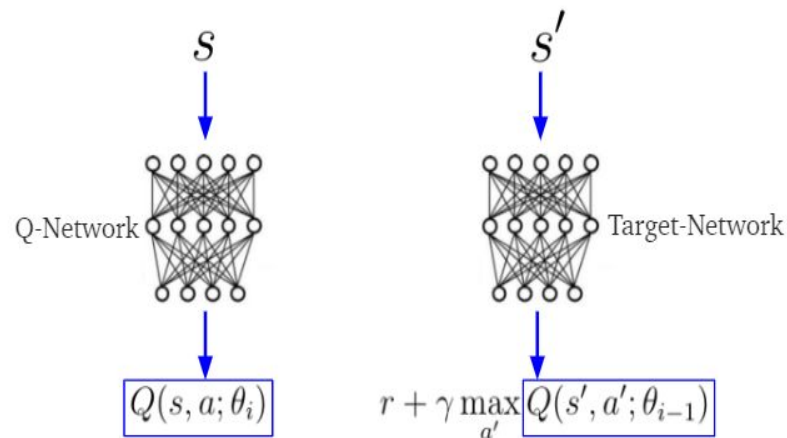


A gas liquefaction unit.

Reinforcement Learning solutions

- Implement off-policy reinforcement learning to find a suboptimal policy.
- **Deep Q-learning** finds a discrete policy.
- Soft Actor-Critic finds a continuous policy.
- Implement algorithms to improve the data efficiency.
- Set policy constraints.

Deep Q-learning.





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Thank you for the attention! Any
questions?

*Supercomputing
shaping the future*