

Paper Title : Prodigy-Towards Unsupervised Anomaly Detection in Production HPC Systems

Paper Link:

<https://dl.acm.org/doi/10.1145/3581784.3607076>

1. Summary

1.1 Motivation : An HPC is built for process, store, and analyze massive amounts of data where sometimes other ML-based frameworks require extensive customization. The solution is the new autoencoder-based anomaly detection framework.

1.2 Contribution : The following variational autoencoder-based anomaly detection framework was accurate 88% in detecting anomalies.

1.3 Methodology : Identify the problem with anomalies in HPC systems. The efficiency for the application performance. Measure key metrics such as recall and F1-score to assess the accuracy of prodigy in detecting performance anomalies. To collect data for applying in real world HPC environment.

1.4 Conclusion : An ease of integration with detailed analysis capabilities position Prodigy as a promising solution for enhancing the efficiency of the real-world HPC environment.

2. Limitation

2.1 First Limitation : For better outcome it only depends on quality and representativeness. false and biased dataset may affect its ability to give a diverse HPC environment.

2.2 Second Limitation : It might need updating on a regular daily basis for optimal performance because it might have some limitations in evolving HPC technologies.

3. Synthesis : Prodigy is an autoencoder-based anomaly detection framework for addressing performance issues in HPC. It's an powerful solution for enhancing the performance of production HPC environments.