

# INTELLIGENCE IN ACTION: AI-DRIVEN NETWORKS

## M2. Project inception and lifecycle architecture

Projeto em Informática 2024/2025

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Mar 2025

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# NWDAF

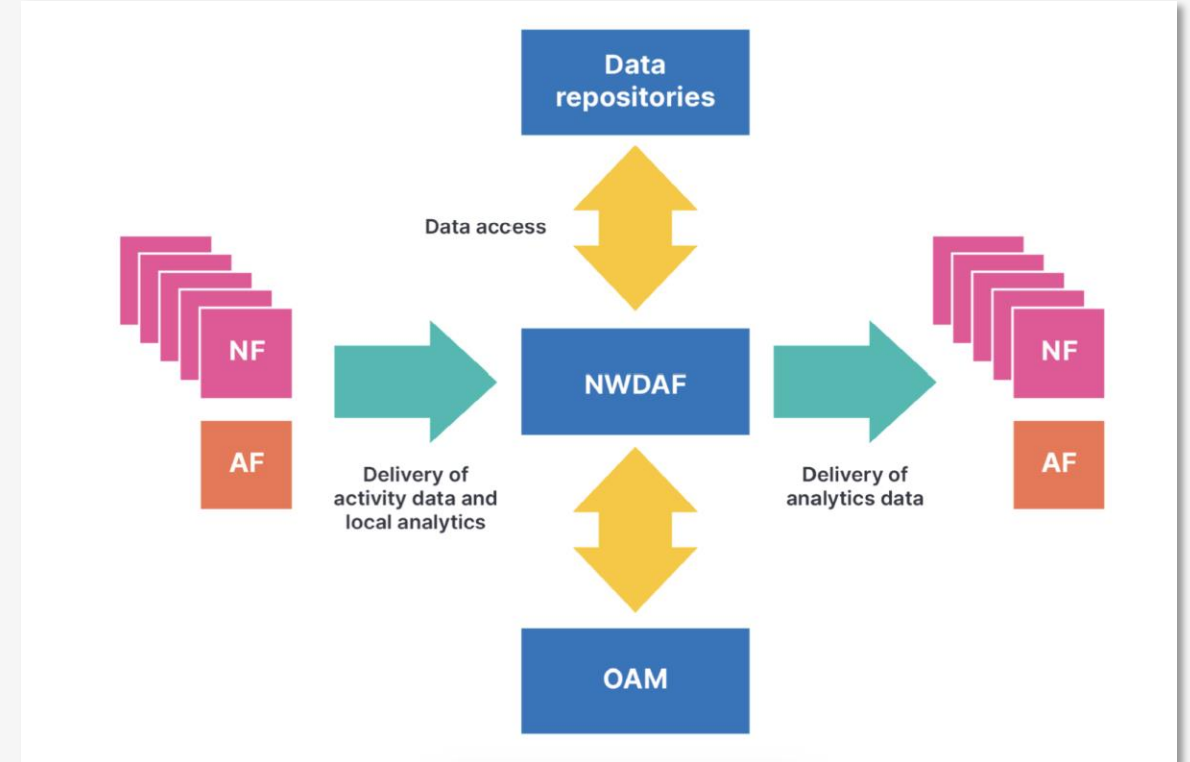
Network Data Analytics Function

- collecting and analyzing network data
- provide predictions for network optimization

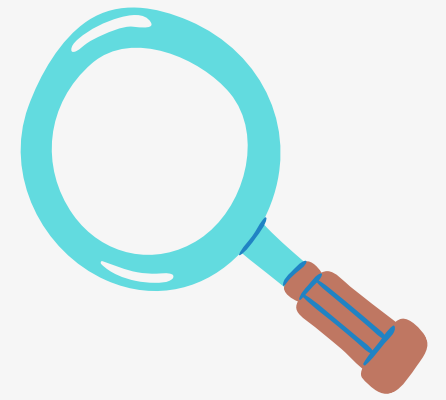
Has three main aspects:

- Data Collection
- Analytics Processing
- Analytics Exposure

**Goal:** Automating the 5G network with machine learning and data analytics



## 2. State of Art



- Machine learning was employed to detect abnormal traffic patterns and potential DDoS attacks. [1]
- The ability of NWDAF to collect data from a 5G network and detect abnormal traffic. [1]
- Neural networks perform better in network load prediction than linear regression. [2]
- The load data of network components can be used for anomaly detection. [3]

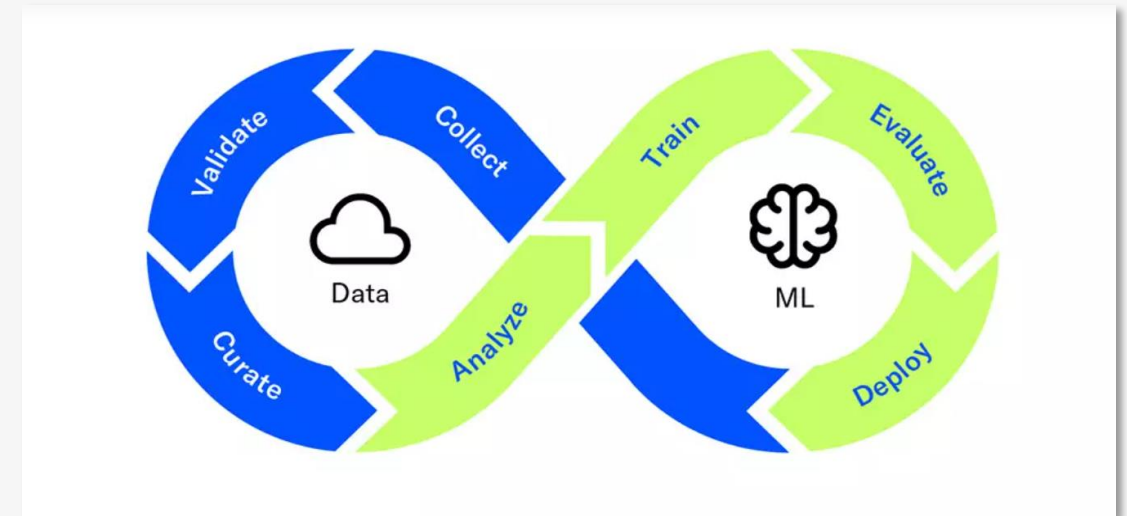
# MLOps Pipeline

MLOps is an extension of DevOps, specifically adapted for machine learning workflows.

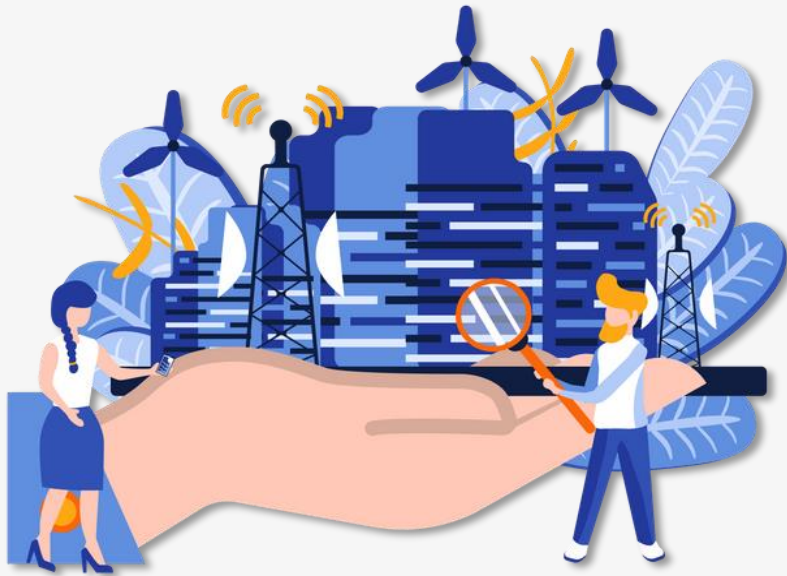
End-to-end machine learning development process.

Aims to unify the release cycle for machine learning.

Enables the application of agile principles to machine learning projects.



### 3. Actors

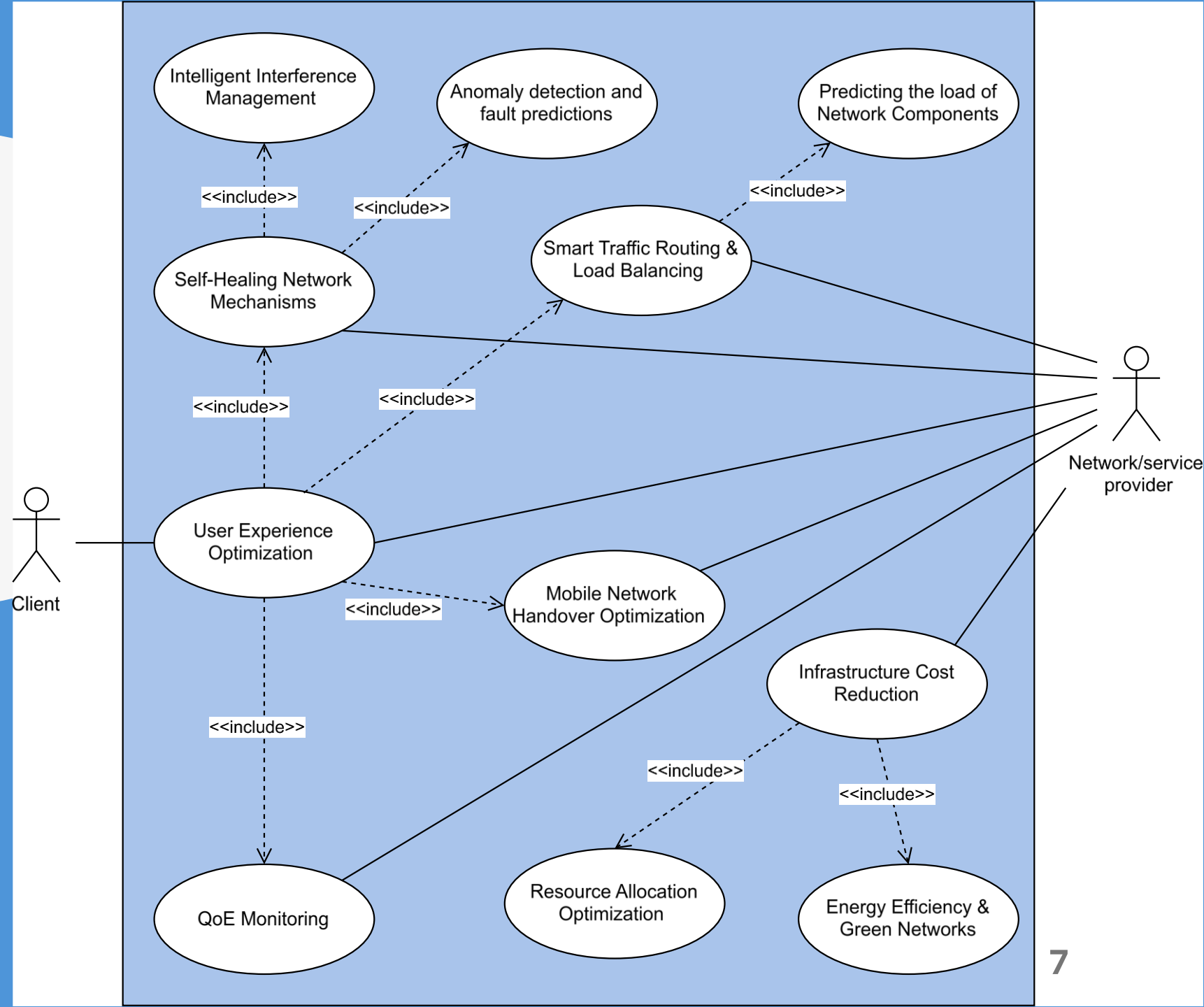


**Network/service provider**



**Service Client**

# 4. Use Cases





# UC: Anomaly Detection & Fault Prediction



Detect abnormal patterns in network behavior.



Predict failures before they occur.



Provide actionable recommendations to prevent service disruptions.



**Real-life scenario:**  
A smart manufacturing facility.

# 5. Requirements Elicitation

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Telecommunications and  
5G research



Machine Learning  
research



State-of-art



Team brainstorming

# Functional Requirements

- Data Collection
- Model
- Monitoring Dashboard
- External Integration



The system must...

# Data Collection

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- Support JSON and CSV data formats.
- Collect and store raw data in a time series data base.
- Perform data pre-processing (cleaning, normalization, and aggregation).



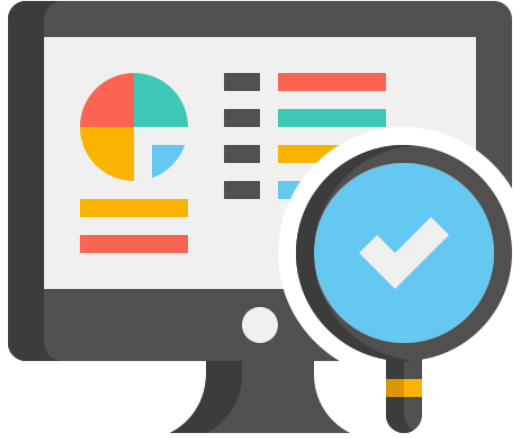
# Model

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- Allow model training with stored data.
- Implement automation in model training, supporting continuous re-training based on new data or data drift.
- Validate and test the models using the obtained metrics.
- Continuous model deployment mechanisms.



## Monitoring Dashboard



- Show anomaly alerts.
- Show relevant metrics,

## External Integration



- Provide APIs so that external functions can subscribe to events.
- Enable compliance with 5G architecture standards.

# Non-Functional Requirements

- Scalability
- Interoperability
- Performance
- Security
- Maintainability



# Non-Functional Requirements

## Scalability

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- The system must efficiently process large volumes of data with a maximum processing latency of 100 milliseconds.

## Interoperability

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- The system must be interoperable, making available APIs and adopting machine learning frameworks.



# Non-Functional Requirements

## Performance

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- To support real-time analytics, data processing should have minimal latency and a response time inferior to 1 millisecond.
- ML inference APIs should provide responses within 1 millisecond for fast decision-making.

## Security

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- The system must be GDPR compliant and keep all data on-premise.

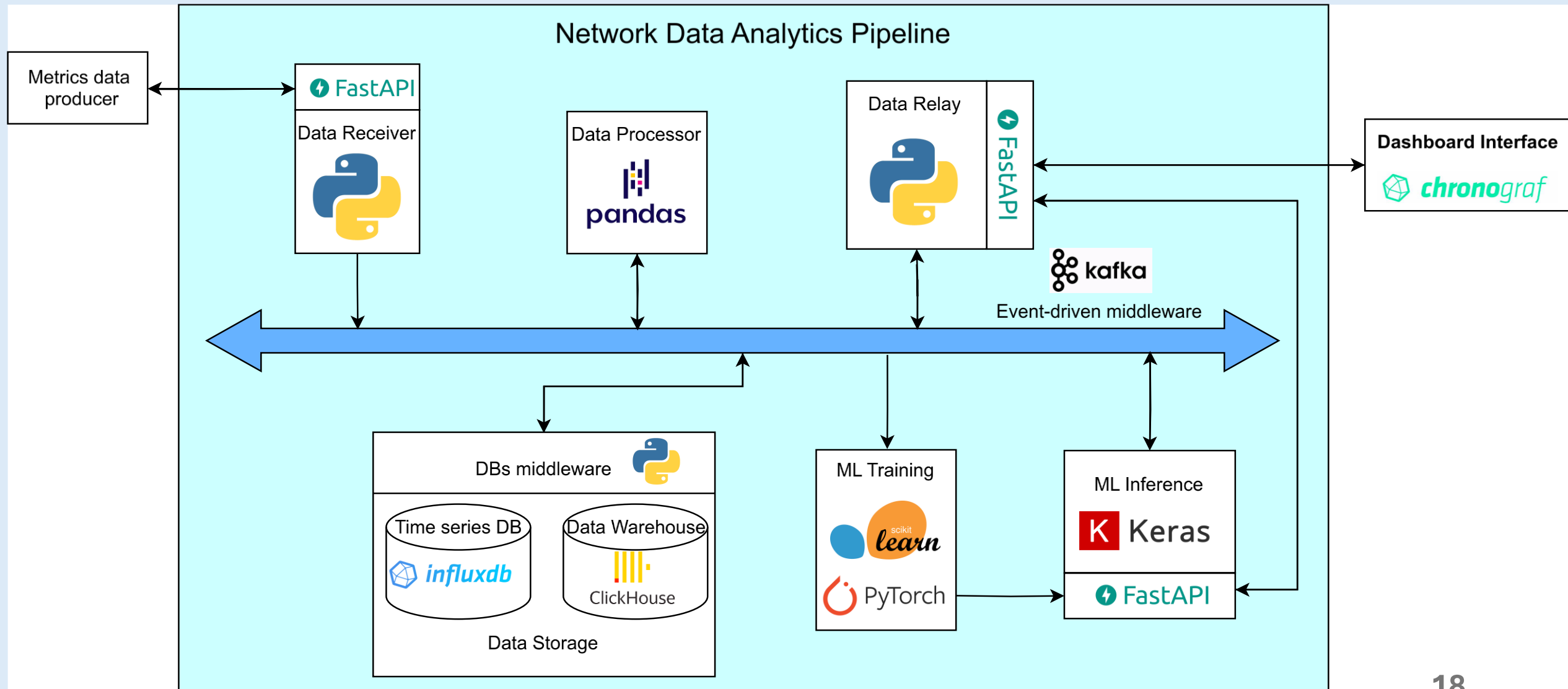
# Non-Functional Requirements

## Maintainability

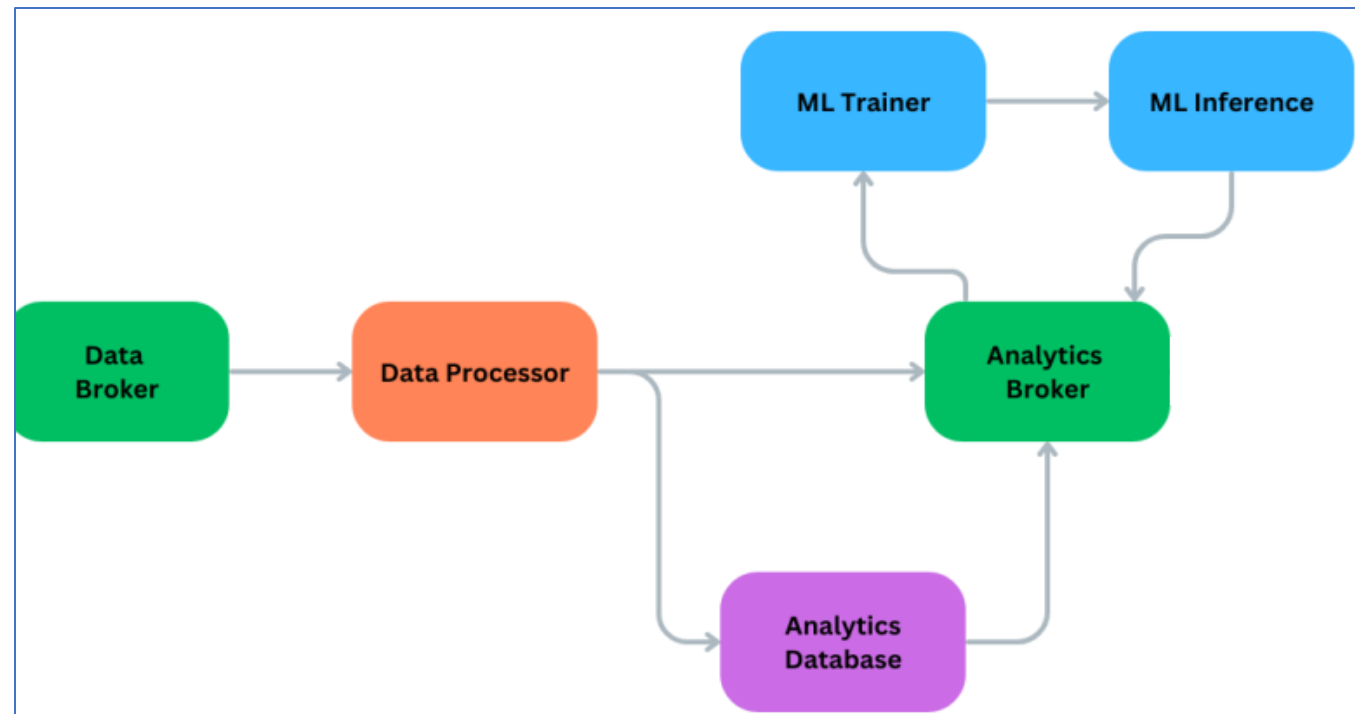
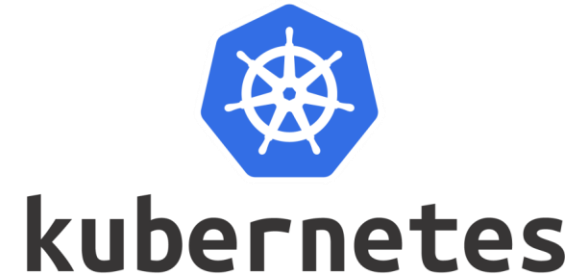
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- The system should use modular components to allow easy updates and debugging.
- The system must allow modules to be replaced by others with higher performance, with minimal impact on other modules.
- The system must be easily adaptable for deployment in several network environments.
- The system must follow good MLOps practices, guaranteeing modularity, reproducibility and complete automation of the ML lifecycle.

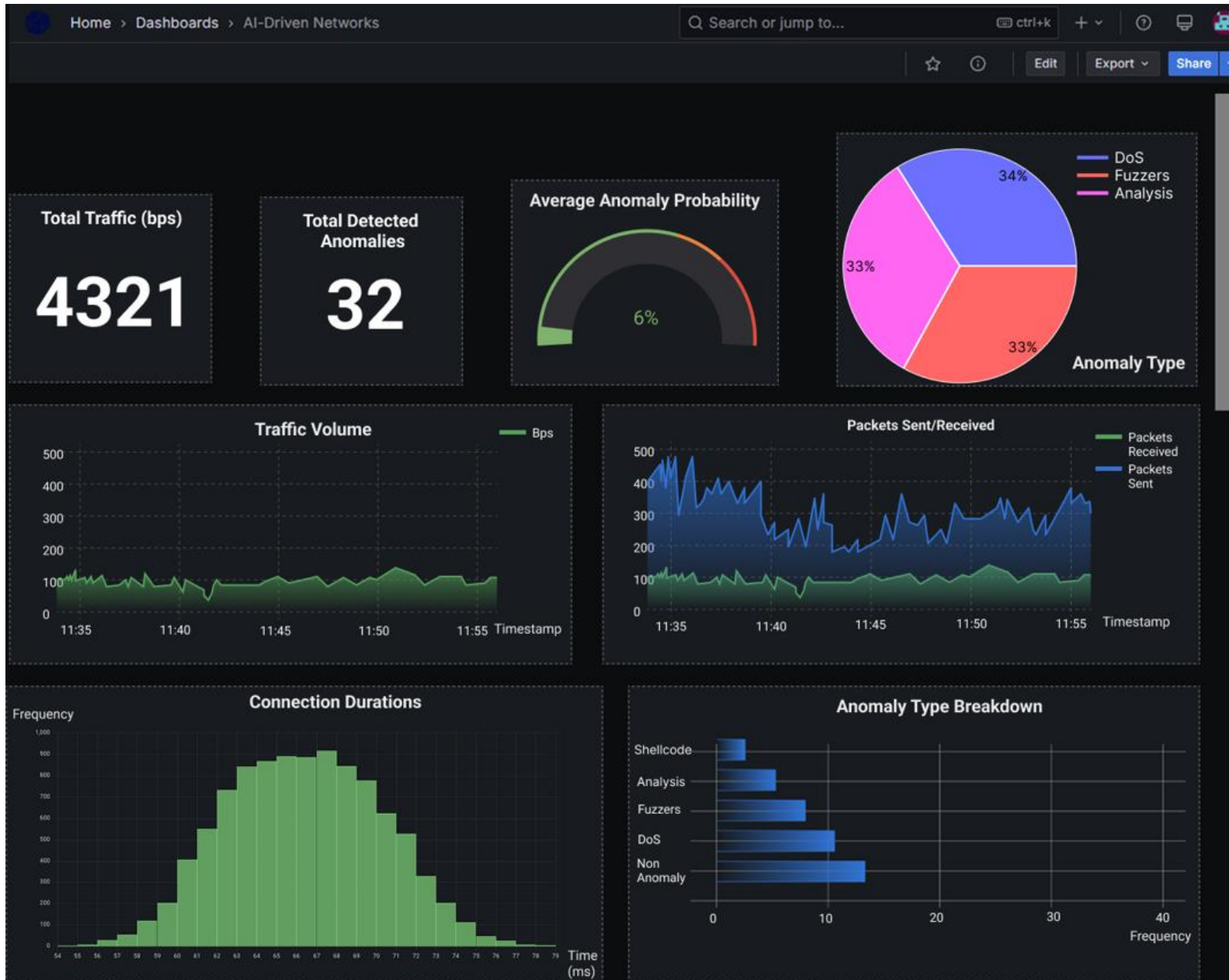
# 6. System Architecture



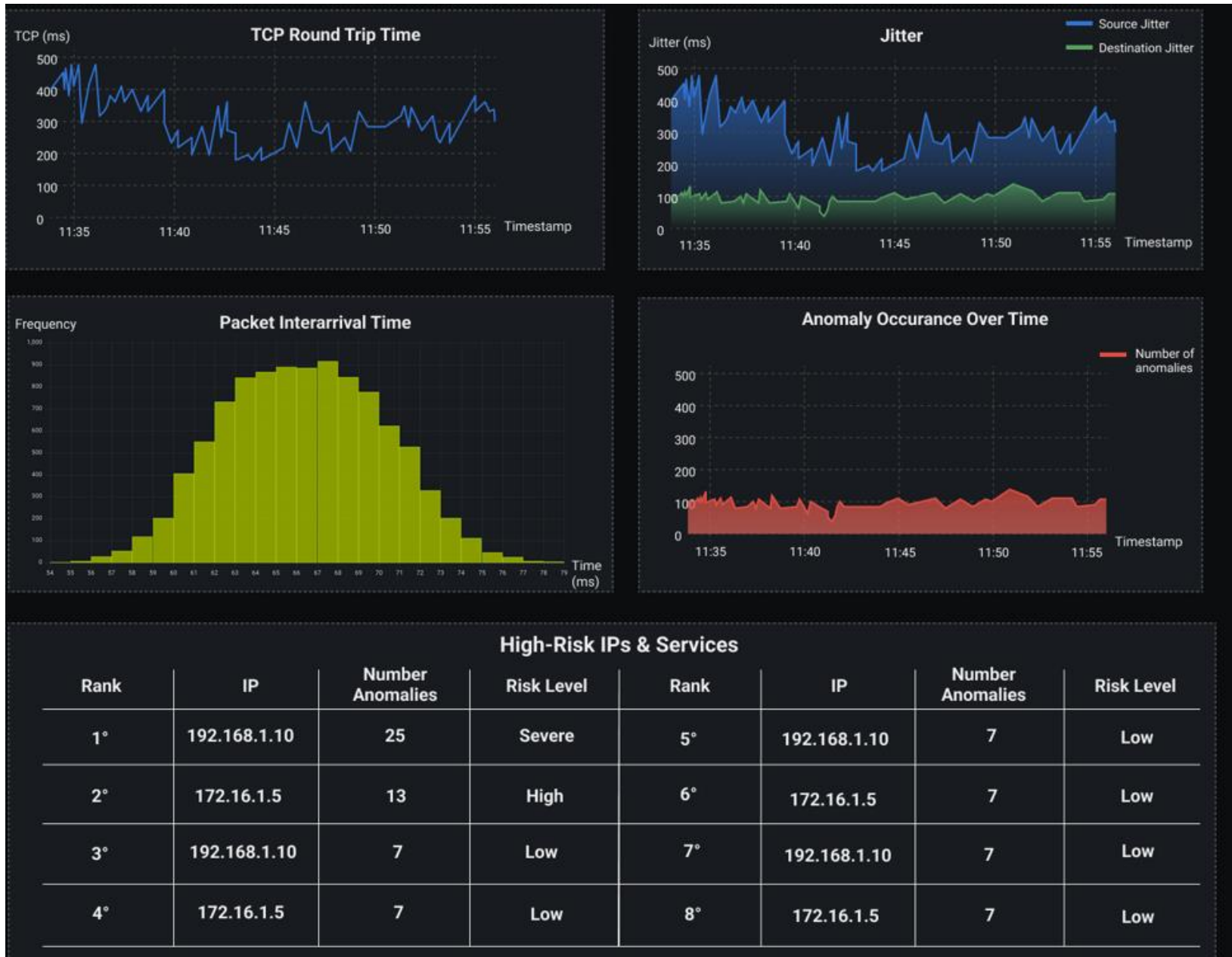
# Deployment



# 7. DASHBOARD MOCKUP - I







# 7. DASHBOARD MOCKUP - II



# 7. DASHBOARD MOCKUP - III



Anomaly Events Table

| Timestamp        | Source IP    | Destination IP | Protocol | Anomaly Type  | Confidence Score | Action Taken | Severity  |
|------------------|--------------|----------------|----------|---------------|------------------|--------------|---|
| 2025-03-08 12:30 | 192.168.1.10 | 10.0.0.5       | TCP      | DDoS          | 79%              | Deny Service |  Extreme |
| 2025-03-08 12:33 | 172.16.1.5   | 8.8.4.4        | TCP      | Login Faliure | 97%              | None         |  Low     |
| 2025-03-08 12:30 | 192.168.1.10 | 10.0.0.5       | TCP      | DDoS          | 79%              | Deny Service |  Extreme |
| 2025-03-08 12:33 | 172.16.1.5   | 8.8.4.4        | TCP      | Login Faliure | 97%              | None         |  Low    |

Scan the QR code to check our documentation website.



Or click [here](#).

THANK  
YOU



# References

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[1]

A. Mekrache, K. Boutiba, and A. Ksentini, "Combining Network Data Analytics Function and Machine Learning for Abnormal Traffic Detection in Beyond 5G," *GLOBECOM 2023 - 2023 IEEE Global Communications Conference*, Dec. 2023, <https://doi.org/10.1109/globecom54140.2023.10436766>.

[2]

N. Nisha, Lakshman K, and R. Kumar, "A Smart Data Analytics System Generating for 5G N/W System Via ML Based Algorithms for the Better Communications," Apr. 2024, <https://doi.org/10.1109/istems60181.2024.10560068>.

[3]

Rui Cruz Ferreira *et al.*, "Demo: Enhancing Network Performance based on 5G Network Function and Slice Load Analysis," Jun. 2023, <https://doi.org/10.1109/wowmom57956.2023.00057>.