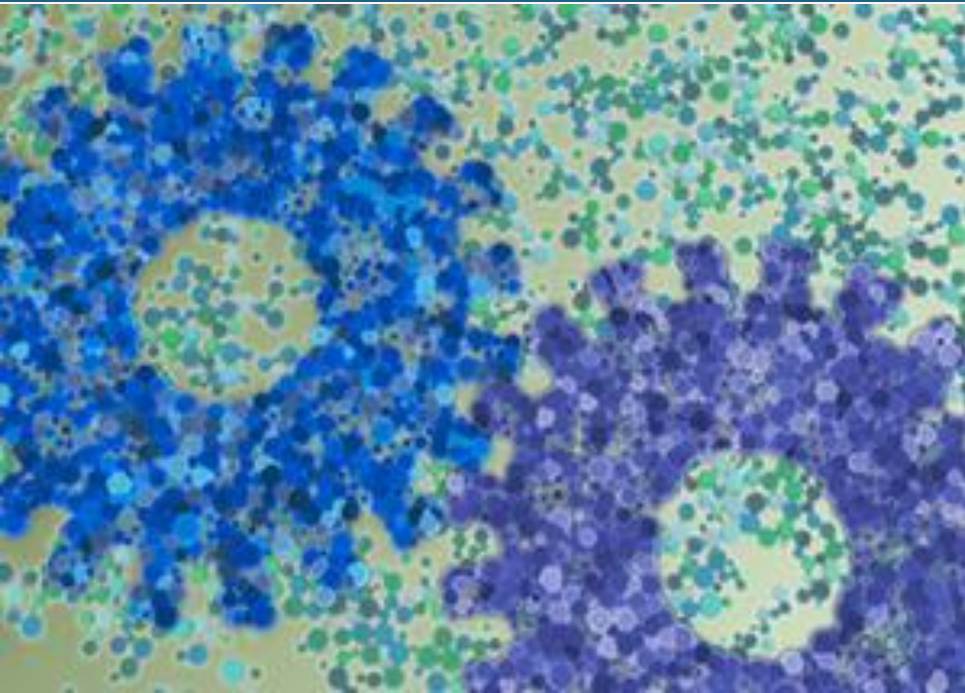


Ecosystem for Collaborative Manufacturing Processes – Intra- and Interfactory Integration and Automation



Big Data Management

The use of Deep learning techniques
in the COMPOSITON platform

Paolo Vergori

ISMB



Co-funded by the
European Union





Agenda

- Why deep learning for factory of the future?
- The COMPOSITION challenge
 - UC-BSL-2 Predictive maintenance
 - UC-ELDIA-1 – Contractual goods and recyclable materials management
- Deep Learning approach in COMPOSITION
 - intra-factory scenario
 - inter-factory scenario
- Future work



Why deep learning for factory of the future?

- Prolific environment
 - large amount of potential data available
 - tangible results achievable in finite time span
 - potential market exploitation
- Overcoming current state of the art
 - outperforming statistical models
 - continuous learning for adapting to dynamic production environment
 - transferrable results in terms of research (not in terms of models or dataset)
- Strong correlation to upcoming existing technologies
 - improving of matchmaking results, supporting decision systems



Deep Learning approach in Composition

- Deep learning toolkit
 - twofold application as both intra and inter-factory intelligent tool
 - intra-factory data analysis for predictive maintenance
 - inter-factory agent-based marketplace market estimation
- Continuous learning
 - triggered by inter-factory agents and learning framework
- Fits in a comprehensive ecosystem



The COMPOSITION's challenge - UC-BSL-2 Predictive maintenance

BSL: predictive maintenance on production equipment

- 2 analyzed reflow ovens
- very complex dataset with ~625,000, ~350,000 samples
 - 60 sensors readings in 20 zones
 - 5 newly deployed sensors linearly independent
- few reported failures ($\sim \ll 20$) over recorded timeframe
- very demanding in terms of accuracy
 - tentative for ~2.5h timeframe
 - more realistically ~5 minutes timeframe



COMPOSITION's challenge - UC-ELDIA-1 – Contractual goods and recyclable materials management

ELDIA: agent-based marketplace estimation

- 4 analyzed raw materials markets
- very small dataset with few samples
 - recording started in 2016 only
- less than 50 reported values
- very demanding in terms of prediction
- steady prices recorded before
- very little contextual data to the analyzed market

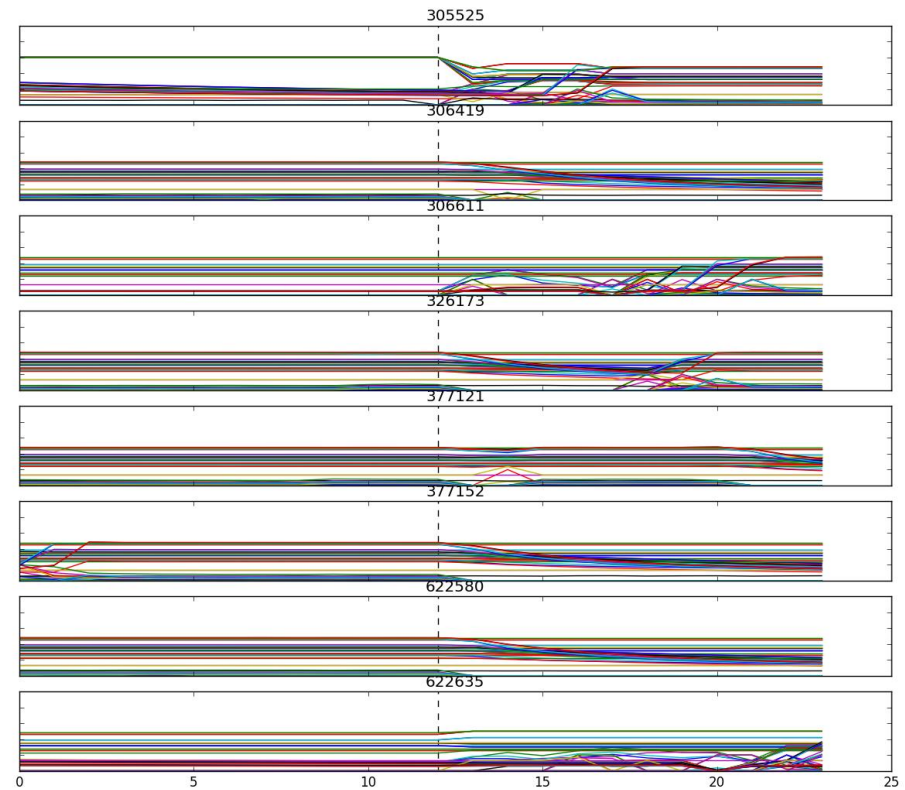
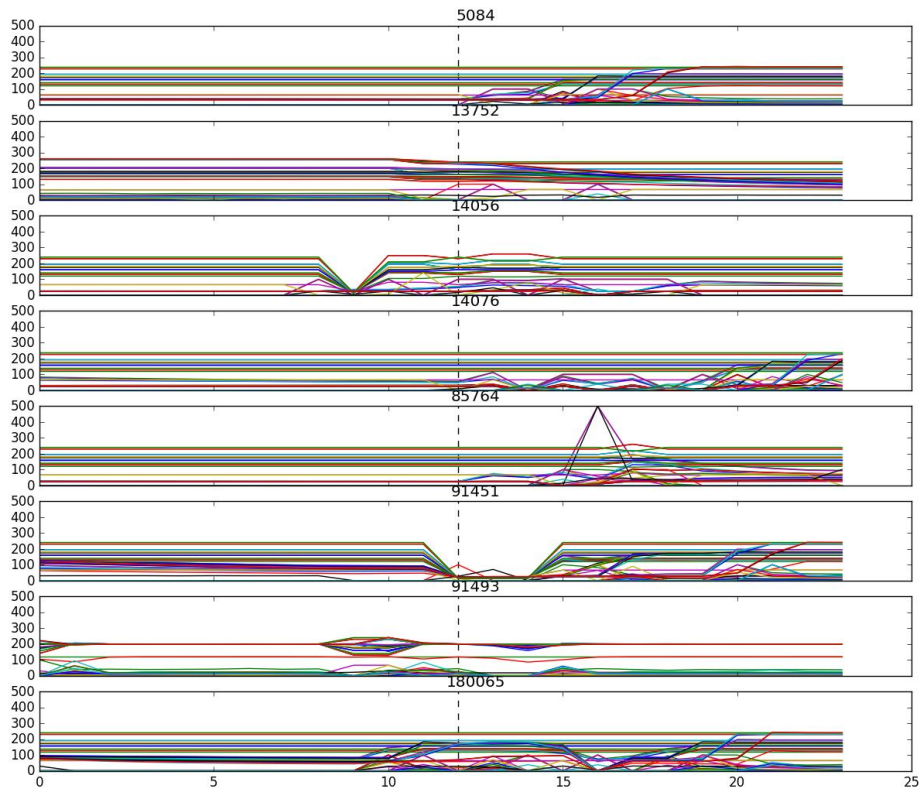


Deep Learning approach in COMPOSITION – intra-factory scenario

- Intra-factory data analysis for **predictive maintenance** from heterogeneous sources
 - data analysis of existing machinery, forming the **historical dataset**
 - **newly** deployed sensors for acquiring more information
- High risk challenge:
 - work with an **incomplete dataset** with no historical information on new measured parameters
 - have **unbalanced classes** in a nearly real-time environment
- Dichotomy of solution:
 - deploy **untrained ANNs** and rely on theoretical convergence forgetting the historical data
 - **balance the classes** and propagate new parameters backwards

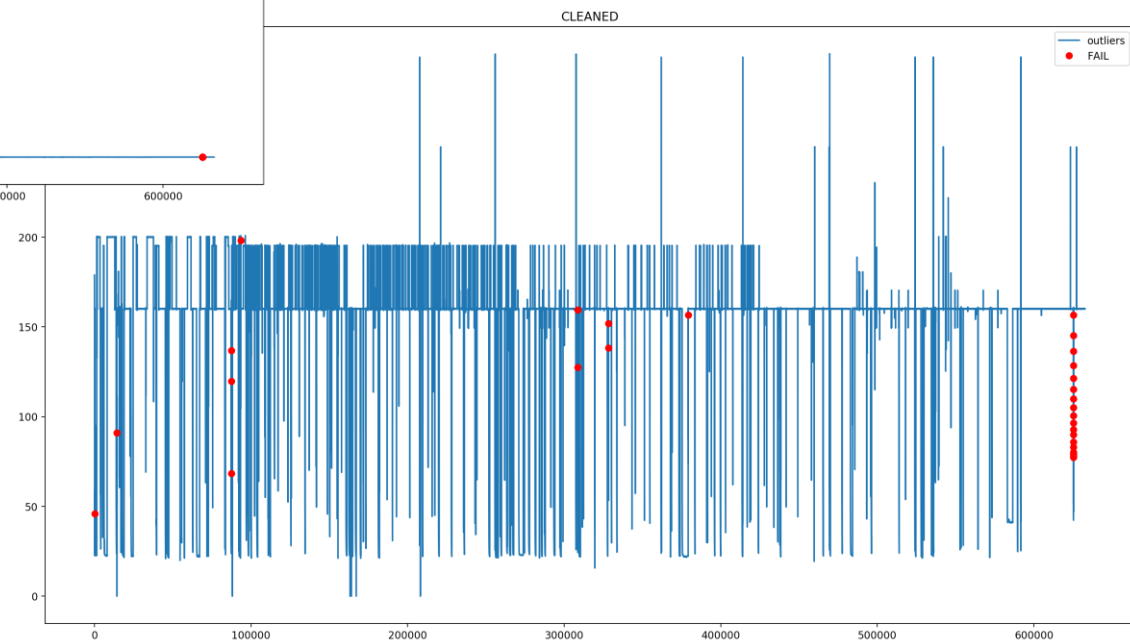
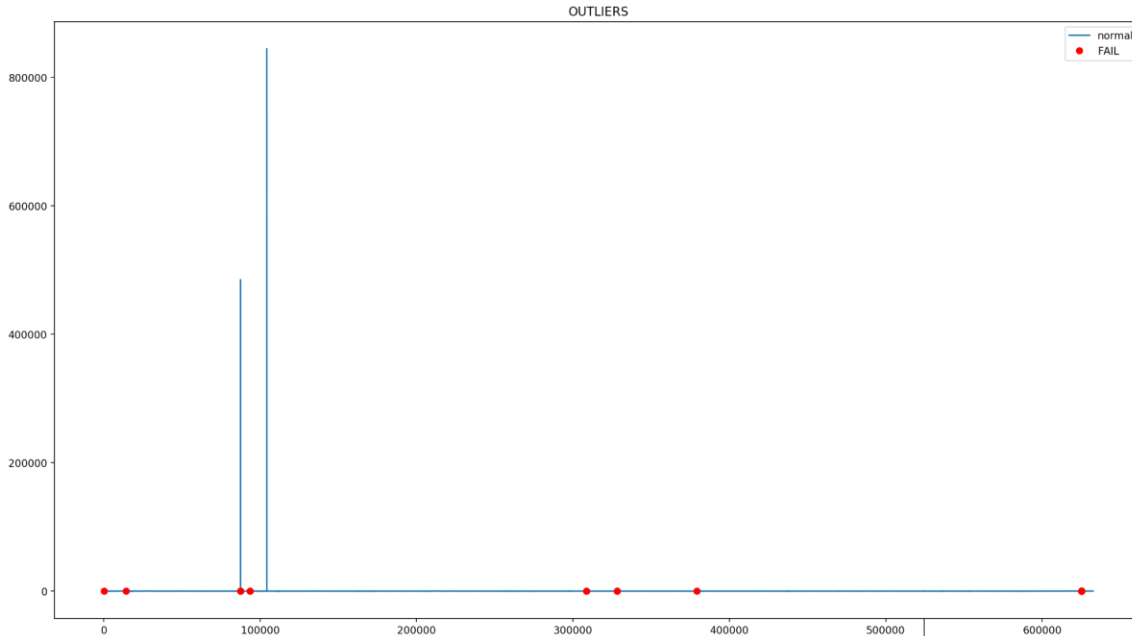


DLT – intrafactory – historical dataset



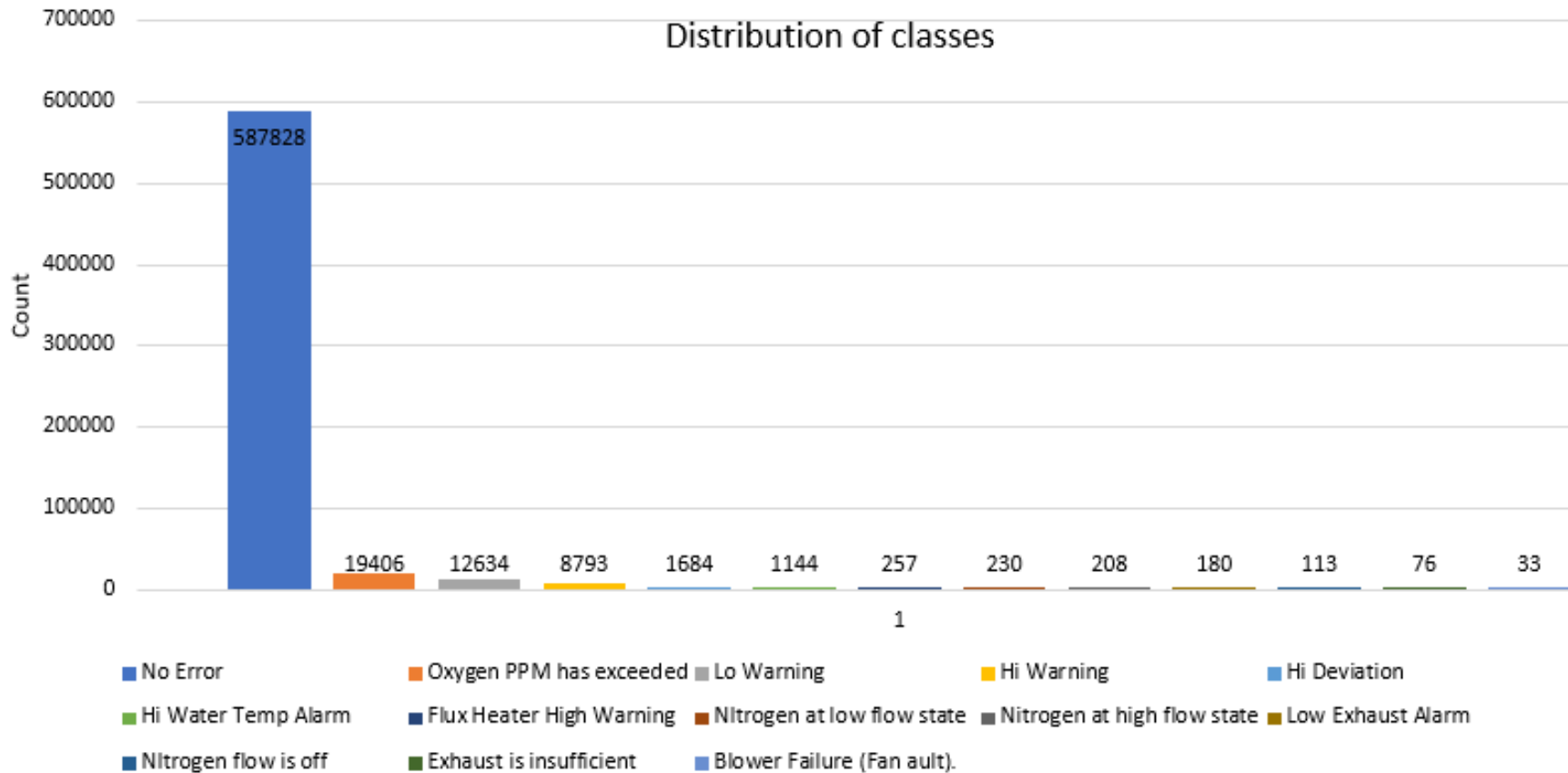


DLT – intrafactory – outliers and cleaned data



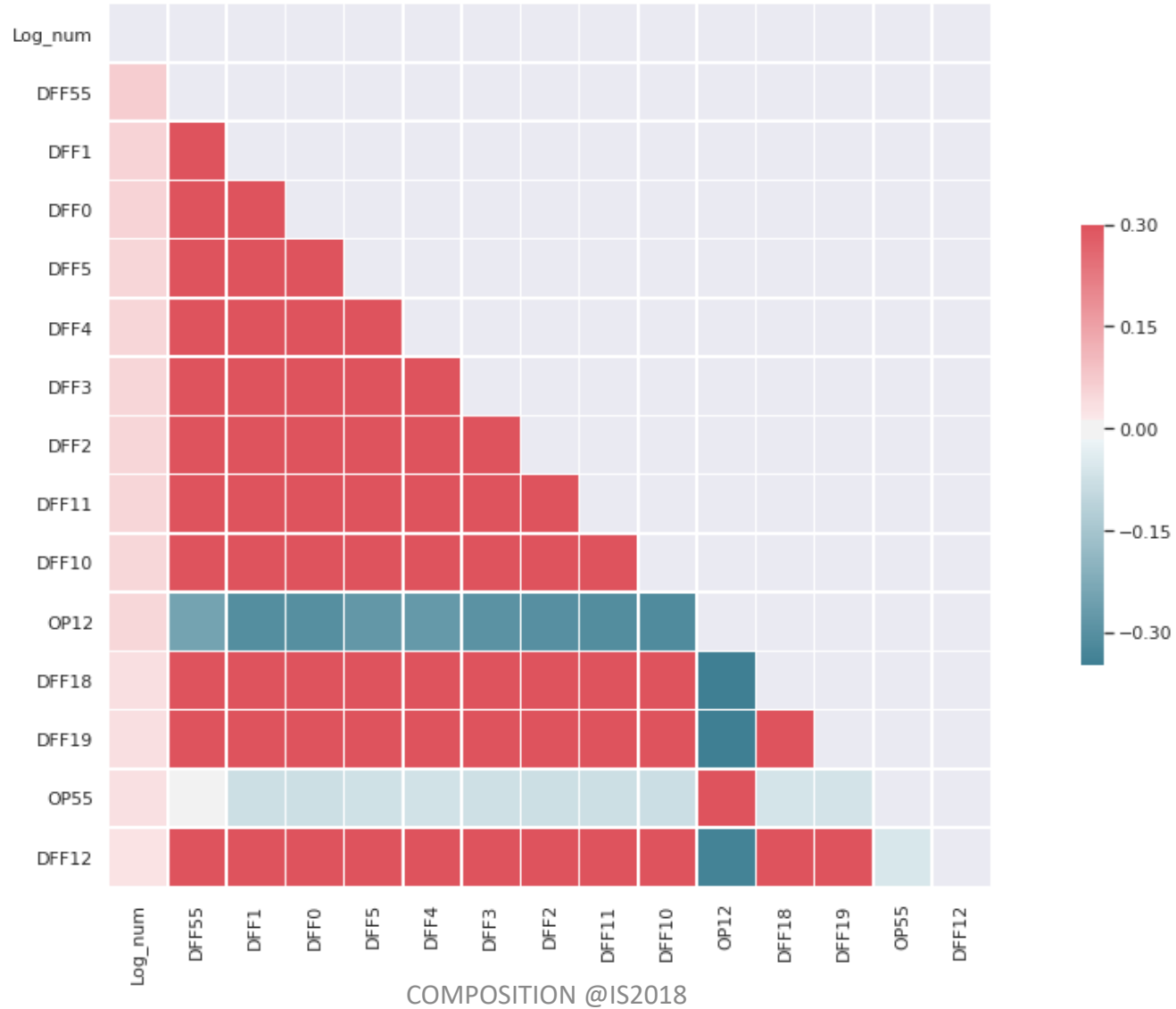


DLT – intrafactory – dataset creation



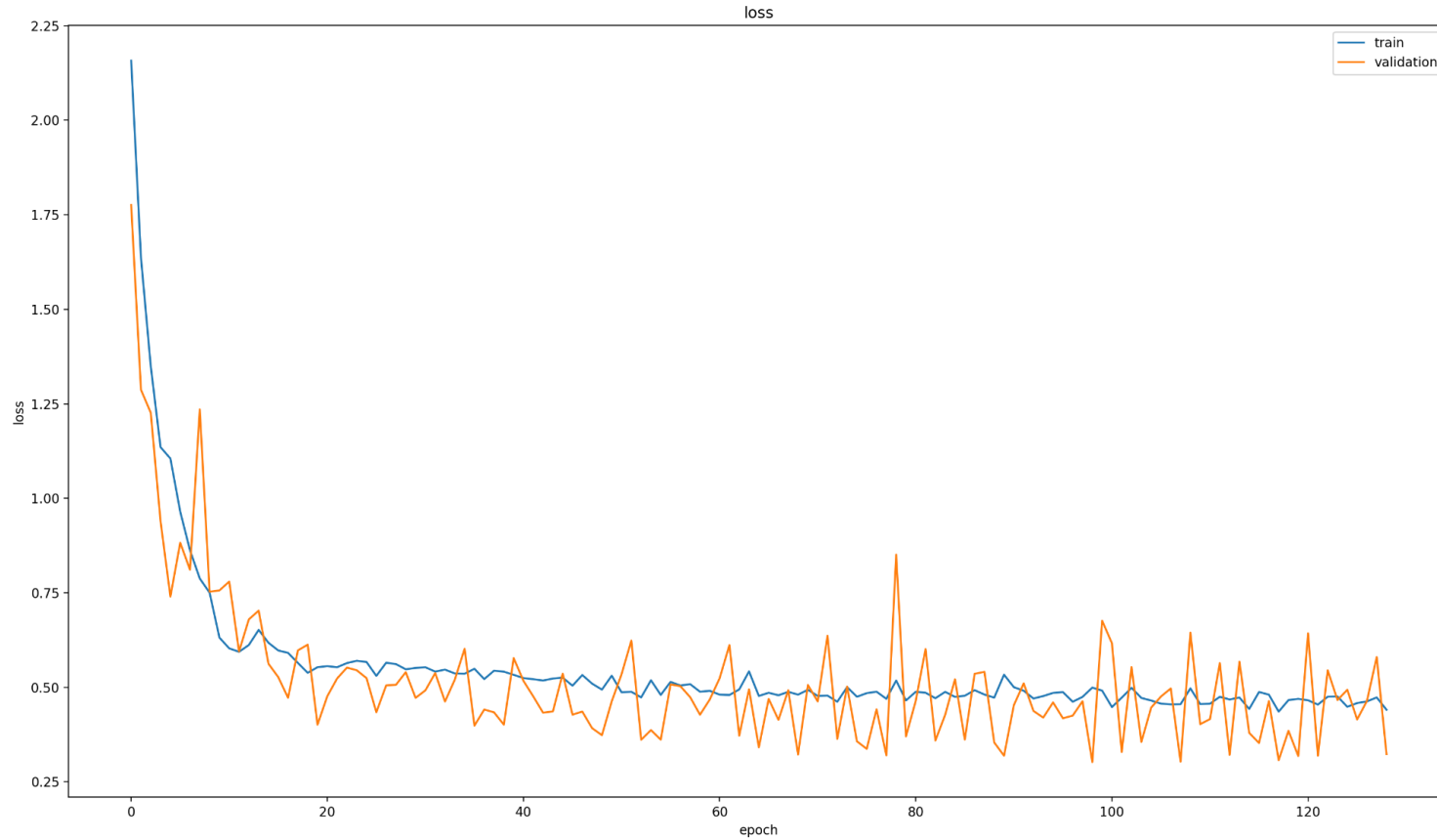


DLT – intrafactory – correlation matrix



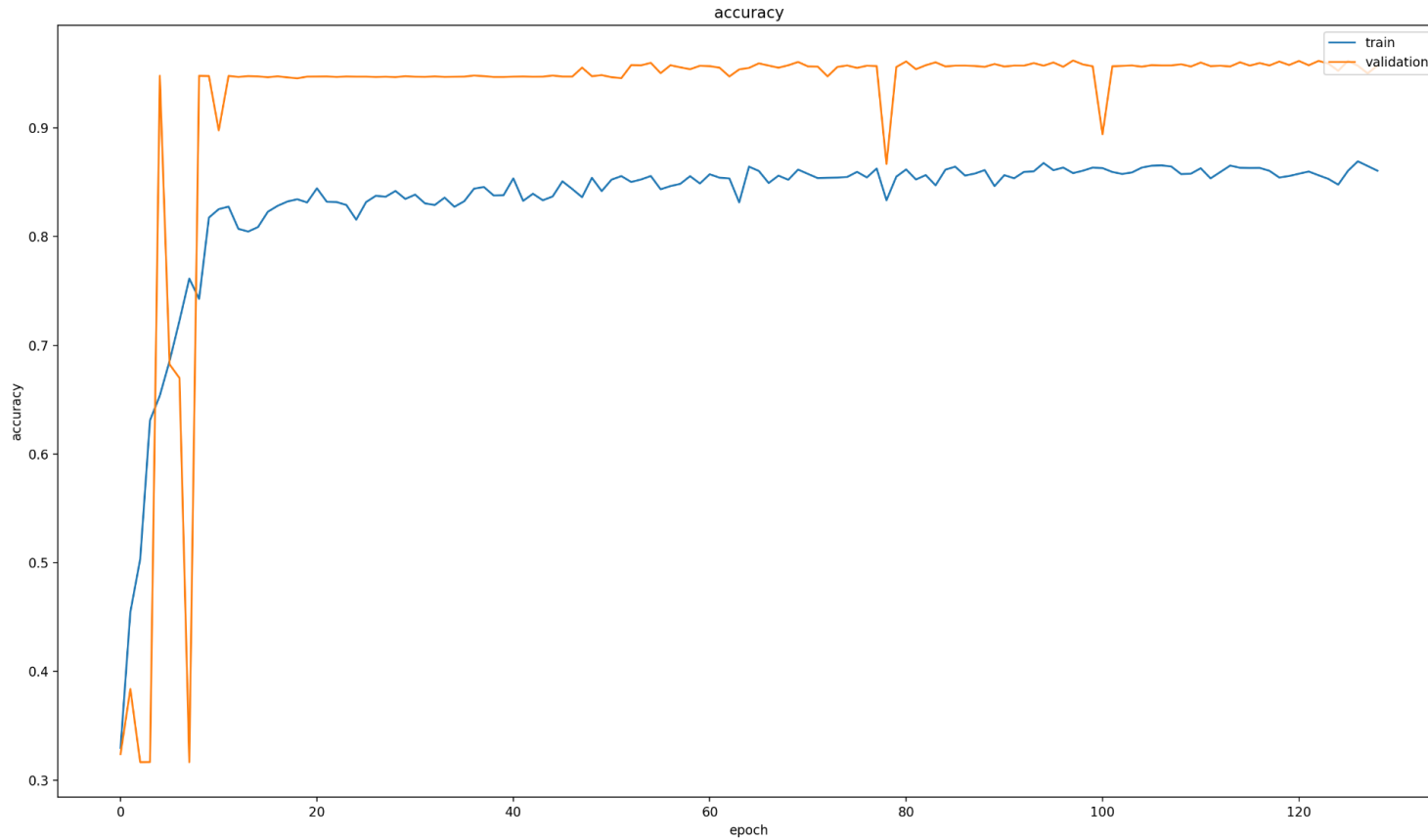


DLT – intrafactory – loss function



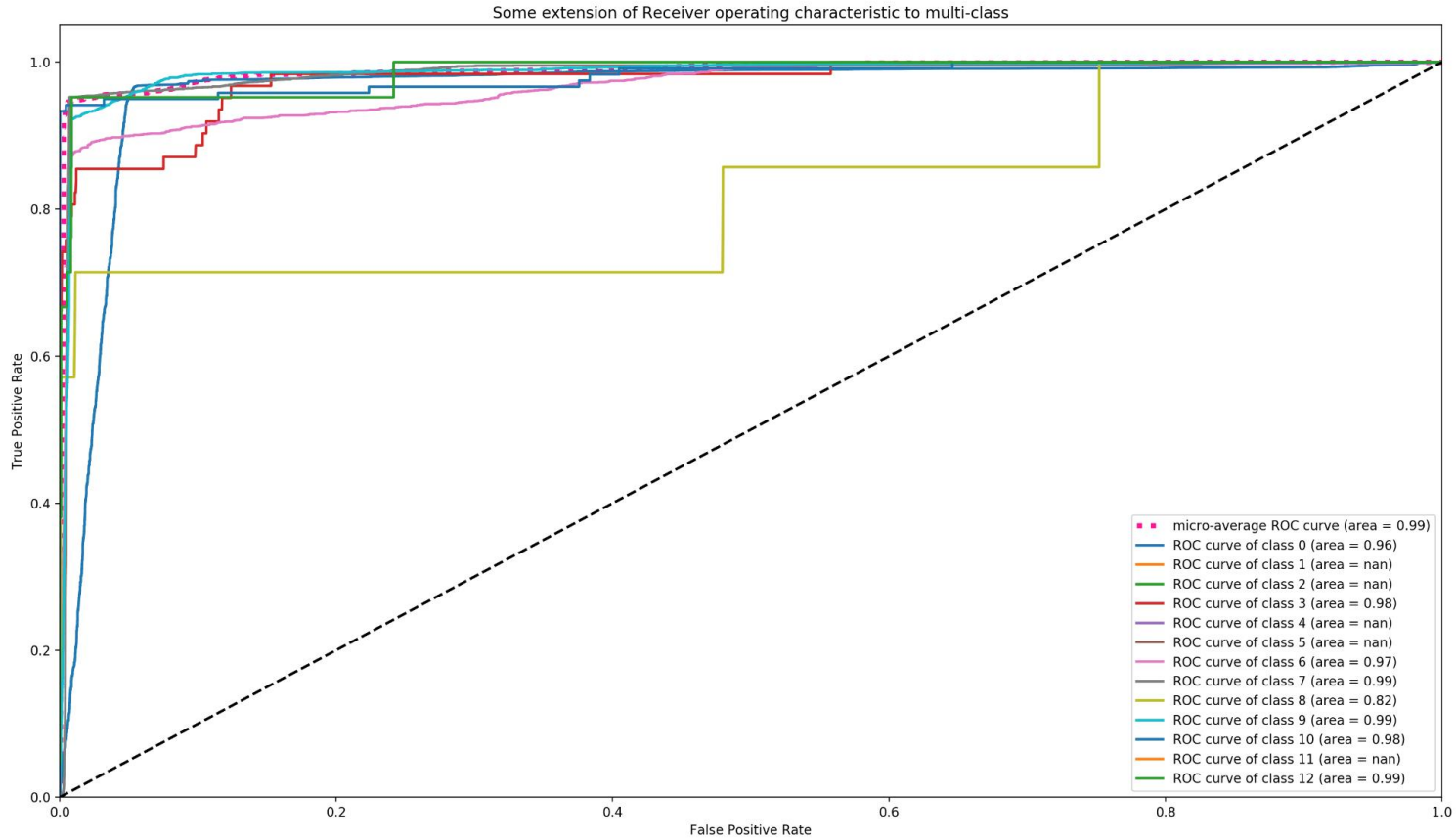


DLT – intrafactory – accuracy



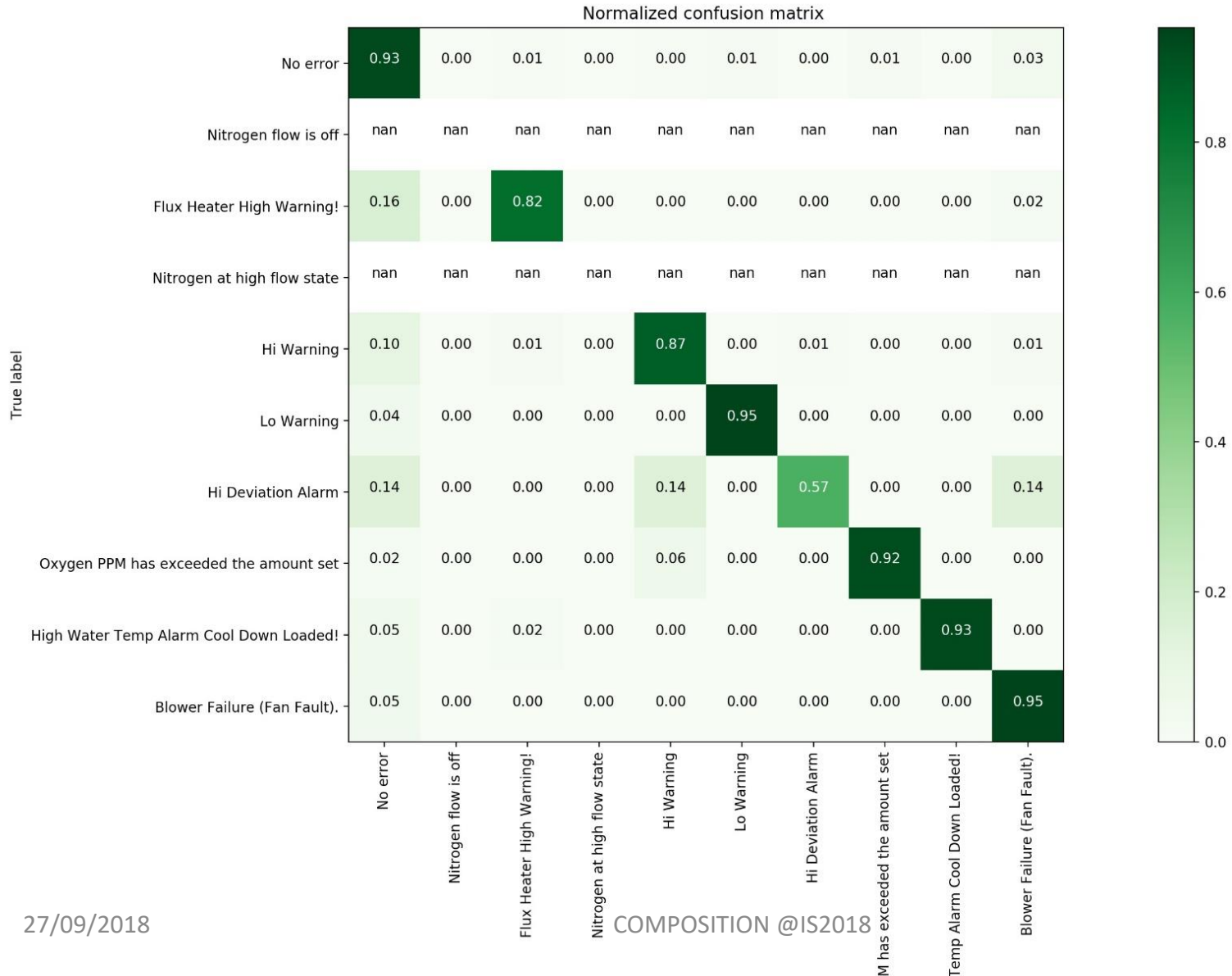


DLT – intrafactory – ROC function



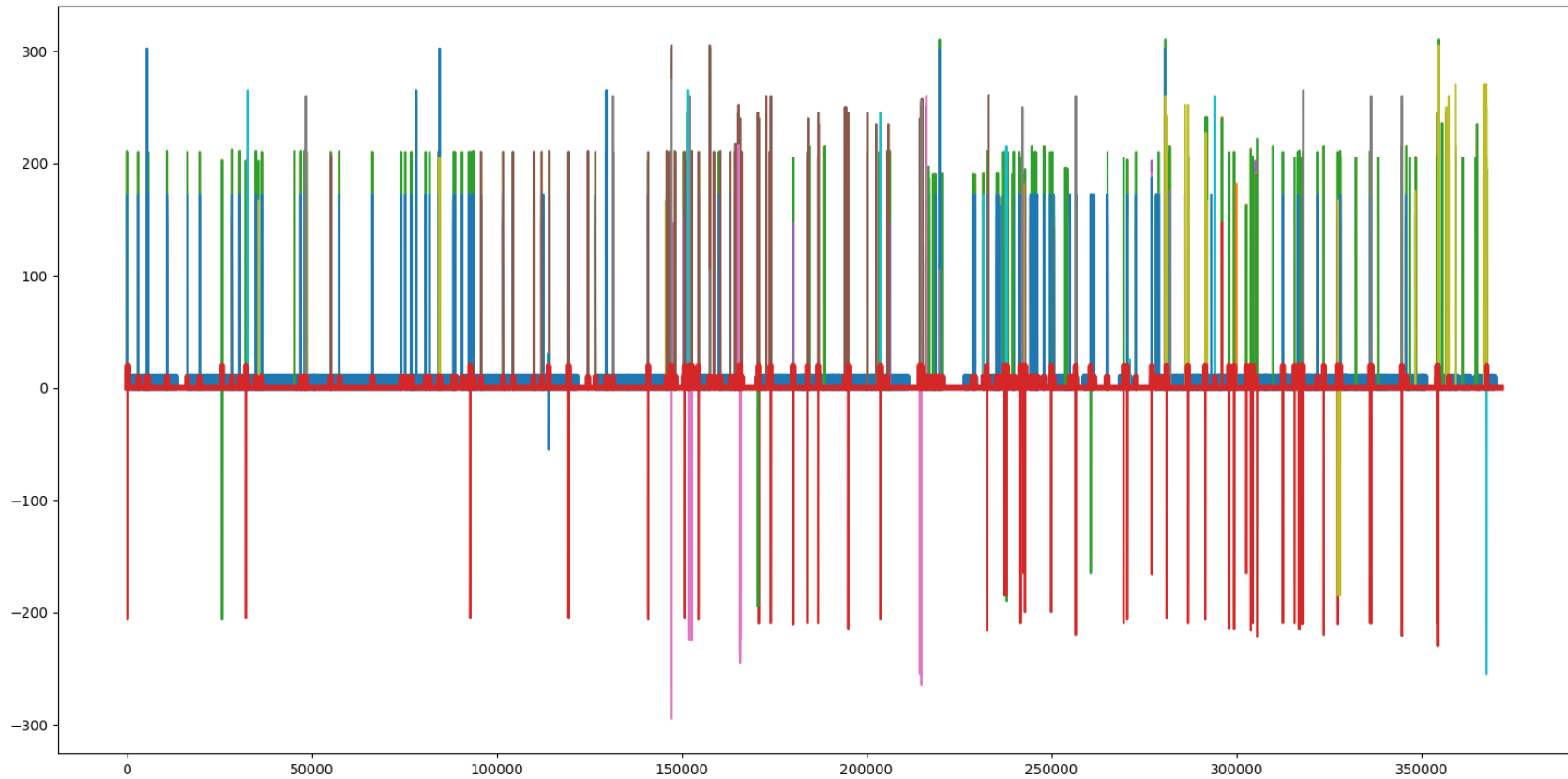


DLT – intrafactory – confusion matrix



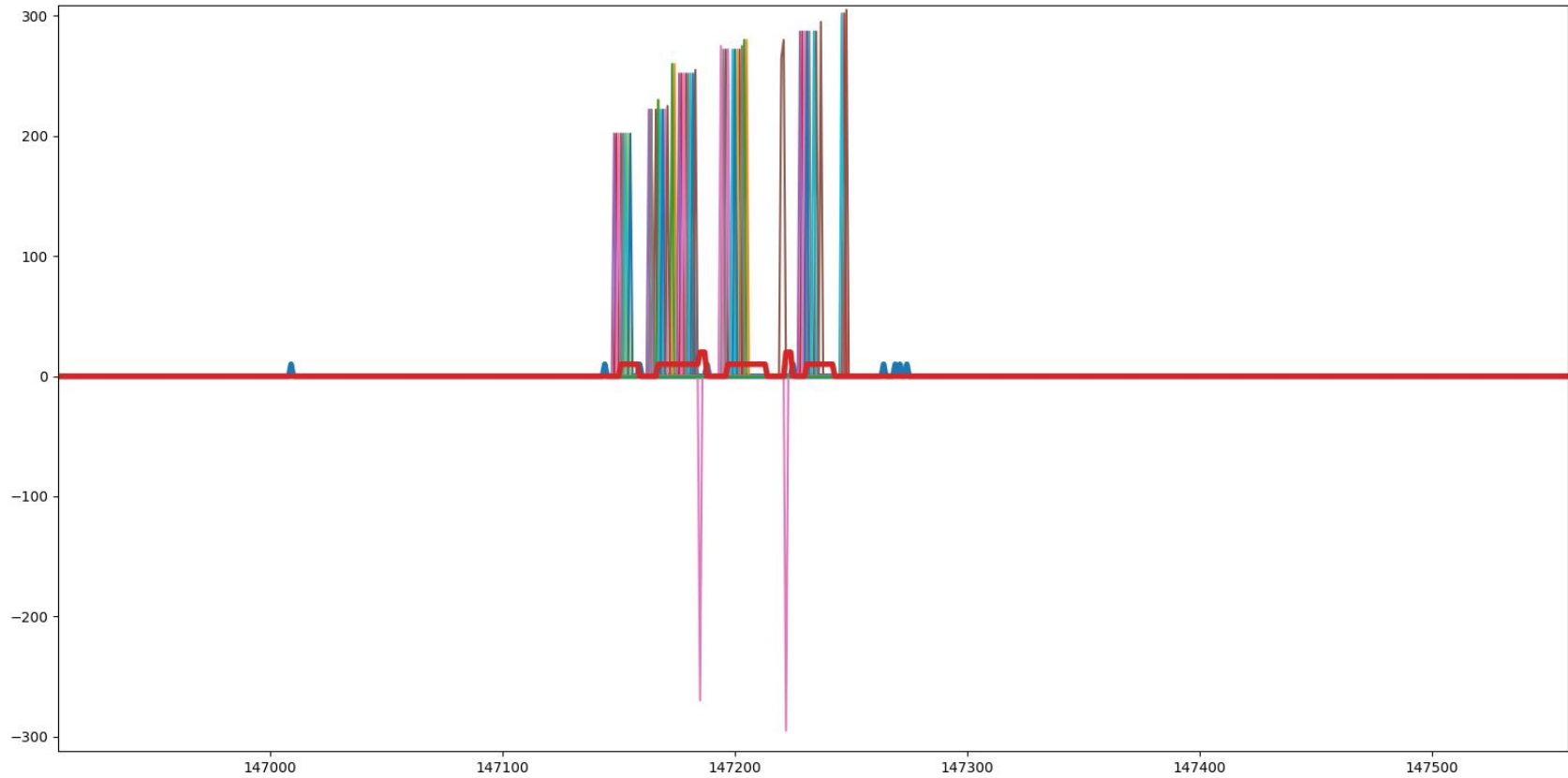


DLT – intrafactory – training dataset [2013-2017]



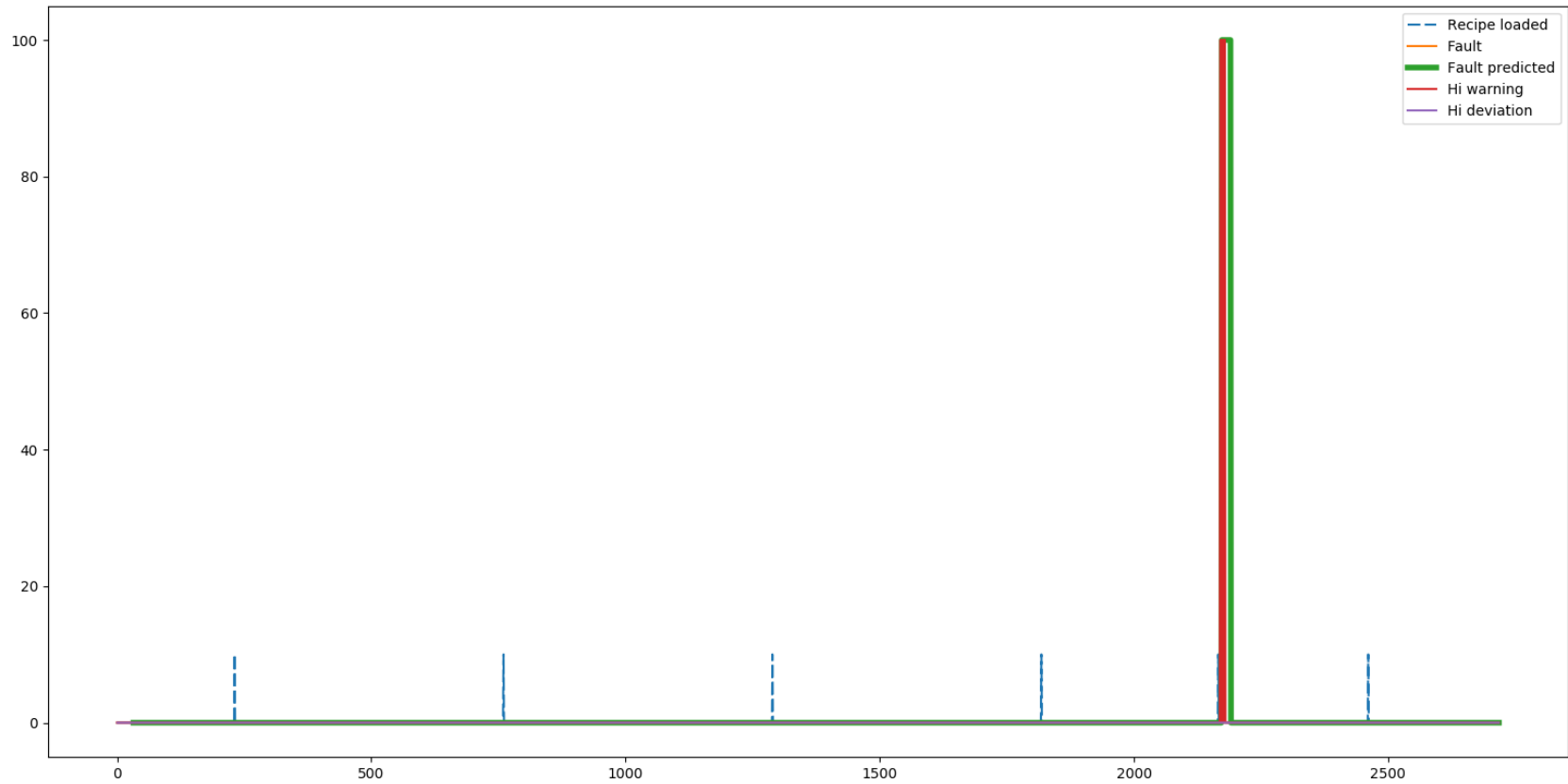


DLT – intrafactory – zoomed training dataset



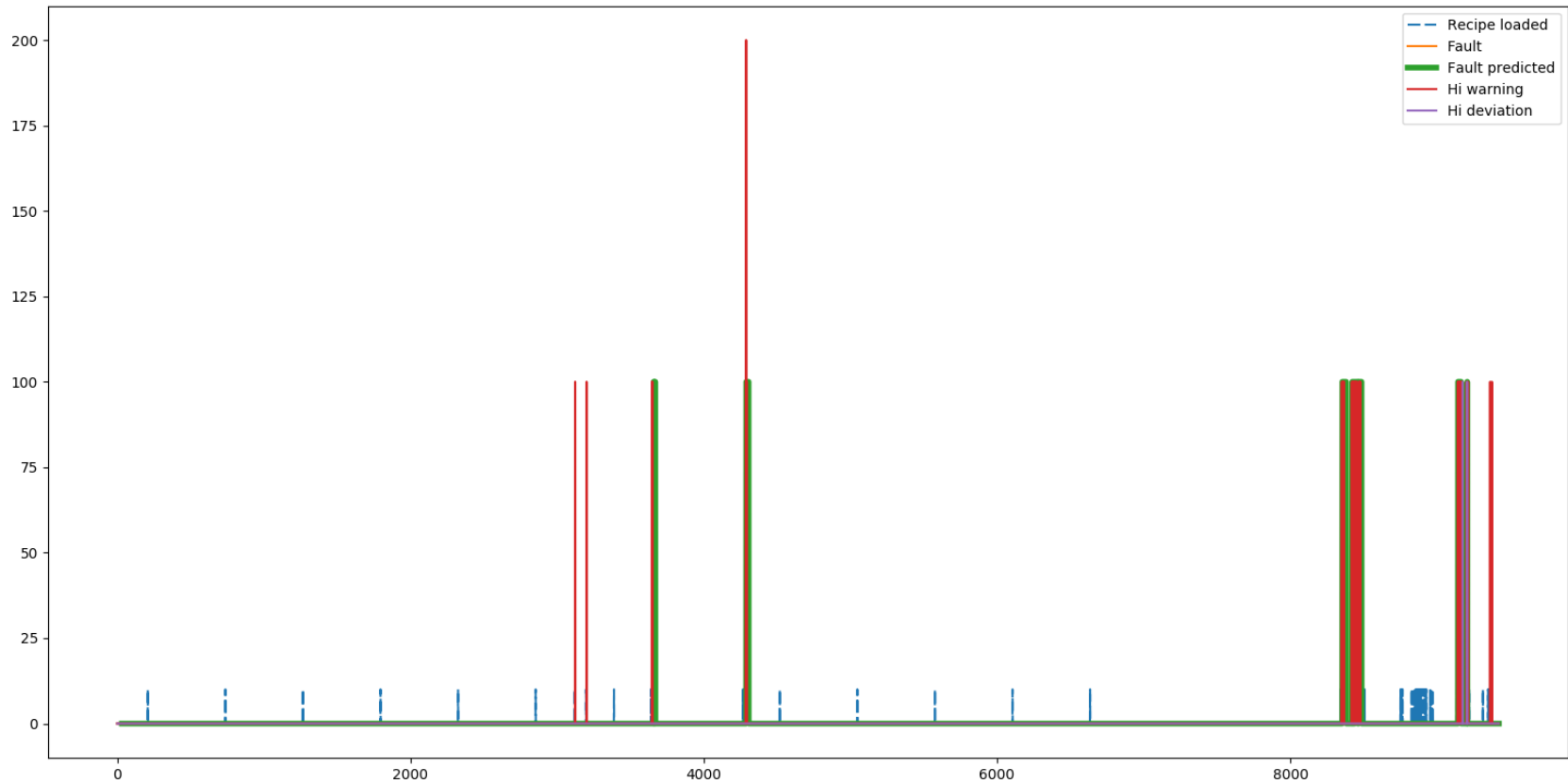


DLT – intrafactory – test dataset – trial 2



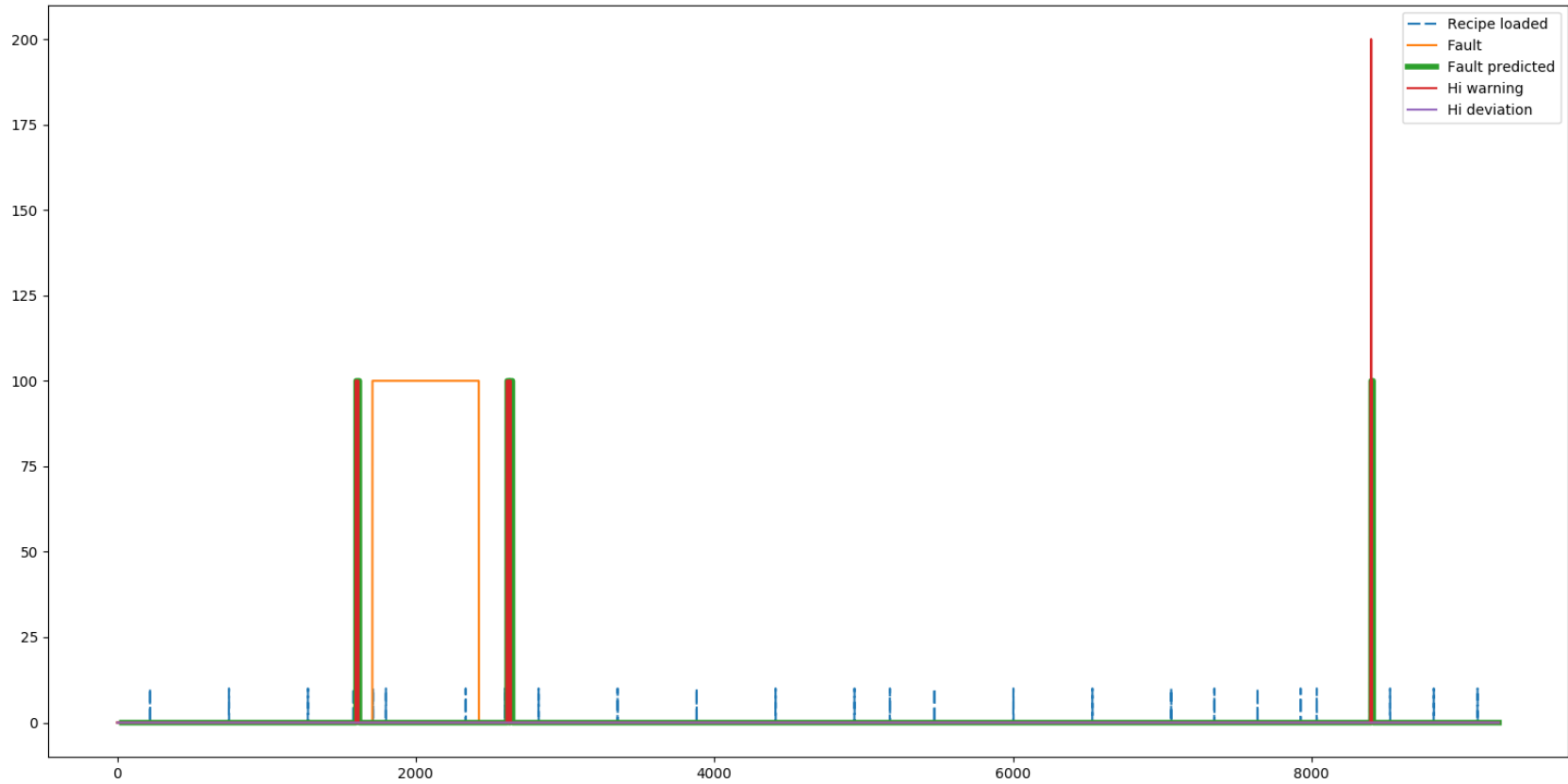


DLT – intrafactory – test dataset – trial 3



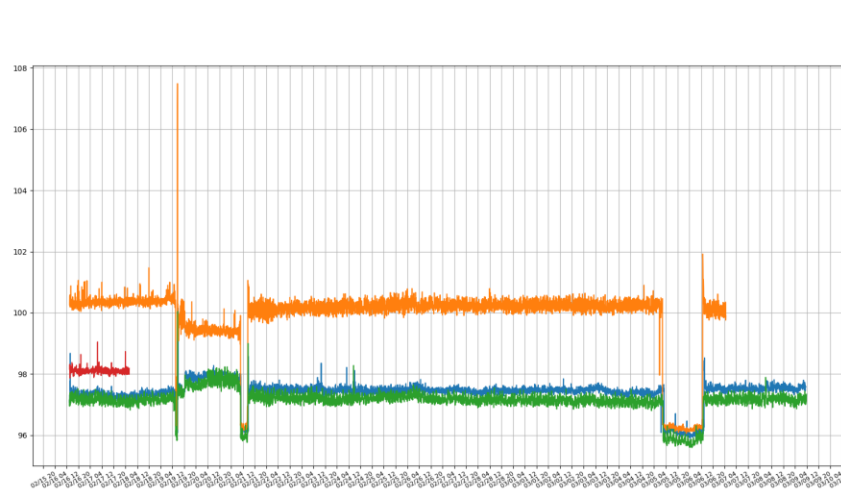
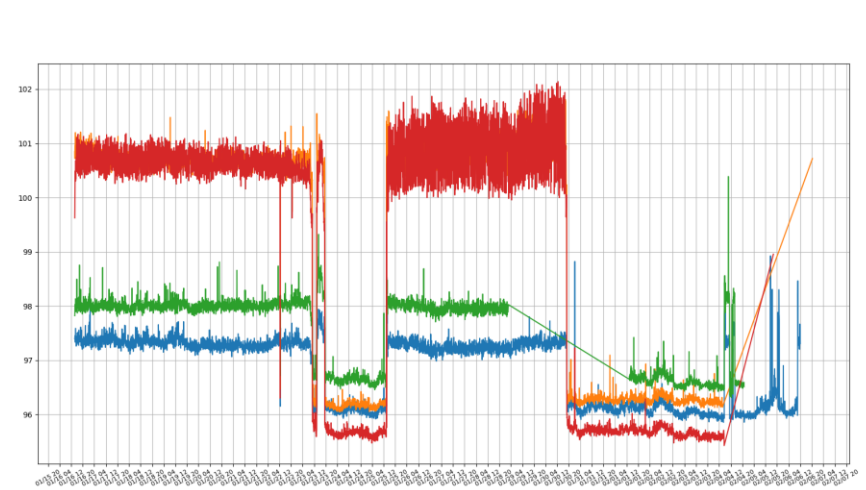


DLT – intrafactory – test dataset – trial 4



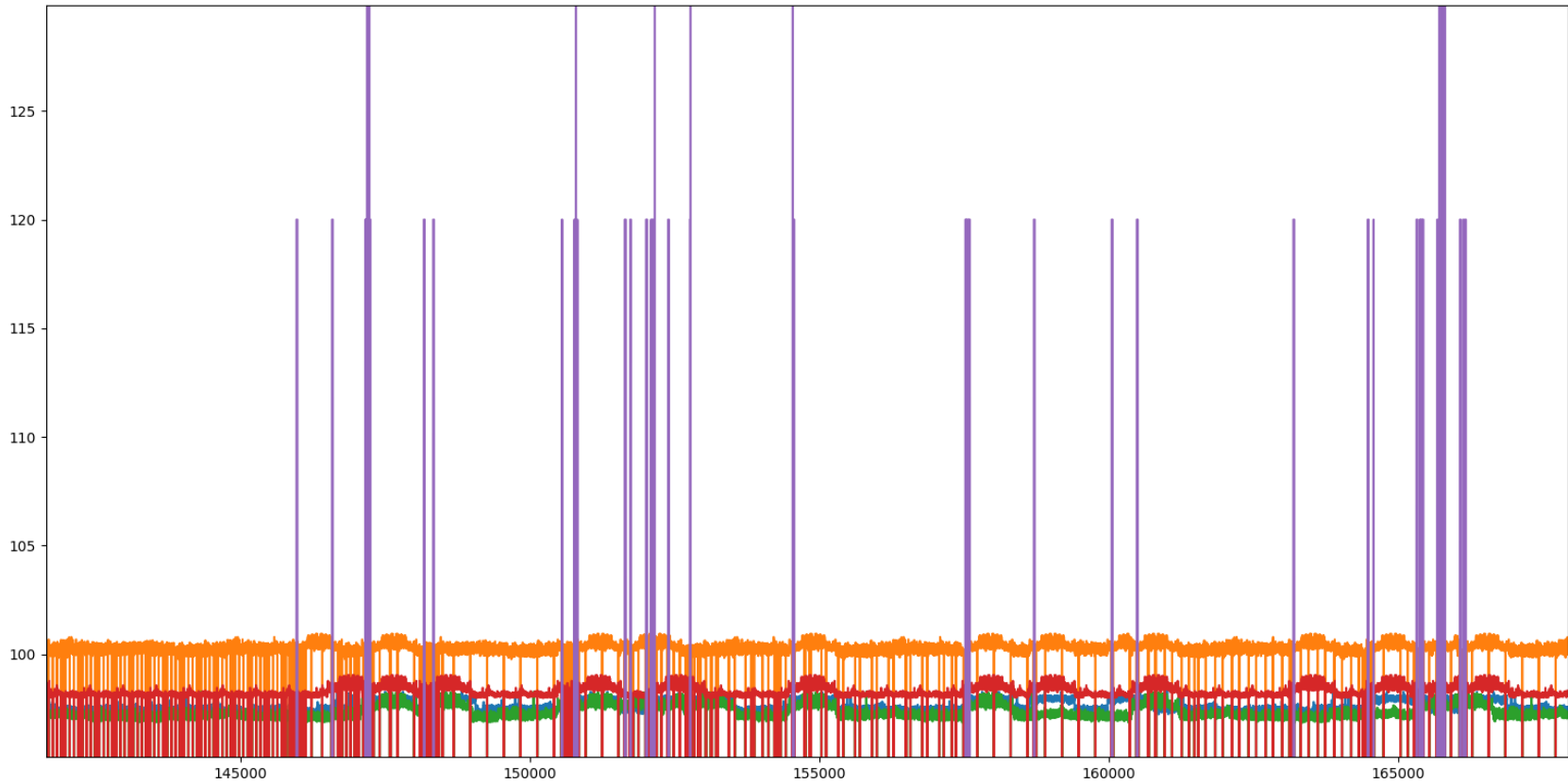


DLT – intrafactory – newly deployed acoustic sensors



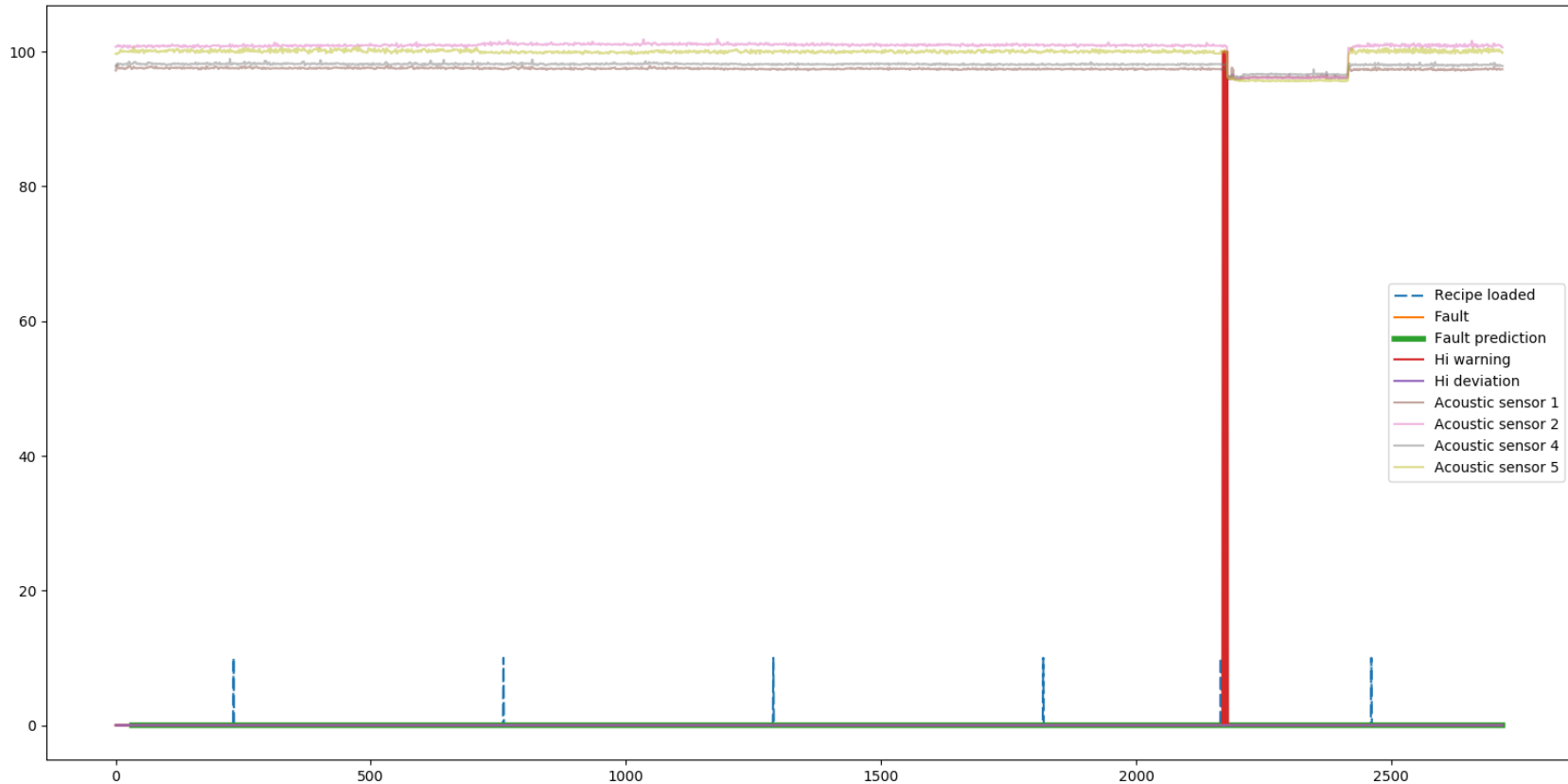


DLT – intrafactory – zoomed training dataset with audio [2013-2017]



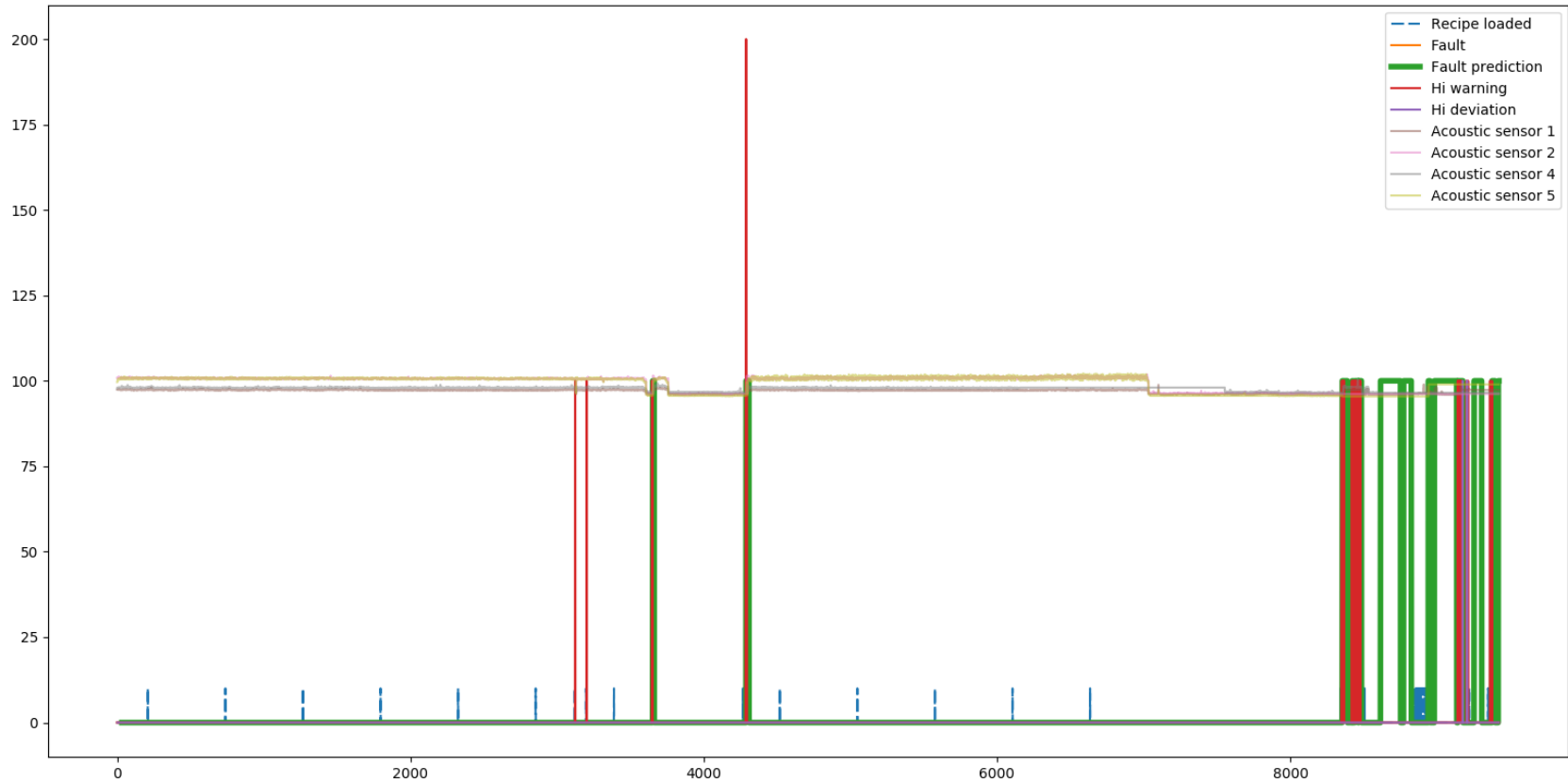


DLT – intrafactory – predictions on test dataset with audio, look ahead 1 sample – trial 2



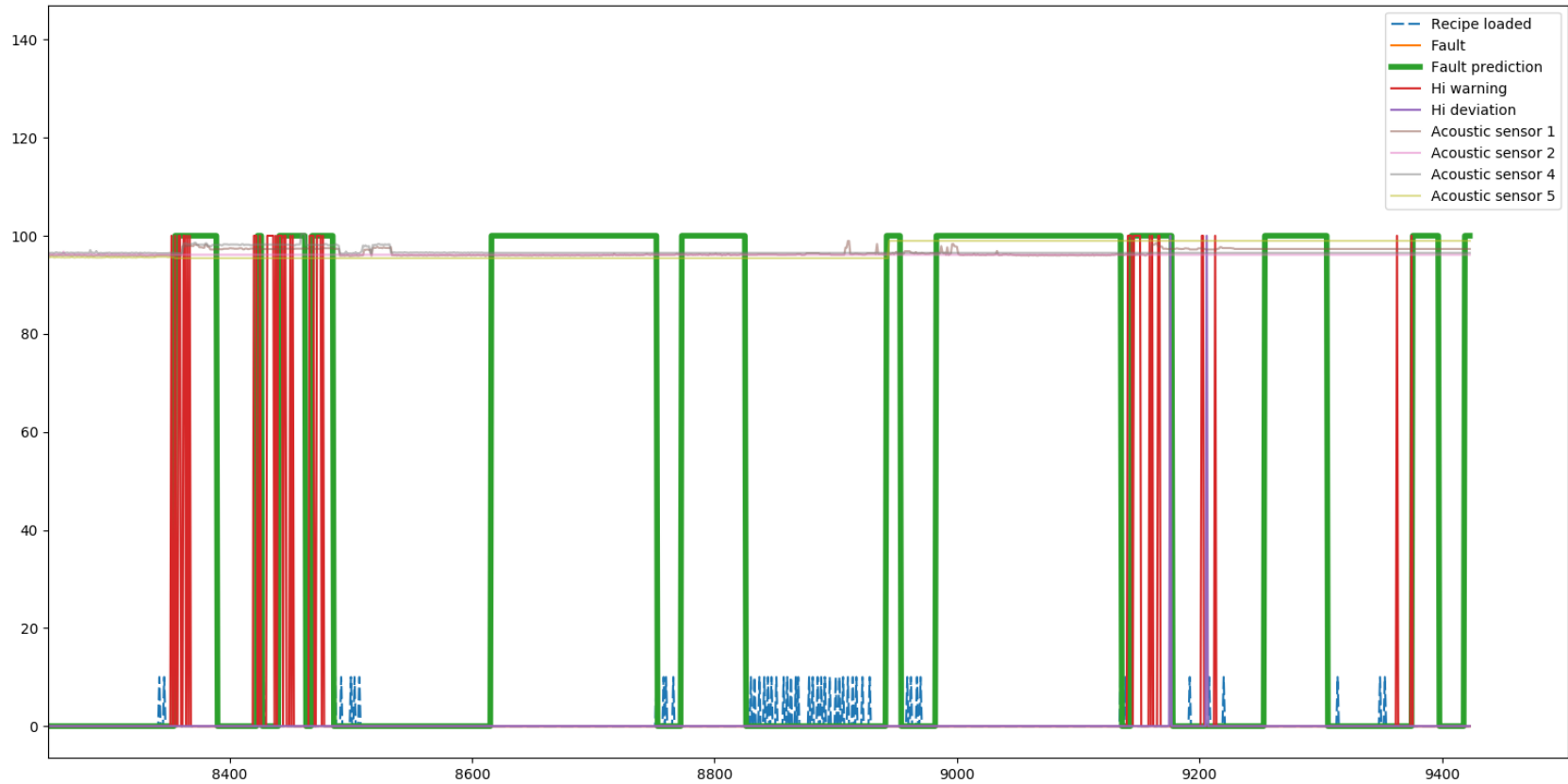


DLT – intrafactory – predictions on test dataset with audio, look ahead 1 sample – trial 3



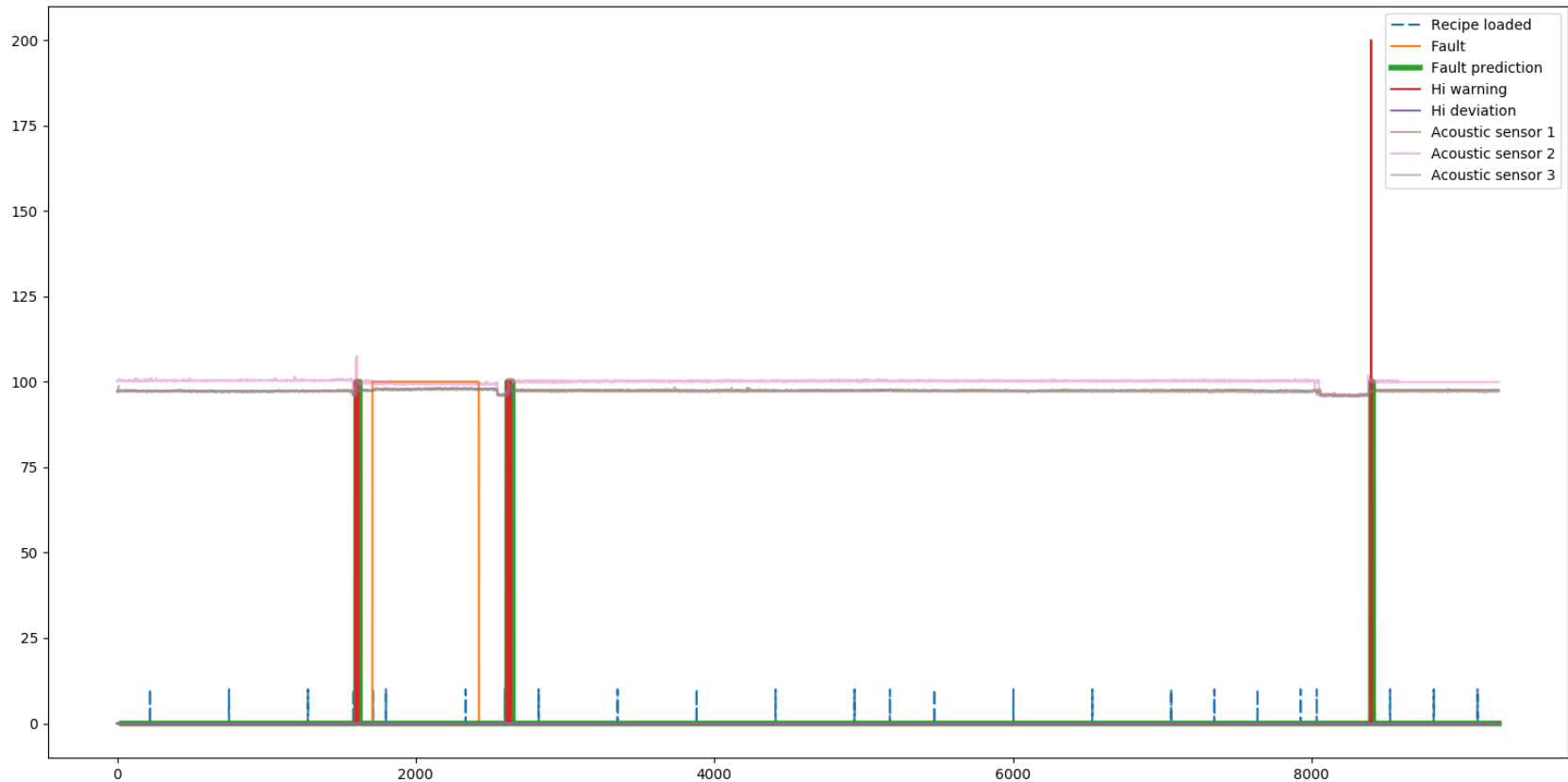


DLT – intrafactory – predictions on test dataset with audio, look ahead 1 sample – trial 3 zoomed



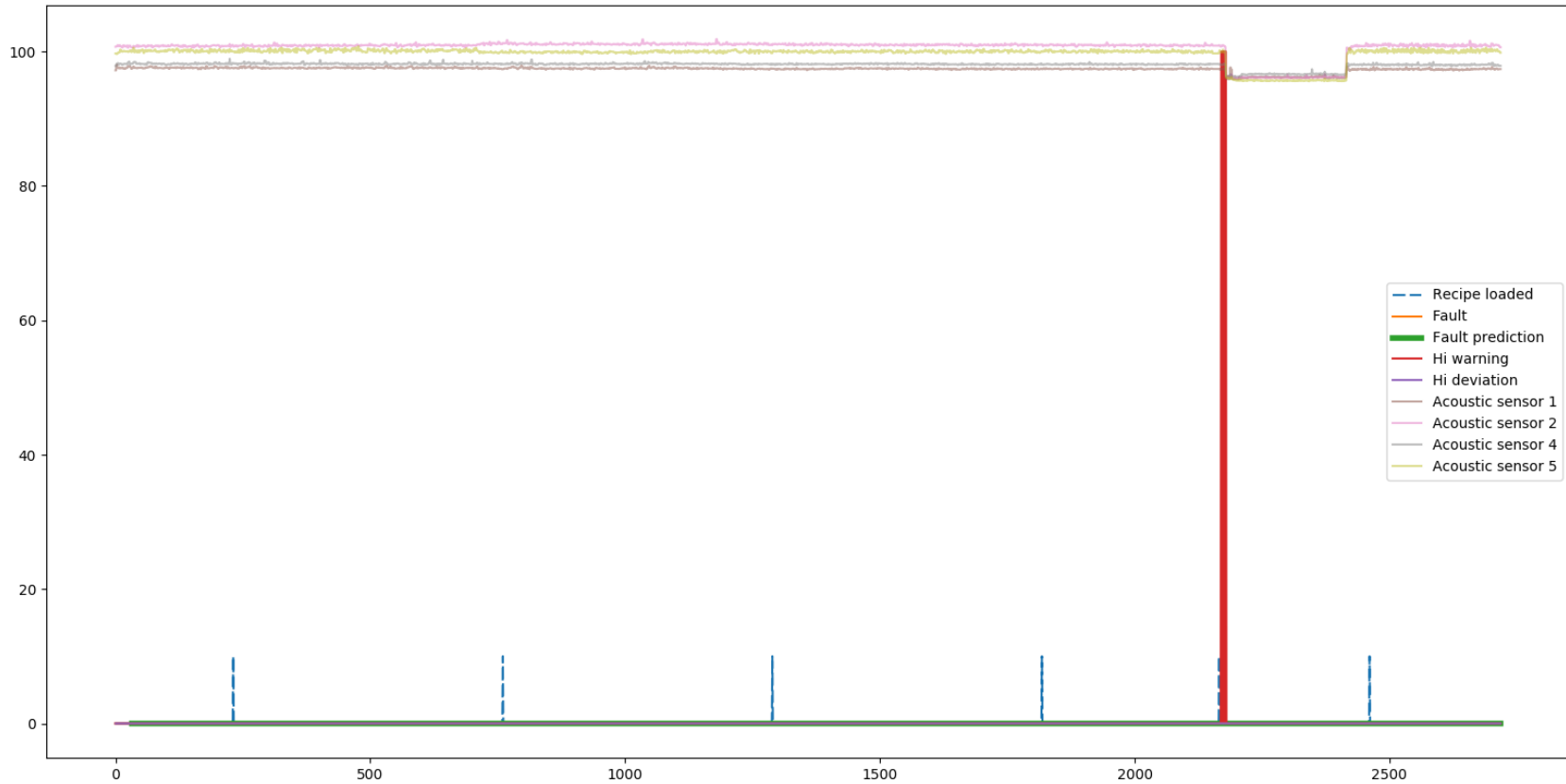


DLT – intrafactory – predictions on test dataset with audio, look ahead 1 sample – trial 4



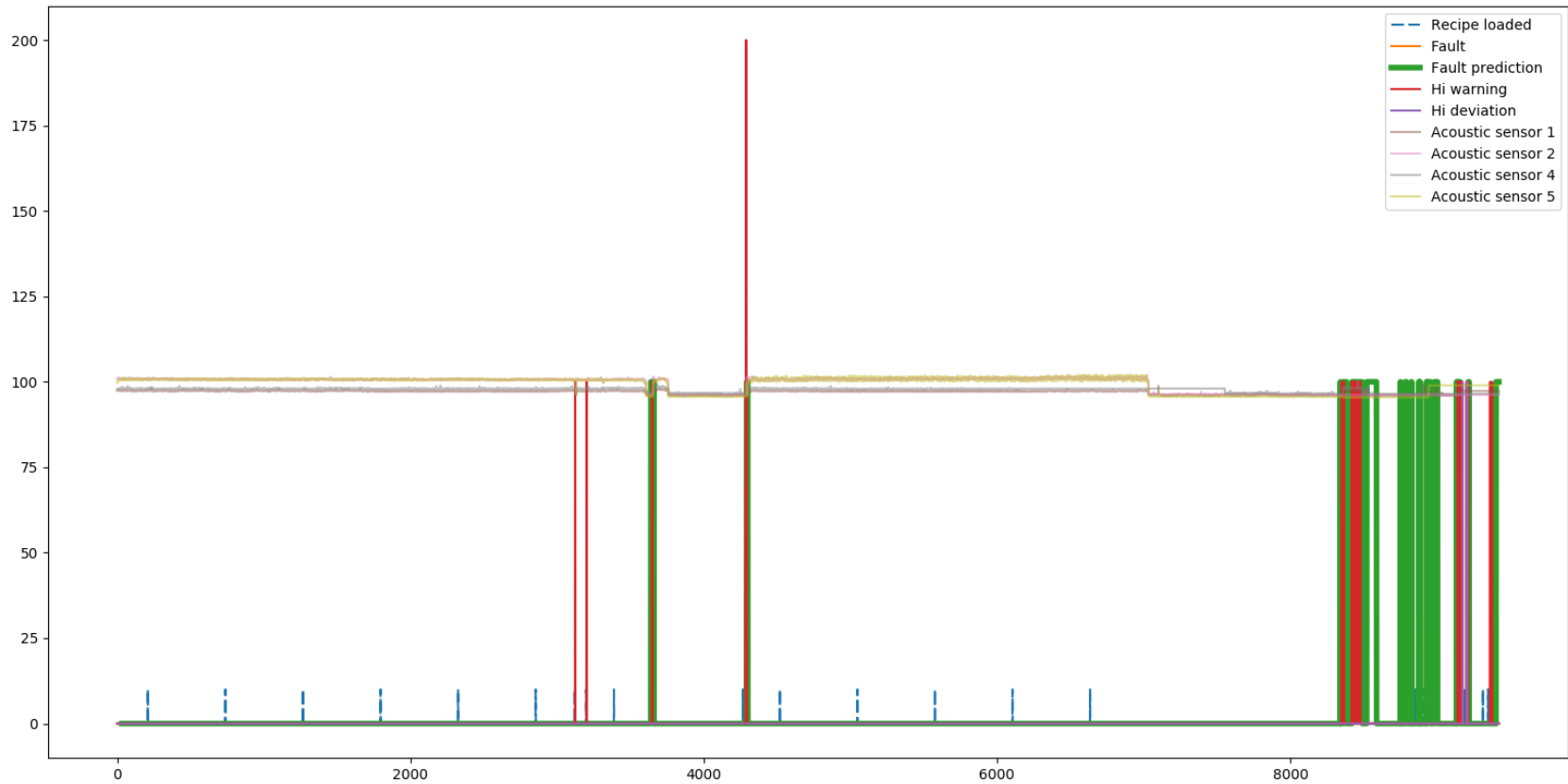


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples - trial 2



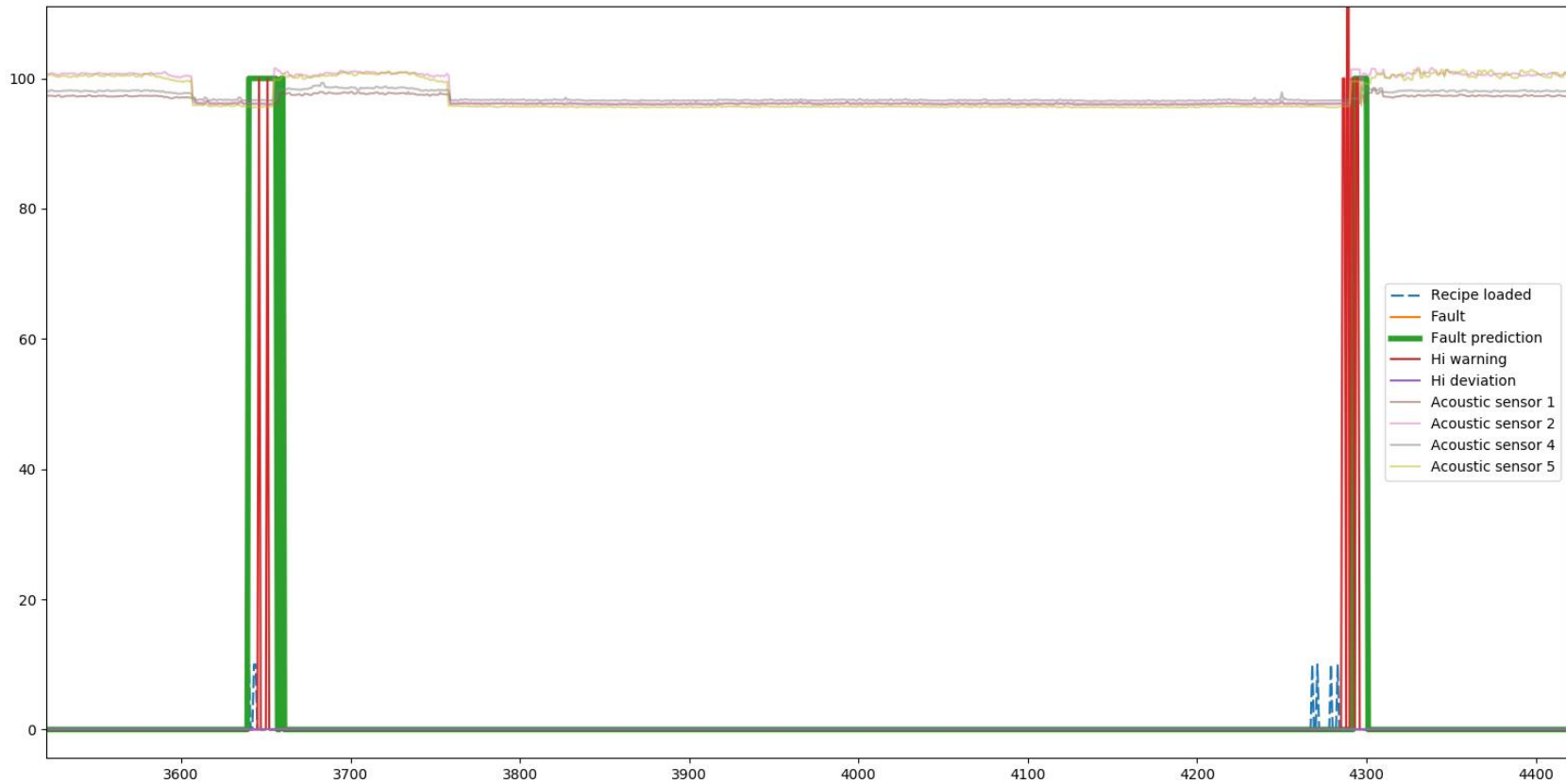


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples – trial 3



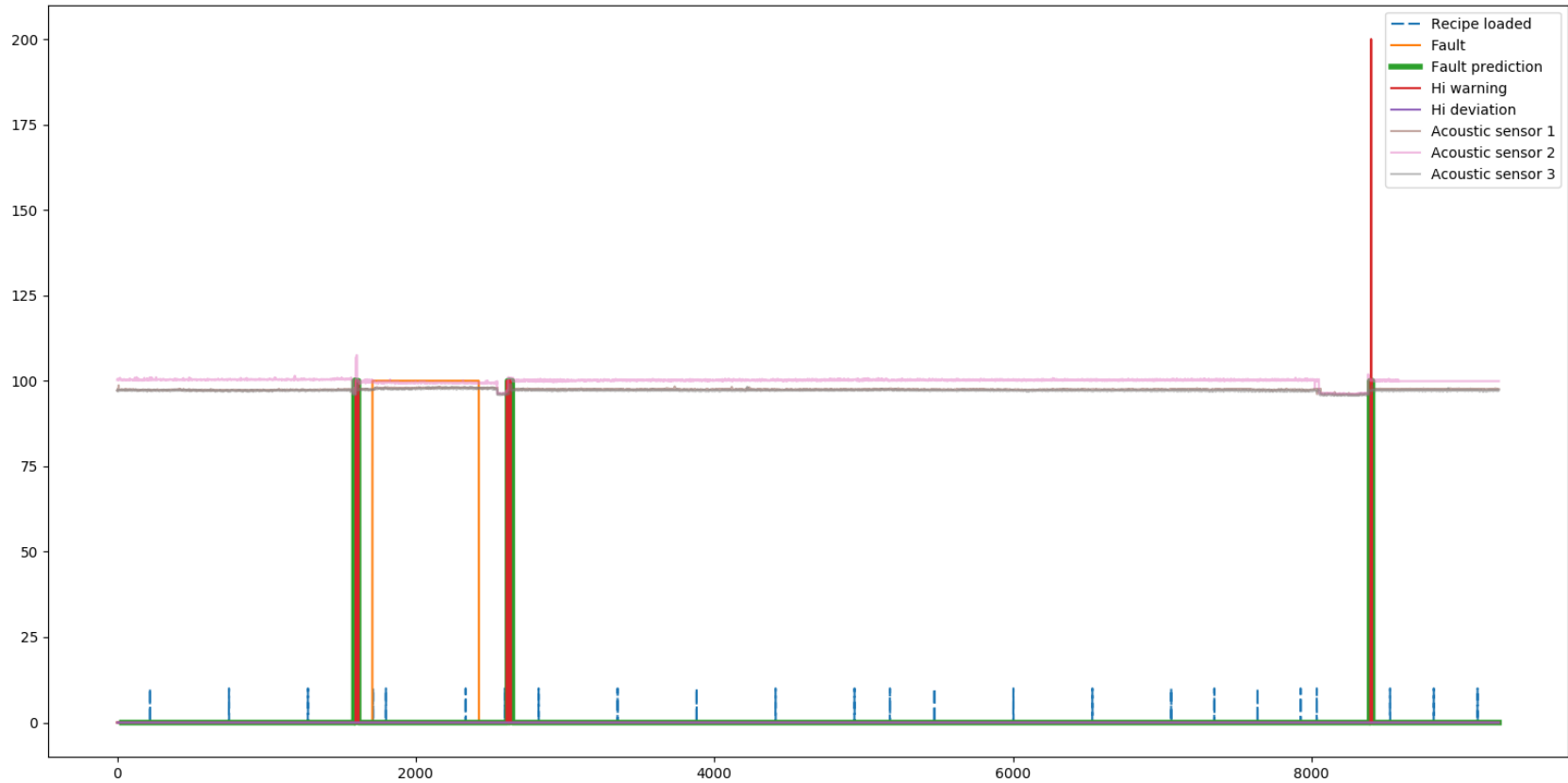


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples – trial 3 zoomed



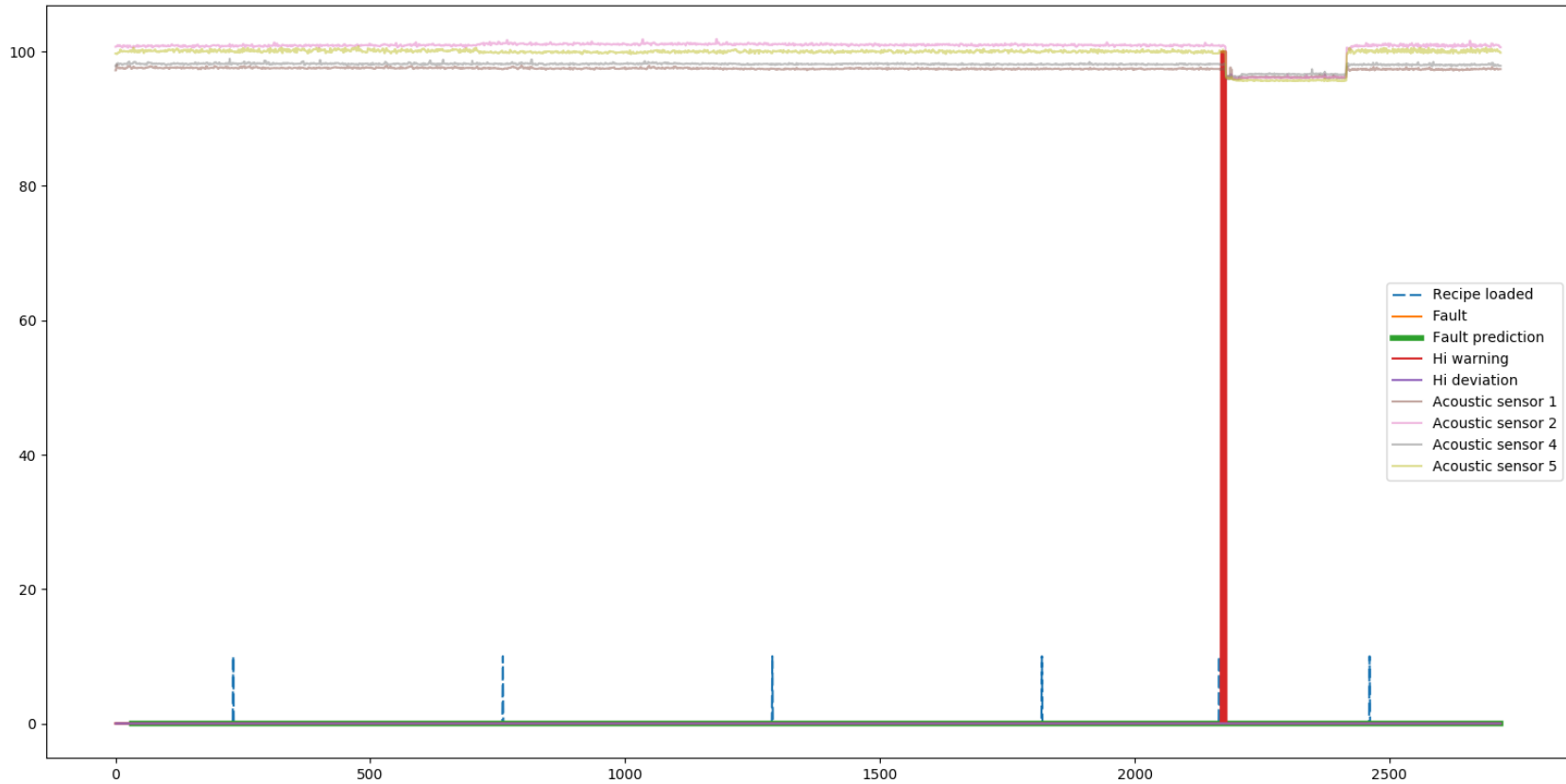


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples – trial 4



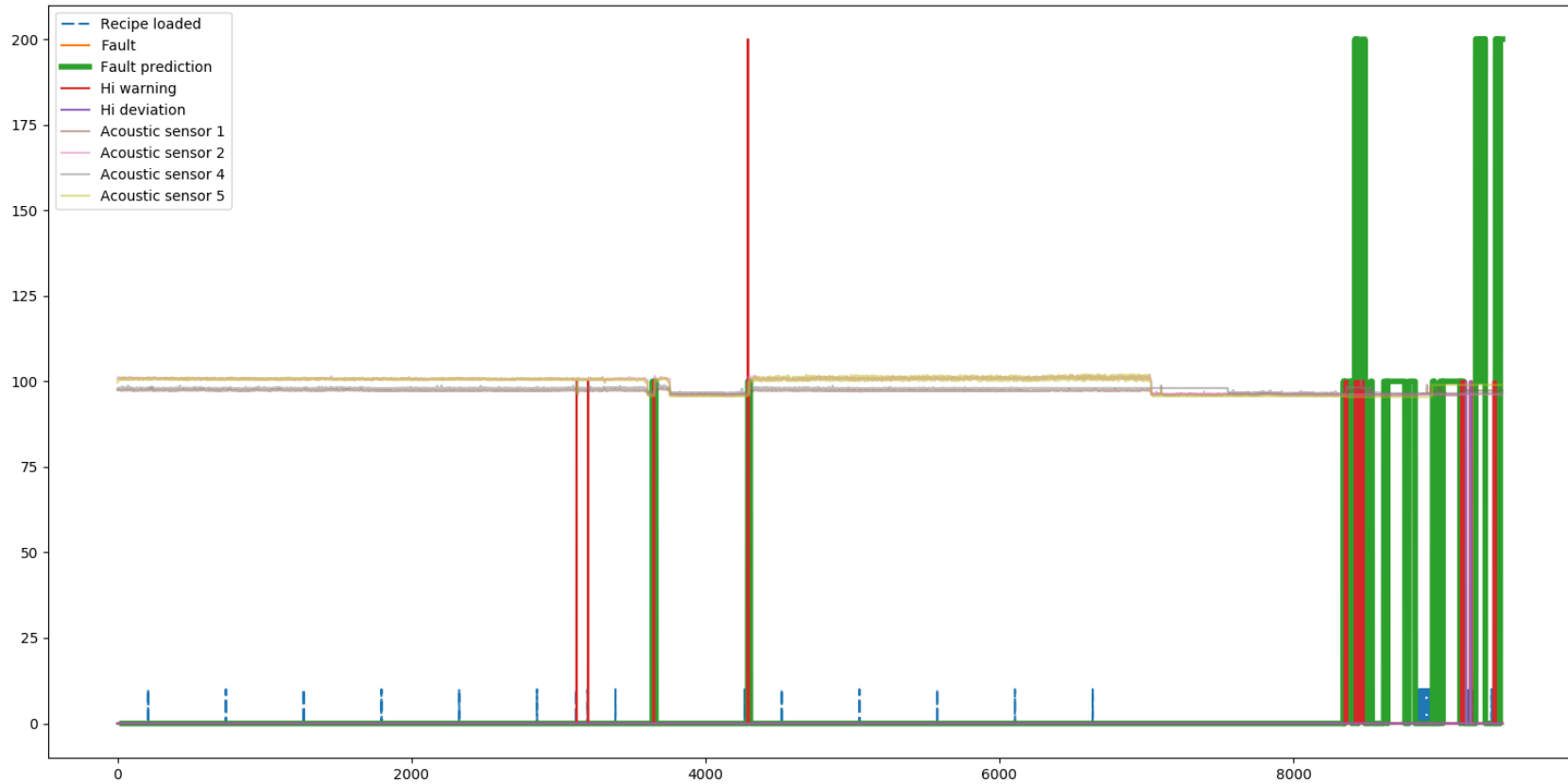


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples weighted – trial 2



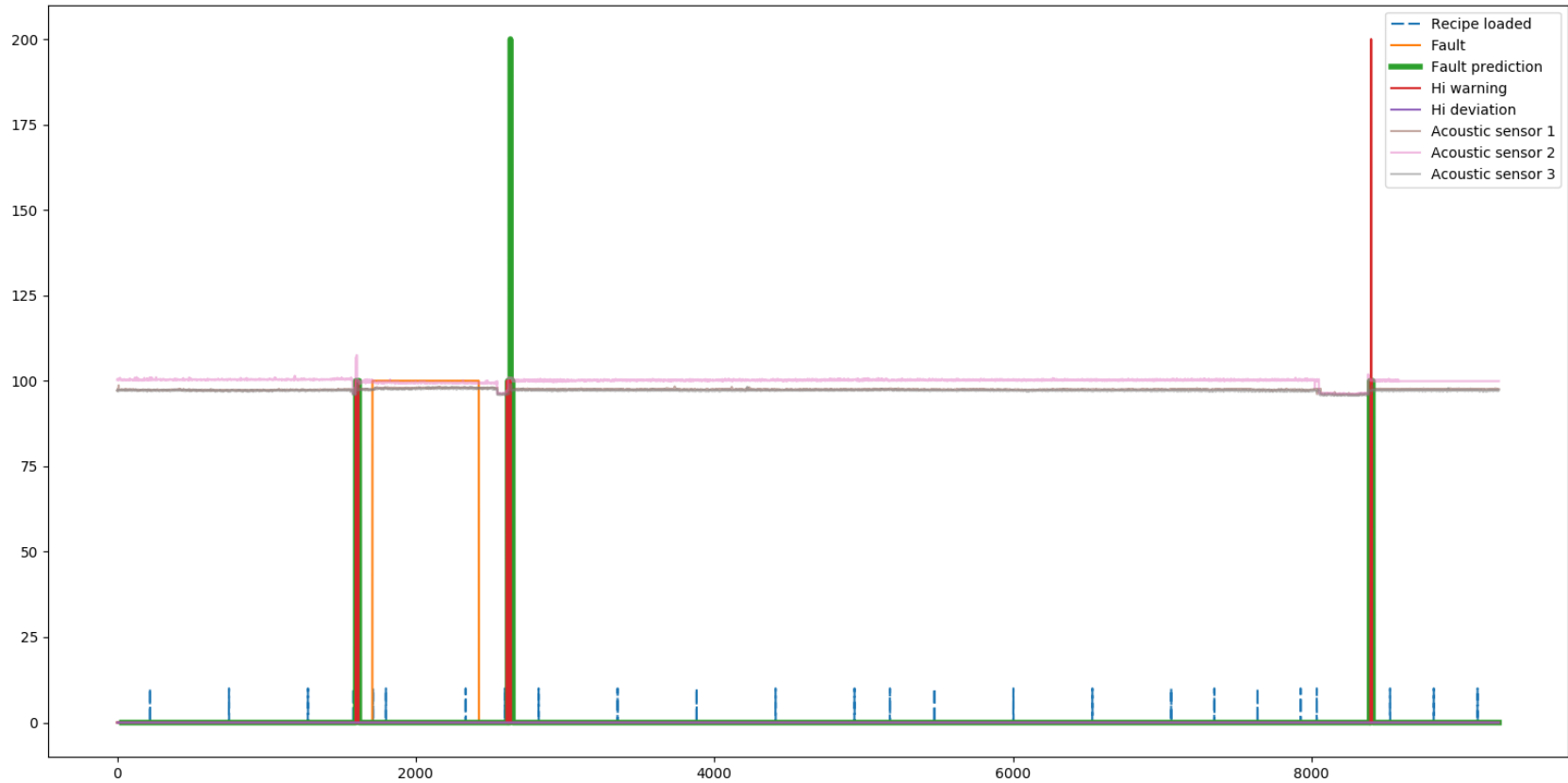


DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples weighted – trial 3





DLT – intrafactory – predictions on test dataset with audio, look ahead 16 samples weighted – trial 4





Deep Learning approach in COMPOSITION – inter-factory scenario (I)

- Provide intelligence to the agents that form the agent-based marketplace
 - marketplace parameters estimation
 - historical data
 - matchmaking policies
 - learned transaction
- Continuous learning
 - triggered by inter-factory agents and learning framework

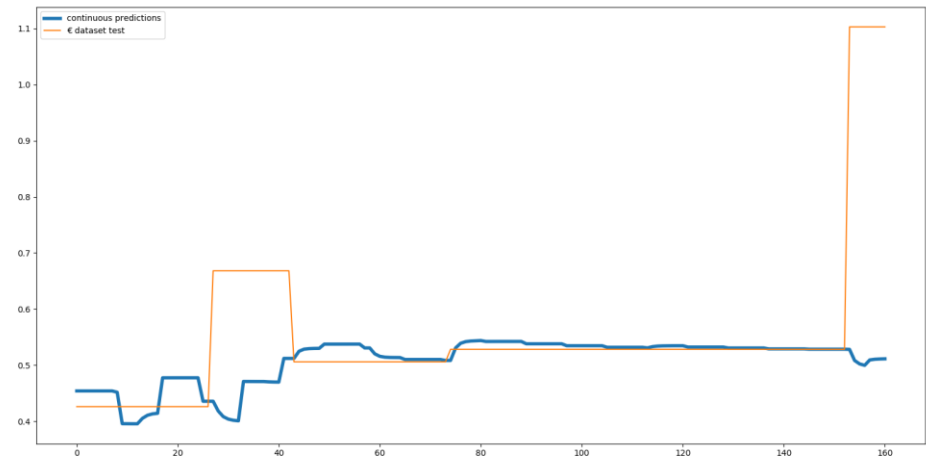
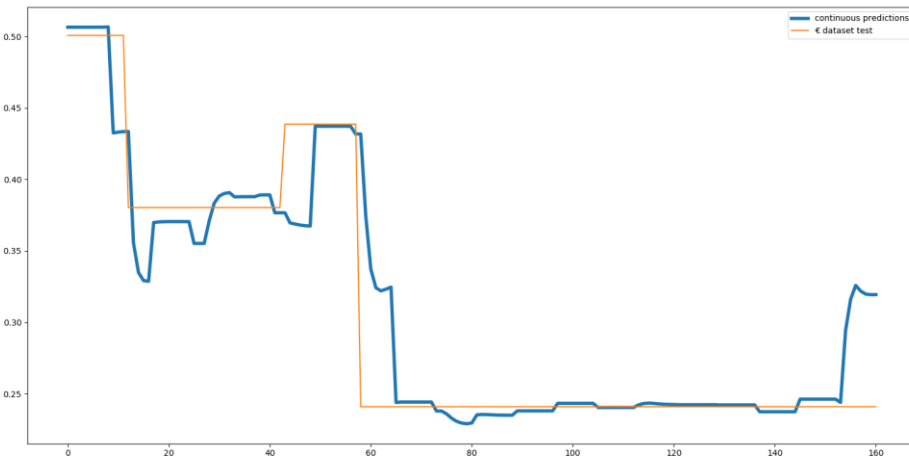
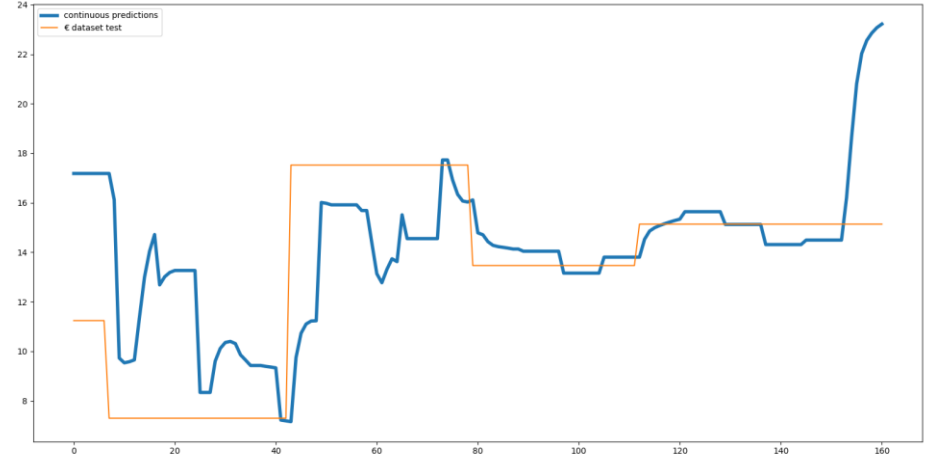
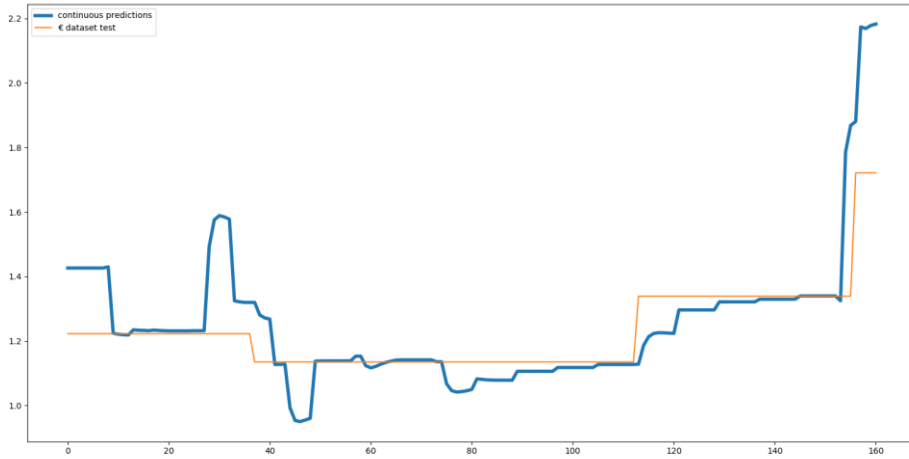


Deep Learning approach in COMPOSITION – inter-factory scenario (II)





Deep Learning approach in COMPOSITION – intra-factory scenario (III)





Future work (I)

- Experiment novel ANNs topologies
- More datasets to be evaluated alongside contextual markets information
- Enhance connection with matchmaking for extending the available information to be evaluated, consolidating the accuracy and improving reliability
- Extend current ontologies for supporting a broader range of possibly market correlated information, coming from heterogeneous sources (e.g. factors that influence stock market)



Future work (II)

Other challenges might be faced in different use cases by the DLT, based on different requirements:

UC-KLE-1: polishing machine predictive maintenance

- existing historical dataset with few recorded parameters
- no reported failures

UC-KLE-4: agent-based marketplace intelligence

- round trip time calculation and estimation by statistical models in simulation and forecasting tools
- bin fill level estimation with new deployed sensors



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