

Elnätens digitalisering

# Maskininläring, elkvalitet och smarta elnät

ENERYIELD

17 juni 2021



# Agenda

1. About Eneryield
2. Project overview and problem outline
3. Outline of work and results
4. Challenges, insights and future work

# Eneryield is a startup founded in 2019, based on several years of research within ML/DL applied in power systems

2013 - today

**Research: Deep Learning based Prediction, LSTM Deep Learning for Classification of PQ Disturbances, Classification of PD using Machine Learning, Deep learning for Earth Fault Classification and Source Localization**



**CHALMERS**  
UNIVERSITY OF TECHNOLOGY

MIDDLE EAST TECHNICAL UNIVERSITY



2018

**Innovation project**



2019

**Company founded**

**ENERYIELD**

# The team possess complimentary knowledge in ML/DL, power systems and business development

## Business development



**Johan Rådemar**

*M.Sc Entrepreneurship  
and Business design*

**Chalmers University of  
Technology**

## Technical development



**Karl Bäckström**

*Ph.D.c Computer  
Science*

**Chalmers University of  
Technology**



**Ebrahim Balouji**

*Ph.D Electrical  
Engineering*

**Chalmers University of  
Technology**



**Matilda Wikström**

*M.Sc Complex Adaptive  
Systems*

**Chalmers University of  
Technology**



**Mariam Helmy**

*M.Sc Physics of Data*

**University of Padova**

## Advisory Board



**Jonas Bergqvist**

**RISE**

**Expert in SaaS  
products**



**Özgül Salor**

**Gazi University**

**Professor Electrical  
engineering**



**David Storek**

**Ph.D Physics**

**Business coach at  
Chalmers Ventures**



# Project overview and problem outline



# Project participants

Project owner

**ENERYIELD**

Partners and reference group

 **Göteborg Energi**

 **Skellefteå Kraft**

  
**Jämtkraft**

**ABB**

 **Tekniska verken**

**e-on**

**VATTENFALL** 

 **HÄRRYDA ENERGI**

 **UNIPOWER**

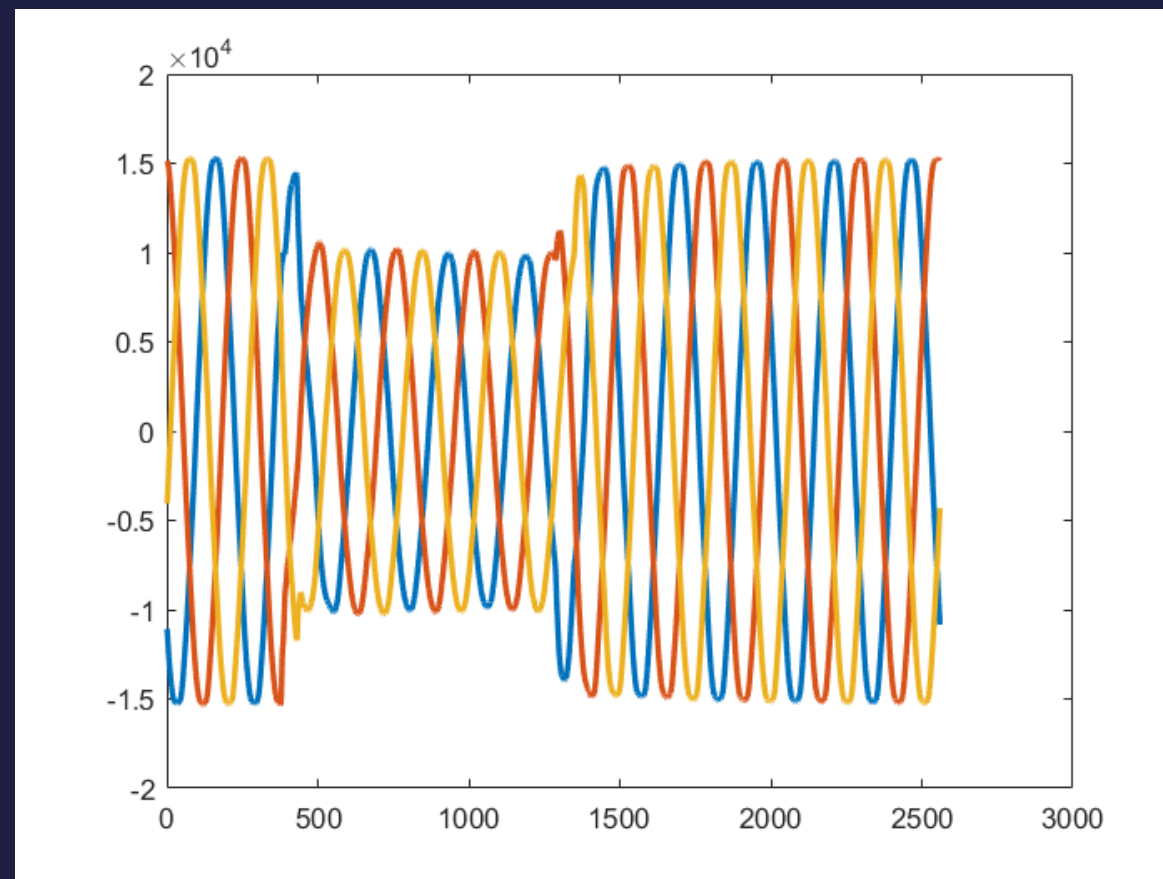


**PQ meters**  
**Event snapshots, RMS values**  
**Comtrade, PQDIF**



# The project targets challenges with power quality disturbances

**Event X**



**What was the  
event type?**

**What was the  
underlying  
reason?**

**Where did it  
occur?**

**Minimize manual time  
consuming fault  
identification**

**Provide deeper  
insights of grids  
health**



# The goal of the project

**Develop a proof of concept for  
more in depth analysis of PQ  
disturbances**

**Accurate and reliable  
classification of PQ  
disturbances**

**Suggest plausible  
underlying reason of  
the disturbance**



**Present the results  
in a PQ report**



# Outline of work and results

# Overview work process and detailed level of analysis

## High level classification

*Sag / Swell / Transient / Earth fault / Interruption*



## Fine-grained PQ event classification / Root cause analysis

*Event is linked to underlying error e.g. phase to ground or motor induction*

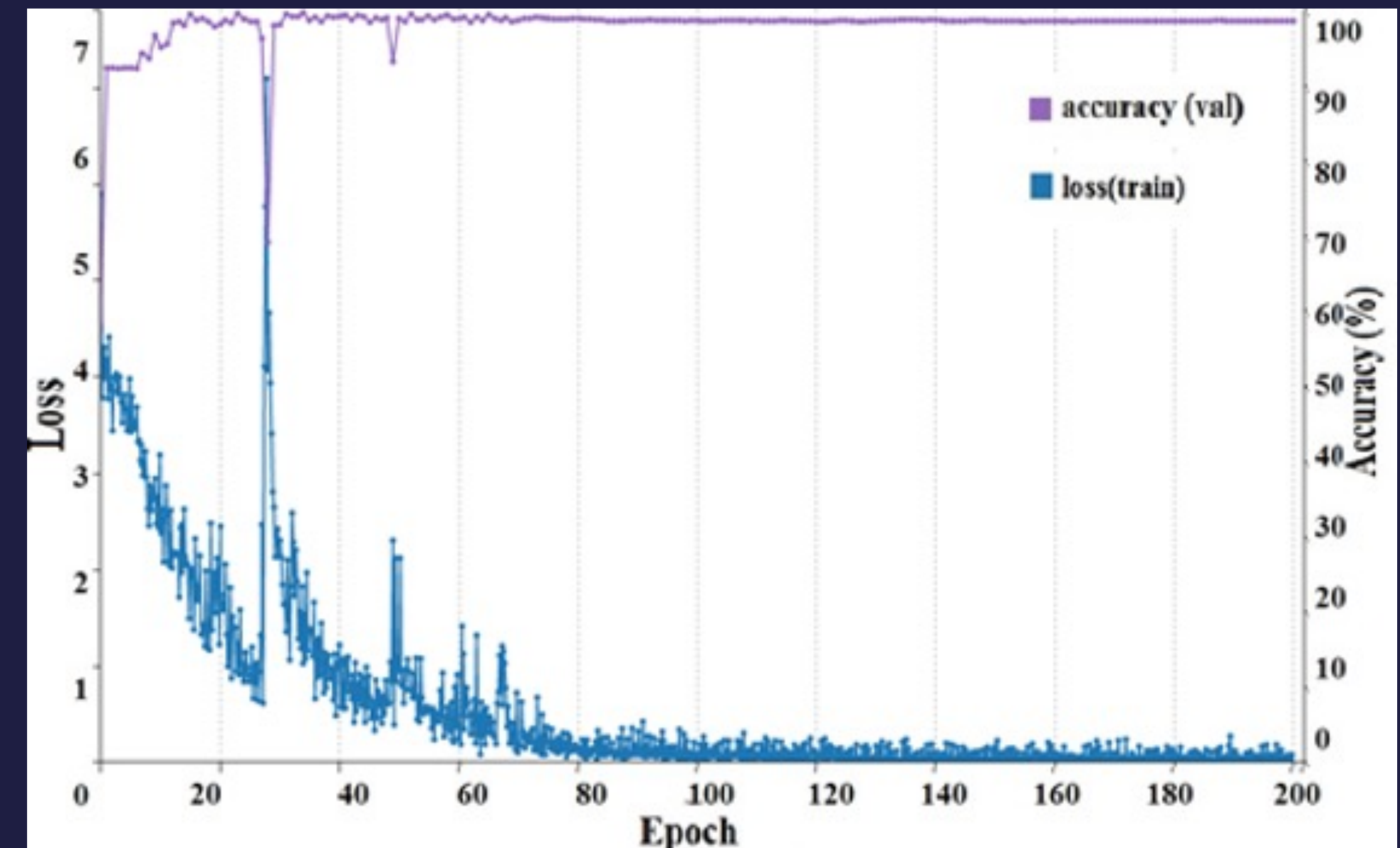
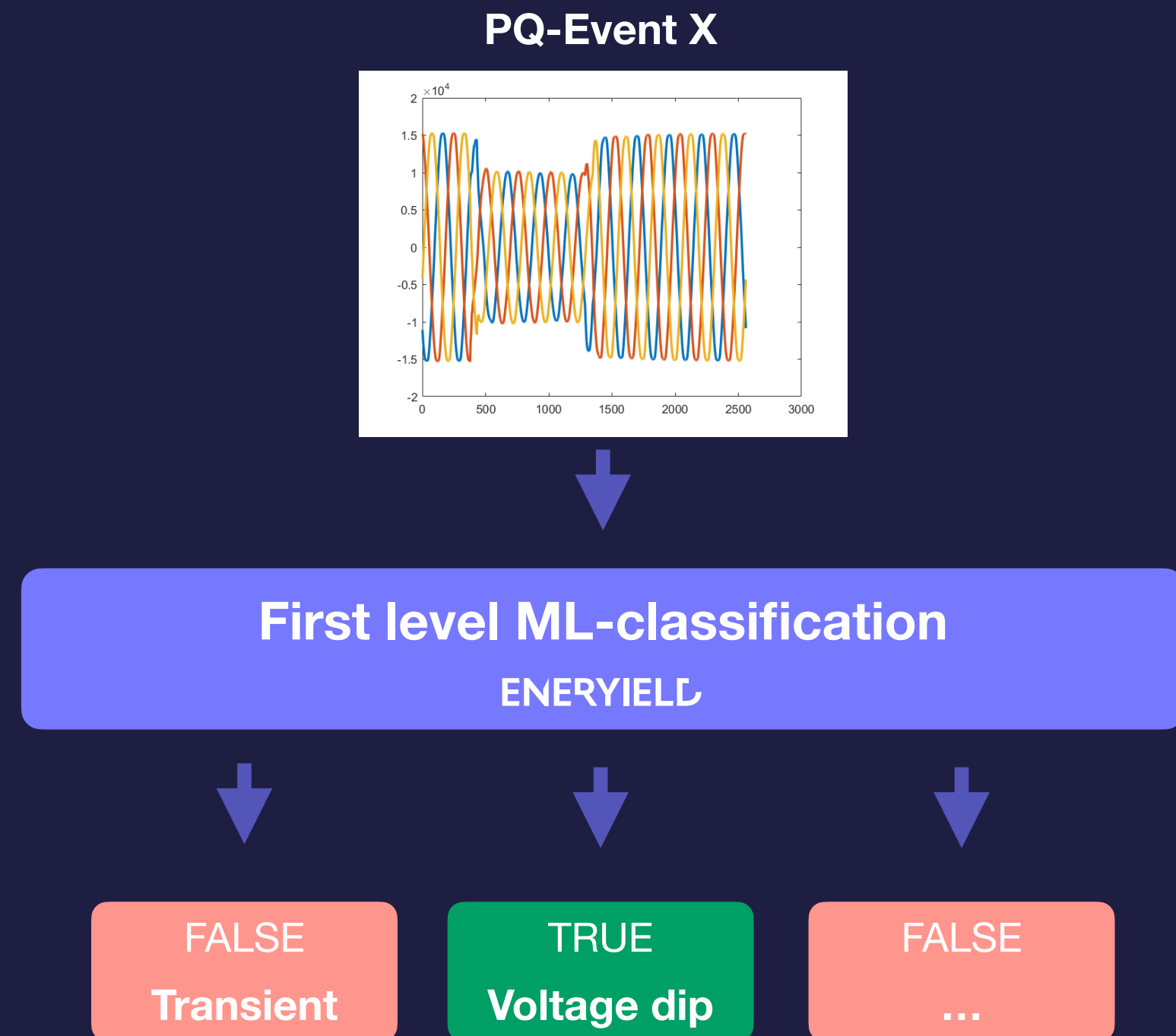


## Automatic identification of direction

*Eg. Determine the direction of events using waveforms & ML*

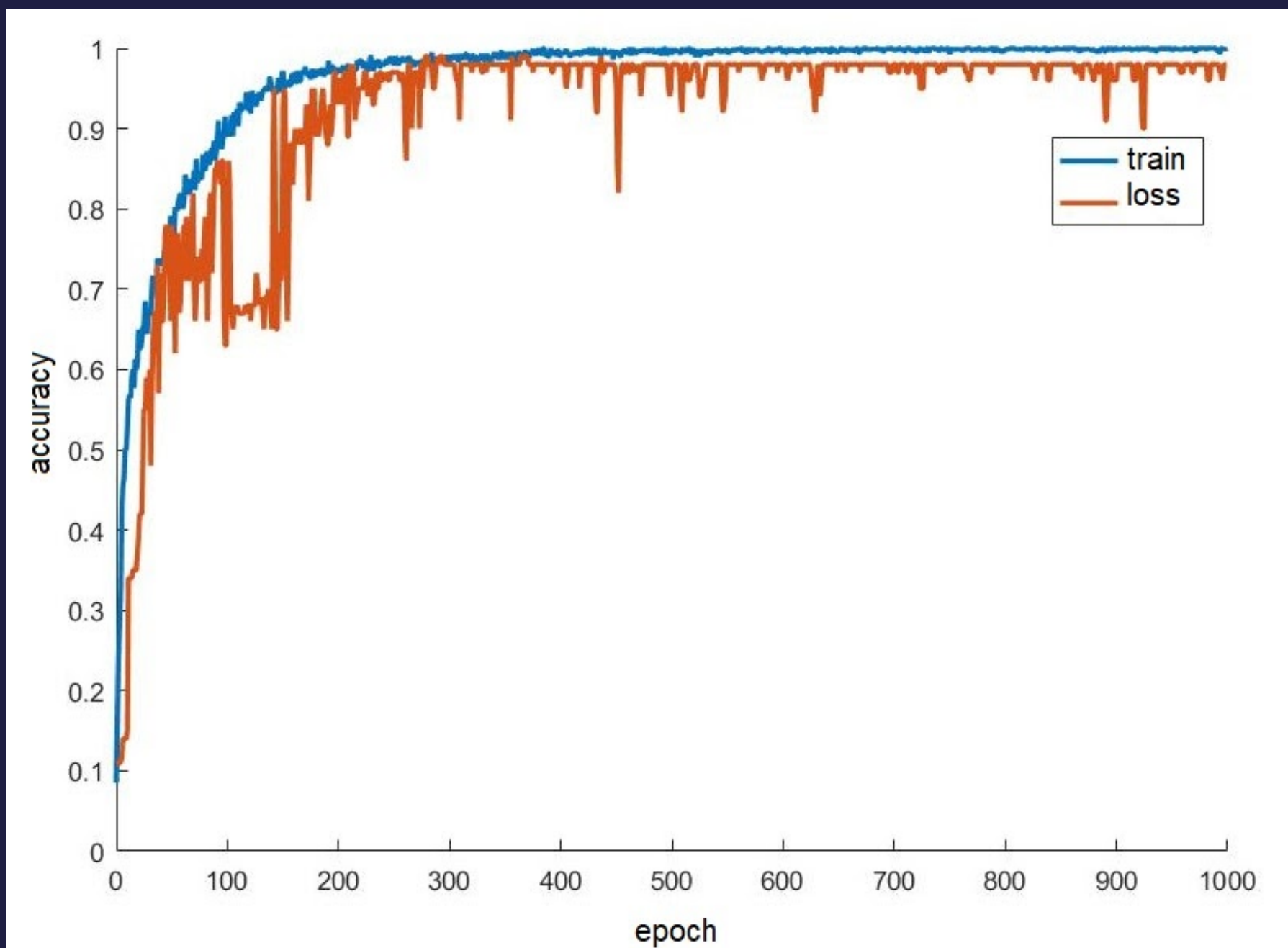
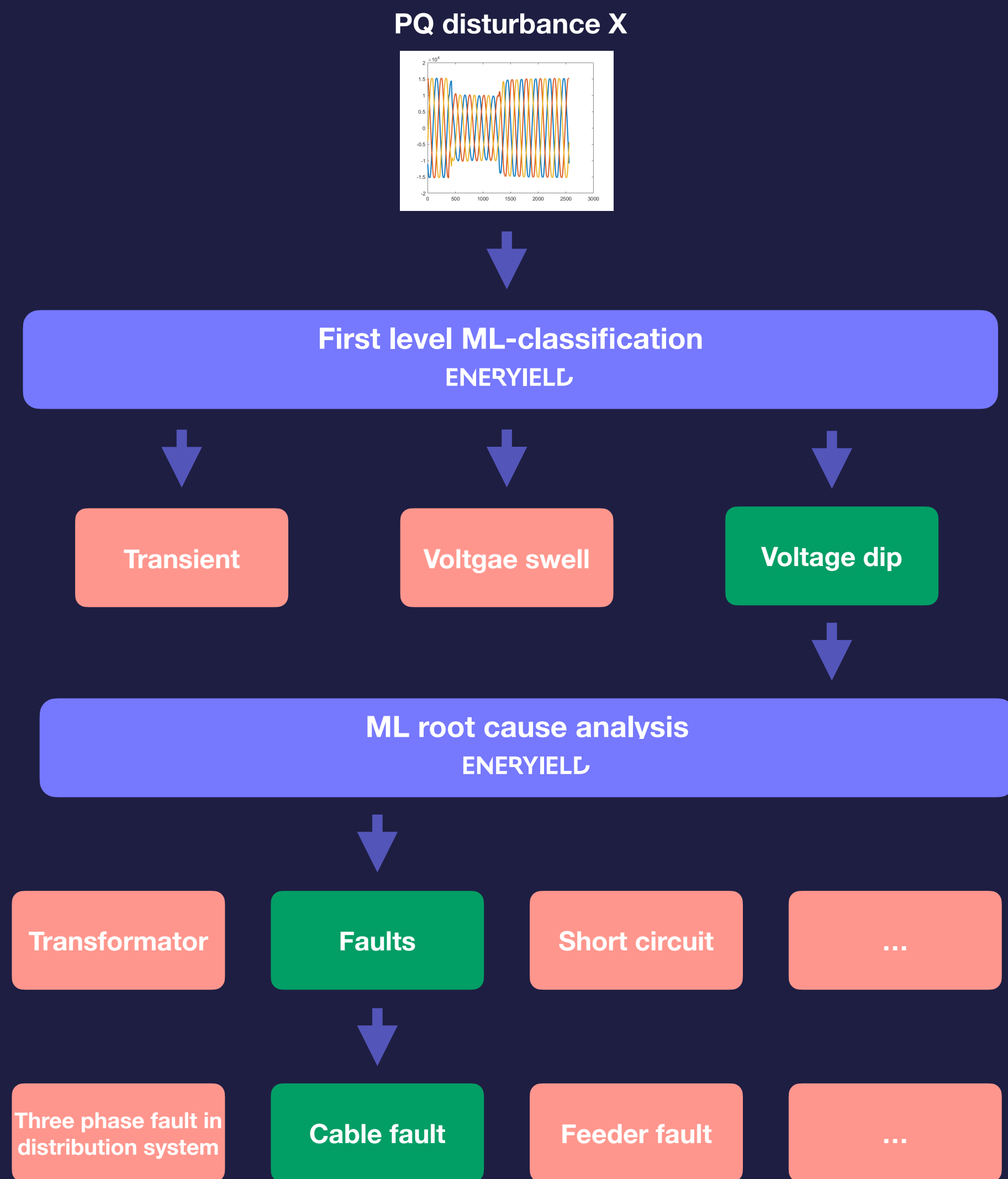


# Method - High level classification of PQ Events



The system quickly reaches high accuracy of 99% on the test data, and low error, with stable convergence.

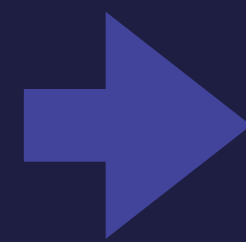
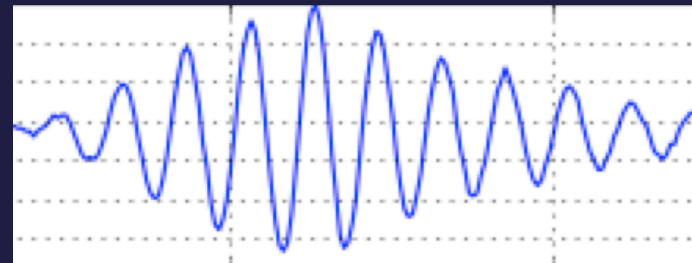
# Method - Root cause analysis of voltage dips



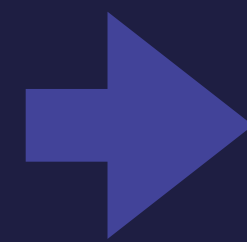
System reaches 95% accuracy on the test data.



# Method - Direction Identification

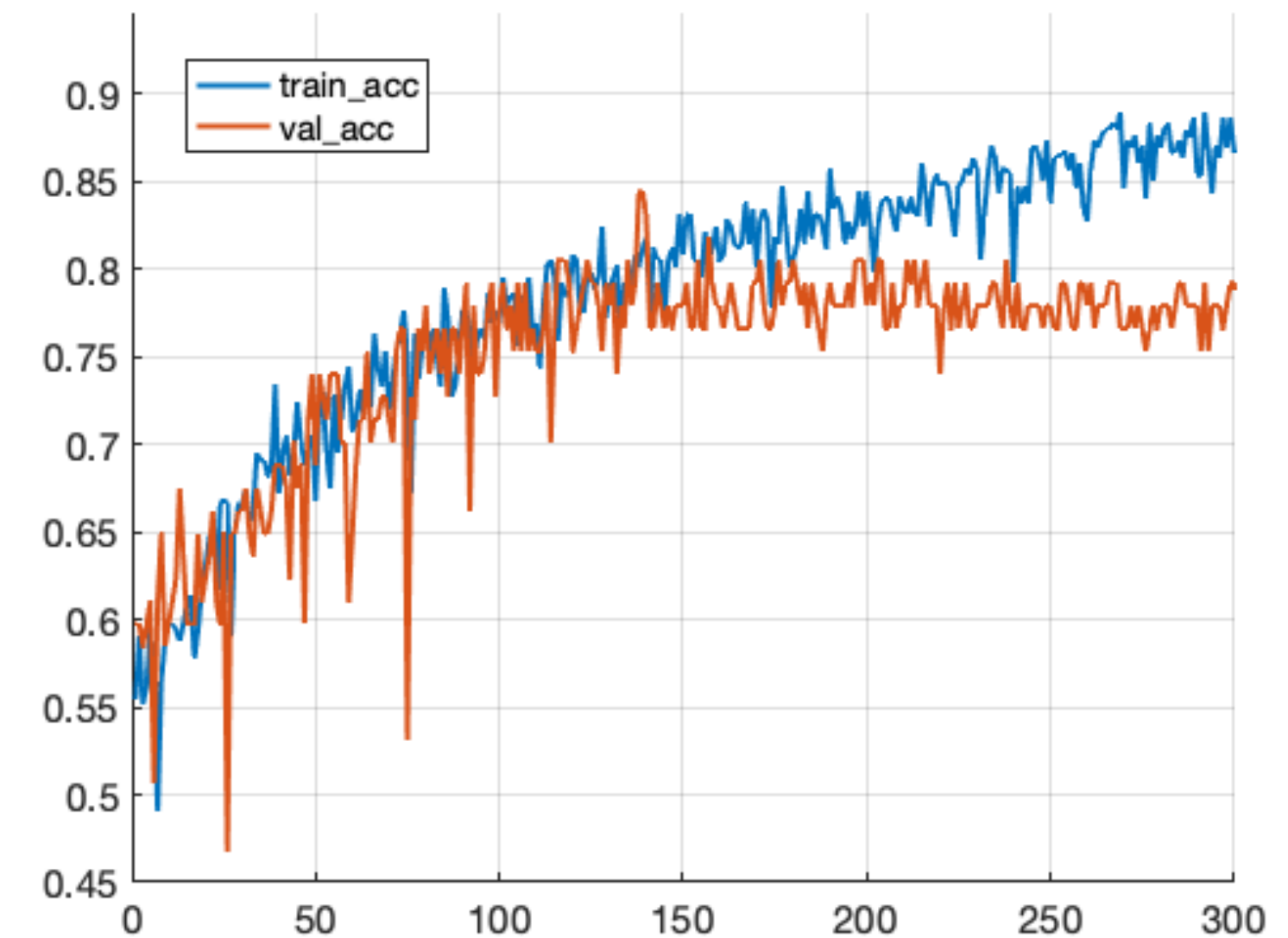
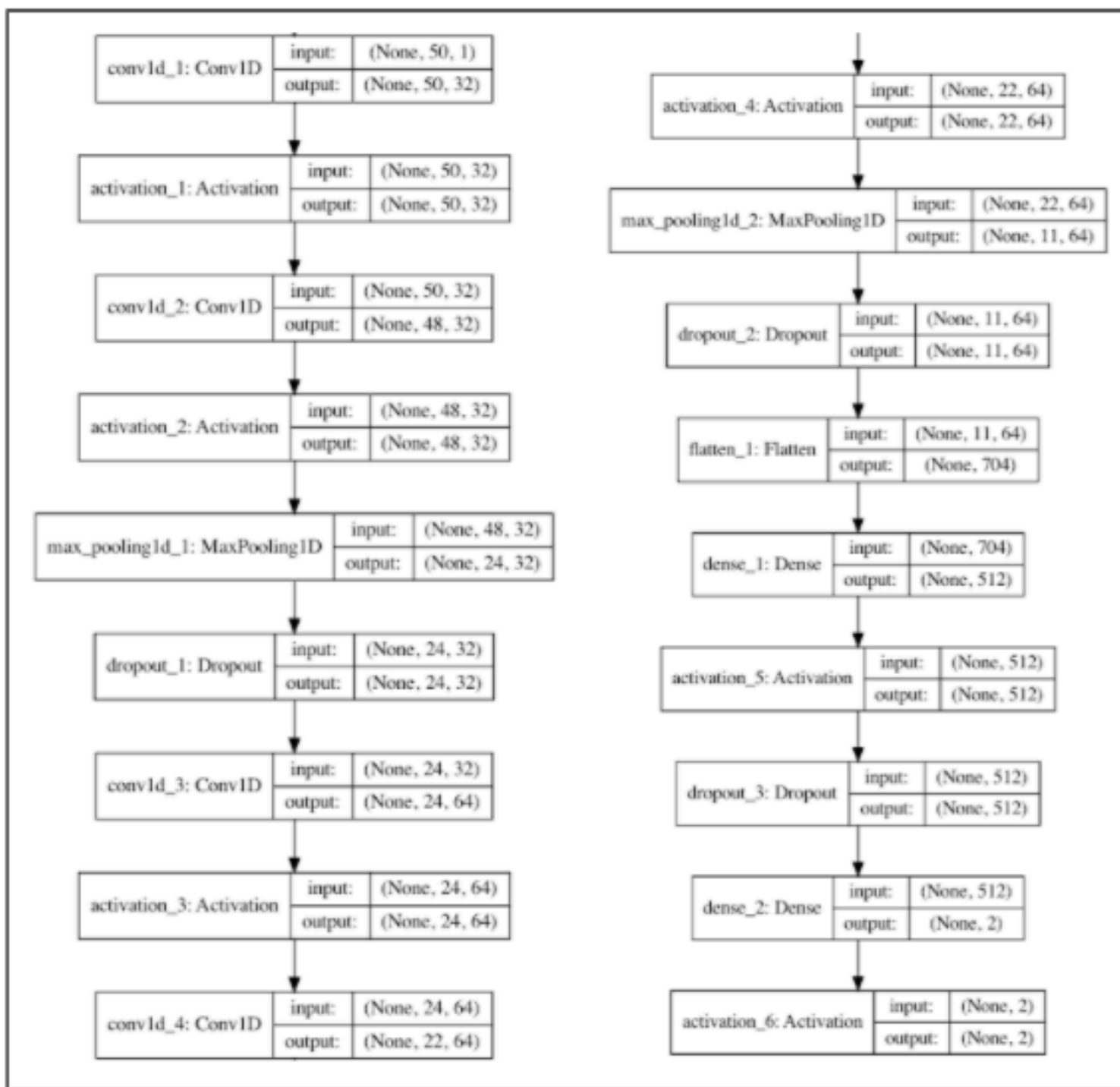


NN



Up/down  
stream

## NN Architecture



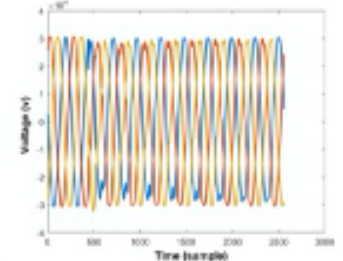
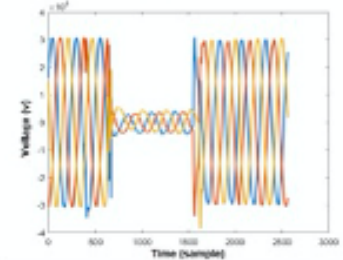
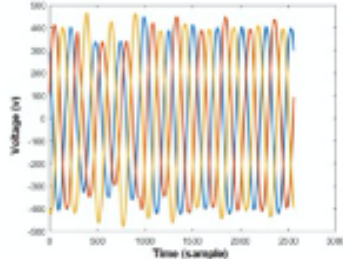
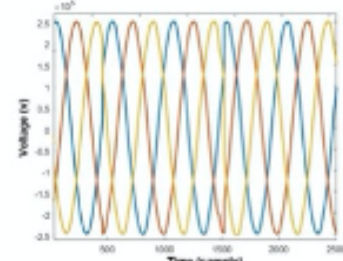
Stable training execution. Test performs  
~85% at best, but stagnates at ~80%

Machine learning based power quality analytics

Classification, root-cause and direction  
identification of power quality events

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Voltage dips

Date & time	Root-cause	Direction	Event figure
2020-01-05 18:45	Transformer energization	Upstream	
2020-01-07 16:14	Three phase fault in distribution system. Followed by transformer saturation	Upstream	
2020-01-13 01:11	Autoreclosure in distribution feeder	Upstream	
2020-01-24 07:23	Transformer saturation	Downstream	



# Challenges, insights and future work

# Challenges and insights - Data export

Data aggregation is a bottleneck and it is important to bridge this challenge for large-scale implementation of future data-driven solutions.

## Today

1. Copy the entire database (SQL file)
2. Manually export PQ events from PQ software (Excel files)
3. Dedicated export module (Comtrade files)

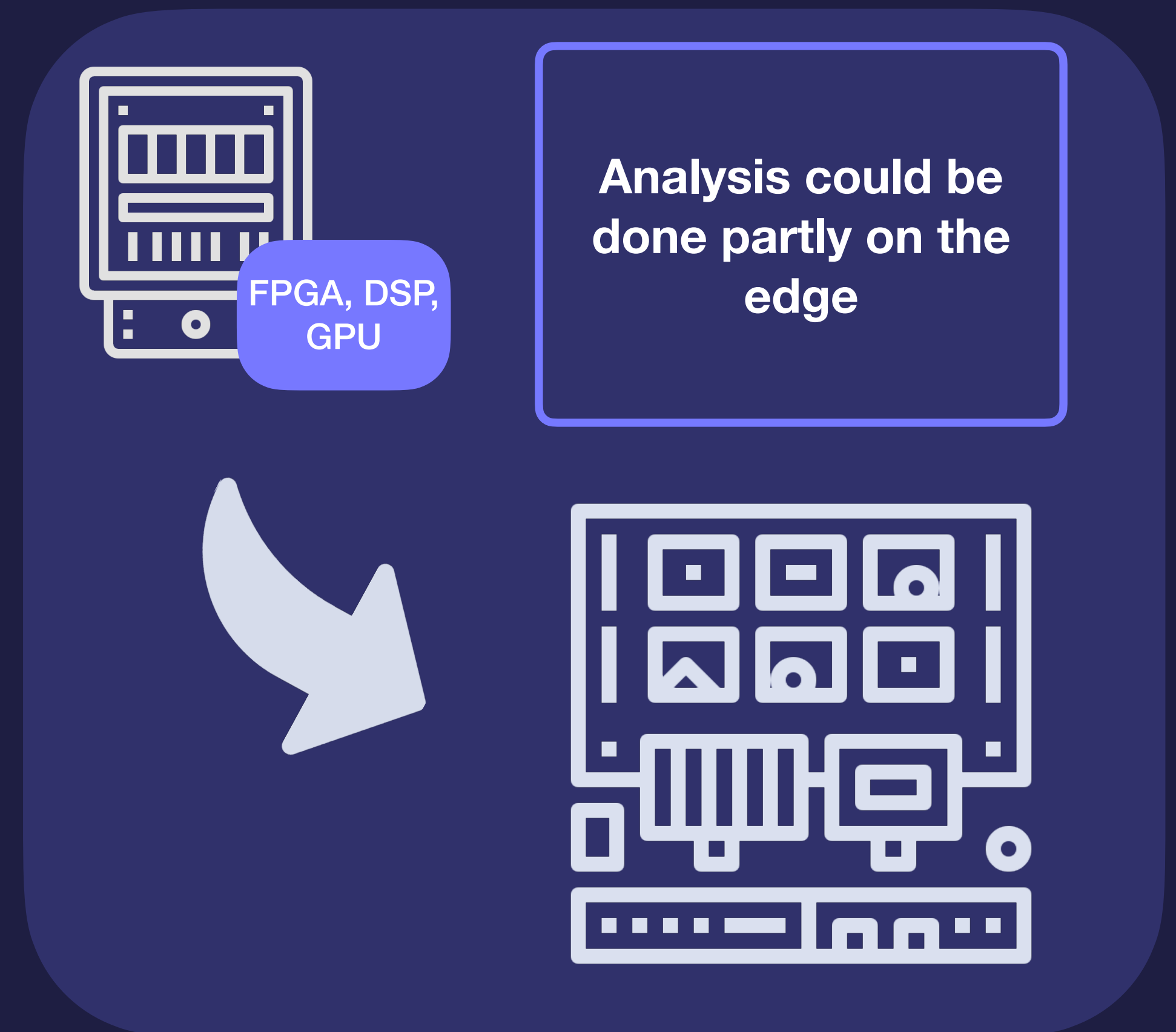
## Future

1. Standardised database with structures for describing meta-information (i.e. frequency, pre-trig time...)
2. Import and export modules that ensure standards are maintained and that the desired recording intervals or batches of data are easily and efficiently exported.



# Challenges and insights - Analysis on the edge

1. Digitised power systems will need continuous improvements of higher sampling frequencies
2. Sufficient technology for measurement and communication do exist
3. Combining Machine Learning on the edge and centrally will optimise tomorrows data driven methods



# Future work

**Developing the root  
cause analysis further**



```
graph LR; A[Developing the root cause analysis further] --> B[Expand with more data, including more type of root causes]; A --> C[Incorporate in existing software or create API]; A --> D[Investigate how data can be more efficiently exported];
```

**Expand with more data,  
including more type of root  
causes**

**Incorporate in existing  
software or create API**

**Investigate how data can be  
more efficiently exported**





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