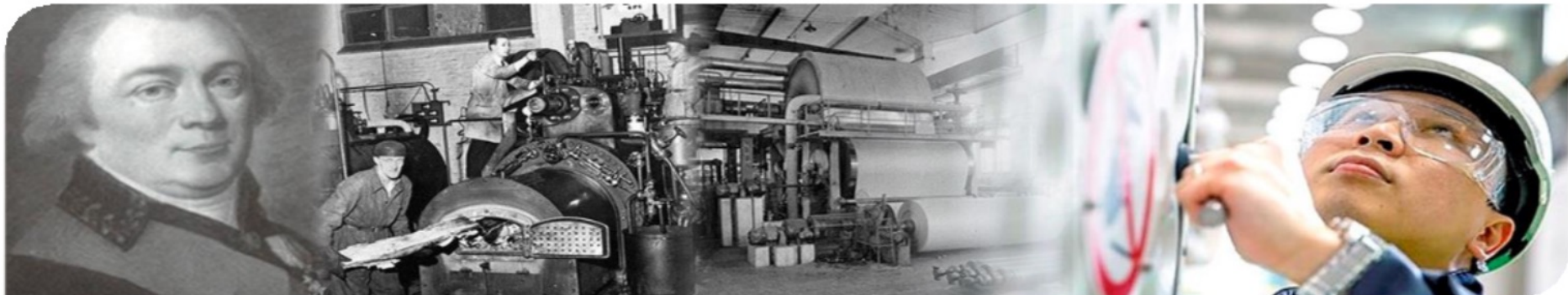
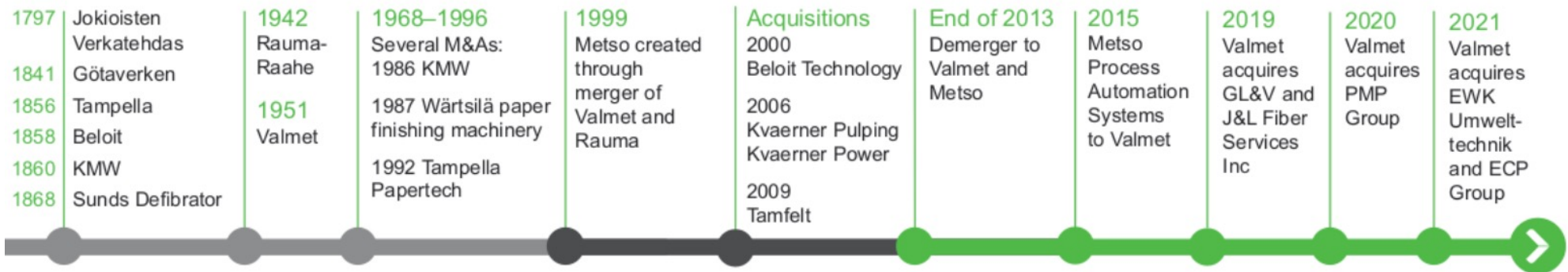


AISA – Demo  
Artificial Intelligence for Situation Awareness  
Digital Twin – Server / Client(s)  
Web based user interface for test, tune & validate

Dr. Karaila – [mika.karaila@valmet.com](mailto:mika.karaila@valmet.com)

# Progress built on 220 years of industrial history

## From cloth making to high-tech processes




# This is Valmet



**Unique offering**

- Market's widest offering combining process technologies, services and automation
- Research and development spend EUR 75 million in 2020



**Market leadership**

- Leading market position in all markets
- Pulp #1–2
- Energy #1–3
- Board #1
- Tissue #1
- Paper #1
- Services #1–2
- Automation #1–3



**Strong global presence**

- Approx. 100 service centers
- 98 sales offices
- 43 production units
- 16 R&D centers
- 14,000 professionals

EMEA	9,200
China	1,900
North America	1,500
Asia-Pacific	900
South America	500

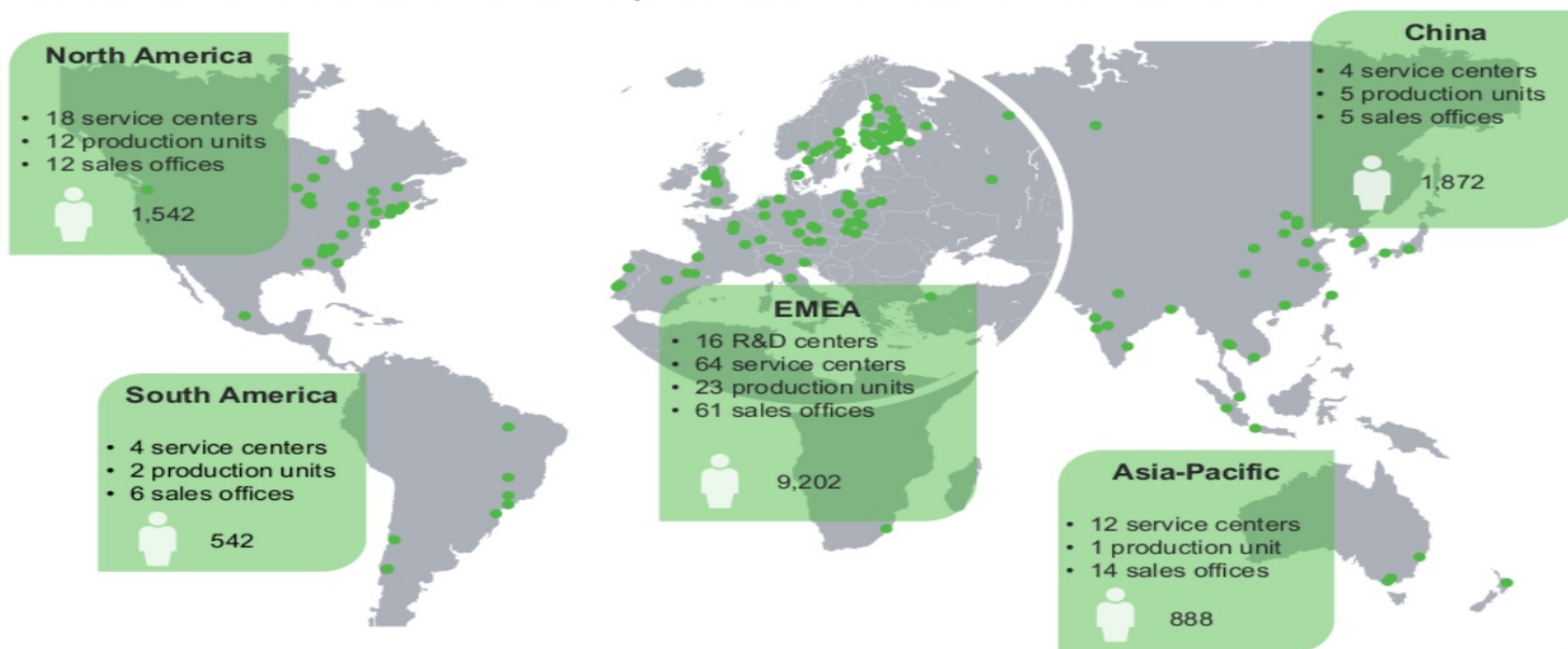


**Leader in sustainability**

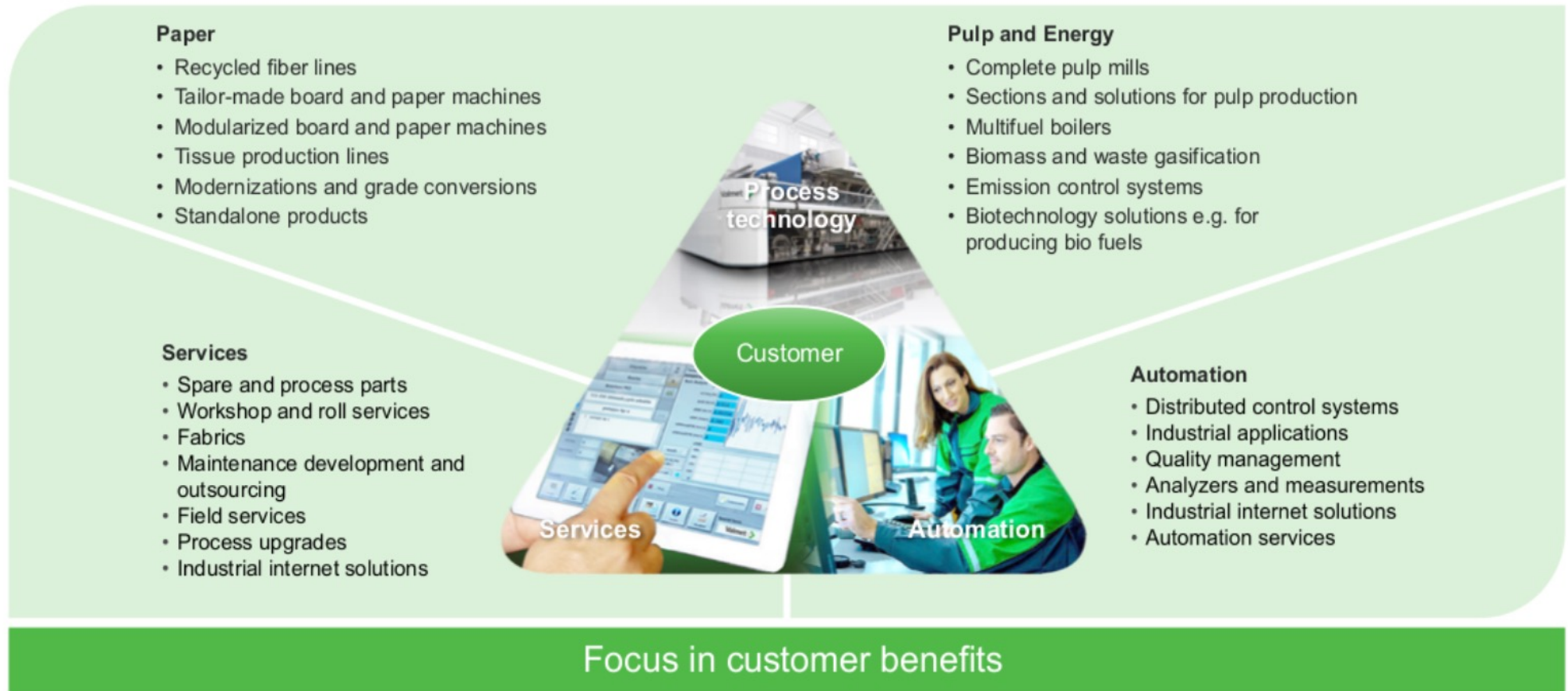
- Eight consecutive years in Dow Jones Sustainability Index
- Highest ranking in Ecovadis sustainability assessment
- B rating in CDP climate program 2021

# Strong global presence is a good platform for growth

100 service centers, 98 sales offices, 43 production units, 16 R&D centers



# Valmet Offerings by business lines

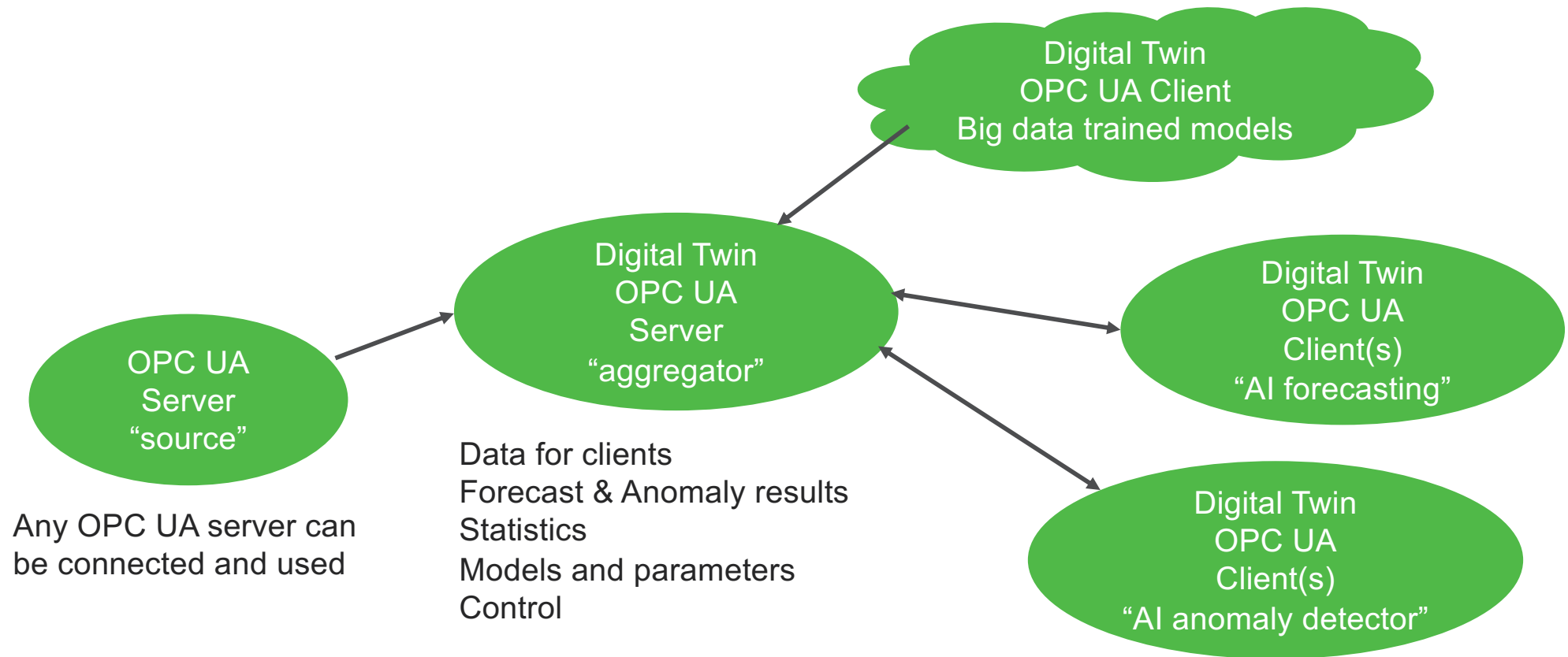


# Digital Twin for Industry

- Easy to install & configure, can be run on site or on cloud, flexible architecture
- Digital Twin Server:
  - Aggregate values
  - Provide historical access to data
  - Digital Twin Alarms for Anomalies
  - Forecast & other results will be stored to server
- Digital Twin Client for forecasting, anomaly detection & regression:
  - Run model: train/fit/predict according data from the Digital Twin Server
  - Write back forecast, accuracy & error
  - Raise: Anomaly detection alarm
  - Implementation with Python / Javascript / other suitable programming language
  - Keeps/hides algorithm/model inside client code

# Digital Twin OPC UA based architecture

OPC UA as enabler: HA communication for data, A&C for anomaly & UA file for model & parameters



# AISA Demo phases

- Phase 1:
  - Server – client infrastructure working
- Phase 2:
  - Clients updating predictions / anomalies to server
- Phase 3:
  - Pretrained models for industry?
  - Federated learning
  - Template for Digital Twin client
    - Forecast
    - Anomaly detector
    - Multivariable regression



# Digital Twin – RNN forecast

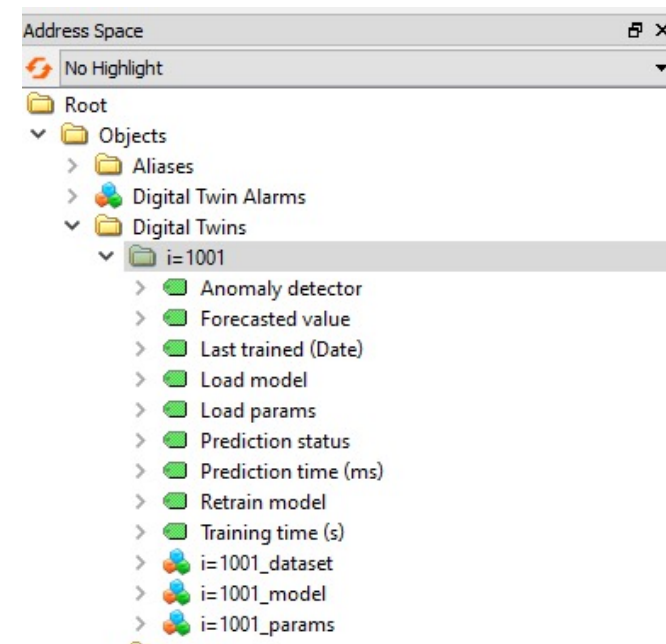
## Findings: TensorFlow (JavaScript)

- First initial training time consuming => run predictions with created model
  - From 40s training -> less than 100ms prediction time
- Re-train model, interval 300s
  - If prediction error with real measurement too big => force re-training
  - In practice NOT USABLE
- RNN parameters needed for each instance
  - General parameters are not matching well => tuning needed for each
  - IDEA: Pre-defined parameters for specific measurements in industry
  - IDEA: Pre-trained models for industry
- TODO:
  - Anomaly detector (Autoencoder first, but it will be not enough, some classification will be needed)
  - Multi-variable regression for complex predictions

# Digital Twin Server – OPC UA based management

Aggregate values from the other server

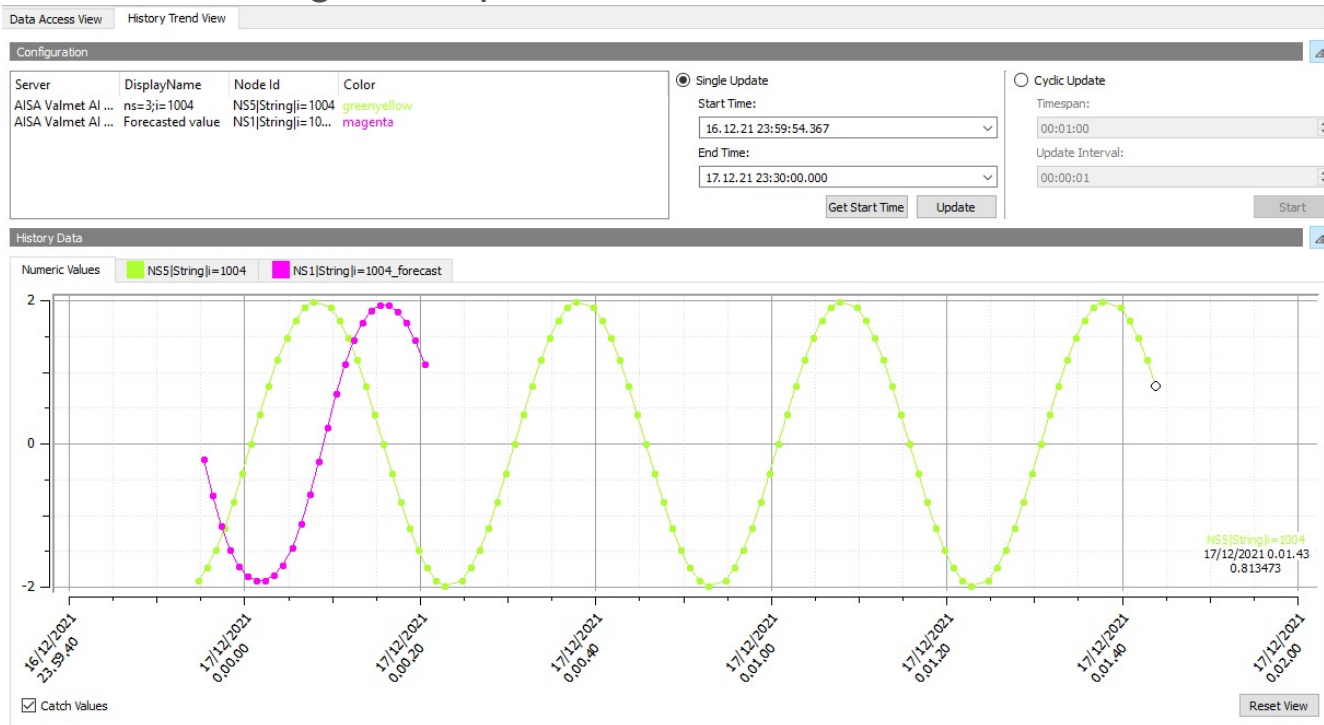
- Address space build for Digital Twin
- Constructed from the configuration
- For the Digital Twin clients variables (nodeId + suffix):
  - `_detector` == Anomaly detection (connected to A&C alarm object)
  - `_forecast` == Predicted value
  - `_last_trained` == Last training time
  - `_load_model` == Request from server to client to load model file (pkl)
  - `_load_params` == Request from server to client to load parameter file (jsonc)
  - `_status` == Client can report status
  - `_prediction_time` == Execution time to make new prediction
  - `_training_time` == Execution time to train model
  - `_model` == Model file
  - `_params` == Parameter file
  - `_dataset` == Dataset (csv)



#	Server	Node Id	Display Name	Value	Datatype	source Timestamp	Server Timestamp	Statuscode
1	Valmet Digital T...	NS1 String 10...	Anomaly detect...	false	Boolean	9.13.02	9.28.04	Good
2	Valmet Digital T...	NS1 String 10...	Forecasted value	20	Float	9.55.13	9.54.44	Good
3	Valmet Digital T...	NS1 String 10...	Last trained (Da...	2022-01-14T09:28:40.456Z	DateTime	9.28.40	9.28.40	Good
4	Valmet Digital T...	NS1 String 10...	Load model	false	Boolean	9.13.02	9.28.04	Good
5	Valmet Digital T...	NS1 String 10...	Load params	false	Boolean	9.13.02	9.28.04	Good
6	Valmet Digital T...	NS1 String 10...	Prediction status	New prediction, time: 2022-01-14 09:58:50.695133	String	9.58.50	9.58.50	Good
7	Valmet Digital T...	NS1 String 10...	Prediction time ...	33	UInt32	9.58.50	9.58.50	Good
8	Valmet Digital T...	NS1 String 10...	Retrain model	false	Boolean	9.28.40	9.28.40	Good
9	Valmet Digital T...	NS1 String 10...	Training time (s)	15	UInt32	9.28.40	9.28.40	Good

# Forecast – First test client

## Time offset – signal shape “correct”



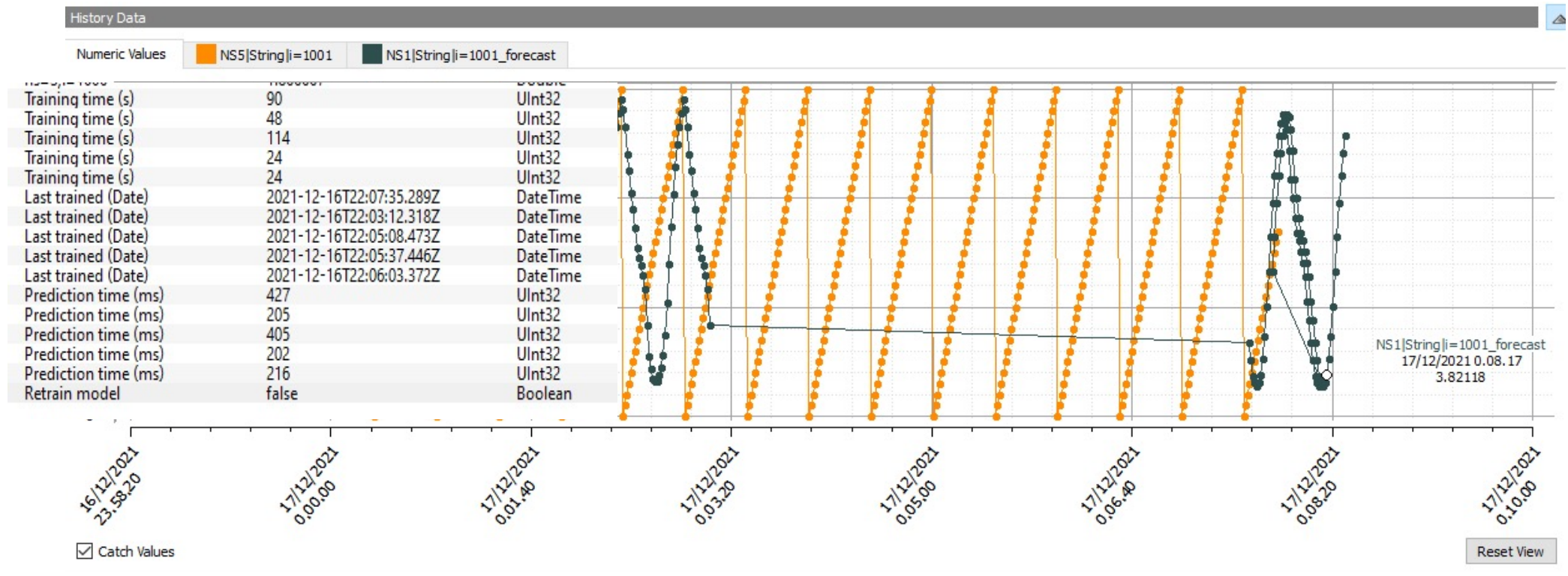
### Example of model parameters:

```
D:\GIT\OPCUA\AISA-DigitalTwin\models>type "ns=5;s=i=1004.jsonc"
{
  // Data points that will be read from the OPC UA HA for the model training
  "trainingLength": 500,
  "trainingsize": 90,
  "n_epochs": 10,
  "learningrate": 0.01,
  "n_hiddenlayers": 4,
  "batch_size": 32,
  // Seconds how long model will be used, then re-trained
  "retrainInterval": 300
}
D:\GIT\OPCUA\AISA-DigitalTwin\models>
```



# Retrain model – interval OR by retrain request

If model does not match to signal shape – add training data or other parameters



# Performance – just some thousands of data points

Digital Twin client will write back to server statistics

Training time (s)	144	UInt32
Training time (s)	48	UInt32
Training time (s)	113	UInt32
Training time (s)	24	UInt32
Training time (s)	24	UInt32
Last trained (Date)	2021-12-16T21:54:33.171Z	DateTime
Last trained (Date)	2021-12-16T21:55:27.215Z	DateTime
Last trained (Date)	2021-12-16T21:57:23.083Z	DateTime
Last trained (Date)	2021-12-16T21:57:51.953Z	DateTime
Last trained (Date)	2021-12-16T21:58:18.173Z	DateTime
Prediction time (ms)	509	UInt32
Prediction time (ms)	202	UInt32
Prediction time (ms)	394	UInt32
Prediction time (ms)	196	UInt32
Prediction time (ms)	194	UInt32

# Tools for tuning & validating Neural Network Overview

**Valmet Digital Twin UI for validating & tuning Forecasting**

Code snippets

- Dataframe information (loads current model & parameters)
- Train & fit (new prediction)
- Cross validation
- Benchmark

Time range:

Time range:

- 2h
- 2h
- 2h
- 15h
- 15min
- 10min
- 5min

Client is connected to endpoint:

src: http://192.168.1.107:8080:80:8080/

Training dataset (csv) filename:

192.168.1.107\_192.168.1.107.csv

Training dataset from file:

**Variables**

Select variable to read time series data from Digital Twin OPC UA Server (modelId):

Server\* model for the variable:

192.168.1.107 10.00

Logged Regressor variable:

0.00

Future Regressor variable:

0.00

Get

Save model parameters

Load model parameters

Load dataset

**Model parameters 1/2:**

growth

linear

changePoints

0.00

n\_changePoints

20

changePoints\_range 0-1, percent

0.50

trend\_reg

0

trend\_reg\_threshold

yearly\_seasonality

true

weekly\_seasonality

true

daily\_seasonality

true

seasonality\_mode

additive

seasonality\_reg

0

n\_forecasts

1

n\_lags

12

**Model parameters 2/2:**

num\_hidden\_layers

4

num\_hidden

None

num\_neurons

30

learning\_rate

0.00

epochs

20

batch\_size

64

loss\_func

Huber

optimizer

AdamW

train\_speed

0.00

normalise

off

impute\_missing

collect\_metrics

periods

20

min\_dataPoints

100



# Tools for tuning & validating Neural Network

## Actions & Variables

Code snippets

- Dataframe information (loads current model & parameters)
- Train & fit (new prediction)
- Cross validation
- Benchmark

Time range:

Time range

- 3h
- 2h
- 1h
- 15h
- 15min
- 10min
- 5min

Client is connected to endpoint:

opc.tcp://H7Q8Q15.VSTAGE.CO:26542/

Training dataset (csv), filename:

Yosemite\_temps.csv

Training dataset from file:

### Variables

Select variable to read time series data from Digital Twin OPC UA Server (nodeId):

Server's model for the variable:

ns=3;pi=1001 15/50

Logged Regressor variable:

0/50

Future Regressor variable:

0/50

Go!

Save model parameters

Load model parameters

Load dataset



# Tools for tuning & validating Neural Network

## Model parameters

### Model parameters 1/2:

growth  
linear

changepts  
0,10

n\_changepts  
30

changepts\_range 0-1, percent  
0,95

trend\_reg  
0

trend\_reg\_threshold

yearly\_seasonality  
true

weekly\_seasonality  
true

daily\_seasonality  
true

seasonality\_mode  
additive

seasonality\_reg  
0

n\_forecasts  
1

n\_lags  
12

### Model parameters 2/2:

num\_hidden\_layers  
4

d\_hidden  
None

sr\_sparsity

learning\_rate  
30

epochs  
10

batch\_size  
64

loss\_func  
Huber

optimizer  
AdamW

train\_speed  
0,10

normalize  
off

impute\_missing

collect\_metrics

periods  
30

min\_deltaeps  
100



# Tools for tuning & validating Neural Network

Show current measurement trend: Yosemite temperatures as example (>40k datapoints)

Datframes description real & forecasted:

	y
count	14,989.0000
mean	18.4023
std	13.8928
min	-8.4000
25%	8.0000
50%	13.8000
75%	27.9000
max	53.2000

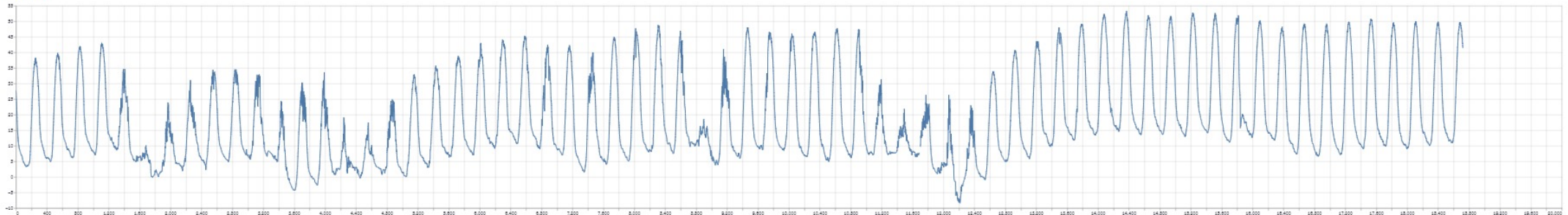
	value
count	3,740.0000
mean	18.8371
std	13.8728
min	-8.0000
25%	8.1000
50%	14.2000
75%	27.8250
max	52.7000

empty

## Trend values

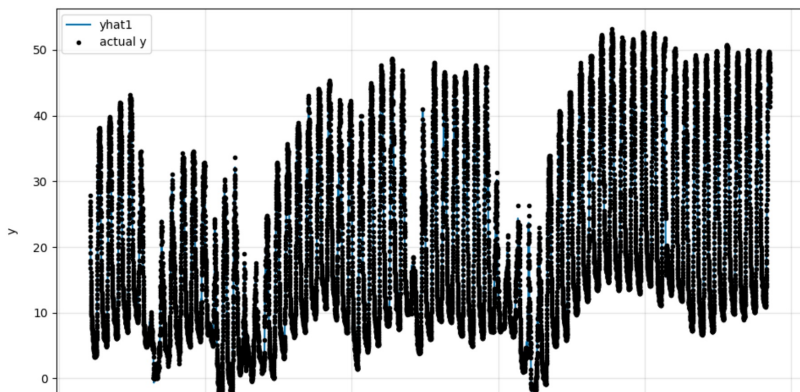
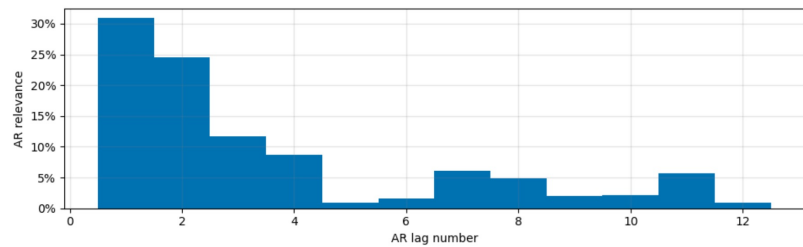
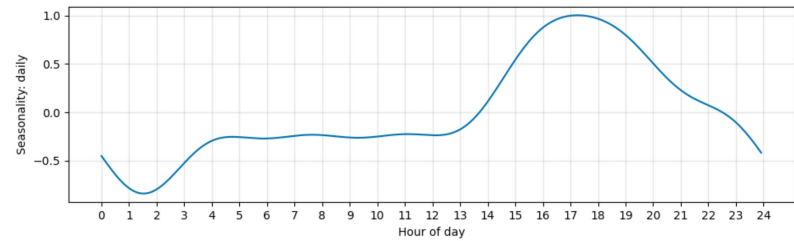
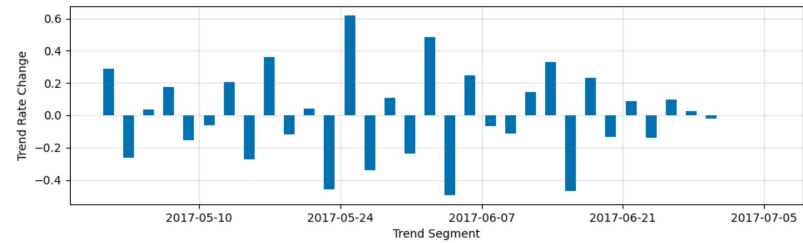
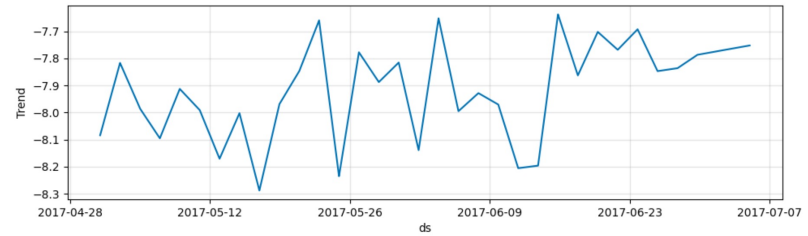
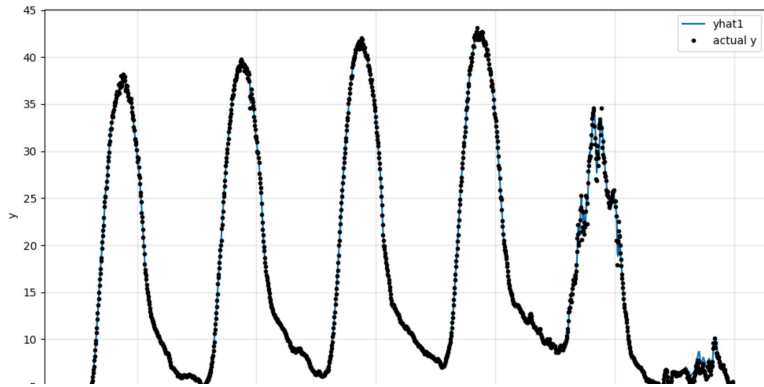
Update trend values

Measurements



# Tools for tuning & validating Neural Network

## Train & fit



# Tools for tuning & validating Neural Network

## Cross validate

### Validation results:

	SmoothL1Loss	MAE	RMSE
mean	0.0000	0.2725	0.4018
min	0.0000	0.2725	0.4018
max	0.0000	0.2725	0.4018

### Dataframes description real & forecasted:

	y
count	14,969.0000
mean	18.4025
std	13.8926
min	-8.4000
25%	8.0000
50%	13.8000
75%	27.9000
max	53.2000

	value
count	3,740.0000
mean	18.6371
std	13.8728
min	-8.0000
25%	8.1000
50%	14.2000
75%	27.9250
max	52.7000

empty
-------

# Tools for tuning & validating Neural Network Benchmark

## Benchmark results, simple experiment:

```
{
  "data" : "ns=5;s=i=1001"
  "model" : "NeuralProphet"
  "params" : "{ 'seasonality_mode': 'multiplicative' }"
  "MASE" : "0.98102695"
  "RMSE" : "19.176846"
}
```

## Benchmark results, cross-validate:

```
{
  "data" : "ns=5;s=i=1001"
  "model" : "NeuralProphet"
  "params" : "{ 'seasonality_mode': 'multiplicative' }"
  "MASE" : [
    0 : "0.9552446"
    1 : "0.9847775"
    2 : "0.99620473"
  ]
  "RMSE" : [
    0 : "18.886786"
    1 : "19.121881"
    2 : "19.44433"
  ]
}
```

# Tools for tuning & validating Anomaly detection

## Overview

### Valmet Digital Twin client for Anomaly detector testing & tuning

Detector

Server extractor: ThresholdAD

Parameters

Select variable to read time series data from Digital Twin OPC UA Server (nodeid): ns5s=i=1001

Server's model for the variable: ns5s=i=1001

Time range: 15.00

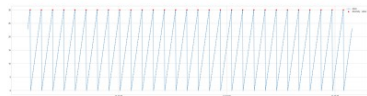
Time range:
 

- 2h
- 1h
- 15min
- 10min
- 5min

Client is connected to endpoint: oee-ops://MQ004\_VST408\_001:8894z/

Low	1.00	→	←	u_min	14	→	←
High	28.00	→	←	u_max	28	→	←
s	5.00	→	←	u_stddev	2	→	←
alpha	1.00	→	←	confInterval	0.05	→	←
size	negative	→	←	target	value	→	←
window	5.00	→	←	k	1	→	←

Server detector: ThresholdAD for tag: ns=5;s=i=1001 parameters: low=-0.1 high=29.5

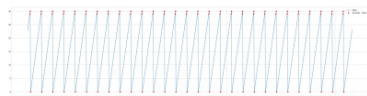


Detailed anomalies by server

Found anomalies:108/1000

time	value
2022-05-12T10:00:05	true
2022-05-12T10:01:28	true
2022-05-12T10:01:37	true
2022-05-12T10:02:18	true
2022-05-12T10:02:58	true
2022-05-12T10:03:00	true
2022-05-12T10:04:01	true
2022-05-12T10:04:32	true
2022-05-12T10:05:03	true
2022-05-12T10:05:34	true

Testing detector: ThresholdAD for tag: ns=5;s=i=1001 parameters: low=1.0 high=28.0



Detailed anomalies with test parameters

Found anomalies:107/1000

time	value
2022-05-12T10:00:34	true
2022-05-12T10:00:55	true
2022-05-12T10:00:58	true
2022-05-12T10:01:03	true
2022-05-12T10:01:28	true
2022-05-12T10:01:37	true
2022-05-12T10:01:58	true
2022-05-12T10:01:57	true
2022-05-12T10:01:58	true
2022-05-12T10:01:57	true
2022-05-12T10:01:58	true
2022-05-12T10:02:01	true
2022-05-12T10:02:01	true

# Tools for tuning & validating Anomaly detection

Testing anomaly detection, select from available detector algorithms (14 implemented)

## Valmet Digital Twin client for Anomaly detector testing & tuning

**Detector**

Default detector: ThresholdAD

**Parameters**

Select variable to read time series data from Digital Twin OPC UA Server (model): ns5.pls1000

Server's model for the variable: ns5.pls1000

Time range:

2h

24h

15min

10min

30min

Low: 1.00

High: 29.00

alpha: 1.00

side: negative

window: 5.00

Time series plot showing a sawtooth pattern.

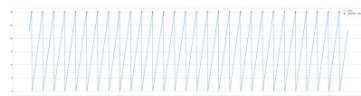
Detector selection dropdown menu:

- PosAD
- QuantileAD
- InterQuartileRangeAD
- GeneralizedESDTestAD
- PersistenceAD
- LevelShiftAD
- SeasonalAD
- AutoRegressionAD
- MinClusterDetector

Parameter configuration table:

r_class	14
det_size	24
r_clusters	3
normalization	0.05
target	value
k	1

Server detector: ThresholdAD for tag: ns=5;s=i=1001 parameters: low=-0.1 high=29.5

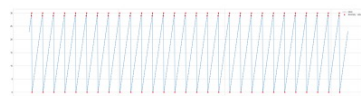


Detailed anomalies by server

Found anomalies: 18 / 100

time	value
2022-01-12T10:00:00	true
2022-01-12T10:01:00	true
2022-01-12T10:02:00	true
2022-01-12T10:03:00	true
2022-01-12T10:04:00	true
2022-01-12T10:05:00	true
2022-01-12T10:06:00	true
2022-01-12T10:07:00	true
2022-01-12T10:08:00	true
2022-01-12T10:09:00	true
2022-01-12T10:10:00	true
2022-01-12T10:11:00	true
2022-01-12T10:12:00	true
2022-01-12T10:13:00	true
2022-01-12T10:14:00	true
2022-01-12T10:15:00	true
2022-01-12T10:16:00	true
2022-01-12T10:17:00	true
2022-01-12T10:18:00	true
2022-01-12T10:19:00	true
2022-01-12T10:20:00	true

Testing detector: ThresholdAD for tag: ns=5;s=i=1001 parameters: low=1.0 high=28.0



Detailed anomalies with test parameters

Found anomalies: 27 / 100

time	value
2022-01-12T10:00:04	true
2022-01-12T10:00:05	true
2022-01-12T10:00:06	true
2022-01-12T10:00:07	true
2022-01-12T10:00:08	true
2022-01-12T10:00:09	true
2022-01-12T10:00:10	true
2022-01-12T10:00:11	true
2022-01-12T10:00:12	true
2022-01-12T10:00:13	true
2022-01-12T10:00:14	true
2022-01-12T10:00:15	true
2022-01-12T10:00:16	true
2022-01-12T10:00:17	true
2022-01-12T10:00:18	true
2022-01-12T10:00:19	true
2022-01-12T10:00:20	true
2022-01-12T10:00:21	true
2022-01-12T10:00:22	true
2022-01-12T10:00:23	true
2022-01-12T10:00:24	true
2022-01-12T10:00:25	true
2022-01-12T10:00:26	true
2022-01-12T10:00:27	true
2022-01-12T10:00:28	true
2022-01-12T10:00:29	true
2022-01-12T10:00:30	true

# Tools for tuning & validating Anomaly detection

## Testing Pca with 5min time range – no anomalies detected

### Valmet Digital Twin client for Anomaly detector testing & tuning

#### Detector

Select detector:

#### Parameters

Select variable to read time series data from Digital Twin OPC UA Server (nodeId):

Server's nodeId for the variable:

Select multivariable regressor, with univariable leave empty (nodeId):

Server's nodeId for the regressor variable:

Server's nodeId for the regressor variable:

Time range:

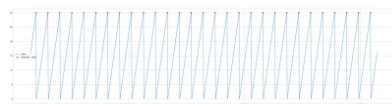
- Time range
- 2h
  - 1h
  - 15m
  - 5min

Client is connected to endpoint:

see: <http://localhost:8080/OPCUA/SYSTEM/00120004/>

low	0.00	-- *	n_base	14	-- *
high	25.00	-- *	abs_size	24	-- *
r	0.00	-- *	n_filters	1	-- *
alpha	1.00	-- *	contamination	0.05	-- *
side	negative	-	target	value	
window	5	-- *	k	1	-- *

Server detector: ThresholdAD for tag: ns=7;s=i=1001 parameters: low=-0.1 high=29.5



Detected anomalies by server

Found anomalies: 0/1000

time	value
2022-02-09T11:47:59	true
2022-02-09T11:48:00	true
2022-02-09T11:48:01	true
2022-02-09T11:48:02	true
2022-02-09T11:48:03	true
2022-02-09T11:48:04	true
2022-02-09T11:48:05	true
2022-02-09T11:48:06	true
2022-02-09T11:50:07	true
2022-02-09T11:50:08	true

Found anomalies: 0/1000

Local detector not executed Wrong anomaly detector parameters!

# Summary

- OPC UA used as standard communication to integrate AI for industrial use
- Python Neural Prophet for forecasting and ADTK Anomaly Detection Tool-Kit anomalies
- Web application build with Streamlit
  - Parameter tuning
  - Visualize model results
  - Test & validate, repeat until satisfied with the results
  - Write new model & parameters back to server with OPC UA
  - Integrated with FreeOpcUa: opcua-asyncio
- Many steps closer to people can use artificial intelligence at industry
  - Tools are first generation, proof of concept build, architecture promising
  - Needs training for people to understand parameters, models & algorithms
  - Multivariable models needed (current solution is simple univariable implementation)



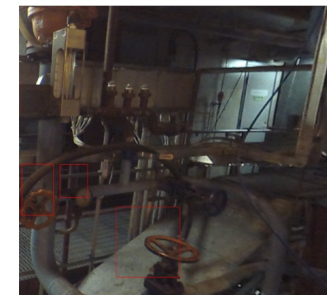
# Next actions with project members

- AWS EC2 instance with Digital Twin server (Valmet)
  - Simulated data (perhaps some historical real data)
  - Some photos (TBD synthetic or augmented)
- Workshop: How to implement Digital Twin client (Python snippets)
  - Digital Twin: [INSTA](#) & [TAU](#)
- [Nokia](#): Test setup for WebRTC & integration
  - Embedded WebRTC (HTML)
  - Field device detector (AutoML TensorFlow model: object detector)
  - [TDS](#): Person detector
- Architecture:
  - [Nokia](#) & [Valmet](#): real-time data & video stream/photo, secure data connection



Select a new file:  Test.png

Results: Found devices: 3



# “Loppukevennys”



