



100 % OIL-FREE INDUSTRIAL
TURBO COMPRESSORS

Machine learning in condition monitoring

Beckhoff; Automation update seminar

23th November 2023

Timo Oksanen



Company purpose

Tamturbo develops, manufactures, and sells industrial direct drive turbo compressors with active magnetic bearings



100 % oil-free



The most environmentally sustainable solution on the market



With the lowest total cost of ownership

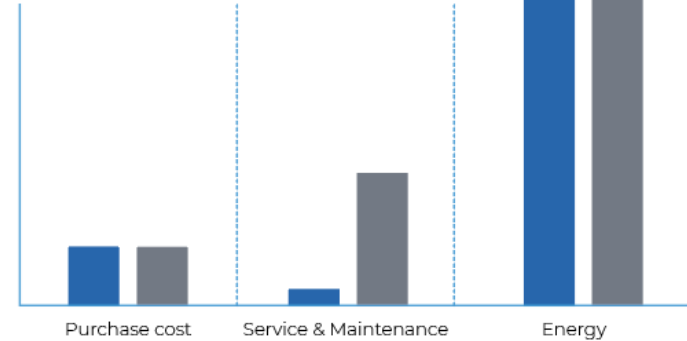


The lowest total cost of ownership

- The product line addresses the most common capacity need (power and volume) for different uses in different industries.
 - From 100 to 350 kW
 - From 10 to 60 m³ per minute



TOTAL COST OF OWNERSHIP

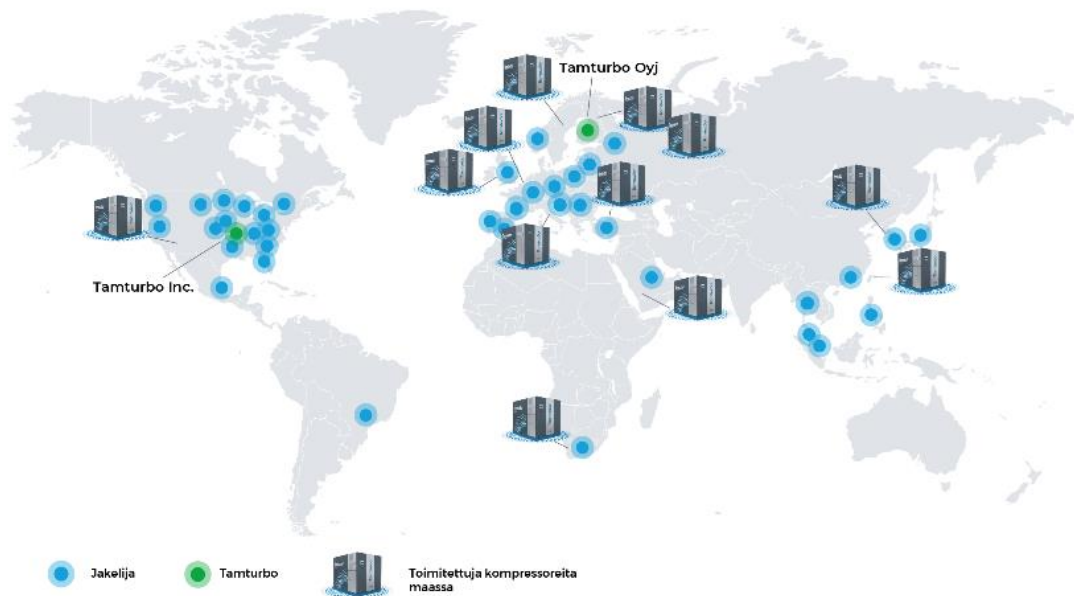


DISTRIBUTORS AND COUNTRIES WITH INSTALLATIONS

N Americas **50%**

Europe **44%**

Others **6%**



REVENUE MODELS

- Most of the compressors are sold with service contract
- Uptime is the most important parameter for customer to buy
- Preventive maintenance helping to do action proactively
- Real-time remote monitoring already used part of the service

Investment + maintenance

Traditional compressor business

Investment + extended warranty

Up to 10 years. All consumable parts, spare parts and labor included

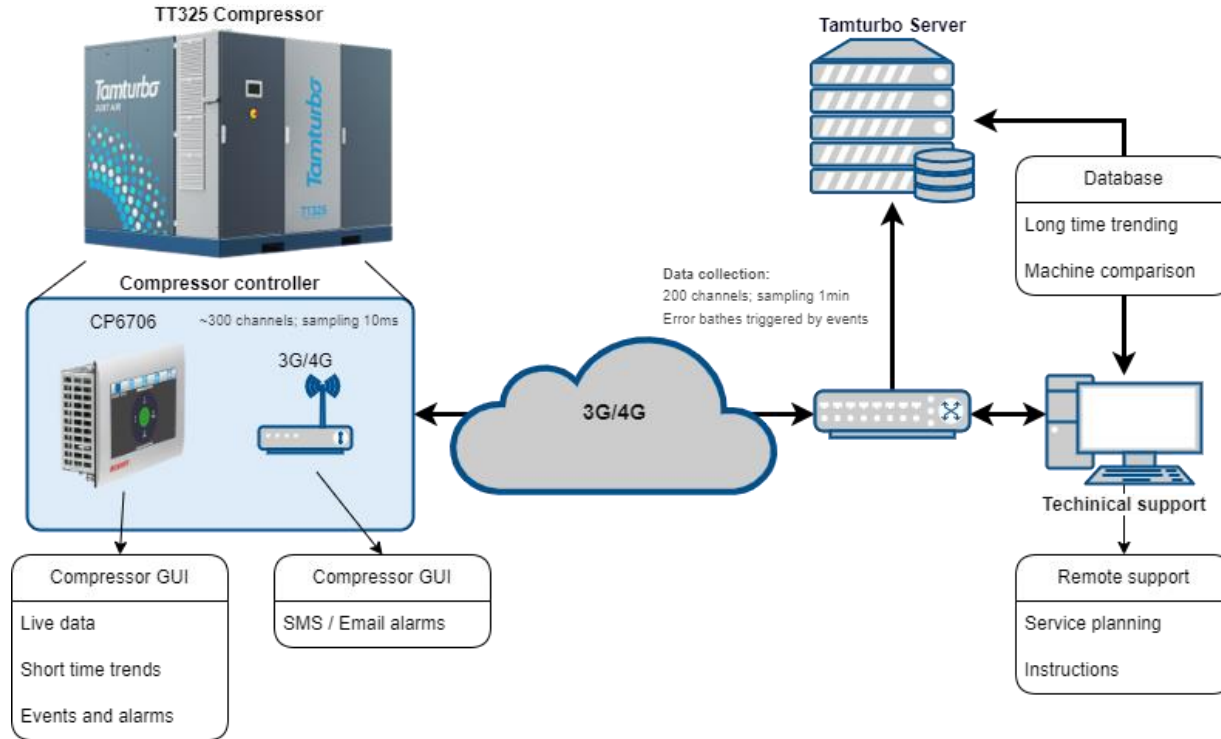
Touch-Free™ Air

Compressor + all related expenses included

Pricing: €/m³, €/kW, or €/month

→ production changes easy, adaptive to customer needs, OPEX

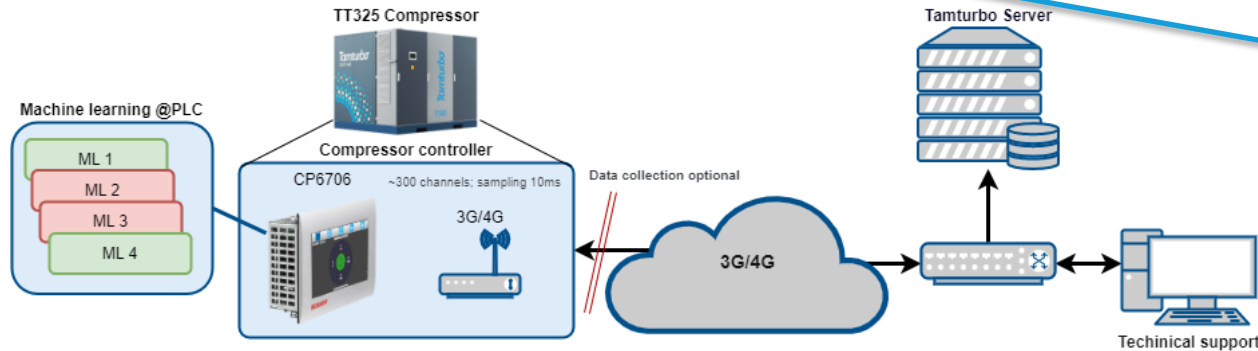
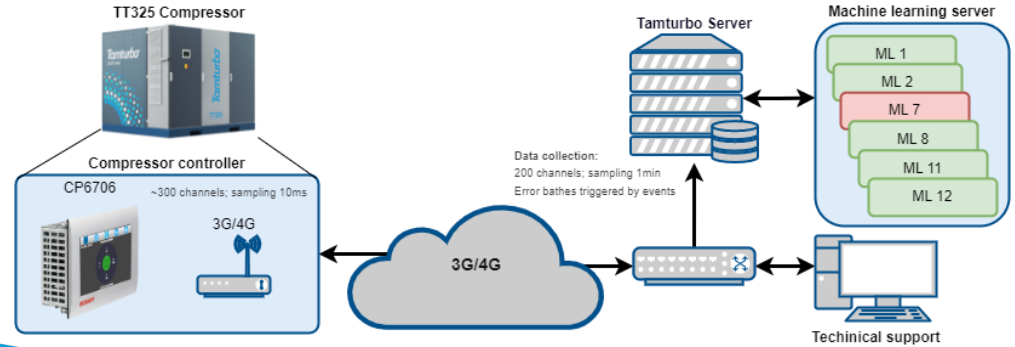
TAMTURBO OPERATION MODEL



MACHINE LEARNING IN ACTION

Remote data:

Local data:



MACHINE LEARNING STEPS TO OPERATION

1. Case to solve with Machine learning; soft sensor, etc.

2. Data for the modelling; field data, test data, and simulation data

3. System physical understanding; design team for process

4. Machine learning model creation

5. Platform to run machine learning model; PLC with Beckhoff library

Presentation: Henri Vainio, TAU

Presentation: Henrik Pammo, Beckhoff
Demo and results: Timo Oksanen, Tamturbo

Mechatronics Research Group

Prof. Kari T. Koskinen, TAU/ENS

Research Questions:

- How to improve and predict **availability, reliability and maintenance** of complex machine systems?
- How to design, analyse and control **autonomous vehicles, robots and multi-robot systems**?
- How to improve the performance of **aircraft systems** and human performance in aviation?
- How to analyse the operation and condition of a machine systems with **knowledge-based analytics**?

Staff: 26 persons including Industry professor, Senior Researchers, University Instructors, Doctoral Researchers and Researchers, Research Assistants

Teaching in Mechanical Engineering –study program: Aircraft Engineering, Mechatronic System Design, Reliability Engineering and Maintenance

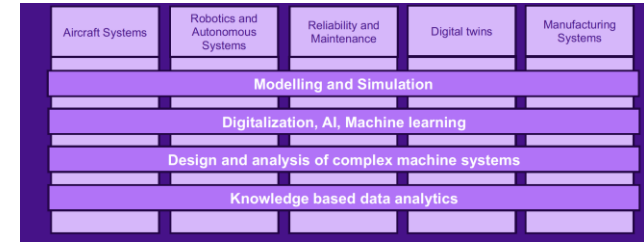
Awards: *Tampere Innovation Award 2020*



Contact: Prof. Kari T. Koskinen kari.koskinen@tuni.fi

Mission:

*More Sustainable
Future Machinery
by Smart Design
and Optimized
Life-cycle*



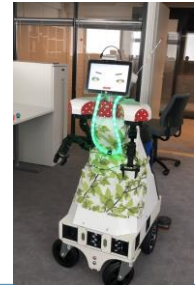
Main projects:

Unexmin (EU-H2020)
Unexup (EIT)
Robominers (EU-H2020)
Dig It (EU-H2020)
PVTO
SMARAGDI
SNOBI
MASTERMINE (HE)
AIRSHIP (HE)

Funding:

EU, EIT, Business Finland, Defence Forces, Industry

Annual project volume: ca. 1 MEURO



Implementation of Machine Learning Algorithms to TwinCAT

The Application of Simulation in Compressor System

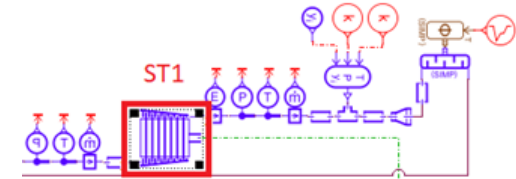
Turbo Compressor Simulation Model

- Dynamic model of physical phenomena in a turbo compressor
- 1D simulation software
- Multiple uses
 - Control design
 - Sales
 - Tech support
 - Condition monitoring, anomaly detection
- Slower than real time

Physical System



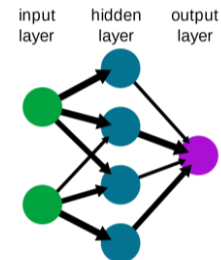
Simulation Model



Surrogate modeling

- Mimics the behavior of the simulation model as closely as possible
- Many times faster than the original model
- Surrogate modeling methods
 - Response surface models
 - Bayesian networks
 - **Neural networks**

Surrogate model: Neural Network

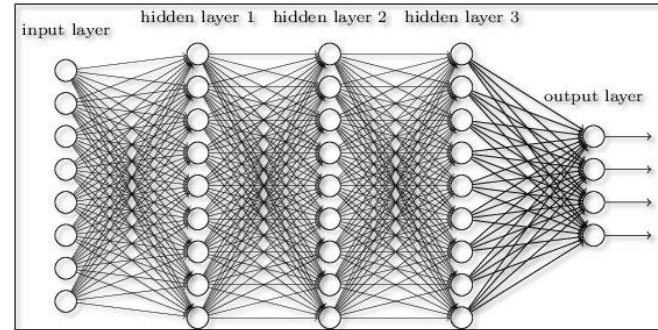
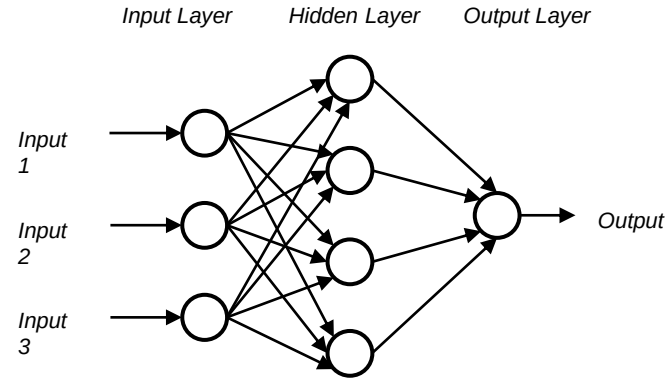


Basic Principle of Neural Networks

- Biologically inspired models
- Learn patterns in the data by updating the parameters of the individual neurons
- Faster than other surrogate methods, and easy to integrate with automation systems

TensorFlow

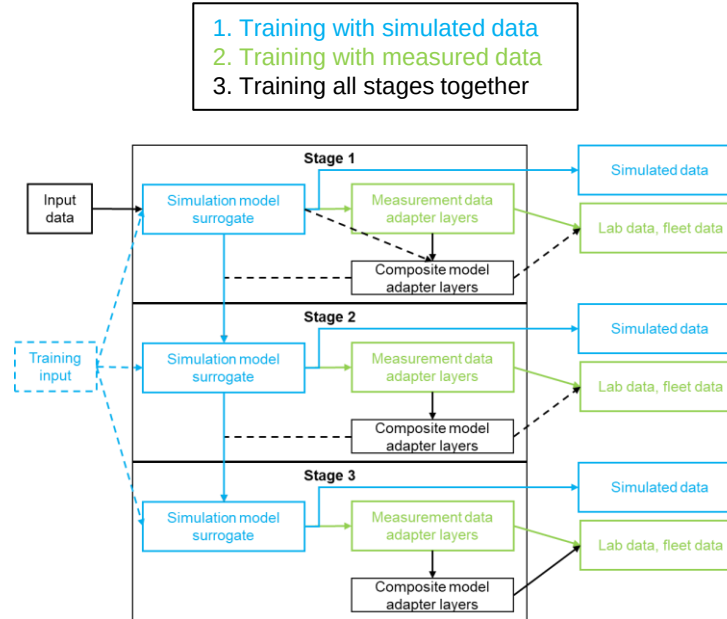
- Google's tensor library, Open source since 2015
- GPU support
- Good support for deploying trained models on various platforms
 - Mobile devices, single board computers etc.



Deep Neural Network = network with many layers

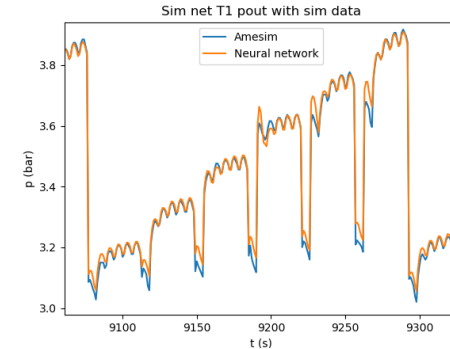
The Application of Surrogate Modeling in Compressor Automation System

- Physical phenomena based modular neural network
- Training of layers of a neural network with simulated data
- Freezing of trained weights
- Training of new layers with data measured from the real machine
- Simulated, laboratory measurement and field measurement data can be used, or all combined
- Sub-models of known physical phenomena make training and re-training easy
- Individual phenomena models can be selected

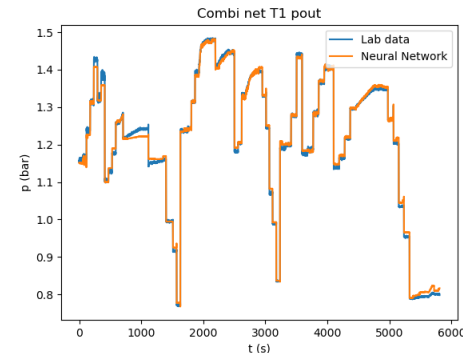


Surrogate model structure:
Turbo compressor with
three turbo stages

Surrogate model simulation results



Neural network test with simulated data



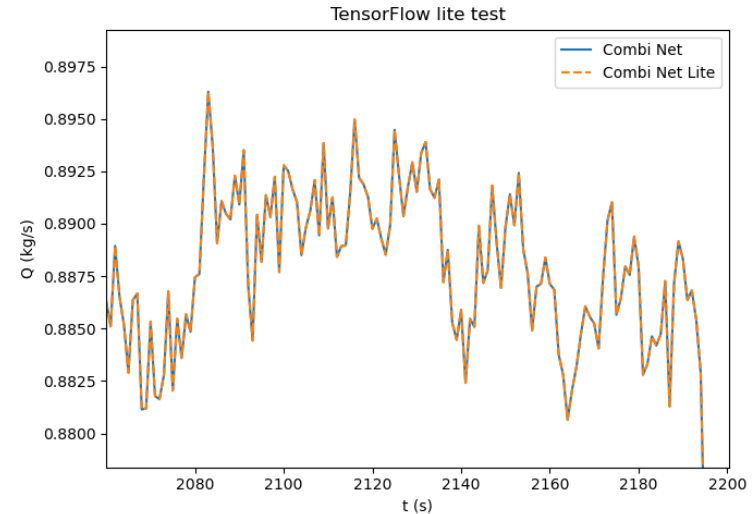
Neural network test with measured data

TensorFlow Lite Library for Implementation

- TensorFlow Lite is a version of TF designed for microcontrollers and mobile devices
- Lighter and faster than TF with some limitations
- Models trained with TF can be converted to TF lite

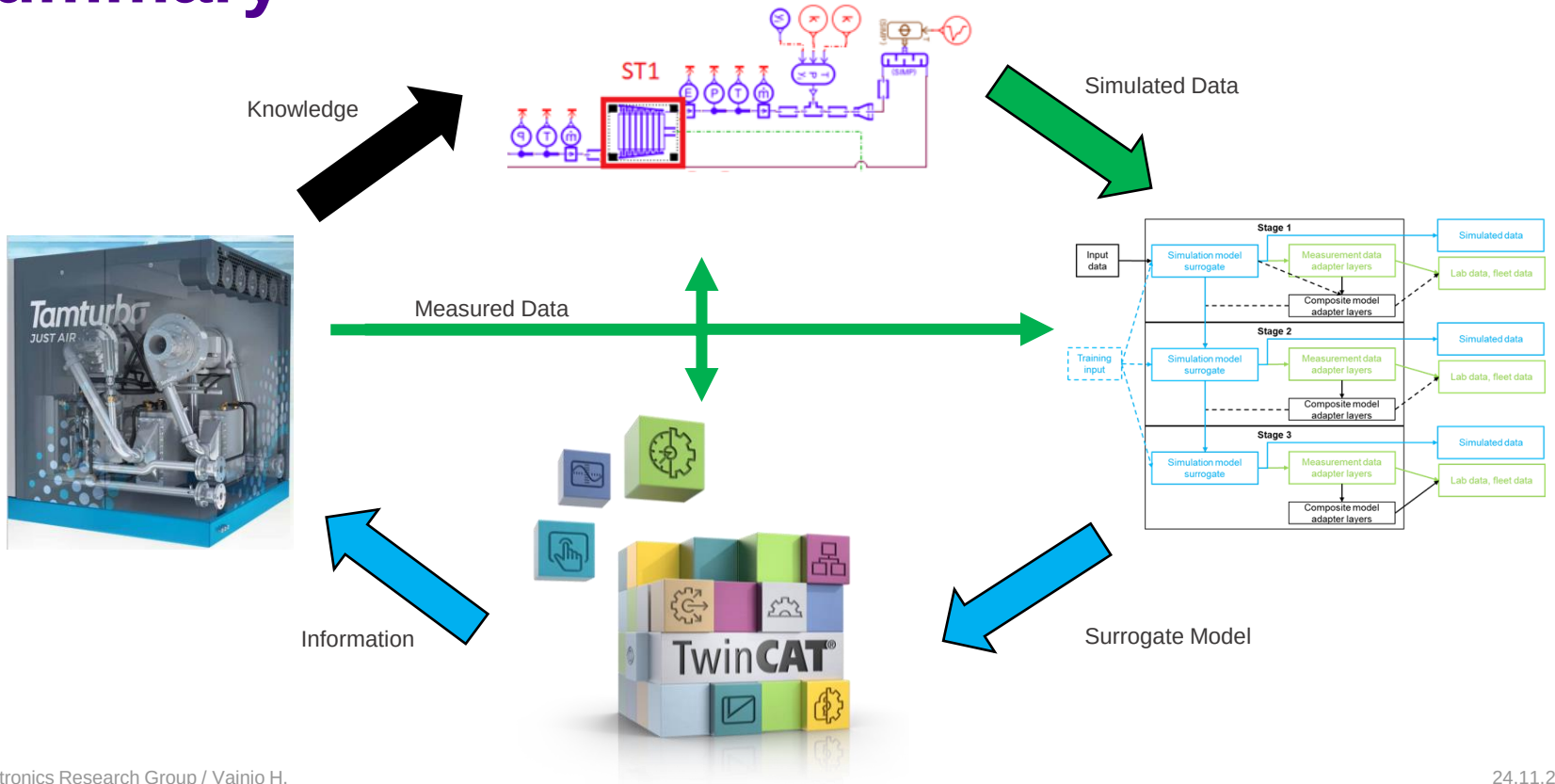
Case example

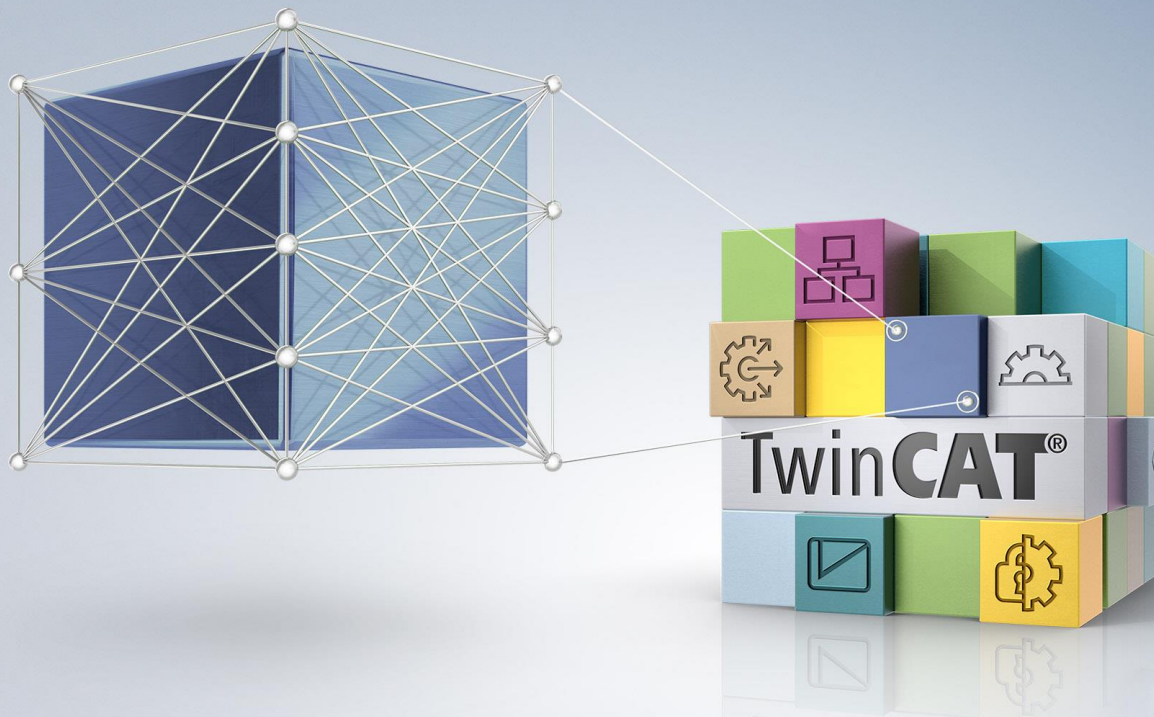
- Simulation times for 5809 steps (i.e. seconds) on a laptop PC:
 - Surrogate network: 4.3125 s (1347 x real time)
 - TensorFlow lite version: 0.015625 s (371776 x real time)
- Note that data pre- and postprocessing (e.g. scaling) may take additional time!

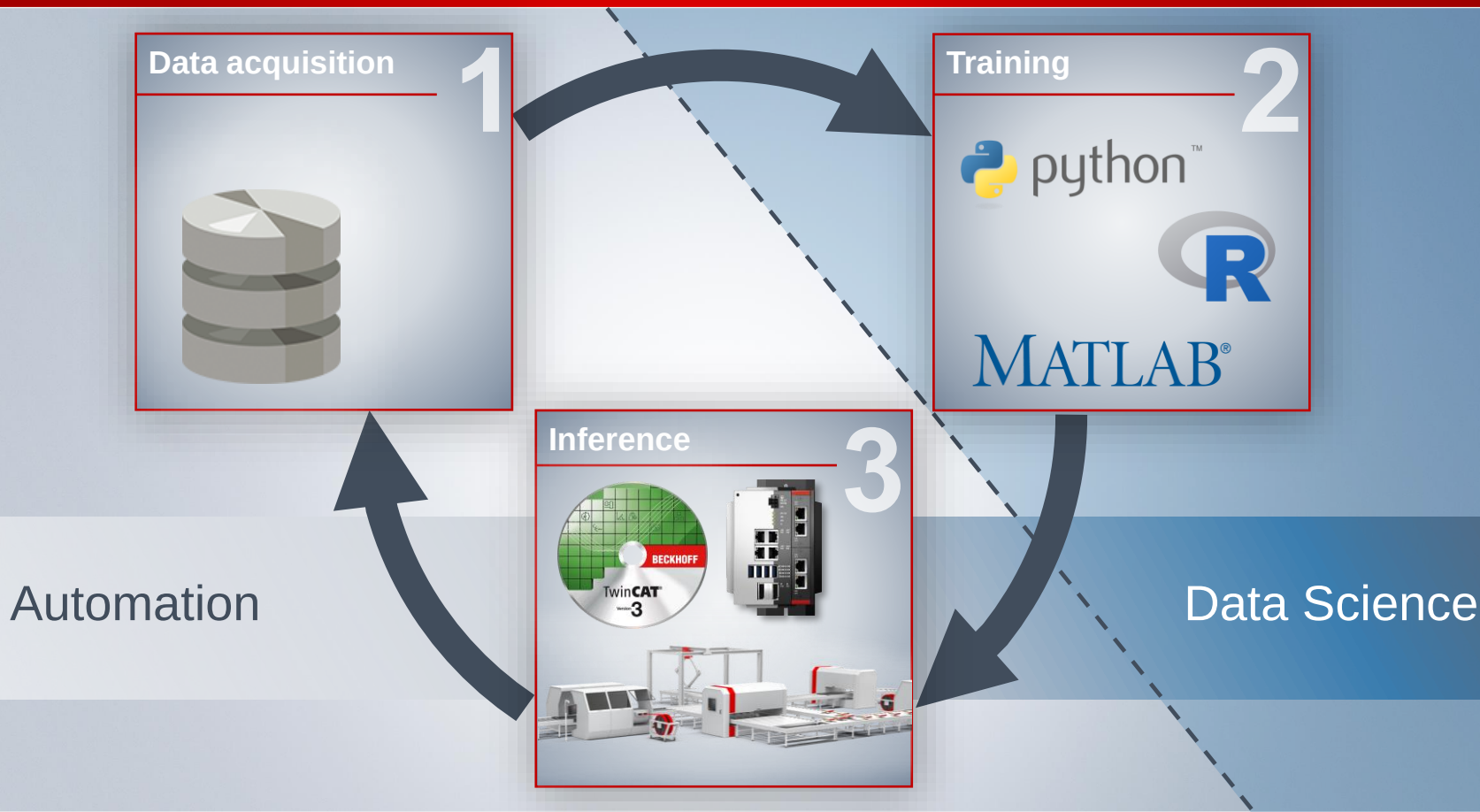


Identical output behavior from both TF and TF lite models

Summary



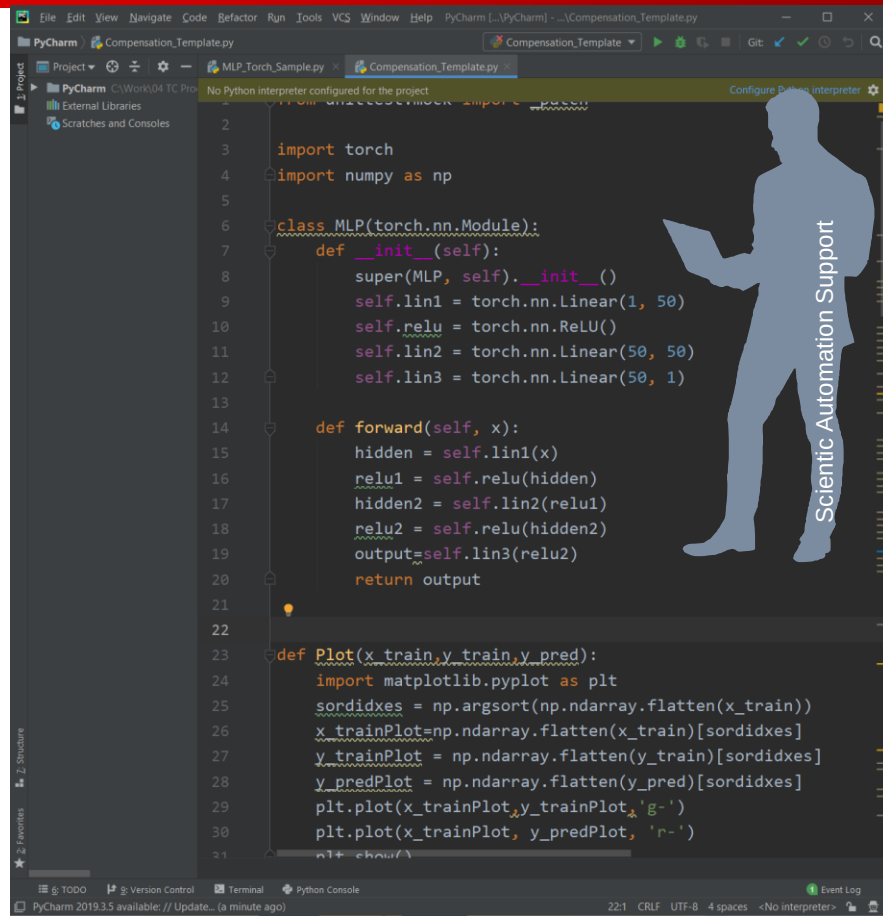




Data modelling, training of AI models

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- more than 80% of all data scientists work with Python or R code
- most of the libraries used for data labeling, cleaning and modelling are free of charge
- heterogeneous AI frameworks are often used even in data science teams of one department
- according to a broad variety of computational costs for AI applications, many target hardware devices exist for deployment
- **an interoperability standard is essential:**



- **Open Neural Network Exchange**
- Open format to represent Neural Networks and Machine Learning models
 - Interoperability between AI Frameworks
 - Easy access to inference hardware
- Fast growing consortium on supported frameworks, runtime converters and compilers

“ONNX is an open format built to represent machine learning models. ONNX defines a common set of operators - the building blocks of machine learning and deep learning models - and a common file format to enable AI developers to use models with a variety of frameworks, tools, runtimes, and compilers.”

<http://onnx.ai/>



ONNX

Frameworks & Converters

Use the frameworks you already know and love.

Caffe2

Chainer

dmk
XGBoost

K

LibSVM

MATLAB

Cognitive
Toolkit

mxnet

Neural
Network
Libraries

PaddlePaddle

PyTorch

SAS

SIEMENS

TensorFlow

TensorFlow

Inference

Deploy your ONNX model using runtimes designed to accelerate inferencing.

BITMAIN

CEVA

habana

intel AI

NVIDIA

ONNX
RUNTIME

MACE

Qualcomm

Rockchip

skymizer

SYNOPSYS

Tencent

tvm

TwinCAT 3

vespa

Windows

Visualize

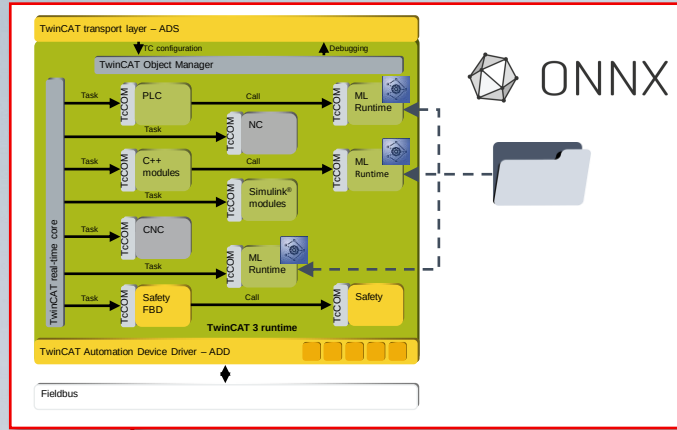
Better understand your model by visualizing its computational graph.

NETRON

Visual DL

TF3800, TF3810 | TwinCAT Machine Learning- and Neural Network Inference Engine for real-time inference in TwinCAT

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www.beckhoff.ai
www.onnx.ai



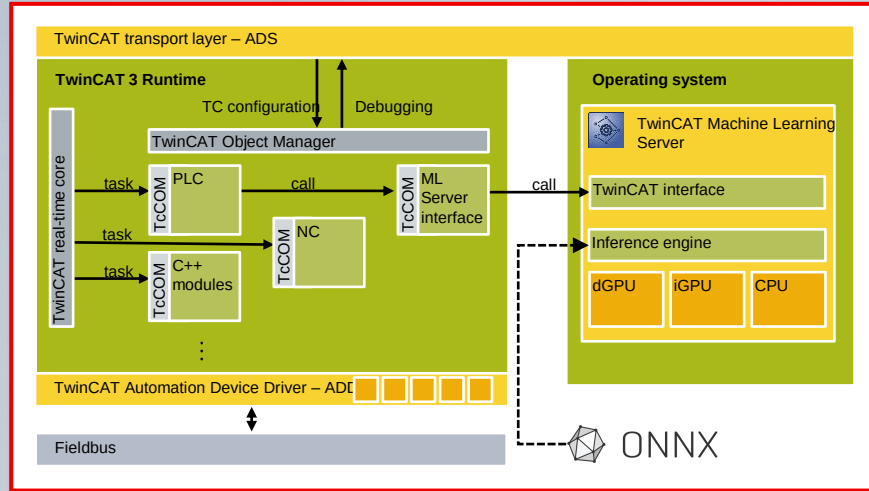
TF3820 | TwinCAT Machine Learning Server

Hardware accelerated inference using TwinCAT 3

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NE

Q2/2024

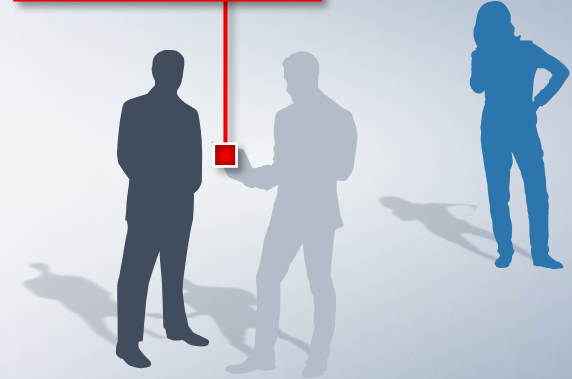


TF3820



C6043-0090

www.beckhoff.ai
www.onnx.ai



Real-time Inference | PLC API

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The screenshot displays the TwinCAT Project11 - TcKaeShell IDE. The left sidebar shows the Solution Explorer with the project structure. The main editor window shows the PLC program code. The code is as follows:

```
PROGRAM MAIN
VAR
    fbprediction : FB_MllPrediction; // ML Interface
    sModelFile   : T_MaxString := 'C:/models/TrigonometryMLP.bml'; // converted onnx file
    nInputDim, nOutputDim : UDINT;
    dtype : EtcMllDataType := EtcMllDataType.E_MLLDT_FP32_REAL;
    nnInput : ARRAY[1..10] OF REAL;
    nnOutput : ARRAY[1..10] OF REAL;
END VAR

CASE nState OF
    0: // idle state
        IF bLoadConfig THEN
            bLoadConfig:=FALSE;
            nState:=10;
        END_IF
    10: // config state
        fbprediction.stPredictionParameter.MIModelFilepath:=sModelFile; // provide model path and name
        fbprediction.stPredictionParameter.MaxConcurrency := 1; // no concurrent calls to this fb
        IF fbprediction.Configure() THEN // load model
            // check input/output dimensions and datatype
            bFailed:=fbprediction.GetInputDim(nInputDim);
            FAILED(bFailed);
            bFailed:=fbprediction.GetOutputDim(nOutputDim);
            FAILED(bFailed);
            bFailed:=fbprediction.CheckPreferredIODataTypes(EtcMllDataType.E_MLLDT_FP32_REAL, EtcMllDataType.E_MLLDT_FP32_REAL, bPreferred);
            FAILED(bFailed);
            bFailed:=fbprediction.CheckSupportedIODataTypes(EtcMllDataType.E_MLLDT_FP32_REAL, EtcMllDataType.E_MLLDT_FP32_REAL, bSupported);
            FAILED(bFailed);
        END_IF
    20: // Predict state
        fbprediction.Predict(
            pDataInp:=ADR(nnInput),
            nDataInpDim:= nInputDim,
            fmtDataInpType:= dtype,
            pDataOut:=ADR(nnOutput),
            nDataOutDim:= nOutputDim,
            fmtDataOutType:= dtype,
            nEngineId:= nCurrentEngineID,
            nConcurrencyId:= 0); // concurrency not in use in this sample. see concurrency sample in infosys for non-blocking parallelization

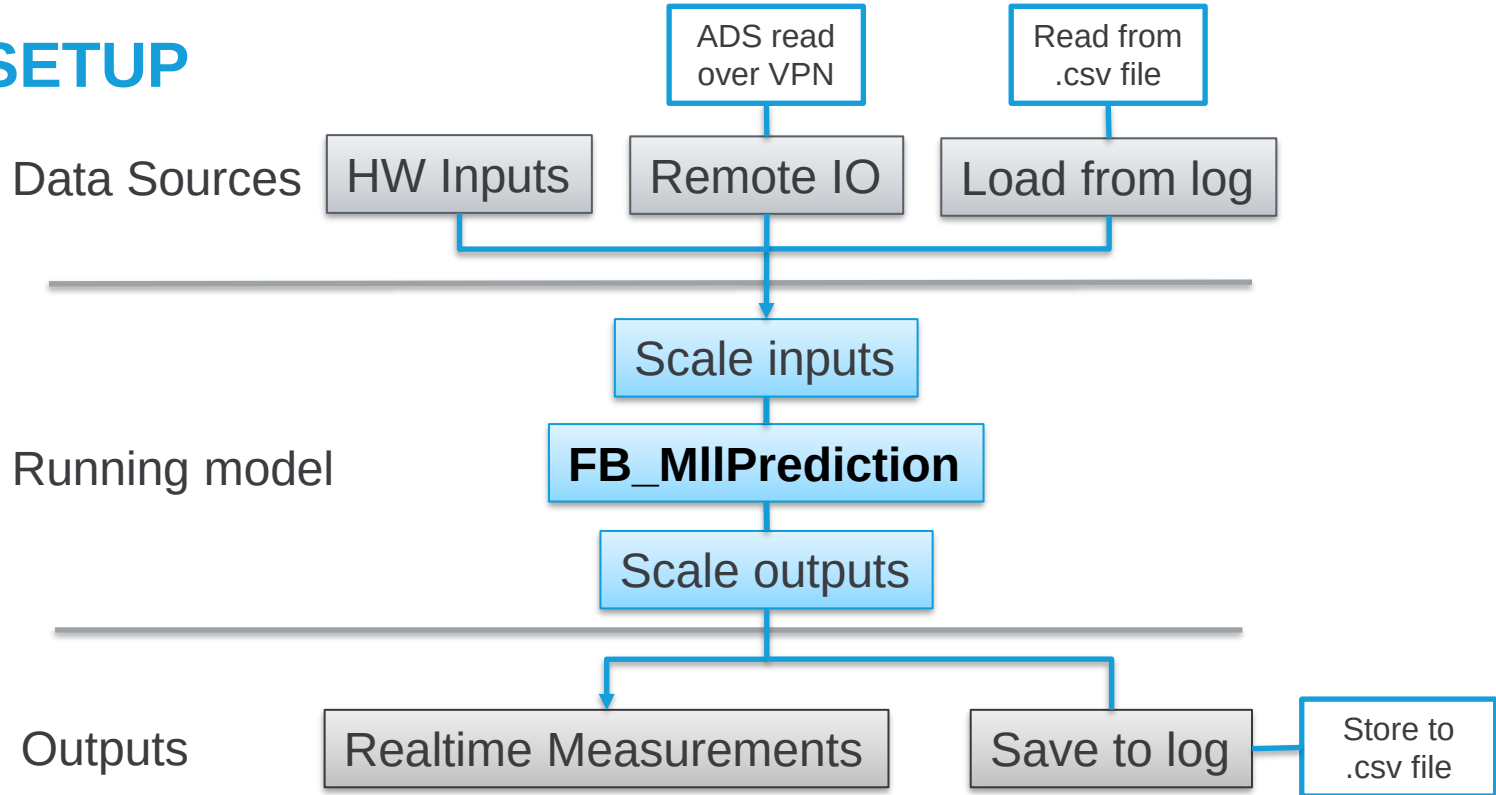
        IF fbprediction.bError THEN // error handling
            nState := 999;
            hrErrorcode := fbprediction.hrErrorcode;
        ELSEIF bLoadConfig THEN // load (updated) model
            bLoadConfig := FALSE;
            nState := 10;
        END_IF;

    999: // Error state
        // add error handling here
END CASE
```

Three red arrows point to specific parts of the code:

- load model**: Points to the `fbprediction.Configure()` call in the `10: // config state` block.
- execute model**: Points to the `fbprediction.Predict()` call in the `20: // Predict state` block.
- update model**: Points to the `ELSEIF bLoadConfig THEN` block in the `20: // Predict state` block.

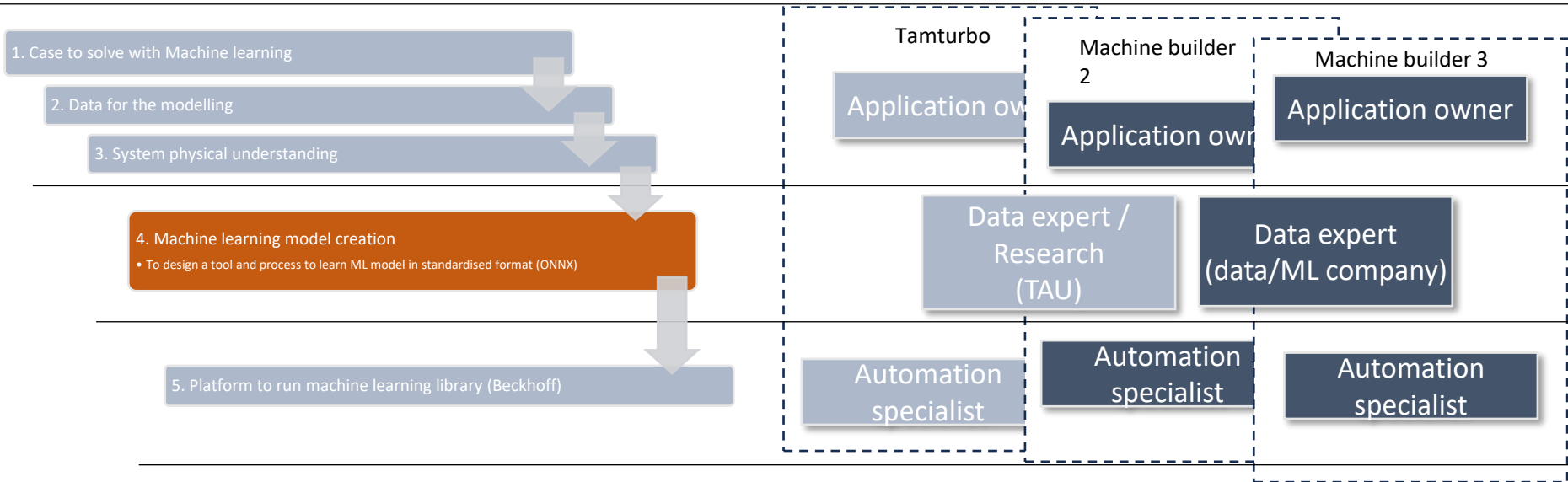
TEST SETUP



FINDINGS

SNOBI_Compressor_machine_learnign_test																																				
PLC_Timestamp	Comp_Turbo1		Comp_Turbo2		Comp_Networ		Comp_M1_RP		Comp_Turbo1		Comp_M2_RP		Comp_Turbo3		Comp_Turbo1		Comp_Turbo1		Comp_Turbo2		Comp_Turbo2		Comp_Turbo3		Sim_Turbo1		Sim_Turbo2		Sim_Turbo2		Sim_Turbo3		Sim_Turbo3			
	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb	IN_T	fb		
2022-11-23-13:54:26.567	22.3	-0.01438334	6.62126	31436	31.90001	33322	34.90002	0.7920106	1.353408	1.30196	4.578506	4.636297	6.644768	0.778457	1.176533	122.80408	0.989043	4.333556	153.1713	4.155183	7.978884	77.01895	1.092245	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:27.567	22.3	-0.014419949	6.62126	31438	31.90001	33322	34.90002	0.7961513	1.353087	1.302139	4.576783	4.635307	6.643585	0.778432	1.176548	122.80236	0.9894938	4.333645	153.1663	4.155325	7.979793	77.02187	1.092119	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:28.568	22.3	-0.014378578	6.62198	31436	31.90001	33322	34.90002	0.7967928	1.353186	1.301874	4.578049	4.635238	6.642288	0.7784527	1.176518	122.80544	0.9894987	4.333529	153.1718	4.155132	7.978931	77.01891	1.092116	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:30.568	22.3	-0.014455276	6.621149	31436	31.90001	33322	34.90002	0.7972362	1.354309	1.30277	4.578057	4.635369	6.642954	0.7784575	1.176653	122.80235	0.9895812	4.334115	153.1694	4.155964	7.979736	77.02526	1.092119	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:31.568	22.3	-0.014509448	6.621233	31438	31.90001	33322	34.90002	0.8016911	1.352873	1.301756	4.577438	4.635144	6.642206	0.778451	1.176674	122.79313	0.9895372	4.333958	153.1588	4.155877	7.979707	77.0289	1.092078	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:32.568	22.3	-0.014381925	6.620332	31436	31.90001	33322	34.90002	0.7988476	1.353018	1.301806	4.577573	4.635128	6.641763	0.7784333	1.176556	122.80618	0.9895633	4.334165	153.1749	4.155877	7.979707	77.0289	1.092156	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:54:33.568	22.3	-0.014321431	6.62137	31436	31.90001	33324	34.90002	0.7972631	1.353413	1.302083	4.577575	4.635106	6.643578	0.7783805	1.176501	122.80651	0.9895633	4.334204	153.1705	4.155877	7.979707	77.0289	1.092156	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
SNOBI_Compressor_machine_learnign_test																																				
HW measurements to model																																				
HW measurement output																																				
Soft sensor output																																				
Input parameters to the model																																				
Data tag	Comp_Turb		Comp_M1		Comp_M2		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb		Comp_Turb			
	o1_IN_P		RPM_fb		RPM_fb		o1_IN_T		o2_IN_T		o3_IN_T		o1_IN_qm		o2_IN_qm		o3_IN_qm		o1_IN_qm		o2_IN_qm		o3_IN_qm		o1_IN_qm		o2_IN_qm		o3_IN_qm		o1_IN_qm		o2_IN_qm		o3_IN_qm	
	mbarg		1/min		1/min		°C		°C		°C		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s		kg/s	
2022-11-23-13:54:26.567	-0.014		31436		33322		22		32		35		0.79		0.78		0.78		0.78		0.78		0.78		0.78		0.78		0.78		0.78		0.78		0.78	
2022-11-23-13:55:03.573	22.3	-0.014327806	6.618016	31440	31.99999	33326	34.94782	0.7929258	1.352465	1.30066	4.574741	4.631937	6.637645	0.7785144	1.175786	122.75703	0.9900708	4.333285	153.1349	4.197156	7.981323	77.03304	1.091798	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:04.573	22.3	-0.014394811	6.618809	31438	31.99999	33326	34.90005	0.7944922	1.353488	1.301964	4.576543	4.636076	6.639706	0.7784814	1.175229	122.74706	0.9898471	4.332052	153.1436	4.195786	7.9808	77.03921	1.091781	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:05.573	22.3	-0.014337778	6.619525	31438	31.99999	33324	34.90002	0.7952643	1.353609	1.302491	4.576336	4.634348	6.641435	0.7784801	1.175141	122.7545	0.9898053	4.331656	153.1485	4.195204	7.98042	77.03436	1.091815	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:06.573	22.3	-0.014210752	6.61977	31438	31.99999	33324	34.90002	0.7934031	1.352489	1.302228	4.575897	4.63395	6.641794	0.7784575	1.174941	122.76025	0.9897109	4.331114	153.1547	4.194316	7.980522	77.03436	1.091815	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:07.573	22.3	-0.014074046	6.620744	31438	31.99999	33324	34.90002	0.7940461	1.352858	1.302451	4.576574	4.634787	6.639424	0.778531	1.175319	122.74769	0.9898481	4.331662	153.1392	4.195486	7.97931	77.03759	1.091746	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:08.573	22.3	-0.014403092	6.621645	31436	31.99999	33324	34.90002	0.7967787	1.353527	1.302772	4.577904	4.635806	6.644067	0.7785509	1.17524	122.75172	0.9898185	4.331283	153.1451	4.194935	7.979707	77.02998	1.091746	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:09.573	22.3	-0.014409093	6.623835	31440	31.99999	33326	34.90002	0.7993334	1.354534	1.301321	4.57901	4.638299	6.64554	0.7785199	1.175102	122.74221	0.9896166	4.329988	153.1289	4.194111	7.979757	77.02678	1.091771	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:10.573	22.3	-0.014521069	6.621715	31440	31.99999	33326	34.90002	0.7999517	1.353313	1.30331	4.579507	4.637464	6.644137	0.7785177	1.175318	122.73819	0.9897719	4.331191	153.1271	4.19503	7.979858	77.03871	1.091708	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:11.573	22.3	-0.014408393	6.620537	31440	31.99999	33326	34.90002	0.7968422	1.352841	1.302239	4.577445	4.635326	6.641823	0.7784814	1.175117	122.74405	0.9897469	4.33124	153.14845	4.194535	7.98042	77.03436	1.091815	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:12.573	22.3	-0.014408385	6.619396	31438	31.99999	33324	34.90002	0.7968528	1.354152	1.302256	4.577299	4.635563	6.640491	0.7785285	1.175389	122.74654	0.9899017	4.332092	153.139	4.196052	7.979714	77.04172	1.091726	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:13.573	22.3	-0.01441782	6.617734	31438	31.99999	33326	34.90002	0.795573	1.35294	1.30151	4.577528	4.635308	6.638468	0.7784736	1.175287	122.74657	0.9899209	4.332441	153.1445	4.196397	7.981399	77.04354	1.091767	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:14.573	22.3	-0.014521885	6.617878	31438	31.99999	33326	34.90002	0.7961167	1.353334	1.301566	4.576522	4.634407	6.638165	0.7784964	1.175443	122.74167	0.9899673	4.332856	153.1388	4.196975	7.981029	77.04485	1.091771	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:15.573	22.3	-0.014521885	6.617878	31438	31.99999	33326	34.90002	0.7961167	1.353334	1.301566	4.576522	4.634407	6.638165	0.7784964	1.175443	122.74167	0.9899673	4.332856	153.1388	4.196975	7.981029	77.04485	1.091771	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:16.573	22.3	-0.014341328	6.61189	31440	31.99999	33326	34.90002	0.7967829	1.352483	1.302897	4.577438	4.634623	6.638468	0.7784814	1.175212	122.75291	0.9900047	4.334173	153.1567	4.198045	7.981399	77.04567	1.091798	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:17.573	22.3	-0.014388563	6.609949	31438	31.99999	33324	34.90002	0.796563	1.35247	1.301379	4.57727	4.634008	6.631288	0.7783901	1.17542	122.75745	0.9900219	4.335476	153.1646	4.19706	7.981733	77.04694	1.091798	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:18.573	22.3	-0.014341324	6.607436	31438	31.99999	33324	34.90002	0.7943117	1.351986	1.30024	4.575214	4.629844	6.629991	0.7783516	1.17547	122.75103	0.9902652	4.332623	153.172	4.200513	7.987844	77.06086	1.091798	197744	7.98122	77.0364	198161	7.982167	77.04072	1982091	7.982167	77.04072	1982091	7.982167	77.04072	
2022-11-23-13:55:19.573	22.3	-0.0																																		

NEW BF Project: Machine Learning to enhance global industrial (remote or edge) services



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