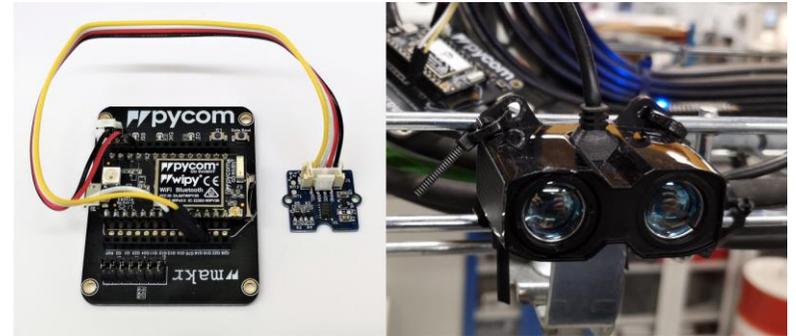


Fuzzy time-series data modeling: theory and anomaly detection on Regatta[®] platform

Mechatronic Circus & Demo Day 2022

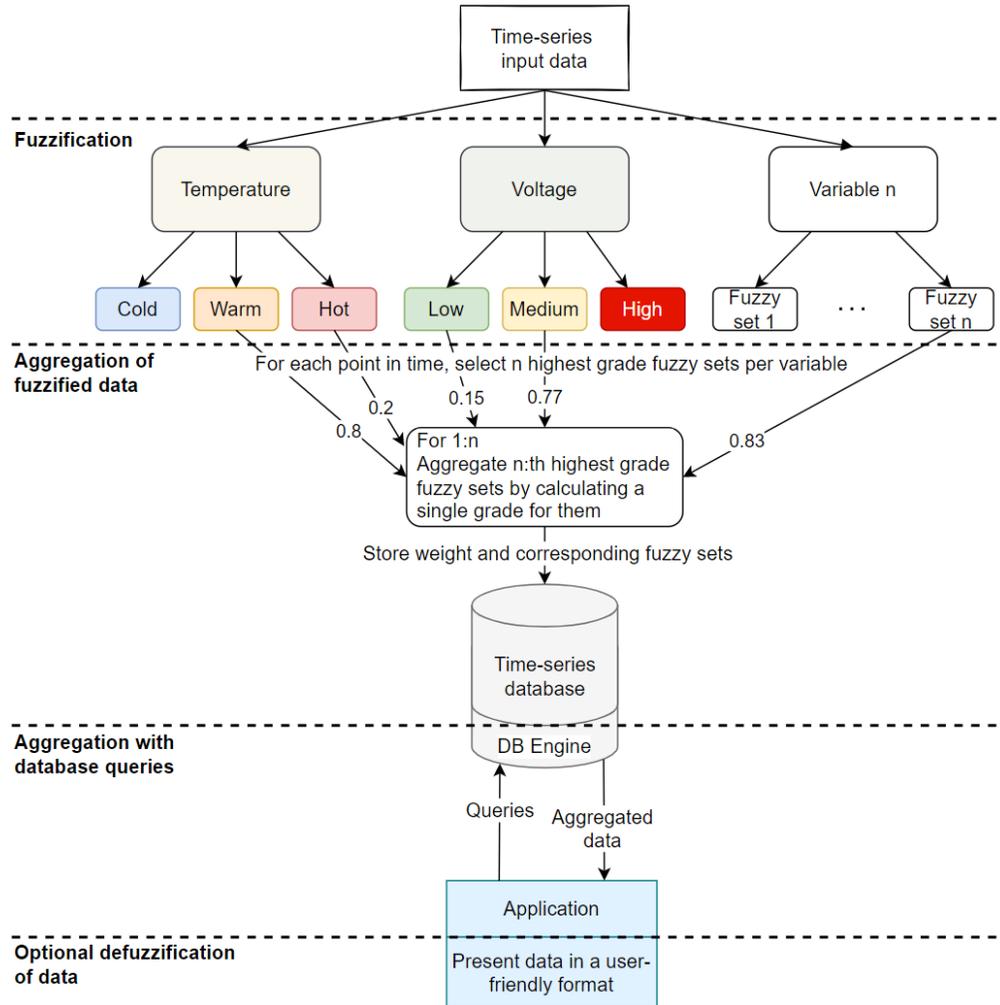
Background: modeling multivariate data

- Modern machines have tens or even hundreds of variables that are measured
- Problem: when the number of variables increases, the computational complexity increases very quickly (even exponentially), problem-specific models
- Solution: Generic method to model machine data enabling diverse instant queries in time-signal space
- Goal: provide information on how machines have been used and conditions that has led to failure



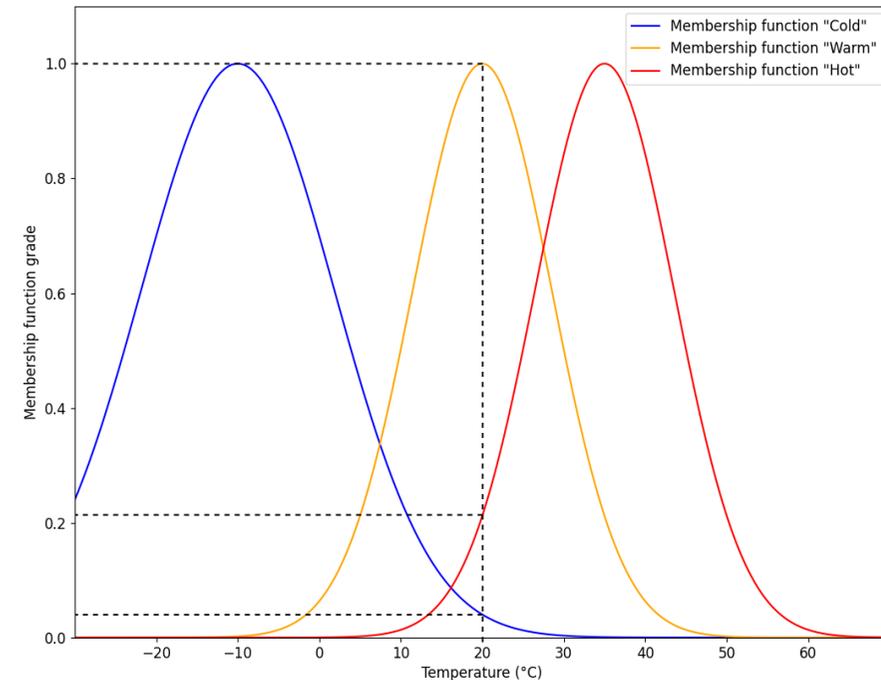
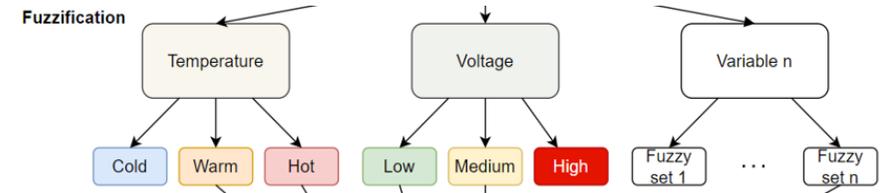
Ala-Laurinaho, R.; Autiosalo, J.; Tammi, K. Open Sensor Manager for IIoT. *J. Sens. Actuator Netw.* **2020**, *9*, 30. <https://doi.org/10.3390/jsan9020030>

Overview of the data modeling method



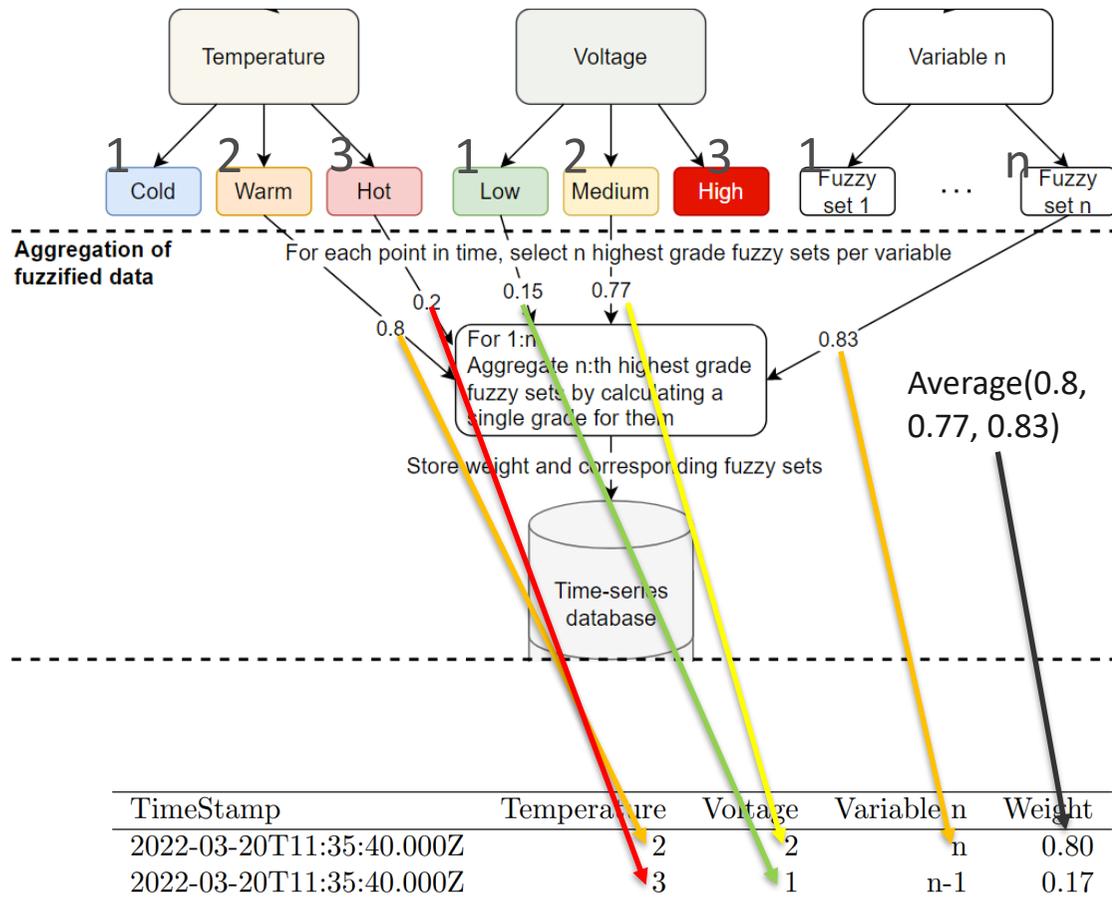
Fuzzification

- **Fuzzy sets** are defined for each variable. For example, cold, warm, and hot for temperature
- Compared to Boolean logic, a value can belong to a fuzzy set with a **continuum of grades from 0 to 1**
- Number of fuzzy sets determines the accuracy of the model
- Using fuzzified variables for time-aggregation in later phases consumes **significantly less memory**



Aggregation to database

- For each point in time, **select** predefined number of the **highest grade fuzzy sets**
- **Combine the grades** of n:th highest grade of fuzzy sets as **weight** with different conjunction methods such as average, product, min, max, sum, etc.
- After spatial aggregation, data can be aggregated in **time-domain** by summing weights together in a **time bucket** -> more efficient queries



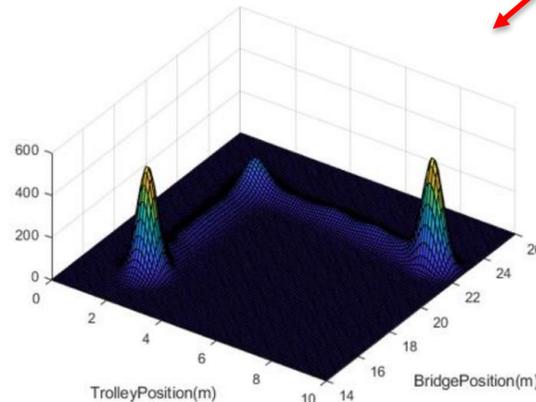
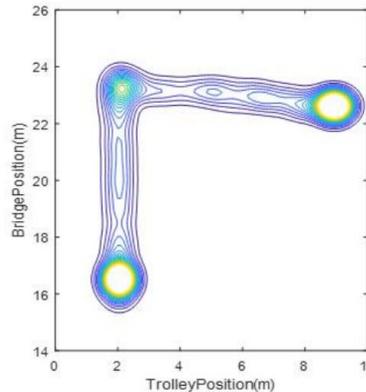
Database queries and visualization

- Queries allow further **aggregation of data**, for example, within certain time range
- Queried data can be used with **visualizations**, such as crane position

```
SELECT "TrolleyPosition", "BridgePosition", SUM(weight) AS sum_weight
FROM craneData
WHERE time BETWEEN '2020-01-30' AND '2022-03-20'
GROUP BY "TrolleyPosition", "BridgePosition"
ORDER BY sum_weight DESC
```



TrolleyPosition	BridgePosition	sum_weight
0	2	341.55
0	1	254.21
1	3	124.47
1	4	105.13
2	0	79.82
2	1	34.68





Analysis

Evaluate raw signal values against data model confidence scores

< 03.01.2022 00.00 - 07.01.2022 12.00 >



Data source

Model

Temperatures

Entity

Reach Stacker

Evaluated signal *i*

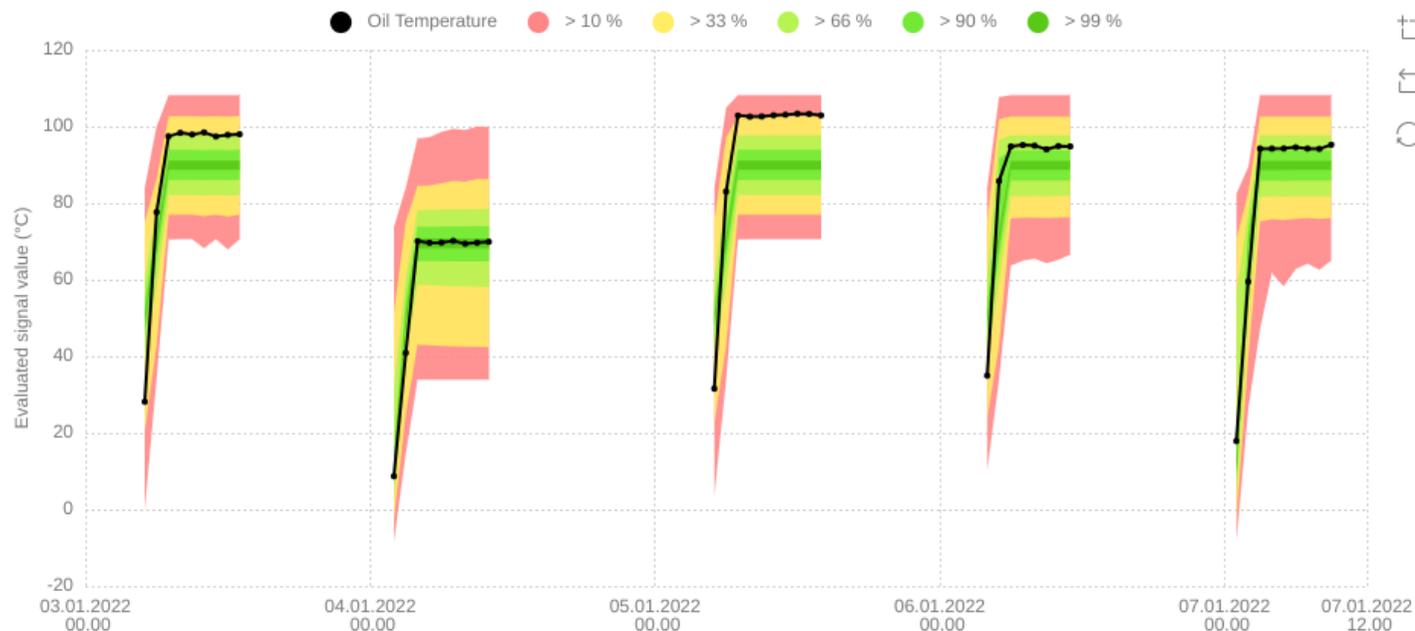
Oil Temperature
on temperature dataset

Explaining signals *i*

Cabin temperature
on temperature dataset

Engine Temperature
on temperature dataset

Time-series analysis Entity distribution



Anomaly detection on Regatta[®] platform

1 Select data

- 10 Reach stacker machines, each operating ~8 hours per day in different climates
- Data from 1 year period, 7.4.2021 – 7.4.2022
- 3 temperature signals reported every 5 minutes



Reach stackers A-J

		Raw temperatures		
Time	Machine	Engine	Cabin	Oil
7.4.2021 19:05	A	26.1	27.2	26.4
7.4.2021 19:10	A	29.2	26.9	28.9
7.4.2021 19:15	A	26.2	26.8	36.3
7.4.2021 16:10	B	-0.8	0.7	-0.6
7.4.2021 16:15	B	6.4	1.8	5.5
7.4.2021 16:20	B	3.8	5.1	9.2

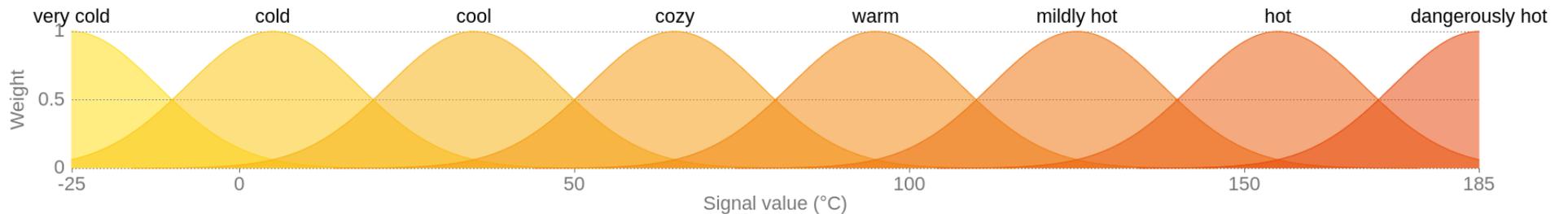
Raw data sample, total of 420k rows

2 Define concepts

- Each signal is defined with Gaussian fuzzy terms
- Linguistic names help in data interpretation

Signal	Linguistic terms
🔥 Oil Temperature	very low low cold normal warm high extremely high
🌡 Cabin temperature	freezing uncomfortable cold warm hot
🔥 Engine Temperature	very cold cold cool cozy warm mildly hot hot dangerously hot

🔥 Engine Temperature
on temperature dataset



3 Setup modeling parameters

Parameters

Model update and aggregation frequency ⓘ *

7 days ⓘ

What is the size of the **time bucket** and **model update frequency**?

Model aggregation function *

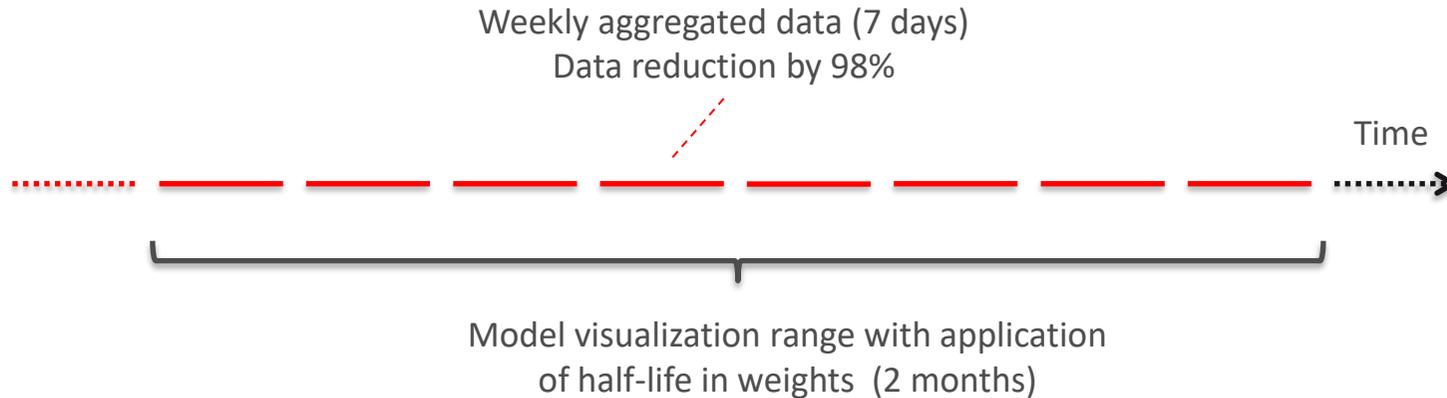
Mean value ▾

What **time-aggregation function** should be used?

Data modeling interval ⓘ *

2 months ⓘ

How old data is included in model queries?



4 Time-series analysis

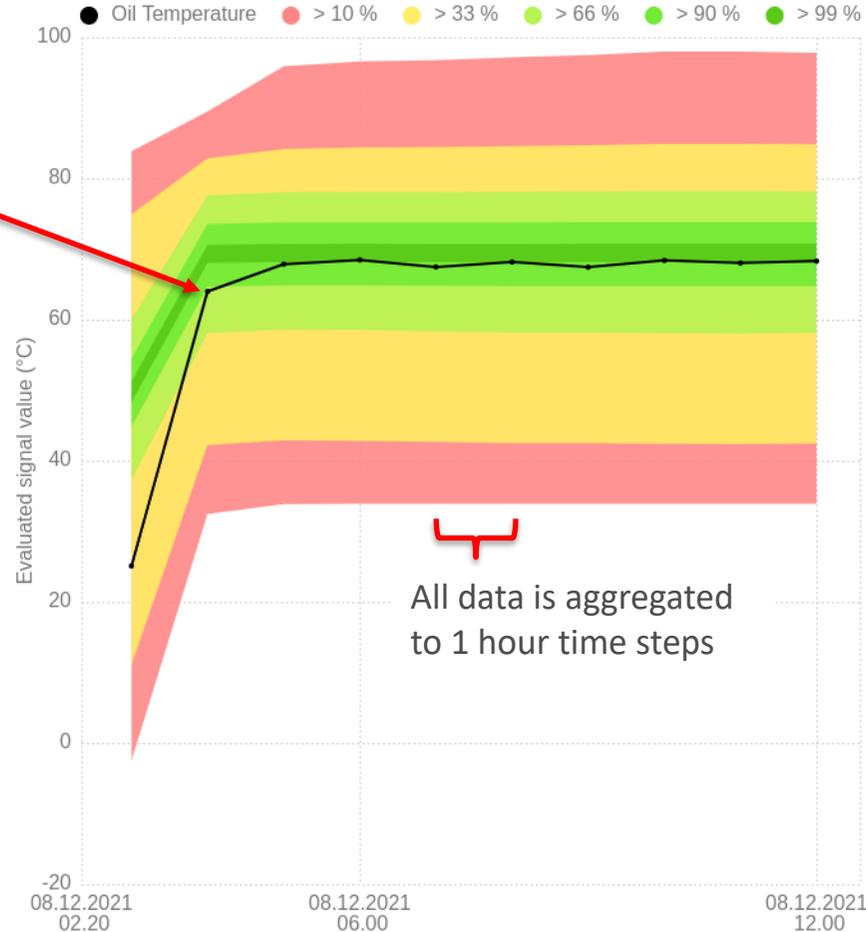
Evaluated raw signal for machine A

Oil Temperature
on temperature dataset



Notes

- Machines have cooled down before start-up and **oil temperature warms up during the first hour of use**
- After that oil temperature maintains a **relatively level temperature**
- The operation of this machine shows **some anomaly during warmup but no anomaly during constant use**



Confidence score distribution

Confidence score ranges (0-100%) are predicted using other raw signals from the model at measurement time



Explaining signals

Cabin temperature
on temperature dataset

Engine Temperature
on temperature dataset

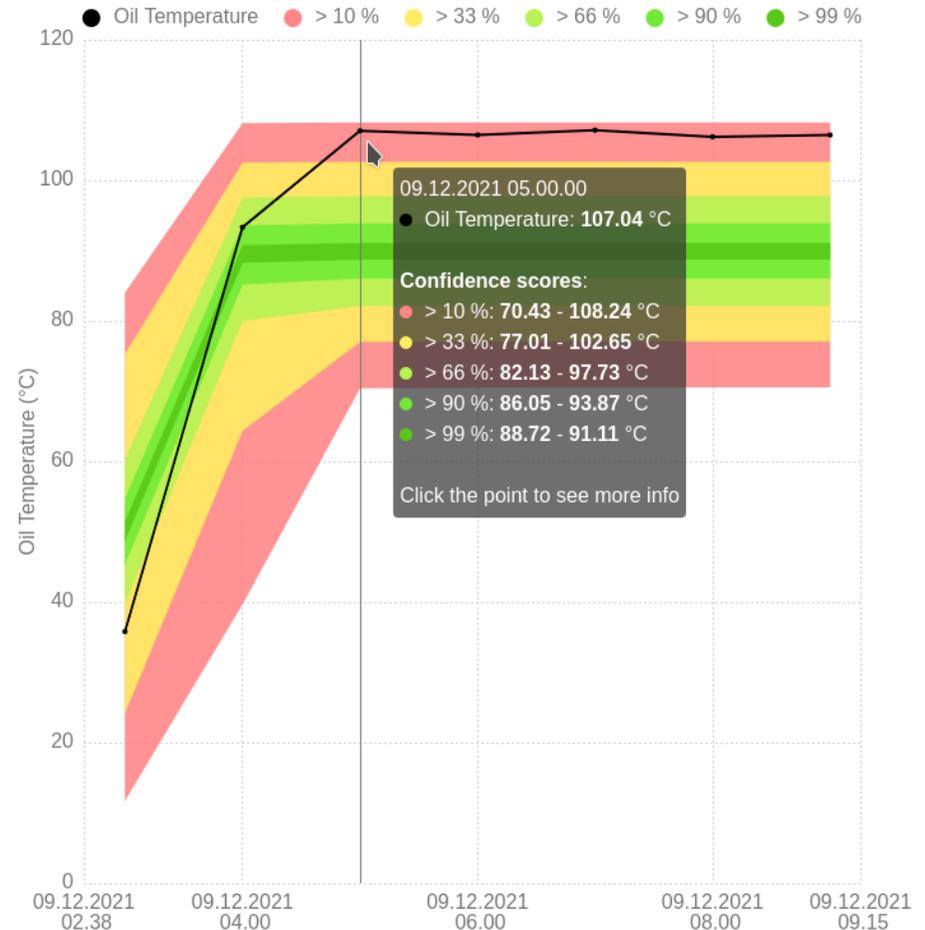
4 Time-series analysis

Notes

- As ambient air temperature changes, the confidence score distribution rises
- Machine shows fairly normal operation during warmup, but oil temperature clearly overheats in constant use

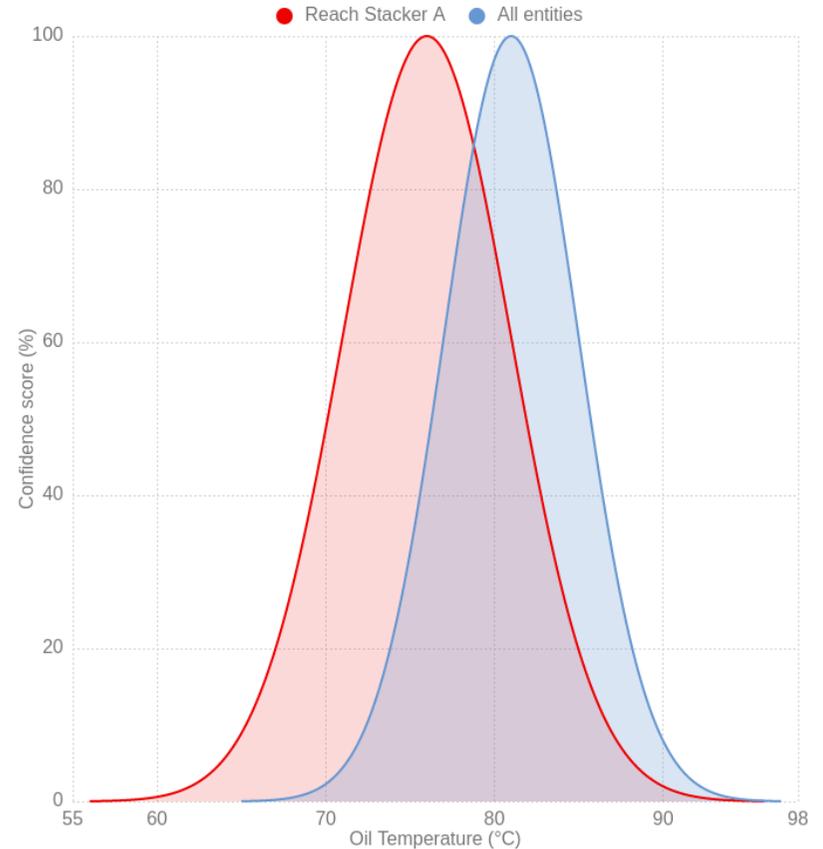
Time-series analysis view

- Allows its users to easily model the raw time-series data and visually explore anomalies using confidence score distributions



5 Model exploration

- As model data is updated periodically in the background, the model can be **visualized in realtime for predictive analytics purposes at fleet level**
- For example, **signal distributions from selected machines can be compared to other machines** operating at similar conditions, enabling early problem detection
- Model changes will enable **signal level predictions and proactive anomaly alerting** from all connected machines



Entity signal-distribution analysis view



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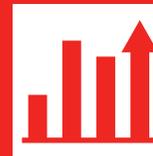
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