

KU LEUVEN



LEUVEN.AI

Advanced Integrated Sensing

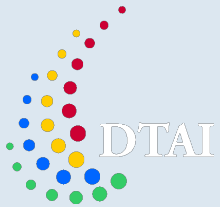
ADVISE

FLANDERS
MAKE
DRIVING INNOVATION IN MANUFACTURING

Data-driven Machine Condition Monitoring in Rotating Machinery

Maarten Meire, Quinten Van Baelen, Lode Vuegen, Peter Karsmakers

*KU Leuven - Department of Computer Science
Declarative Languages and Artificial Intelligence (DTAI)*



peter.karsmakers@kuleuven.be



<https://dtai.cs.kuleuven.be>

1

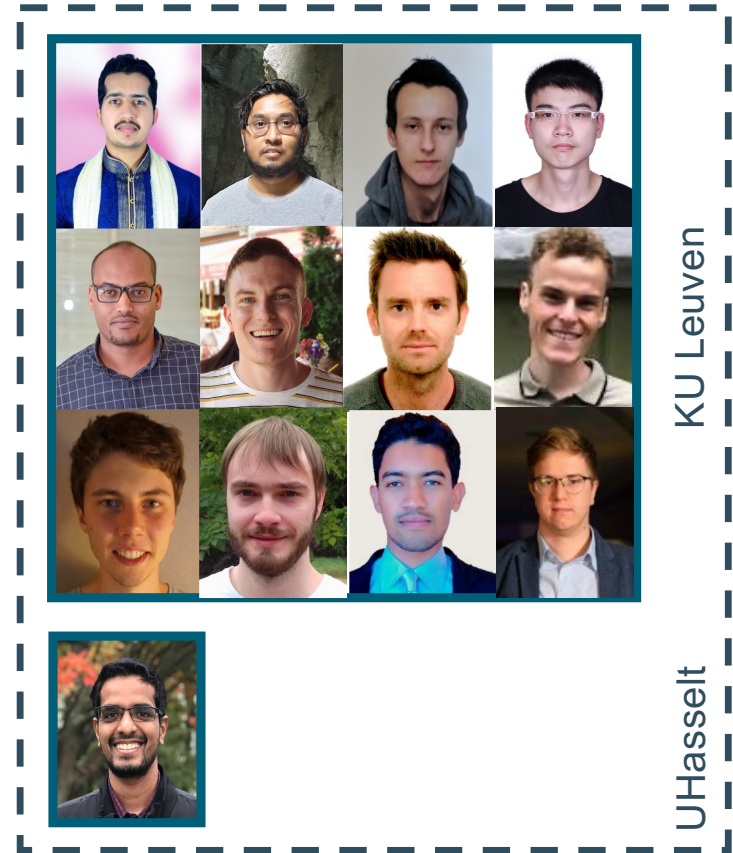
Overview

- DTAI@Geel
- Constraint guided deep learning
- Use-case: condition monitoring

DTAI@Geel



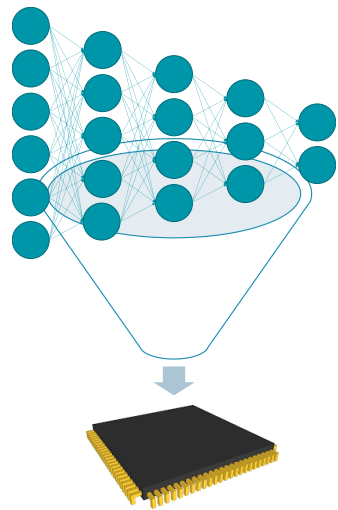
- DTAI@Geel people
 - 1 Professor
 - 2 FTE Postdoc
 - 11 FTE PhD
- Flanders Make Core Lab



DTAI@Geel

Common theme

Machine learning under limiting conditions

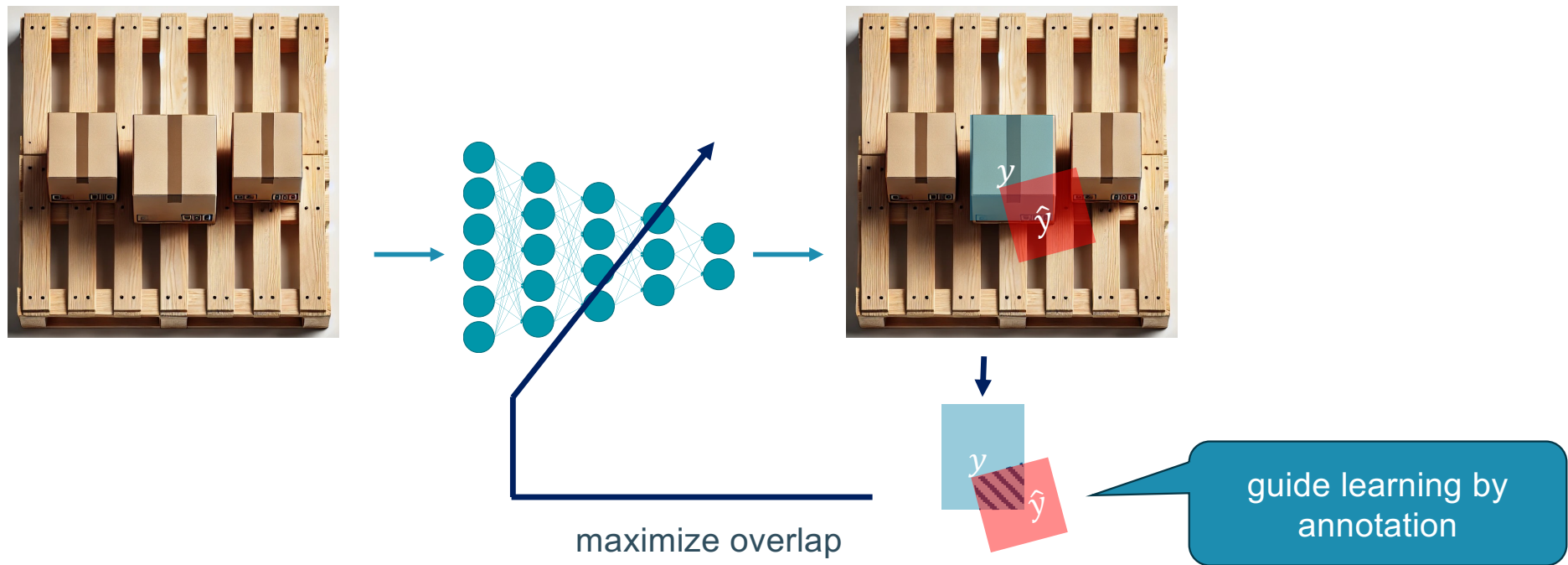


with a focus on (acoustic) time-series signals

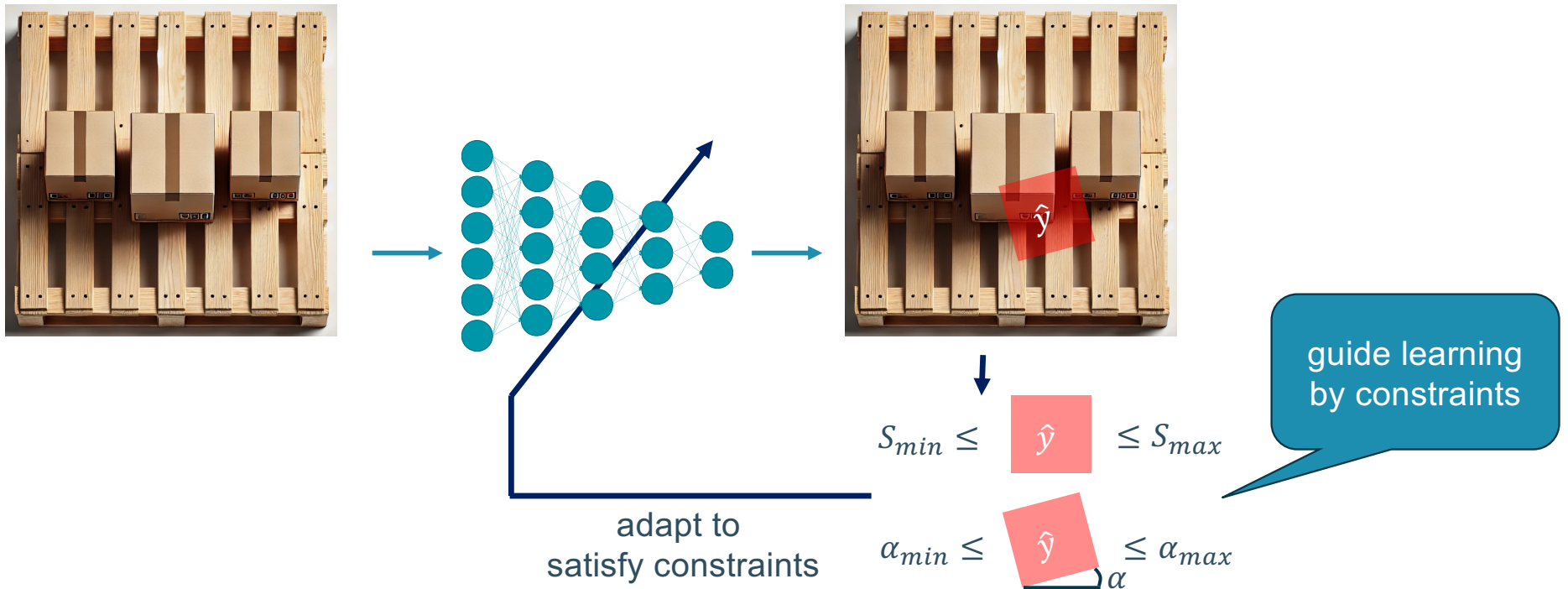
Overview

- DTAI@Geel
- **Constraint guided deep learning**
- Use-case: condition monitoring

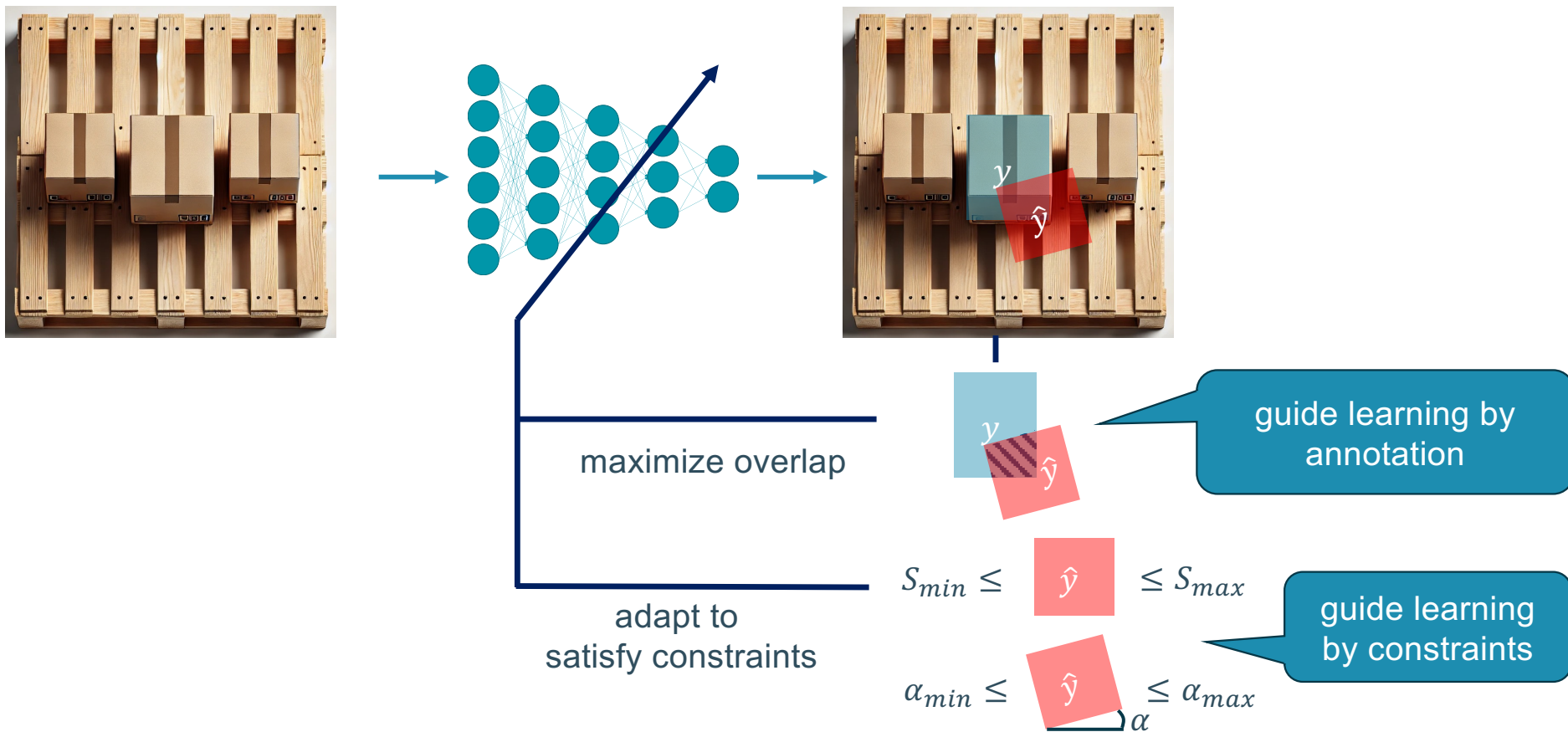
Annotation guided (supervised) deep learning



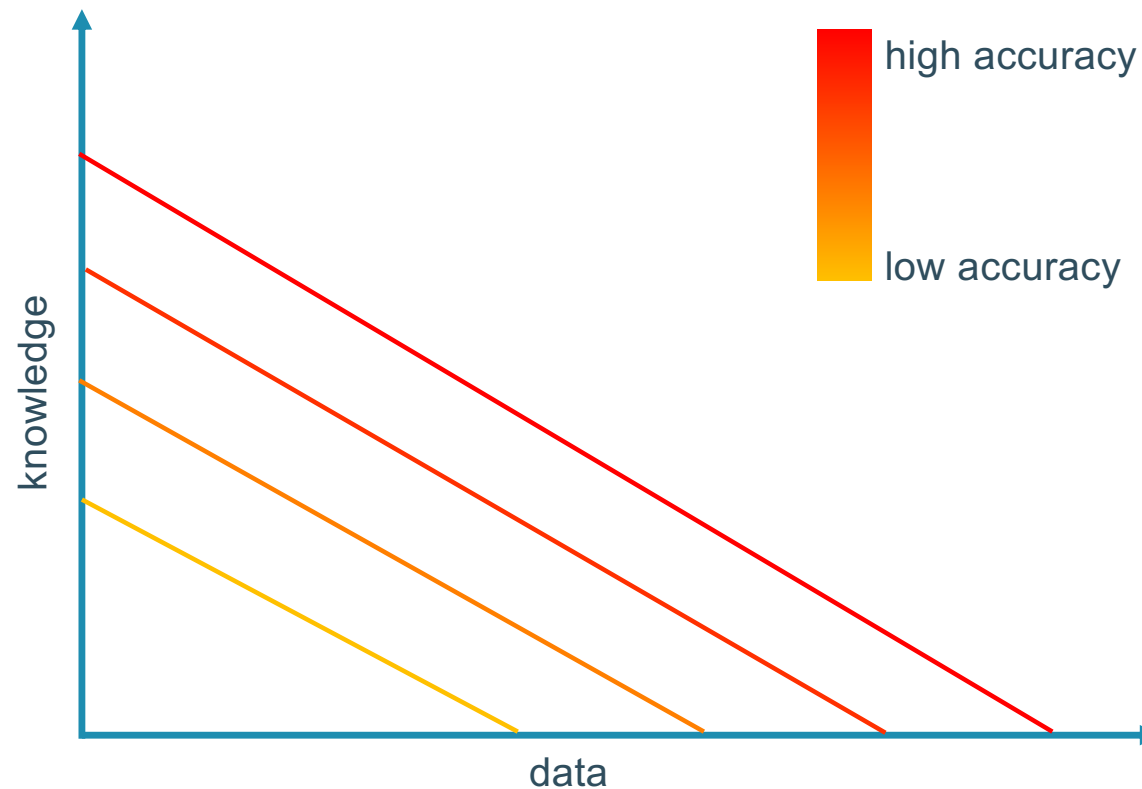
Constraint guided deep learning



In-house framework: *Annotation and constraint guided deep learning (CGGD)*



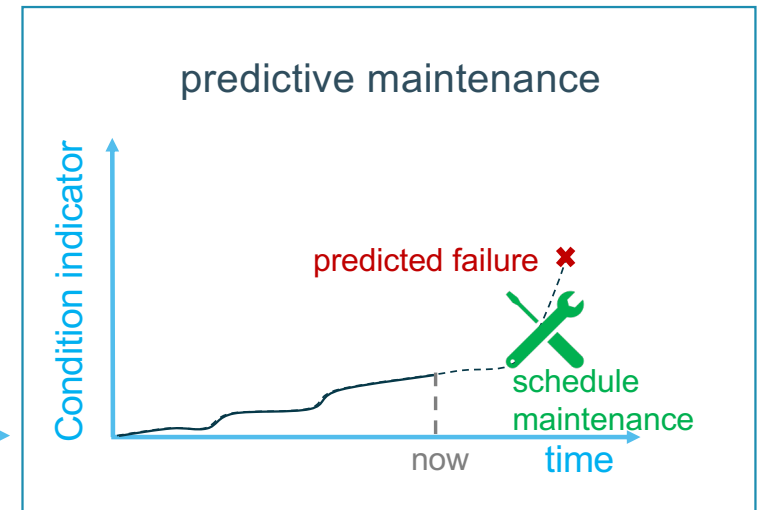
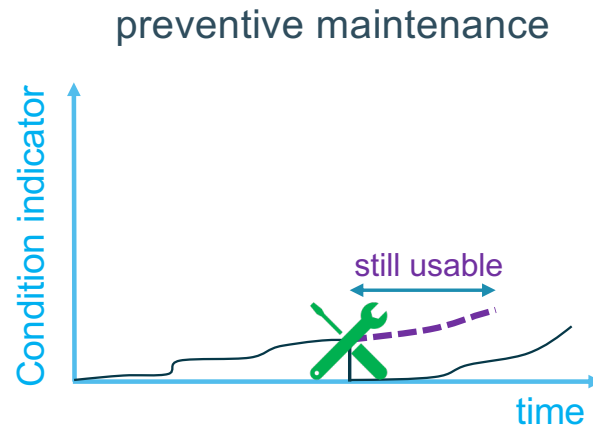
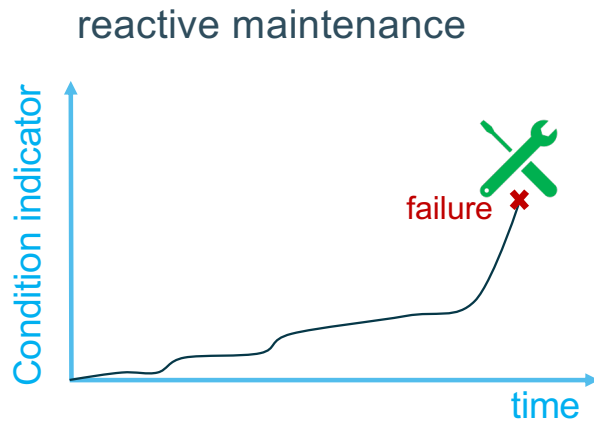
In-house framework: *Annotation and constraint guided deep learning (CGGD)*



Overview

- Research focus
- Constraint guided deep learning
- **Use-case: condition monitoring**

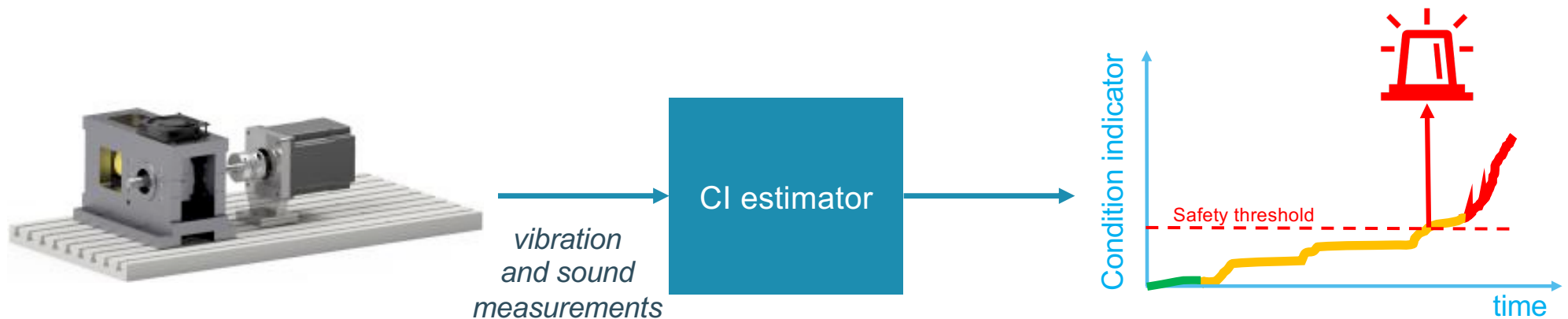
Use-case: condition monitoring



our focus

Use-case: condition monitoring

Goal: data-driven learning of a (high quality) condition score to enable incipient defect detection in rolling element bearings

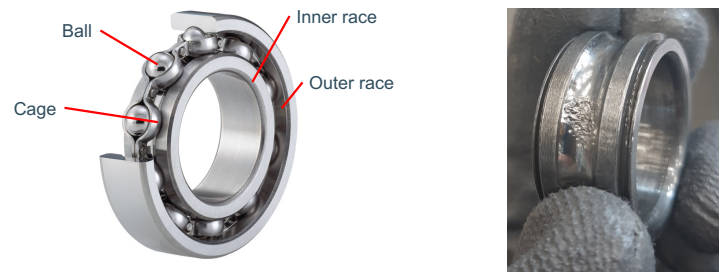
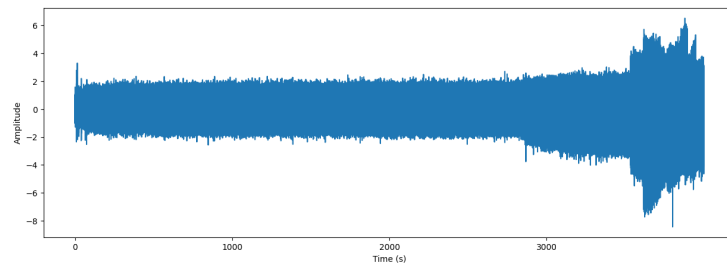


Use-case: condition monitoring

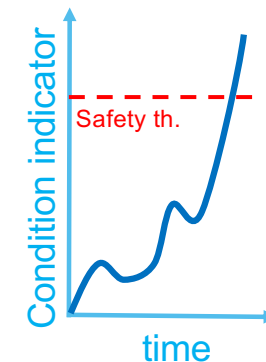
Baseline: handcrafted condition indicators

Hand-crafted approach:

- set of rules based on domain knowledge
- simple (univariate) checks
- might be time consuming



Handcrafted condition indicators
RMS
Kurtosis
BPFI
BPFO
BSF
...

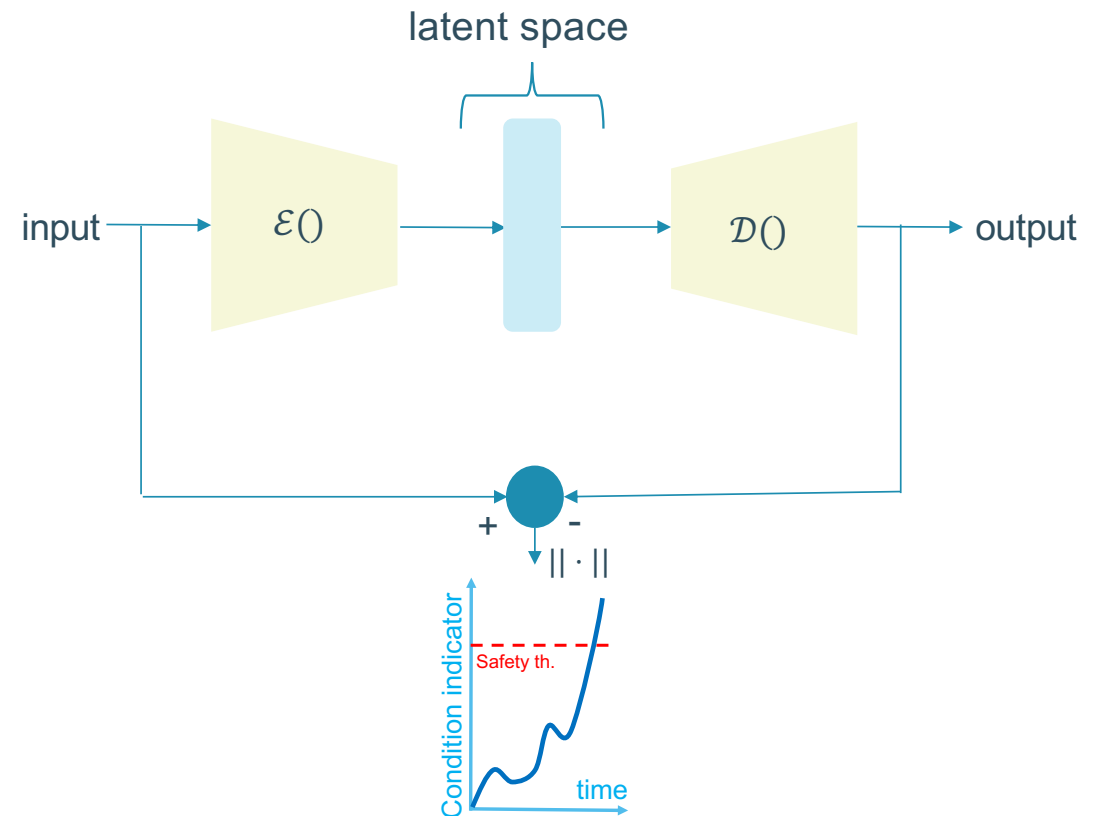


Use-case: condition monitoring

Baseline: data-driven based condition indicator

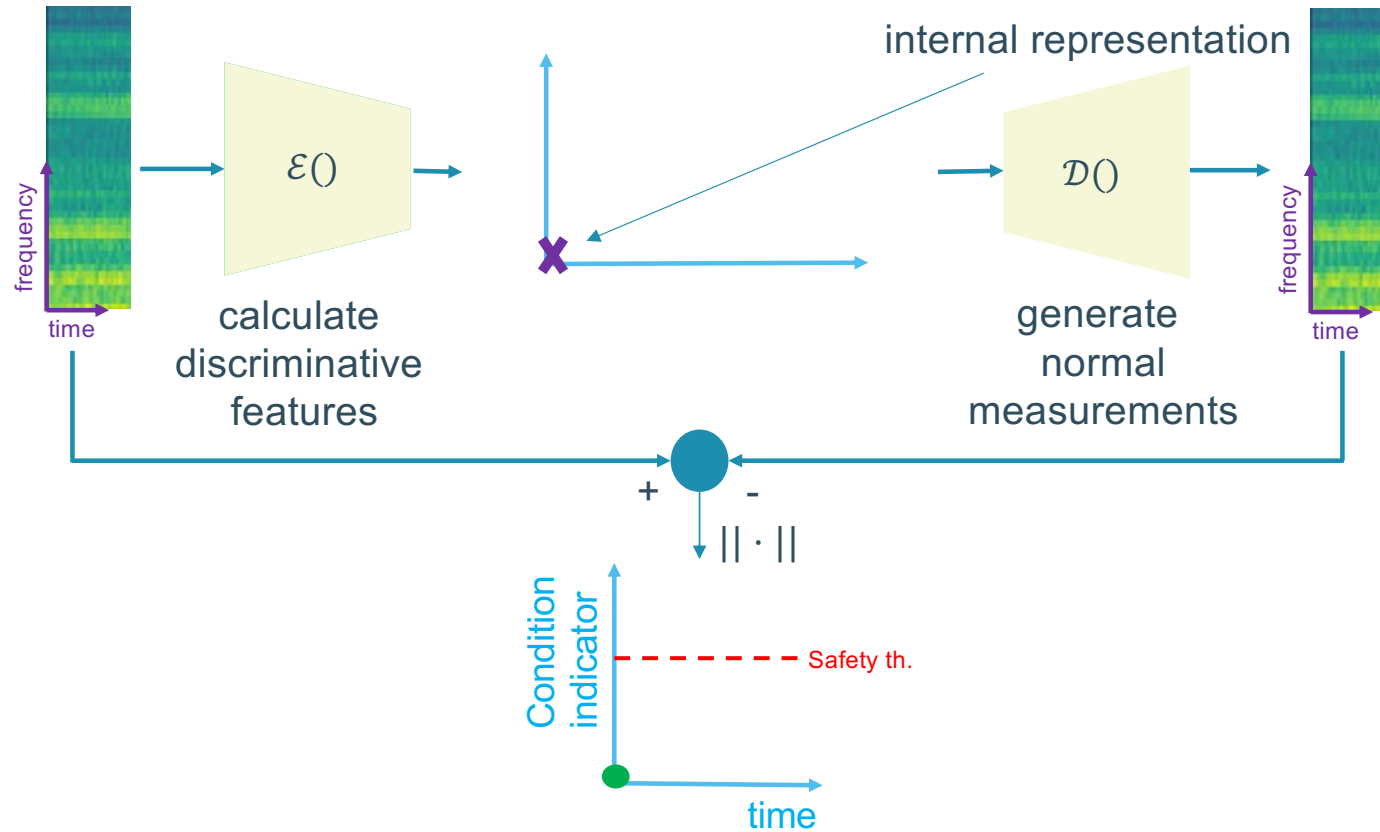
Data-driven approach:

- learn patterns that indicate condition automatically
- more advanced multivariate checks possible
- less engineering time
- e.g. AE based CI estimation



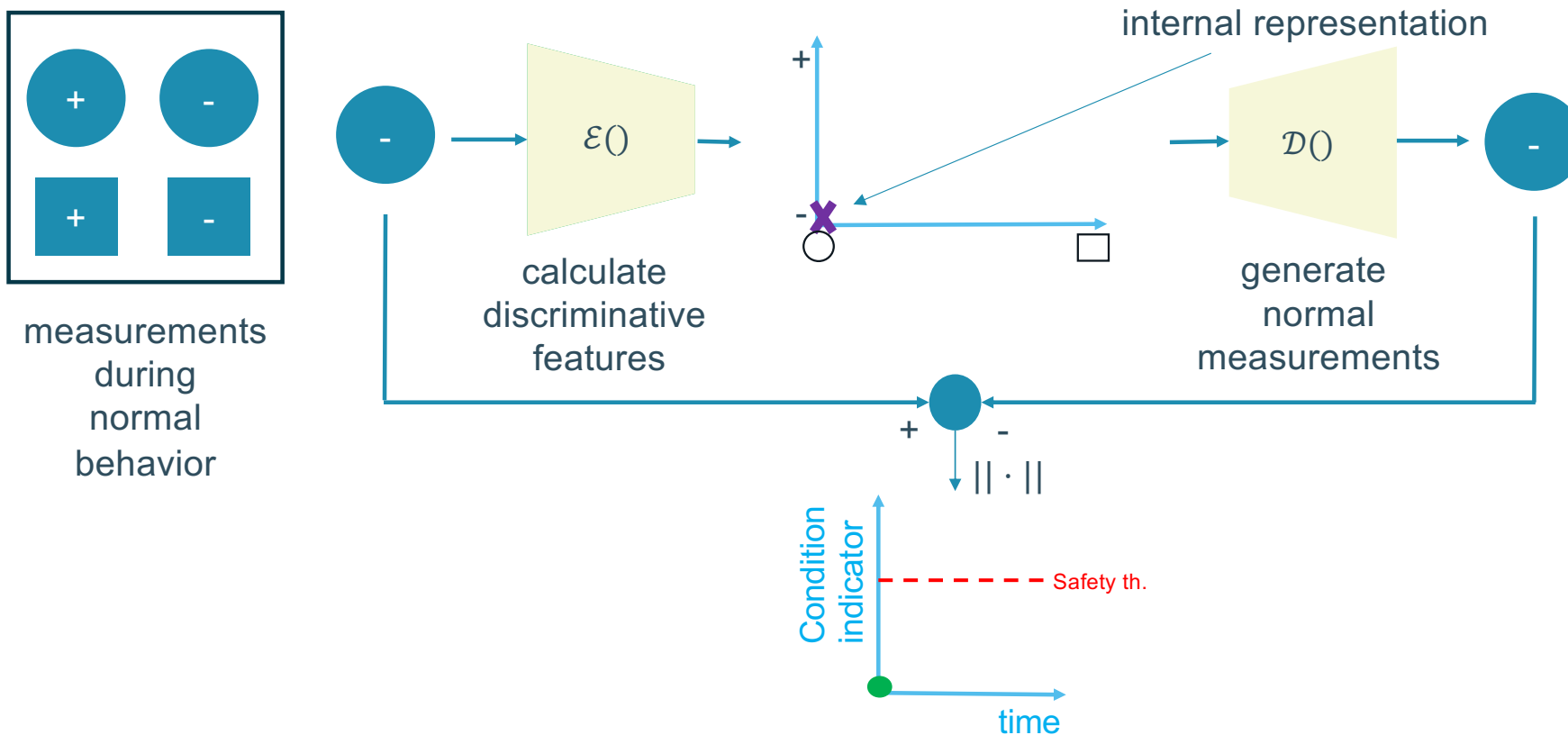
Use-case: condition monitoring

Baseline: data-driven based condition indicator



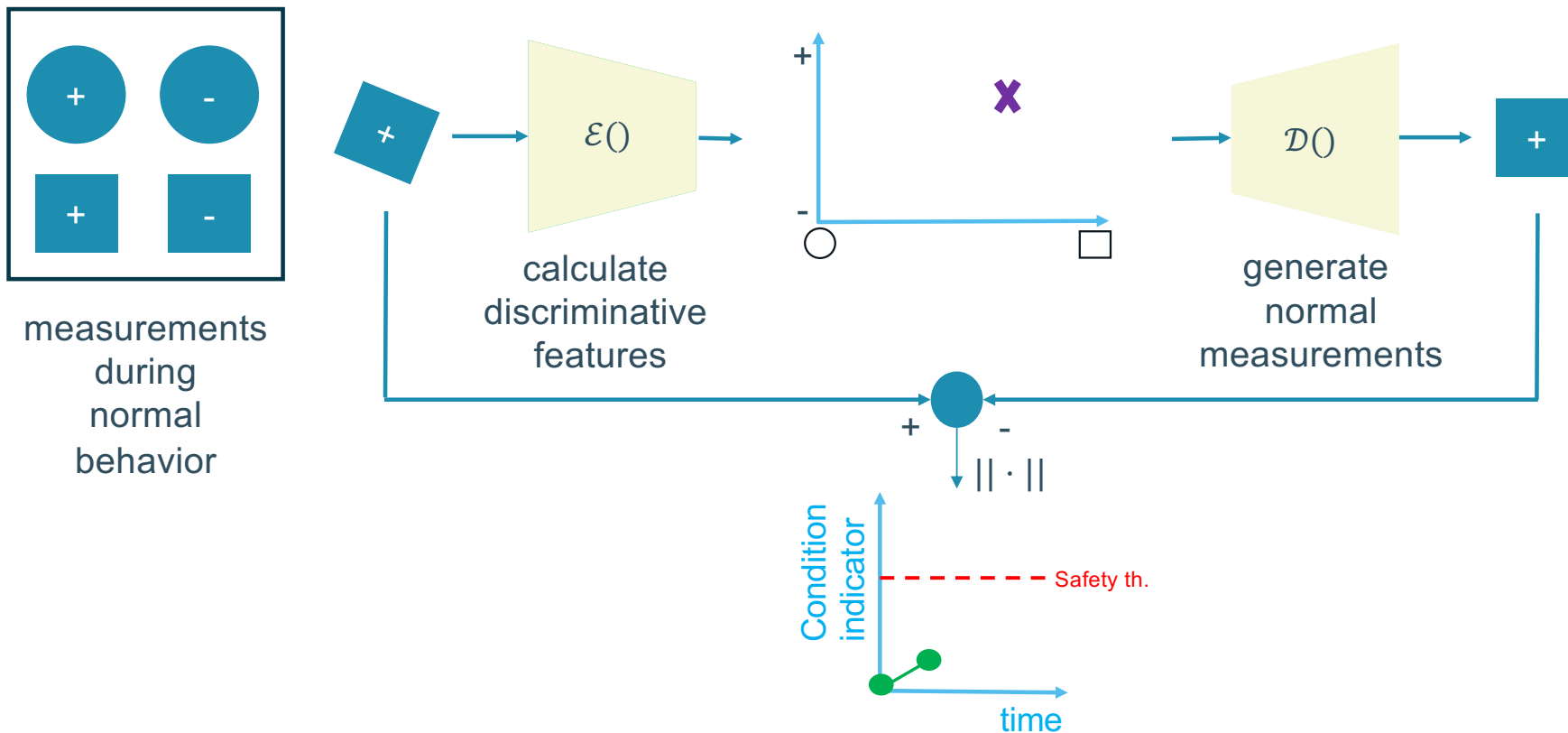
Use-case: condition monitoring

Baseline: data-driven based condition indicator



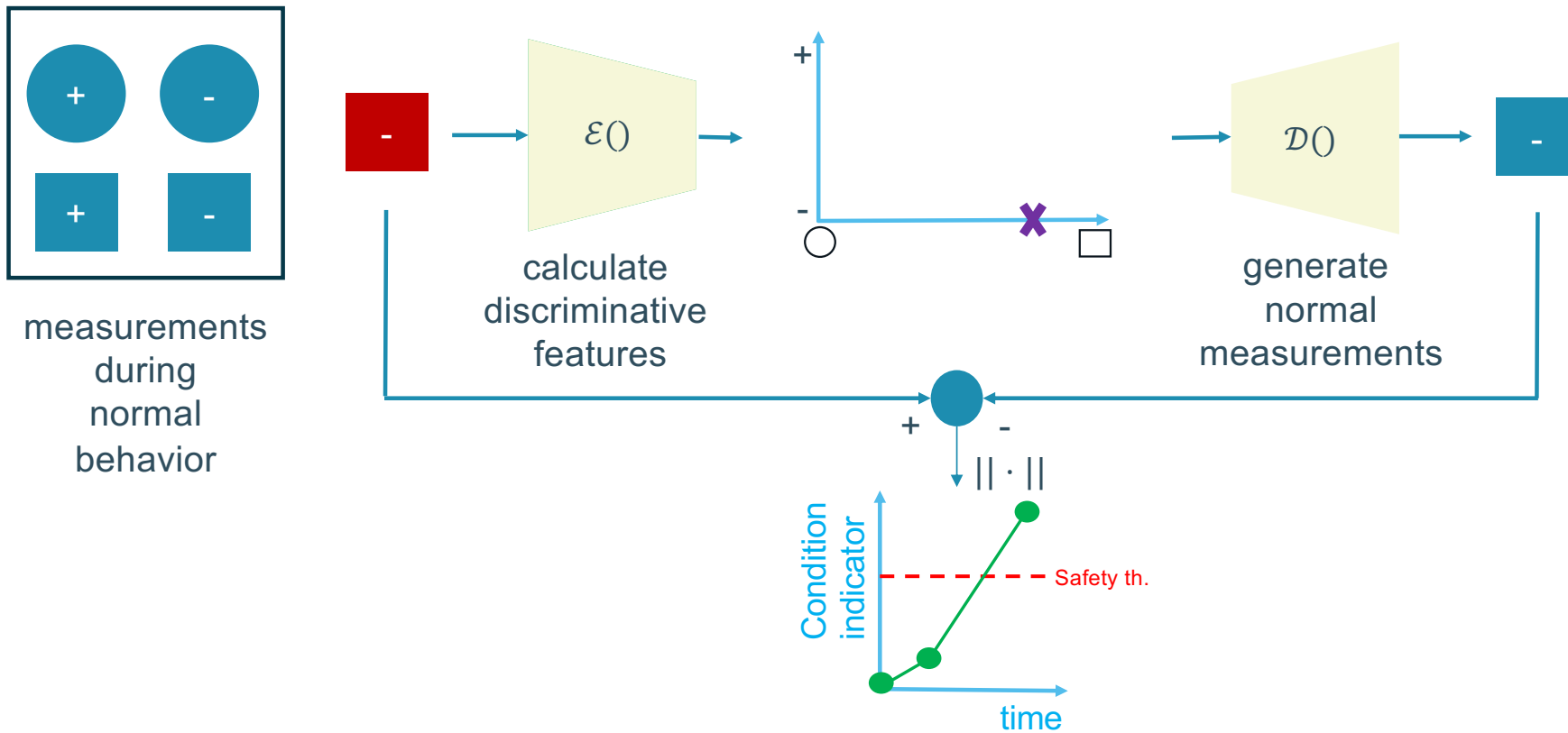
Use-case: condition monitoring

Baseline: data-driven based condition indicator



Use-case: condition monitoring

Baseline: data-driven based condition indicator

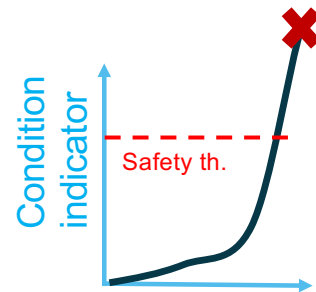


Use-case: condition monitoring

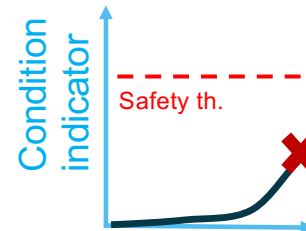
Baseline: data-driven based condition indicator



Asset A



Asset B

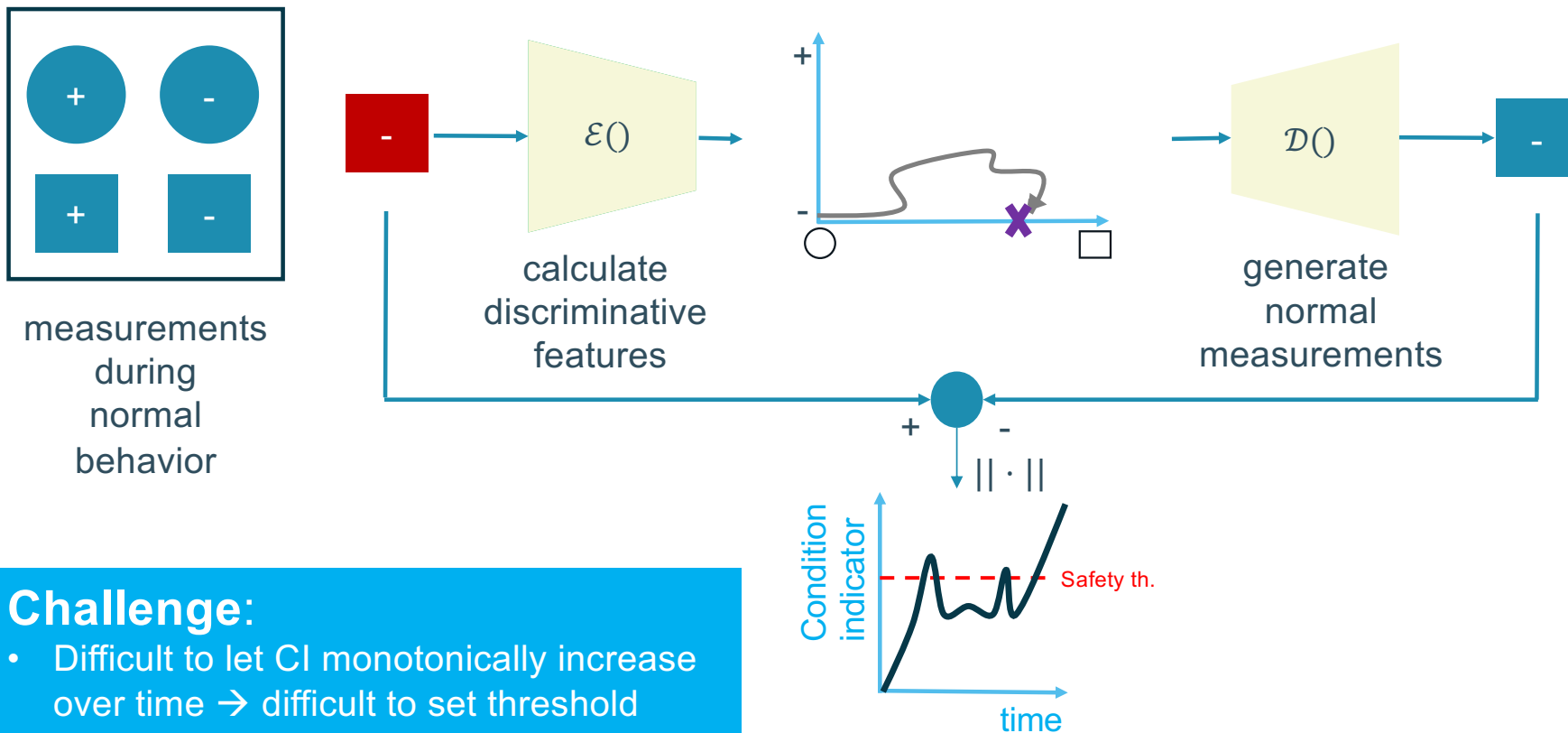


Challenges:

- CI range depends on machine
→ difficult to set threshold

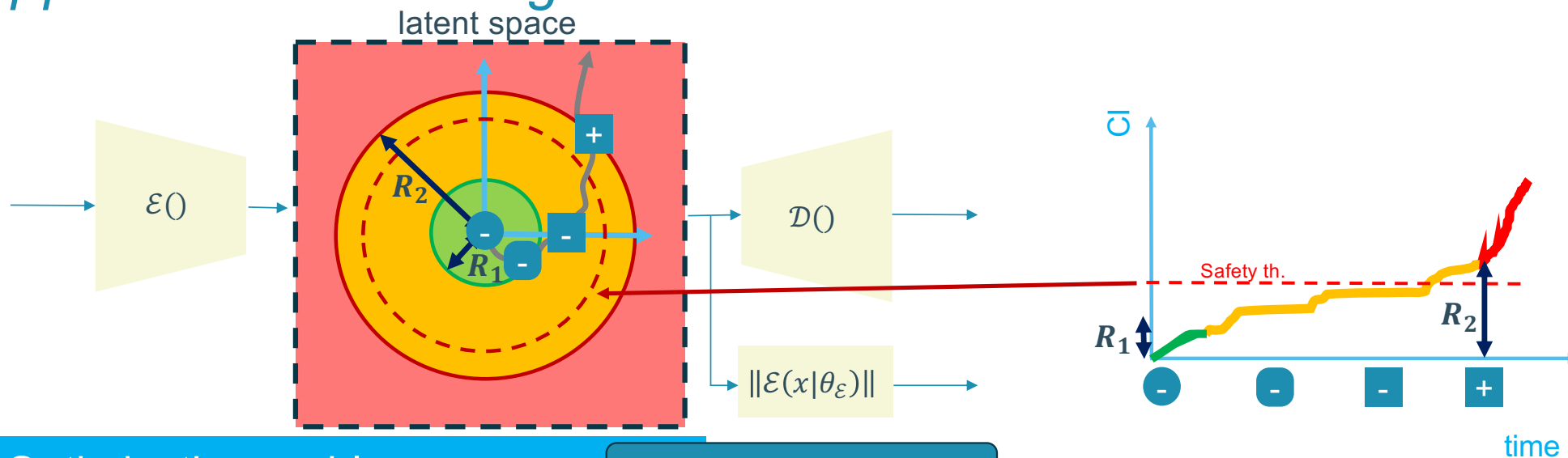
Use-case: condition monitoring

Baseline: data-driven based condition indicator



Use-case: condition monitoring

Approach: constraint guided DL CI estimation



Optimisation problem:

Minimize *reconstruction error*

Subject to *structure constraints*

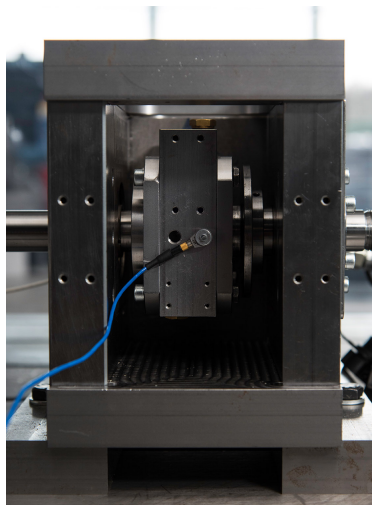
deep learning

deep learning
&
constraint satisfaction
→ Use CCGD

Use-case: condition monitoring

Constraint guided DL CI estimation: snapshot results

Vibration based bearing monitoring

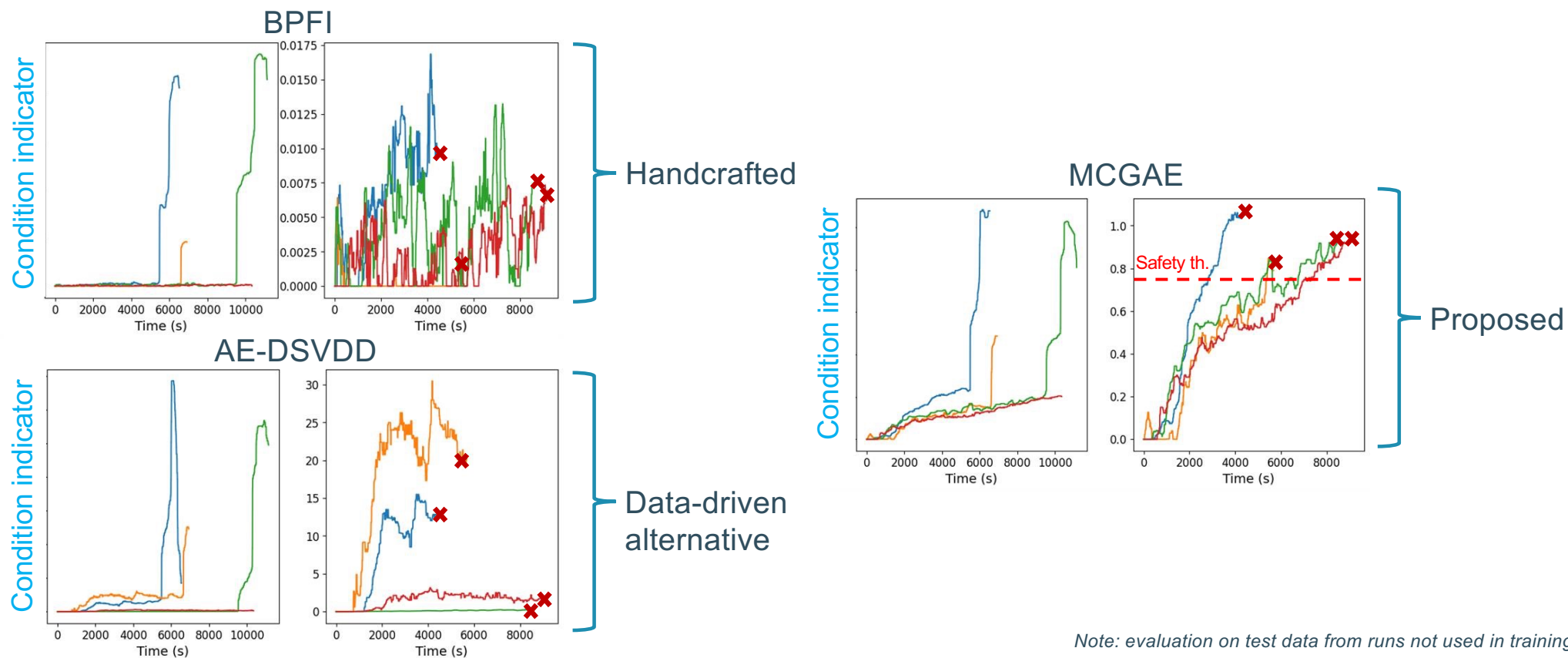


Sound based bearing monitoring



Use-case: condition monitoring

Constraint guided DL CI estimation: snapshot results



Note: evaluation on test data from runs not used in training

Conclusion

- Constraint guided DL enables to learn powerful CI estimators from data for predictive maintenance without much engineering effort
- Adding structure to DL model via constraints results in:
 - a CI estimator with better generalisation properties
 - allowing a fixed, asset independent, safety threshold to be set