



# Data- en expertgedreven aanpak voor efficiënte én effectieve installaties: Van onderzoek naar maatschappelijke impact

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# Inhoud

## Ontwikkelingen en vooruitzichten:

- Brains4Buildings project
- IEA EBC Annex 81: Data-driven smart buildings
- Eindhoven Engine B4B-APK project



Prof.em. Wim Zeiler

# Belangrijkste uitdagingen



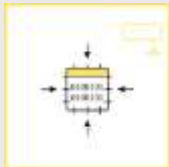
Energieverspilling door slechtwerkende gebouwinstallaties



Balanceren vraag en aanbod bij inzet hernieuwbare energie

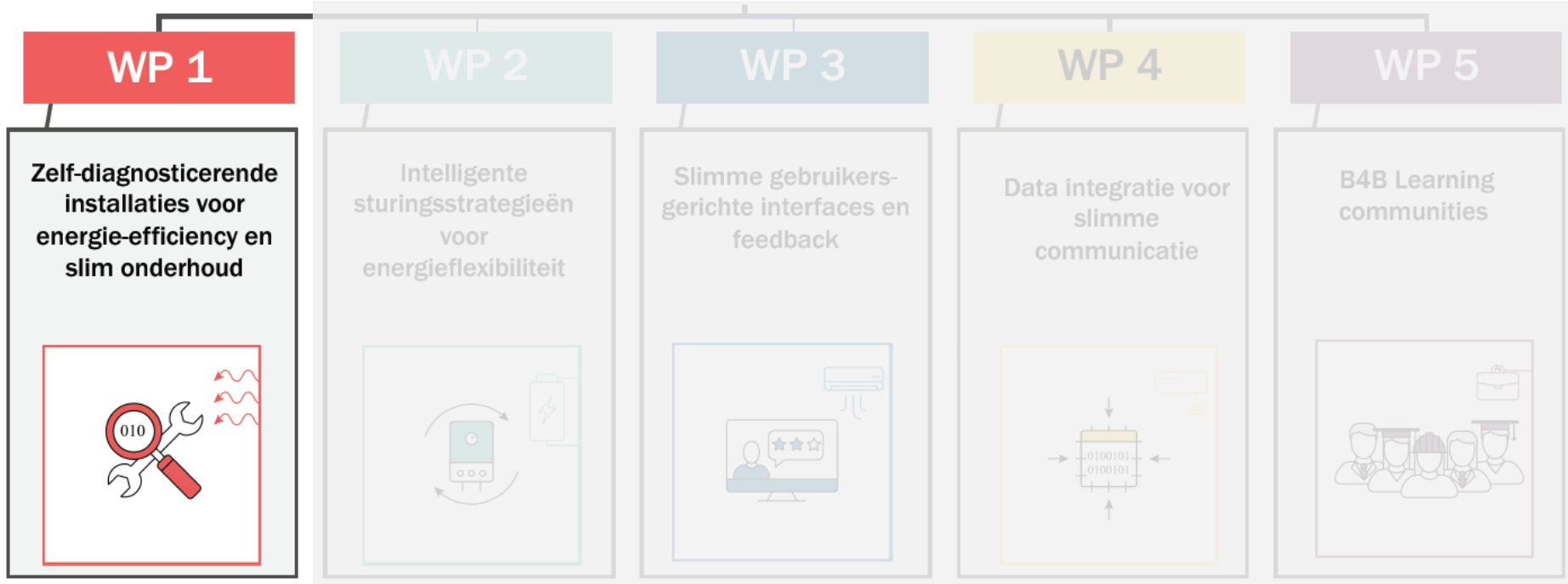


Klachten rondom comfort en welbevinden

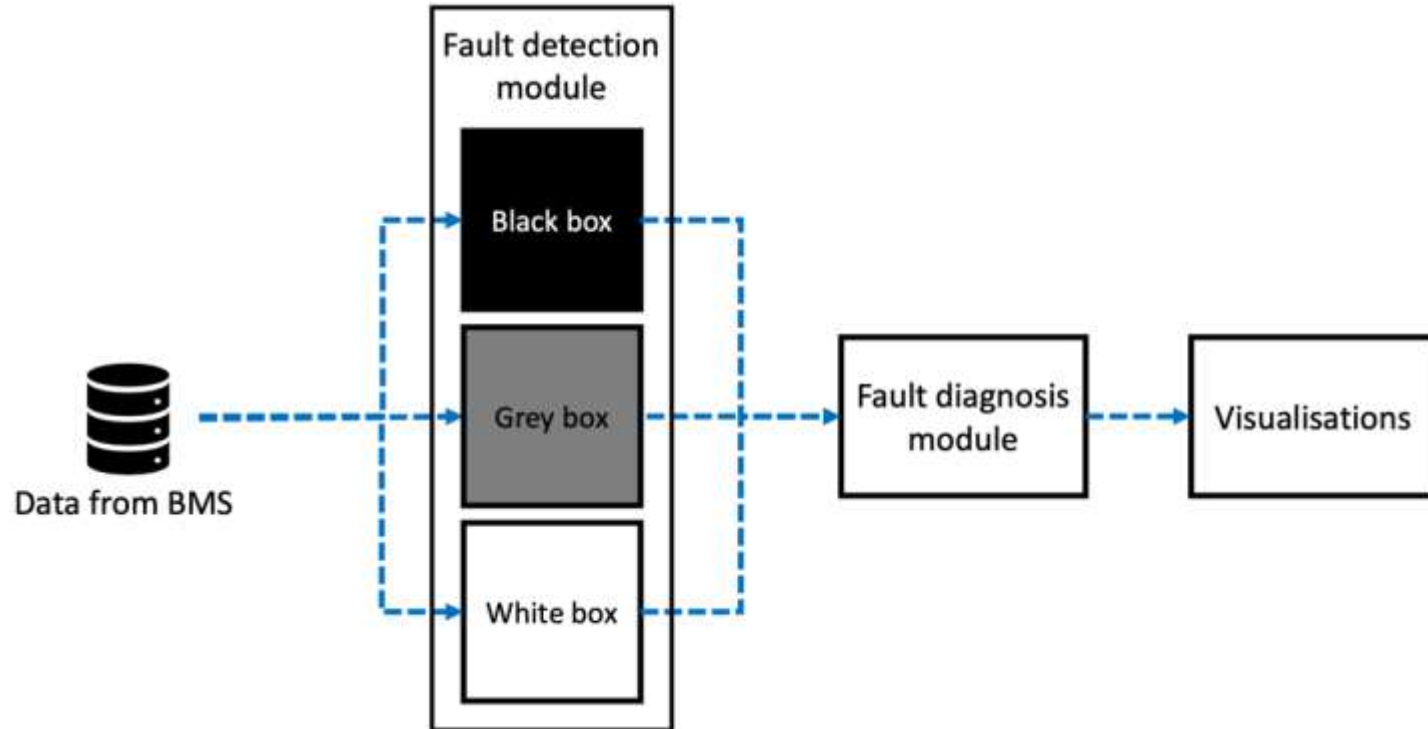


Gebrek aan interoperabiliteit van bestaande systemen

# Brains4Buildings: Beoogde oplossingen



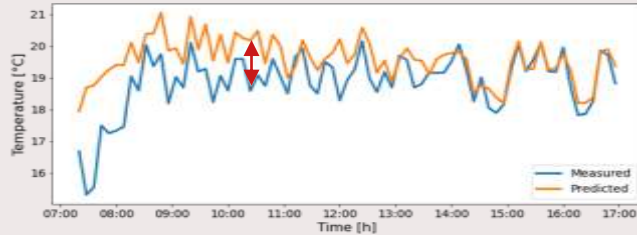
# Automatische foutdetectie én foutdiagnose



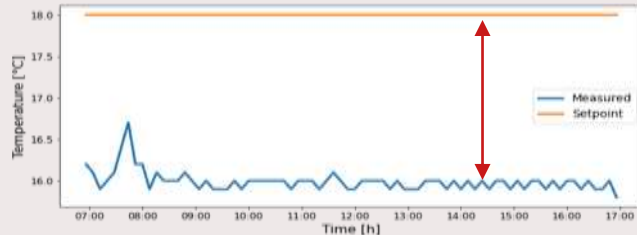
# Automatische foutdetectie én foutdiagnose

## Machine Learning: symptoom-DETECTIE

ML-Based  
Fault  
Detection

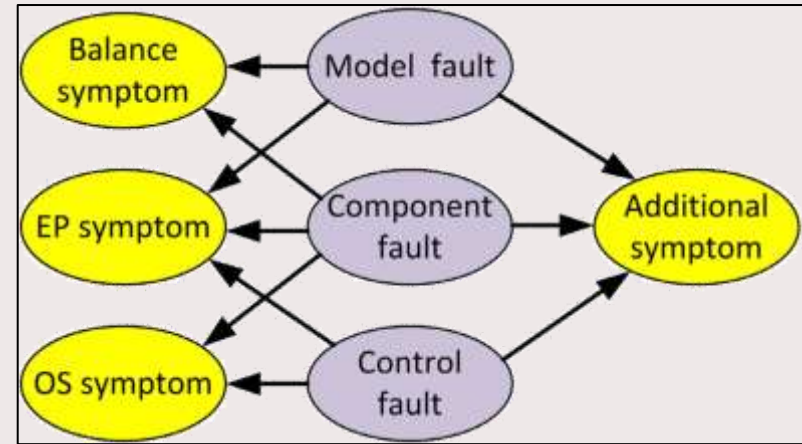


Rules-Based  
Fault  
Detection



(A. Thamban, 2021 , TUE)

## Bayesiaans Netwerk: fout-DIAGNOSE



(A. Taal, proefschrift, 2021, TUE)

# Automatische foutdiagnose: Bayesiaans netwerk

Diagnose ziekten

Bijvoorbeeld:

A: Koorts

B: Hoesten

C: Vermoeidheid

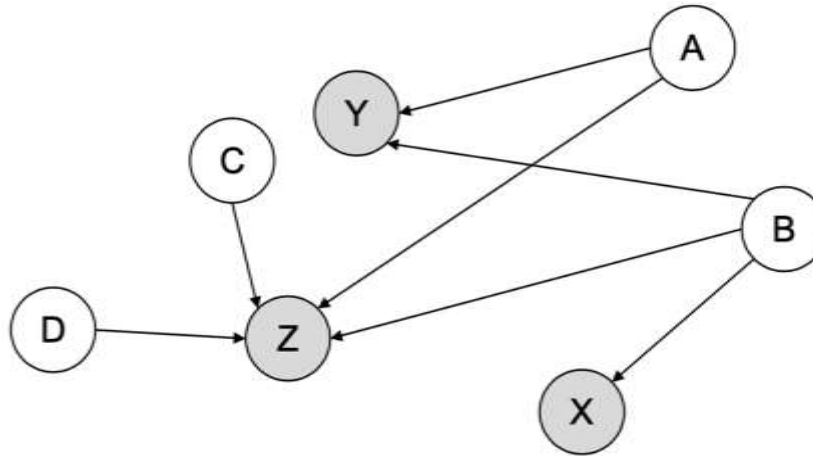
D: Smaakverlies

Diagnose:

x: Verkoudheid

y: Griep

z: Covid





# Automatische foutdiagnose: Bayesiaans netwerk

Diagnose ziekten

Bijvoorbeeld:

A: Koorts

B: Hoesten

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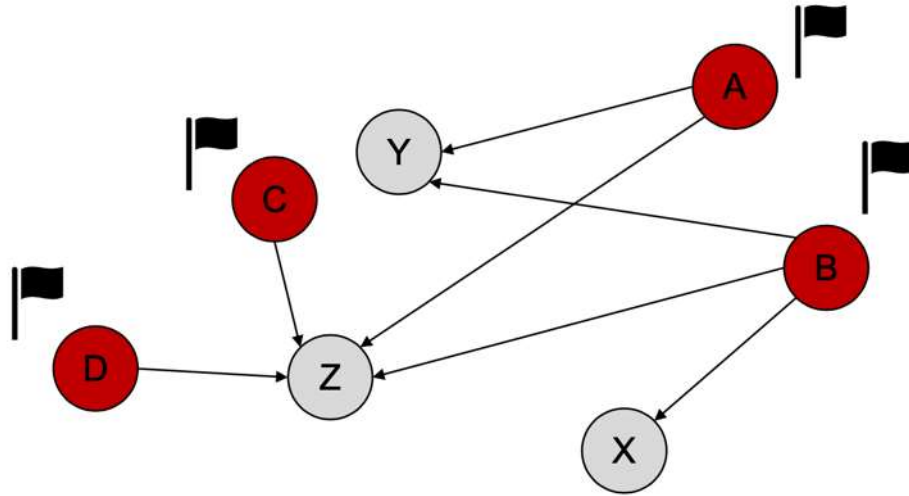
D: Smaakverlies

Diagnose:

x: Verkoudheid

y: Griep

z: Covid



Koorts + Hoesten + Vermoeidheid + Smaakverlies → COVID



# Automatische foutdiagnose: Bayesiaans netwerk

## Voorbeeld luchtbehandelingskast:

A: Status LBK aan/uit

B: Aandrijfriem gebroken

C: Filter gescheurd

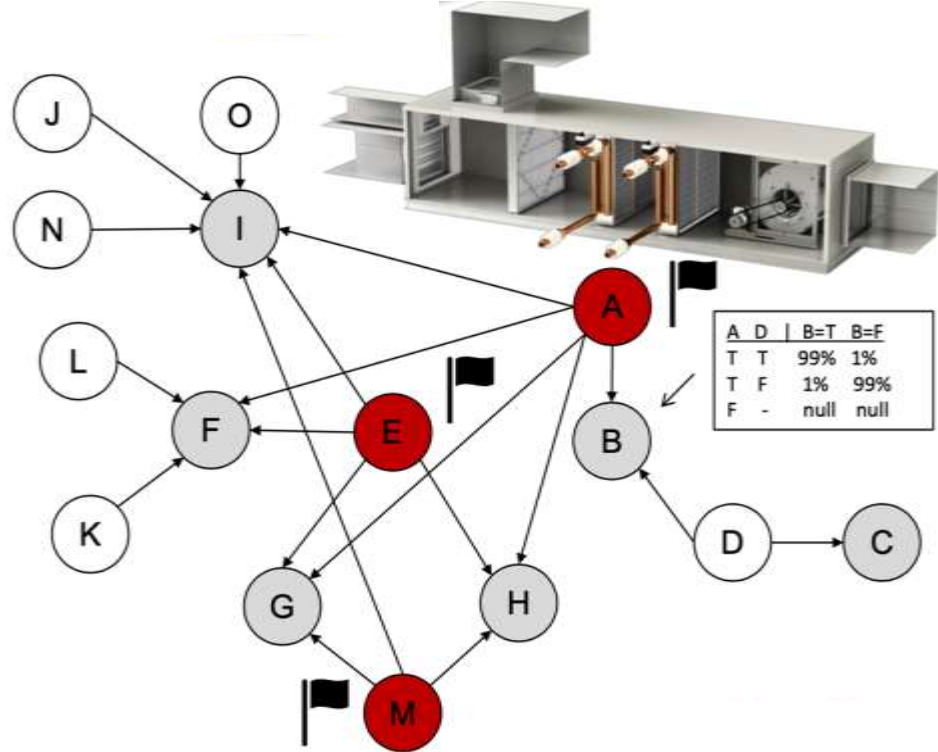
D: Druk over filter = 0

E: Te weinig capaciteit

F: Hittegolf

G: Vastzittende koelklep

H: Veel bezoekers



# Ingrediënten

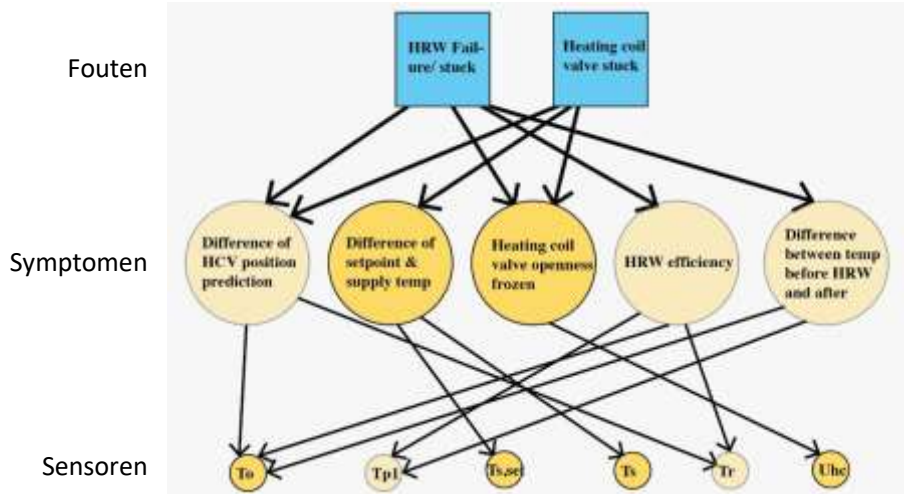
- Sensor-technologie
- Data-technologie
- Cloud: o.a. t.b.v. Machine Learning
- Dashboards: Interactief, KPI's.



# Meten is weten ...

# DBN met diverse sensor-configuraties

- Living Lab: LBK in gebouw Kropman, Breda
- Voorbeeld: Foutdiagnose verwarming in LBK



| Case | Fault         | Limited Sensor Configuration |                  | Standard/Redundant Sensor Configuration |                  |
|------|---------------|------------------------------|------------------|-----------------------------------------|------------------|
|      |               | Post Probability             | Diagnosis Result | Post Probability                        | Diagnosis Result |
| 1    | HRW 80% stuck | 0%                           | Wrong            | 99%                                     | Correct          |
| 2    | HRW 50% stuck | 0%                           | Wrong            | 89%                                     | Correct          |
| 3    | HRW 30% stuck | 0%                           | Wrong            | 76%                                     | Correct          |
| 4    | HRW 10% stuck | 0%                           | Wrong            | 67%                                     | Correct          |
| 5    | HRW failure   | 0%                           | Wrong            | 65%                                     | Correct          |
| 6    | HCV 40%       | 67%                          | Correct          | 76%                                     | Correct          |
| 7    | HCV 70%       | 75%                          | Correct          | 67%                                     | Correct          |
| 8    | Normal        | 100%                         | Correct          | 100%                                    | Correct          |

(Z. Wang, C. Lu, L. Itard, 2023, TUD)

# LBK sensor-configuratie: Situatie in Nederland

- ASHRAE (VS): Basiseisen.
- ISSO (NL): Uitgebreide aanbevelingen.

| Sensor                                   | A | I | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12  | 13 | 14 | 15  | 16  | 17  | 18 |
|------------------------------------------|---|---|---|---|---|---|---|---|---|---|---|----|----|-----|----|----|-----|-----|-----|----|
| Outdoor air temperature                  | ✓ | ✓ | ✓ |   |   |   | ✓ |   |   | ✓ |   |    | ✓  | ✓   |    | ✓  | ✓   | ✓   | ✓   | ✓  |
| Outdoor air relative humidity            |   | ✓ |   |   |   |   |   |   |   |   |   |    |    |     |    |    | ✓   | ✓   | ✓   | ✓  |
| Preheated air temperature                | ✓ | ✓ |   |   | ✓ |   |   |   |   |   |   |    |    | ✓   |    |    | ✓   | ✓   | ✓   | ✓  |
| Preheated air relative humidity          |   | ✓ |   |   |   |   |   |   |   |   |   |    |    | ✓   |    |    | ✓   | ✓   | ✓   | ✓  |
| Supply air temperature                   | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |    | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Supply air relative humidity             |   | ✓ |   |   | ✓ | ✓ | ✓ |   | ✓ | ✓ |   | ✓  | ✓  |     |    | ✓  | ✓   | ✓   | ✓   | ✓  |
| Return air relative humidity             |   | ✓ |   |   | ✓ | ✓ | ✓ |   | ✓ | ✓ |   | ✓  | ✓  | ✓   |    | ✓  | ✓   | ✓   | ✓   | ✓  |
| Exhaust air temperature                  | ✓ | ✓ |   |   |   | ✓ | ✓ |   |   |   |   |    |    | ✓   |    | ✓  | ✓   | ✓   | ✓   | ✓  |
| Exhaust air relative humidity            |   | ✓ |   |   |   |   |   |   |   |   |   |    |    | ✓   |    |    | ✓   | ✓   | ✓   | ✓  |
| Supply water temperature                 | ✓ | ✓ |   |   |   |   |   |   |   |   |   |    |    | ✓   | ✓  |    | ✓   | ✓   | ✓   | ✓  |
| Return water temperature                 | ✓ | ✓ |   |   |   |   | ✓ |   |   | ✓ | ✓ | ✓  | ✓  | ✓   |    |    | ✓   | ✓   | ✓   | ✓  |
| Pressure difference at supply air filter | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Pressure difference at return air filter | ✓ | ✓ |   |   | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ |   | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Pressure difference at supply air on fan |   | ✓ |   |   | ✓ |   |   |   |   |   |   | ✓  | ✓  |     | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Pressure difference at return air on fan |   | ✓ |   |   |   |   |   |   |   |   | ✓ | ✓  |    |     |    |    | ✓   | ✓   | ✓   | ✓  |
| Supply air flow rate                     | ✓ | ✓ |   | ✓ |   |   |   |   |   |   |   |    |    |     |    |    | ✓   |     |     |    |
| Return air flow rate                     | ✓ | ✓ |   |   |   |   |   |   |   |   |   |    |    |     |    |    |     |     |     |    |
| Coil valve control signal                | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Supply fan control signal                | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Return fan control signal                | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Supply damper control signal             |   | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Return damper control signal             |   | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Air quality sensor                       |   | ✓ |   |   |   | ✓ |   | ✓ |   | ✓ |   |    |    |     | ✓  |    | ✓   |     |     |    |
| Temperature sensor after coil            | ✓ | ✓ |   |   |   |   | ✓ | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Coil water flow                          |   | ✓ |   |   |   |   |   | ✓ | ✓ | ✓ | ✓ | ✓  | ✓  | ✓   | ✓  | ✓  | ✓   | ✓   | ✓   | ✓  |
| Fit ASHRAE recommendation                | - | - |   |   |   |   |   |   |   |   |   |    |    | Yes |    |    | Yes | Yes | Yes |    |

Slechts 4 LBK's conform ASHRAE standaard!

Definitie van sensor-configuraties:

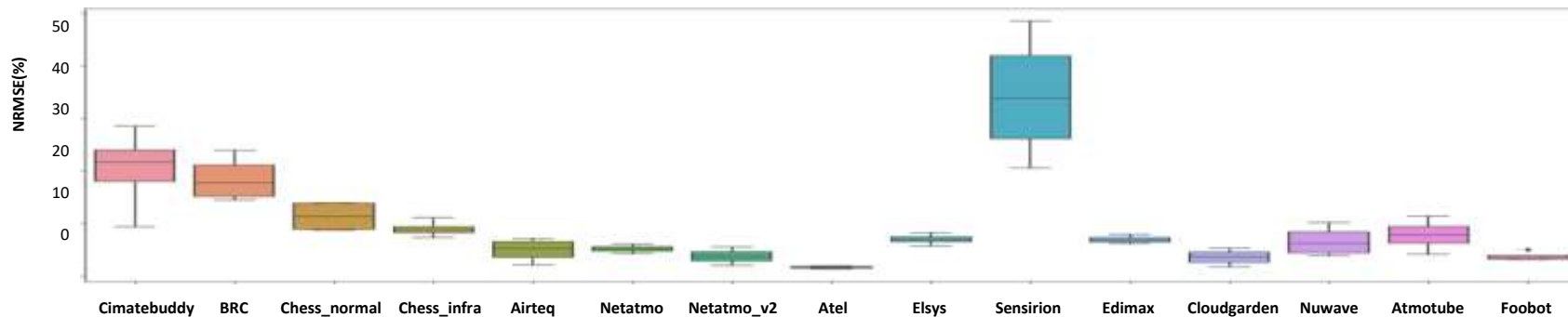
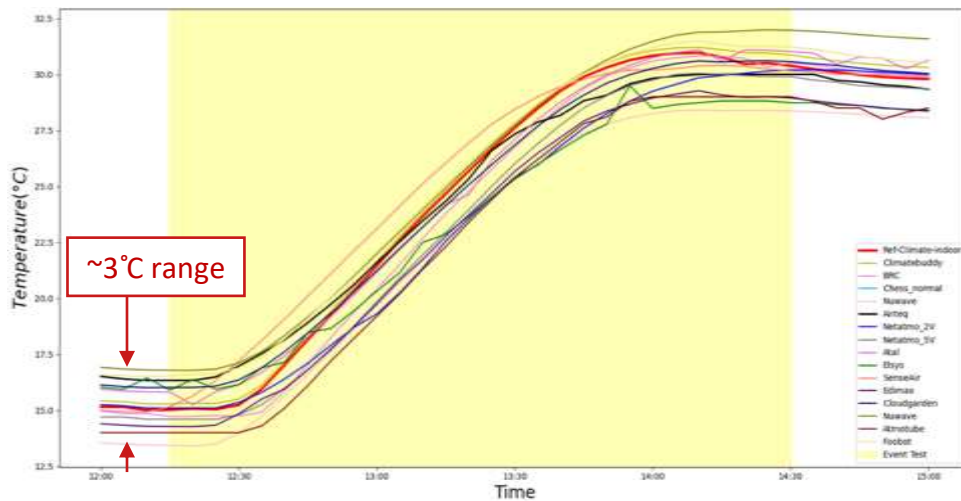
- ❑ **Limited:** LBK's niet conform ASHRAE standaard
- ❑ **Standard:** LBK's conform ASHRAE standaard
- ❑ **Redundant:** LBK's met meer sensoren dan ASHRAE

(Z. Wang, C. Lu, L. Itard, 2023, TUD)

# Smart Buildings: Sensoren

- Low-cost monitors
- Virtuele (soft) sensoren

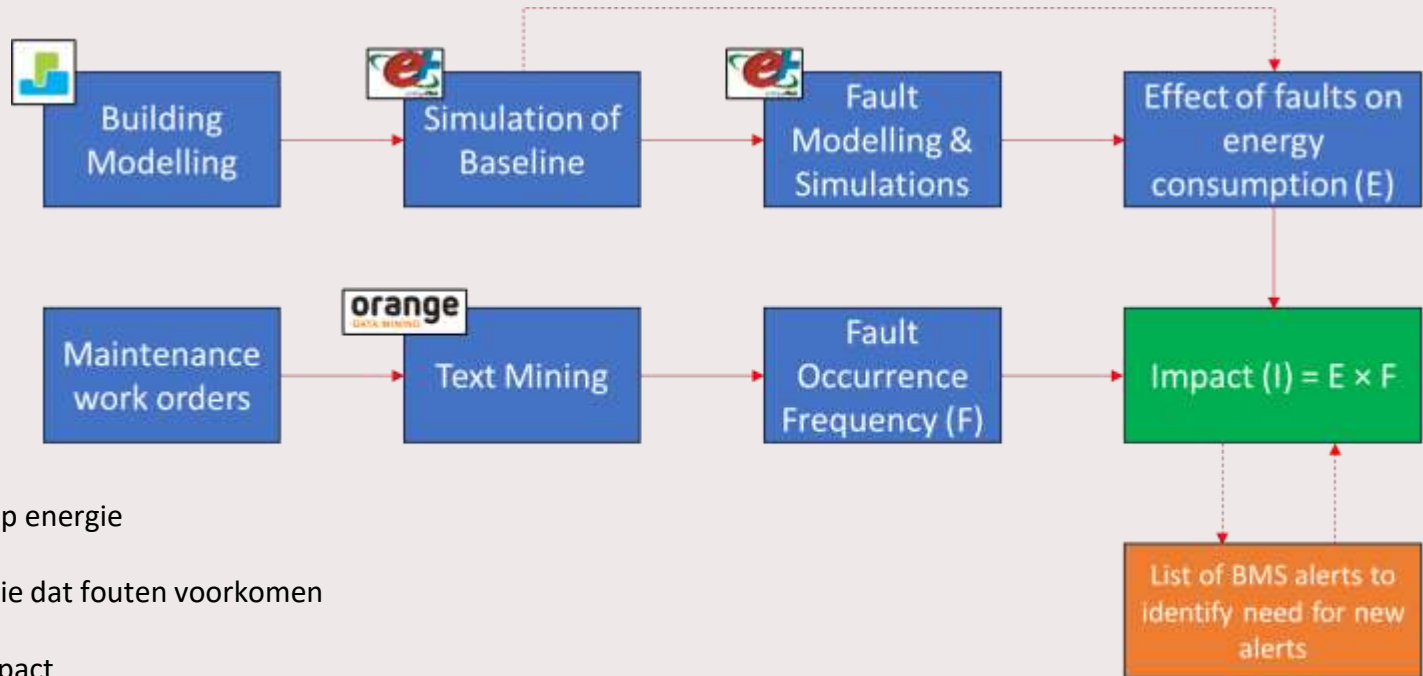






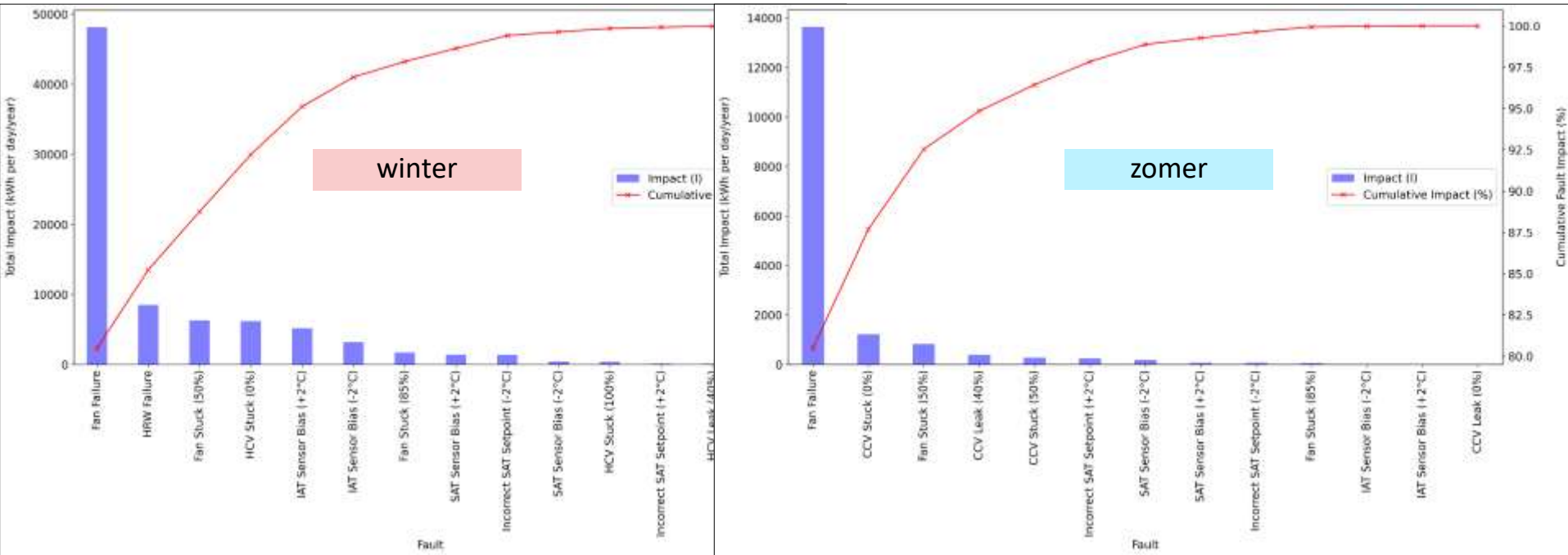
# Van data naar symptoomdetectie

# Fouten prioriteren



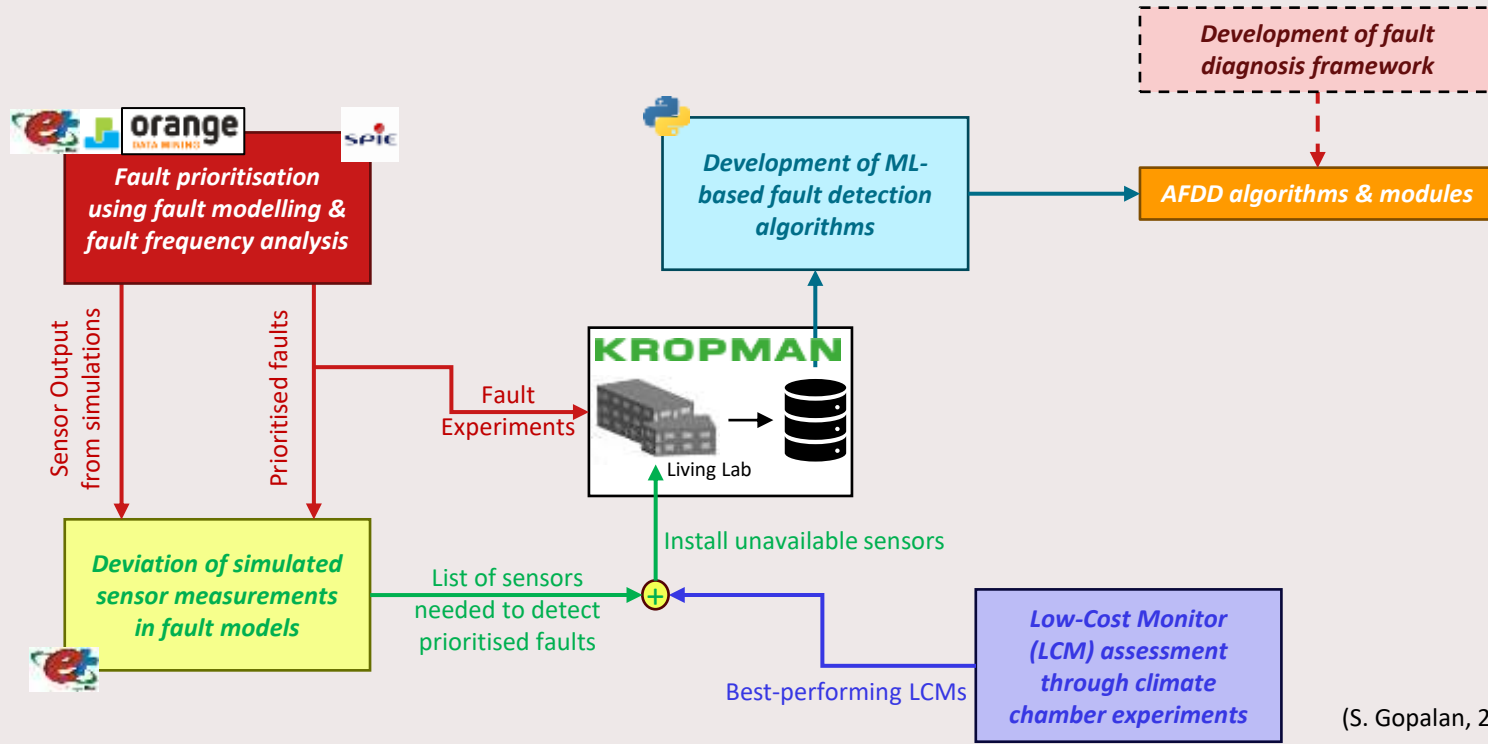
- $I = E \times F$
- 'E' = impact op energie
- 'F' = frequentie dat fouten voorkomen
- 'I' = totale impact

# Fouten prioriteren



(S. Gopalan, 2024, TUE)

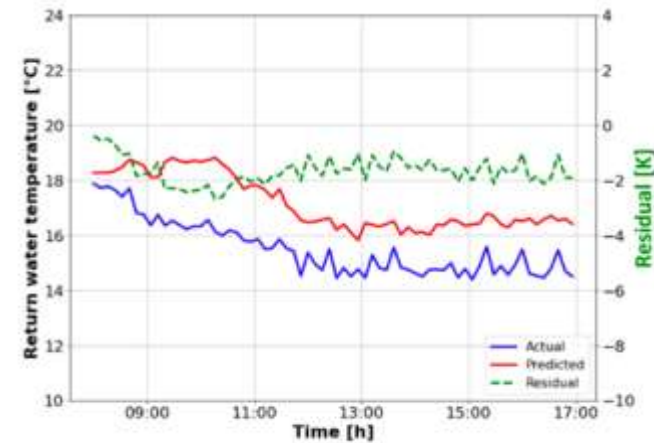
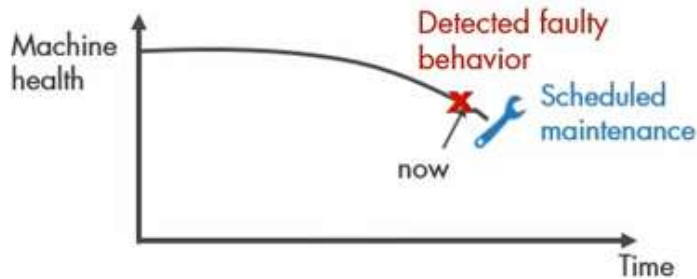
# Foutdetectie: Van simulatie naar de praktijk



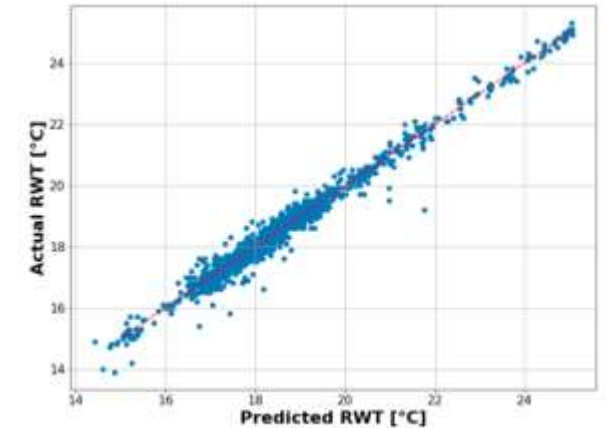
(S. Gopalan, 2024, TUE)

# Foutdetectie: Volgende stappen

- Vergelijken ML-modellen: XG Boost, SVR, ANN.
- Off-line testen op data set Living Labs.
- Koppelen aan diagnose modules.
- Conditie-afhankelijk onderhoud.



*Fault-free prediction (9 hours) using ML model*



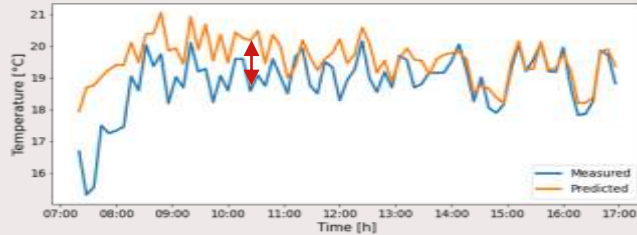
*Performance evaluation of ML model (3 months fault-free data)*

# Van symptoomdetectie naar foutdiagnose

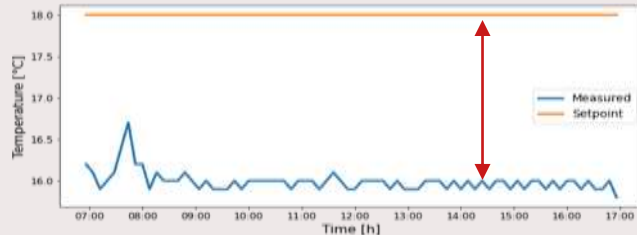
# Automatische foutdetectie én foutdiagnose

## Machine Learning: symptoom-DETECTIE

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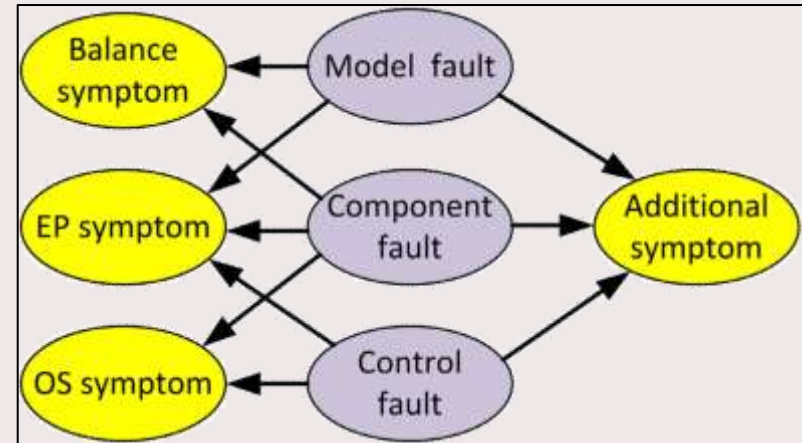


Rules-Based  
Fault  
Detection



(A. Thamban, 2021, TUE)

## Bayesiaans Netwerk: fout-DIAGNOSE



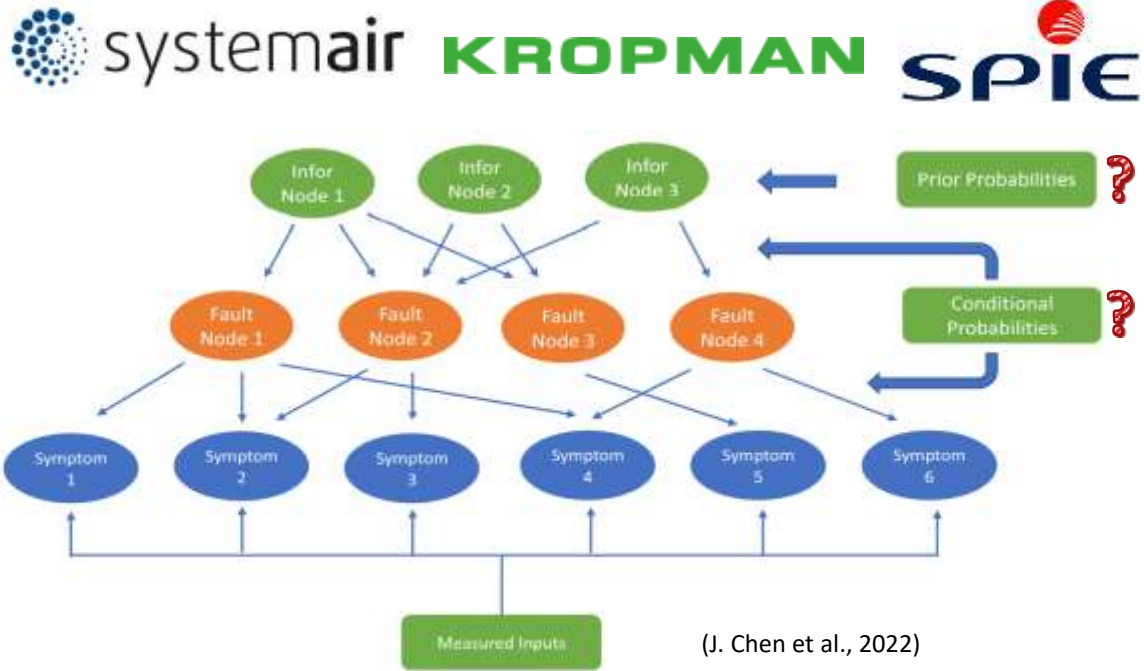
(A. Taal, 2021, TUE)



# Data-gedreven en op kennis gebaseerde DBN's

Interviews met experts:

- DBN structuur
- Prior probabilities
- Conditional probabilities



# Gebouwgebruiker centraal: Subjectieve data

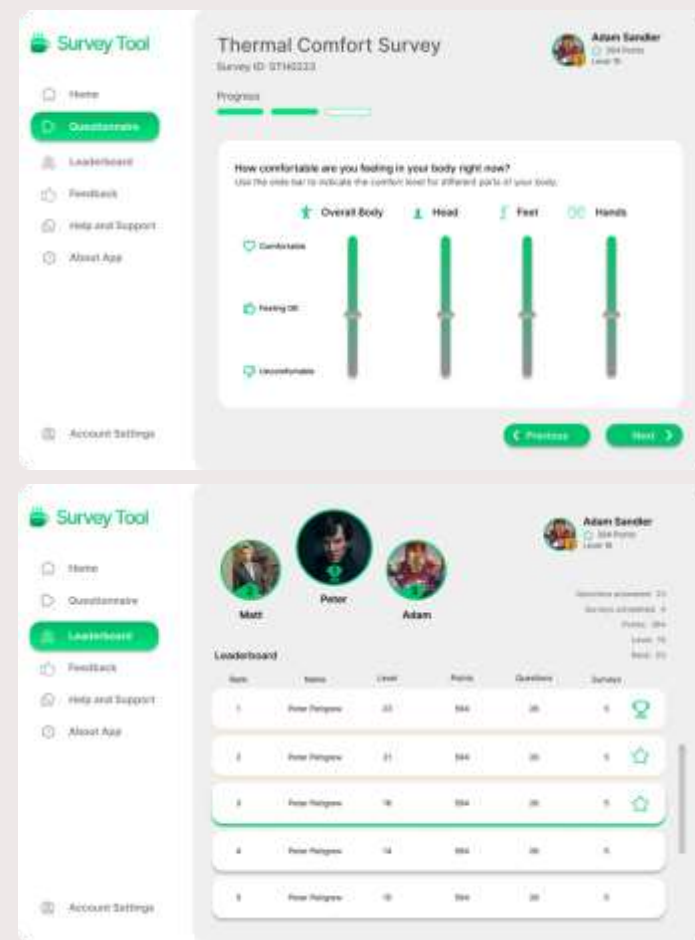
- Recent veldonderzoek met 3892 pers. in 60 kantoorgebouwen in de VS: Ongeveer 40% ontevreden over thermische omgeving, ver boven doelstelling van 20% (ASHRAE 55). (L. Pérez-Lombard, J. Ortiz, C. Pout- 2008, Z. Wang, B. Lin, Y. Zhu - 2015, Y. Long, S. Liu, L. Xie, K.H. Johansson – 2014, Y. Long, S. Liu, L. Xie, K.H. Johansson-2016)
- Integreren gebruikersfeedback kan thermisch comfort verbeteren én energie besparen met wel 20%-40% in kantooromgevingen. (Y. Murakami, M. Terano, K. Mizutani, M. Harada, S. Kuno – 2007, Jazizadeh Farrokh, Ghahramani Ali, Becerik-Gerber Burcin, Kichkaylo Tatiana, Orosz Michael – 2014)

# Survey tool

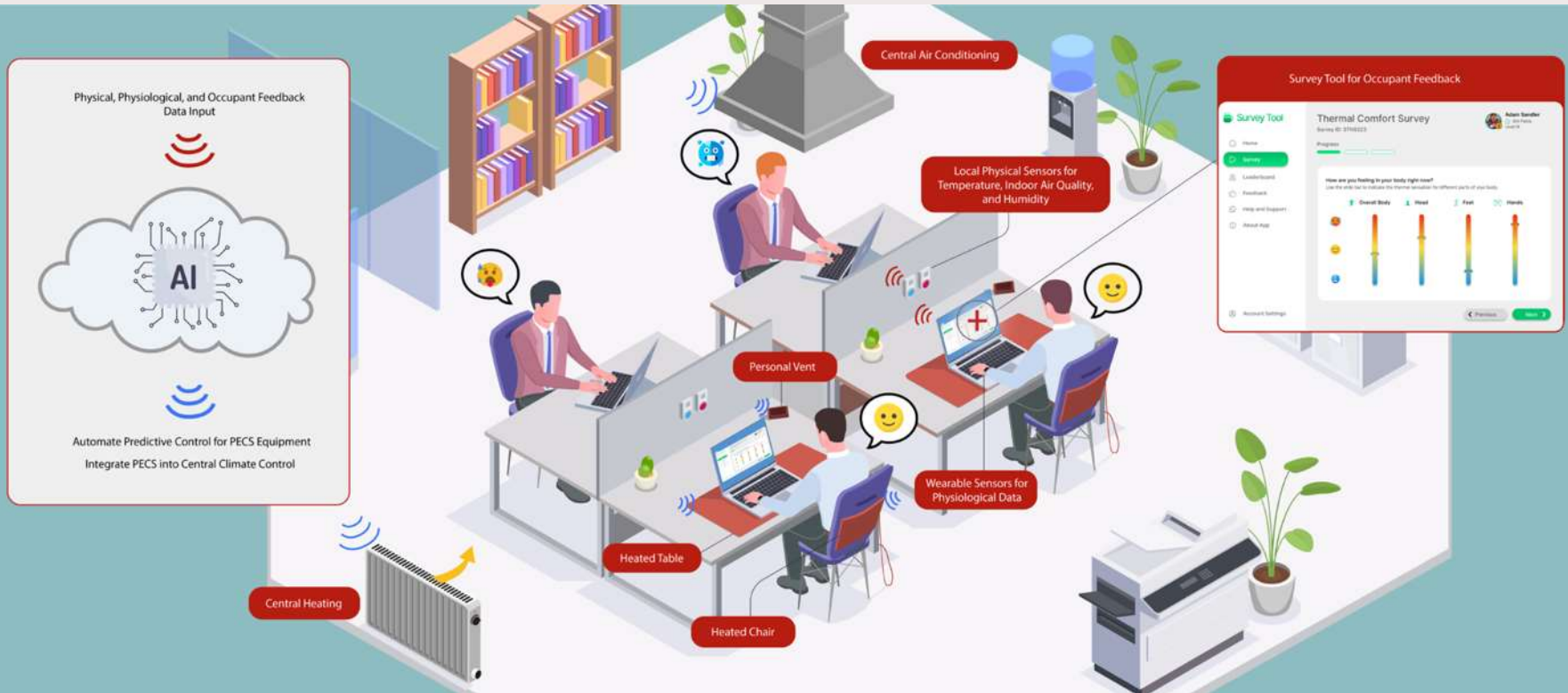
## Engageren van gebouwgebruikers

- Responsieve vragenlijsten: Gericht vragen.
- Automatische notificaties: Alleen wanneer nodig.
- Gamification: Verbeterde betrokkenheid.

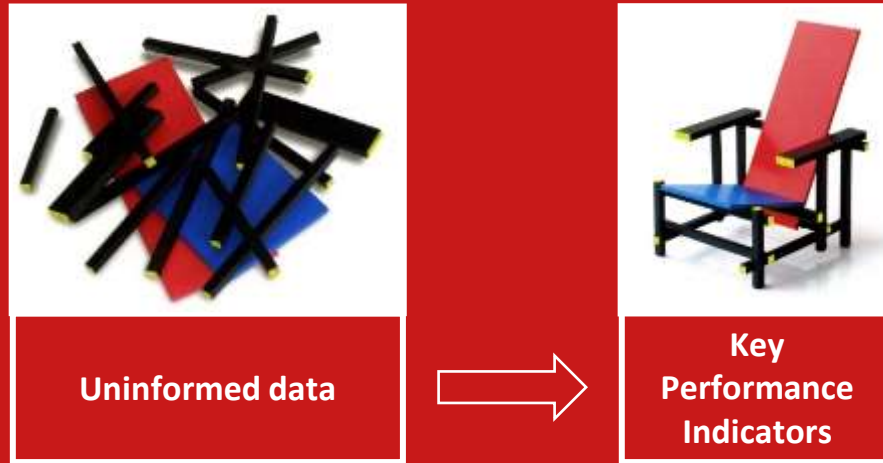
(S. Atulugama, EngD-kandidaat, 2024, TUE)



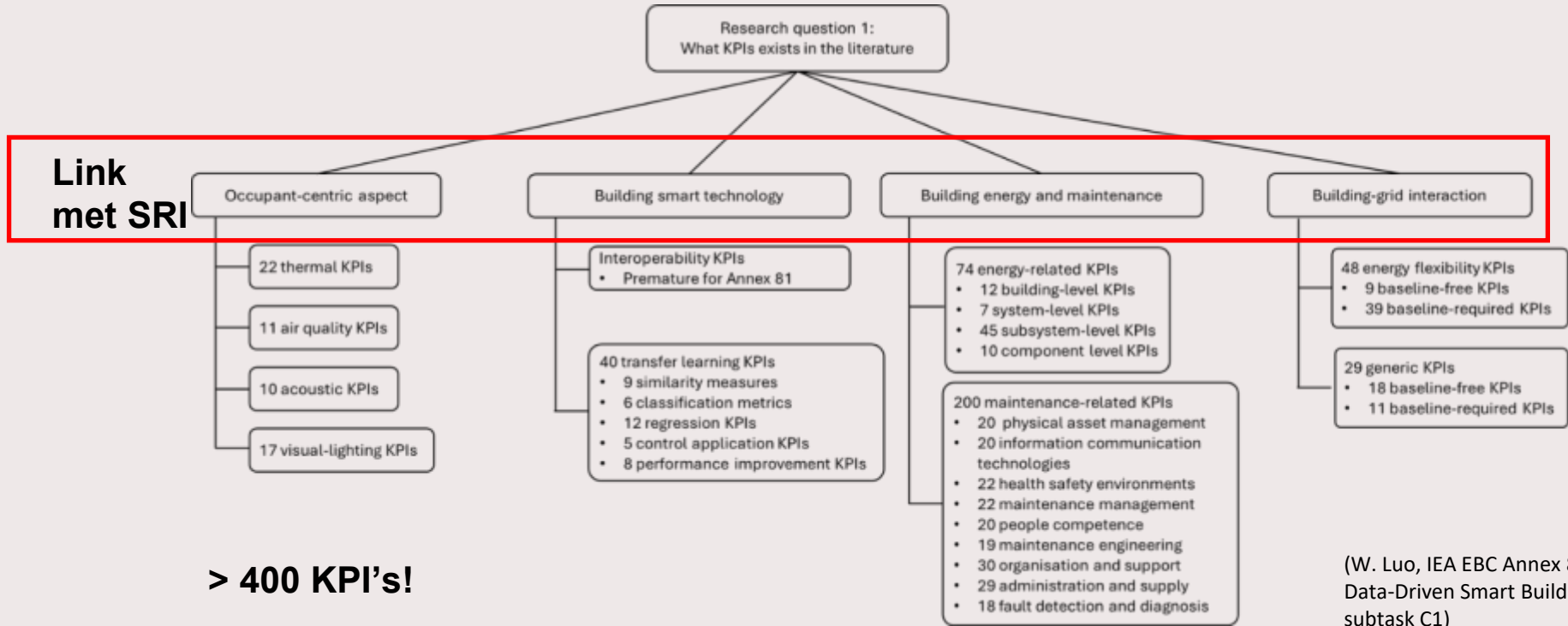
# Vooruitzicht



# Van data naar KPI's: Voor wie en wanneer?



# Bestaande KPI's: Literatuurstudie



(W. Luo, IEA EBC Annex 81 - Data-Driven Smart Buildings, subtask C1)

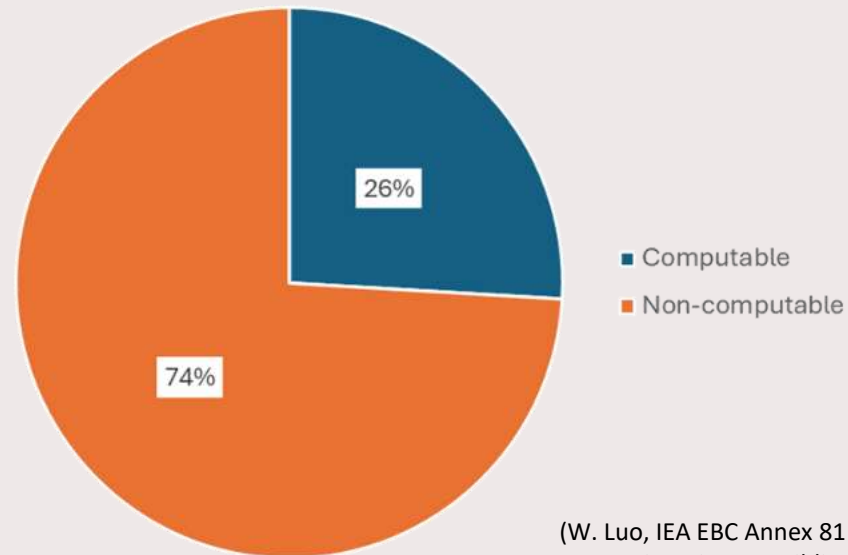
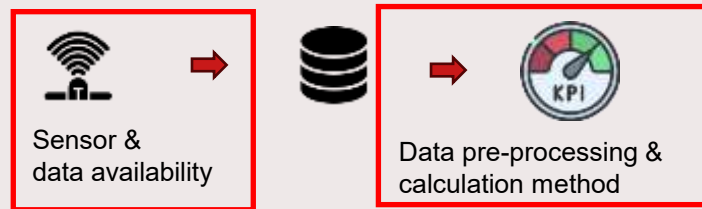
# Welke KPI's kunnen bepaald worden?

Gebaseerd op GBS data

Gebouwen:

- 4 in Nederland
- 1 in Zwitserland
- 2 in Denemarken
- 7 in Noorwegen

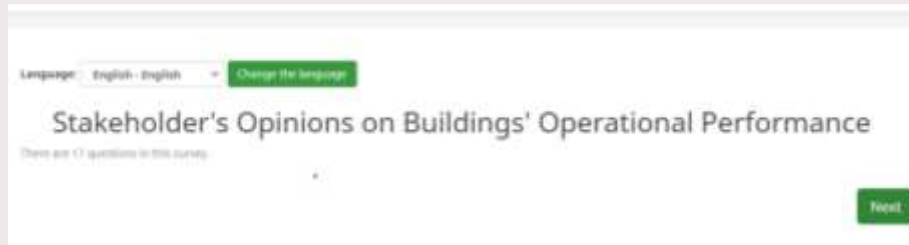
Uitdagingen:



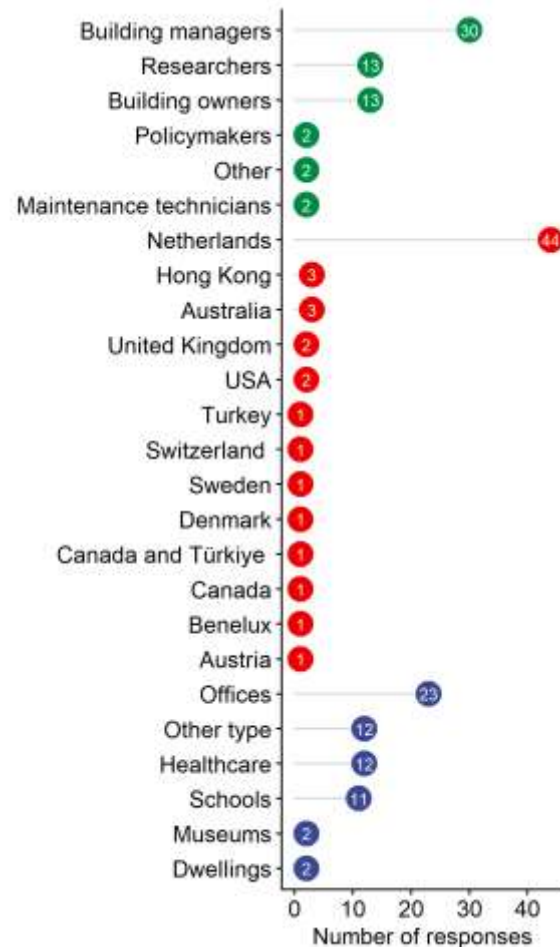
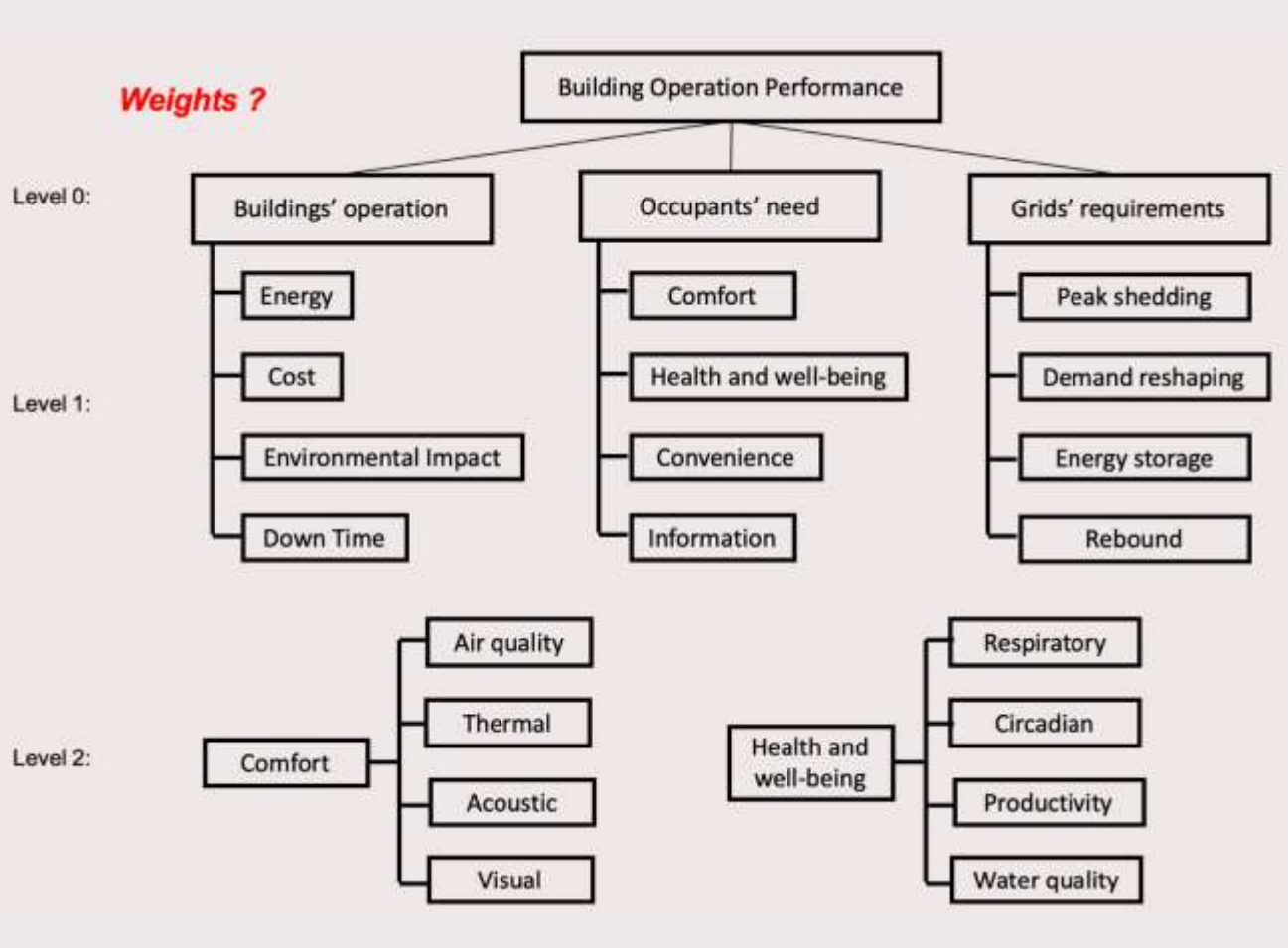
(W. Luo, IEA EBC Annex 81 - Data-Driven Smart Buildings, subtask C1)

# Welke (meta)KPI's zijn belangrijk en voor wie?

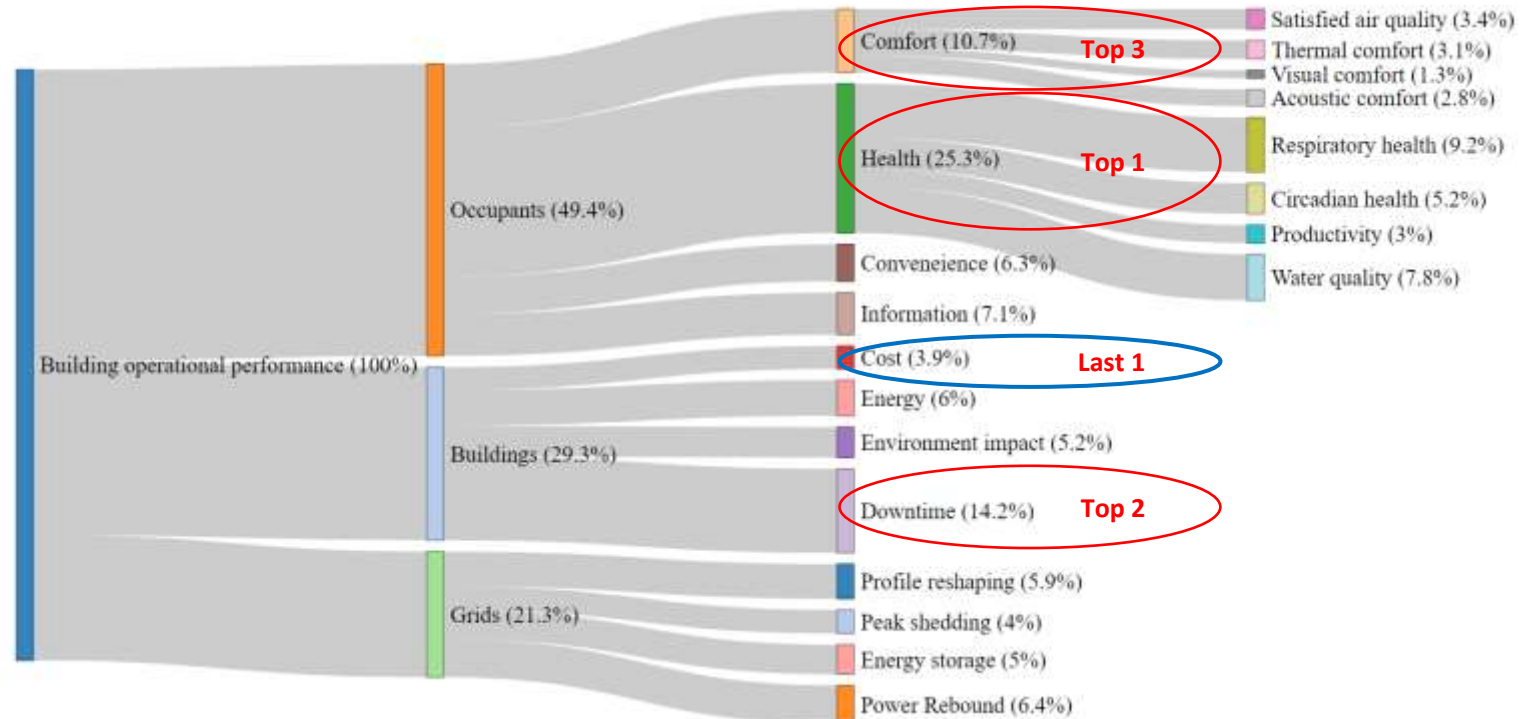
Vragenlijst uitgezet door Building G100:



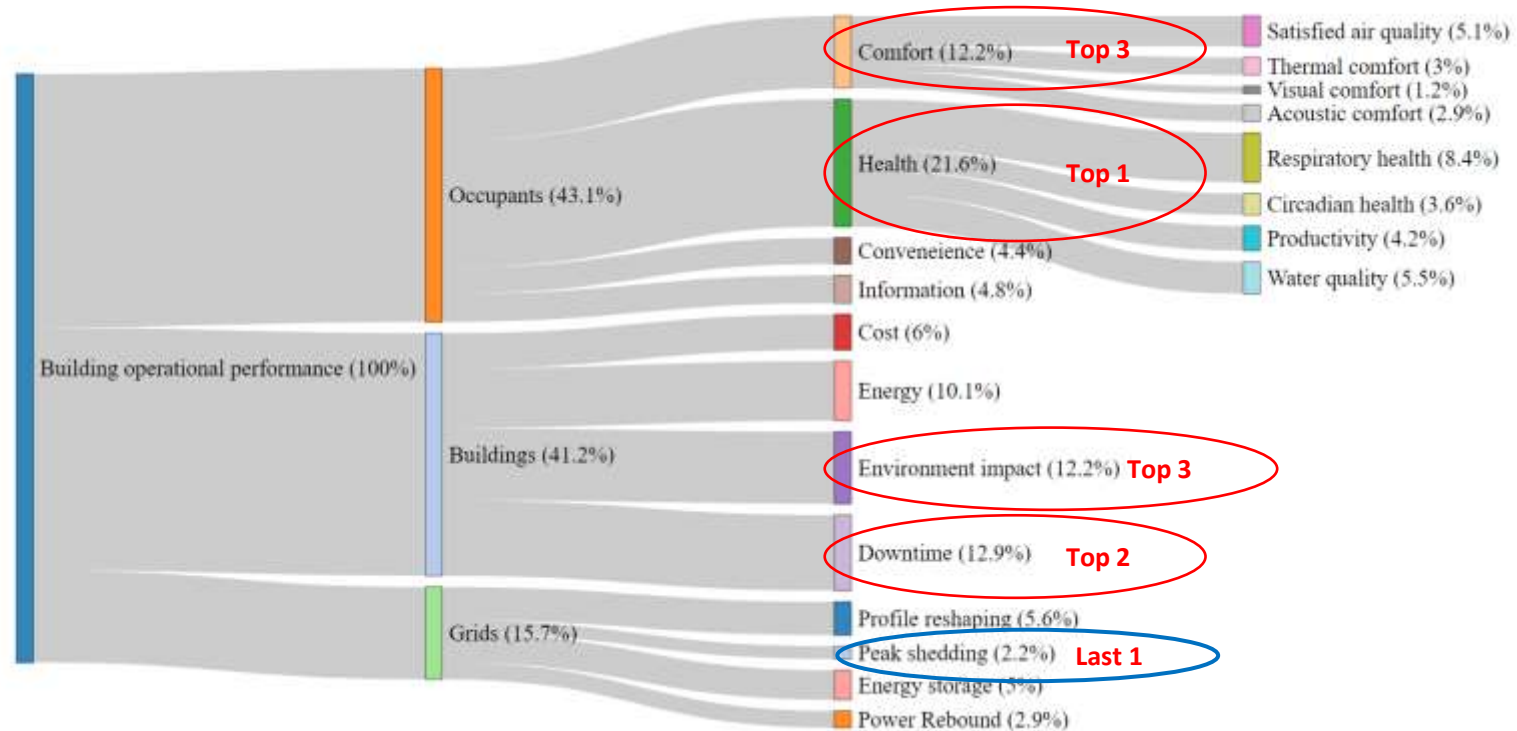




# Resultaten gebouwbeheerders (N=27)



# Resultaten gebouw eigenaren (N=13)



# Afsluitend

Partners: Kennisinstellingen, installatiebedrijven, adviesbureaus, platform & interface-ontwikkelaars, gebouweigenaren en -beheerders, leveranciers en branche-organisaties.



# Ter discussie

- Data maakt veel mogelijk, maar gerichte aanpak is essentieel:
  - Wat moeten we meten en waar?
  - Hoe halen we informatie uit data?
  - Hoe presenteren we die informatie: Voor wie?
- Machine Learning maakt veel mogelijk, maar:
  - Hoe borgen we interpreteerbaarheid en schaalbaarheid?
  - Hoe sluiten we aan met (beroeps)opleidingen?
  - Hoe sluiten marktpartijen aan? Nieuwe business modellen?



# Data- en expertgedreven aanpak voor efficiënte én effectieve installaties: Van onderzoek naar maatschappelijke impact

