

AIVC Workshop "Climate Change, Ventilation and Resilience", Madrid, 2026

Power outage profiles generation to assess thermal and indoor air quality resilience

Peihang An (安沛航), KU Leuven



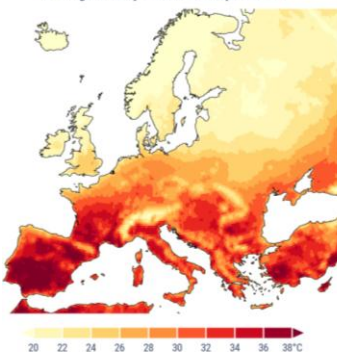
1

BACKGROUND

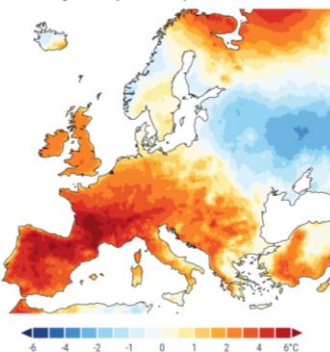
Surface air temperature during the August 2025 heatwave (8–18 August)

Data: ERA5 • Reference period: 1991–2020 • Credit: C3S/ECMWF

Average of daily maximum temperatures



Average of daily mean temperature anomalies



Heatwaves(HW) in Europe have become more frequent and intense:

- Europe came across multiple HW events from **April** to **September** 2025.
- Maximum temperature in both Portugal and Spain broke **46 °C** [1].
- The 2025 European heatwaves caused over **24,000** heat-related deaths [2].

[1] <https://www.theguardian.com/environment/2025/jun/30/spain-records-high-of-46c-and-france-under-alert-as-europe-swelters-in-heatwave>

[2] <https://www.euronews.com/2025/09/17/alert-killer-climate-change-led-to-16500-more-heat-deaths-in-europe-this-summer-study-say>

2

BACKGROUND



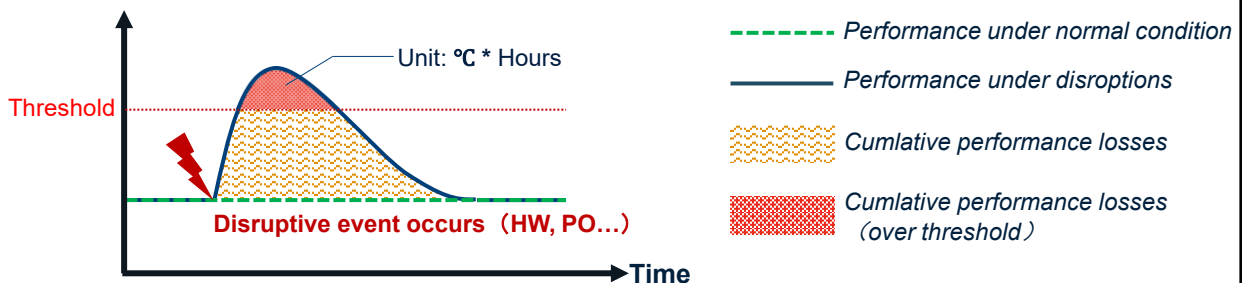
- AC adoption in Europe is estimated to rise from 20% today to **35%** by 2030 [1], with total units expected to more than **double** by 2050 [2]. Over **40%** of households have AC in Spain [3].
- PO is more frequent and more challenging than expected. Belgium is estimated to have over **3,000** individual PO events last summer based on PO data provided by local grid operator

3 [1] https://energy-efficient-products.ec.europa.eu/product-list/air-conditioners-and-comfort-fans_en
 [2] <https://www.ica.org/reports/the-future-of-cooling>
 [3] <https://www.idealista.com/en/news/property-for-sale-in-spain-2024/07/30/818081-41-of-the-housing-stock-in-spain-has-air-conditioning>

BACKGROUND

In general, **resilience** refers to a system's capacity to absorb disturbances, sustain its critical functions under stress, and recover efficiently from disruption with acceptable losses in performance, time, cost, and risk.

Performance indicator (e.g. indoor temperature)



GAP & AIM

- There is no methodology to quantify PO events in current studies. e.g. some studies on resilience:
 - Sheng et al. [1] assumed the PO to have the **exact start and end time** as the extreme weather (duration: over **multiple days**)
 - Sengupta et al. [2] assumed a **24-hour** PO to occur on the exact **hottest day**.
 - Al-Assaad et al. [3] assumed a **1-hour** PO during HW referring to historical PO data.
 - ...

**Worst cast scenario lack of data-basis;
Limited use of the PO data(lack of probability estimation)**

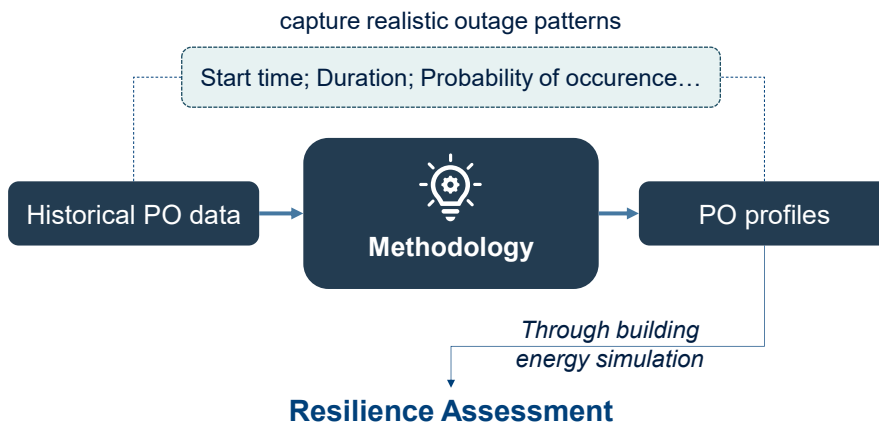
[1] M. Sheng, M. Reiner, K. Sun, and T. Hong, Assessing thermal resilience of an assisted living facility during heat waves and cold snaps with power outages

[2] A. Sengupta et al. Evaluation of thermal resilience to overheating for an educational building in future heatwave scenarios.

[3] D. Al Assaad et al. Demand-controlled ventilation in educational buildings: Energy efficient but is it resilient

GAP & AIM

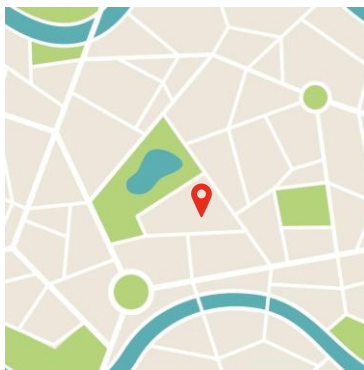
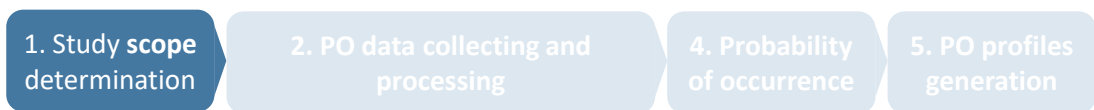
- **AIM:** To develop a methodology that generates PO profiles from historical data.



METHODOLOGY



METHODOLOGY



Target building

The building that we would like to study its resilience under POs

The “scope ”is geographical area covering the target building, referred to as the **spatial boundary (SB)**

METHODOLOGY

1. Study scope determination


2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation



 Target building

 Spatial boundary(SB)

Extreme example: the SB include **only** the Target building

SB is too small :

Local PO data is more representative for the target building. However, require **more data in temporal dimension** to create the profiles.

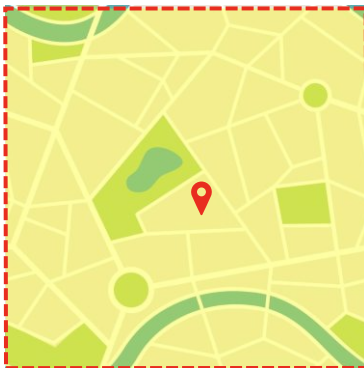
METHODOLOGY

1. Study scope determination


2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation



 Target building

 Spatial boundary(SB)

PO occurs in another city

SB is too big :

Local PO data has higher variety in spatial distribution and is **less representative for target building**.

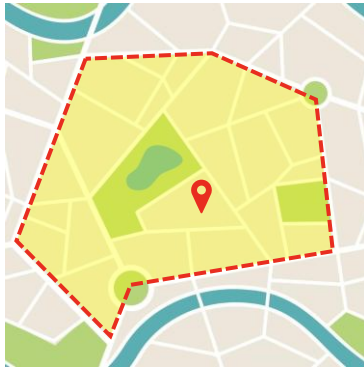
METHODOLOGY

1. Study scope determination

2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation



Target building

Spatial boundary(SB)

Rule of Thumb: Same grid, no bigger than one administrative region

SB should :

be set at an appropriate size to **represent the local situation** around the target building, also comply with the **data availability**.

METHODOLOGY

1. Study scope determination

2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation

Example of the PO dataset

Start date (yyyy-mm-dd)	Start time (hh-mm)	Duration (hours)	Impact users ($n_{impacted\ users}$)	Location
2024-06-17	08:30	2.5	1,500	xx street...
2024-06-18	07:45	3.2	4,800	xx street...
...

Within **SB** or not ?

- Remove the entries occurs outside the SB



METHODOLOGY

1. Study scope determination

2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation

Example of the PO dataset

Start date (yyyy-mm-dd)	Start time (hh-mm)	Duration (hours)	Impact users ($n_{impacted\ users}$)	P_{PO}
2024-06-17	08:30	2.5	1,500	0.005
2024-06-18	07:45	3.2	4,800	0.016
...

e.g. There are **300,000** electricity users inside SB; a single PO impacted **1,500** users; we can assume there is a probability of $1,500/300,000 = 0.005$, that this PO impact the target building

METHODOLOGY

1. Study scope determination

2. PO data collecting and processing

4. Probability of occurrence

5. PO profiles generation

PO profile generation through **Monte Carlo Sampling (MCS)**

Start date	Start time	Duration	Impact users	P_{PO}	$Rand$	$\leq P_{PO} ?$
				0.005	0.241	FALSE
				0.016	0.012	TRUE
				

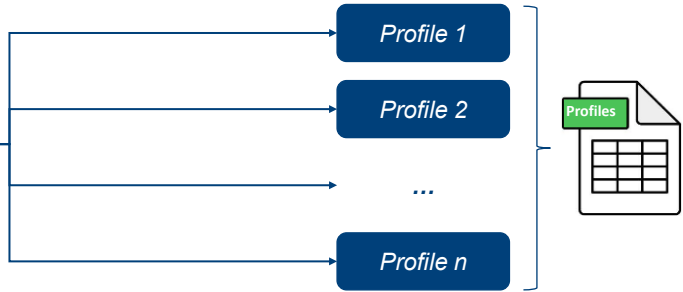
- A column of **random** rational numbers between 0 and 1 is generated;
- If the $Rand \leq P_{PO}$, this PO is considered to **occur and affect the target building**
- All the PO entries labeled **"TRUE"** are selected to combine a **"Profile"**

METHODOLOGY



Processed PO dataset

Monte Carlo Sampling



PO dataset:

- Actual collected historical PO records;
- **One timeline only**;
- The timeline records all POs within SB.

PO profiles:

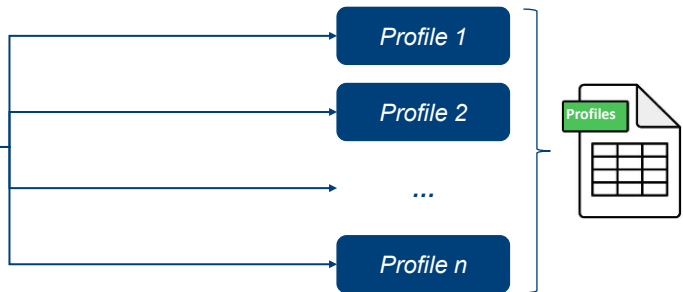
- Generated based on PO dataset;
- Each profile is an **individual possible scenario**, with an **individual timeline**.
- In each profile, the timeline only contains the PO that impact the **Target building**

METHODOLOGY



Processed PO dataset

Monte Carlo Sampling



When to stop ?



Consider the computational cost (not only for MCS, but also for Building Simulation)

CASE STUDY

Ghent, capital city of the East Flanders, Belgium

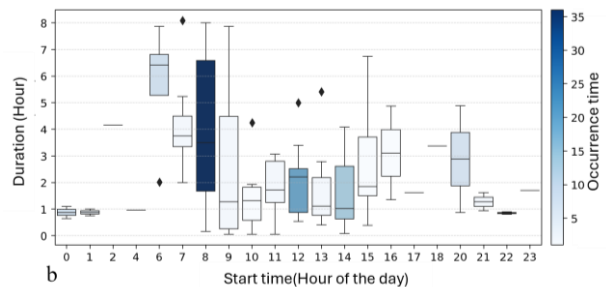
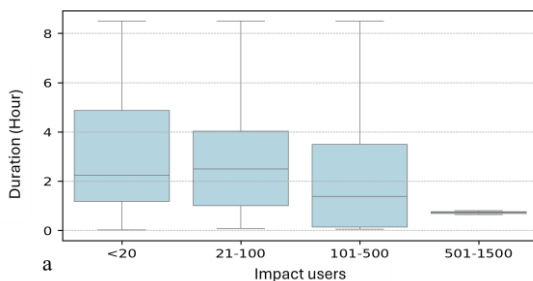
Lecture room in KU Leuven, Ghent campus



- **Fluvius** is the sole grid network operator in Ghent, holding a 100% market share (legal monopoly in Flanders).
- **Fluvius** provides daily updated PO data.

CASE STUDY

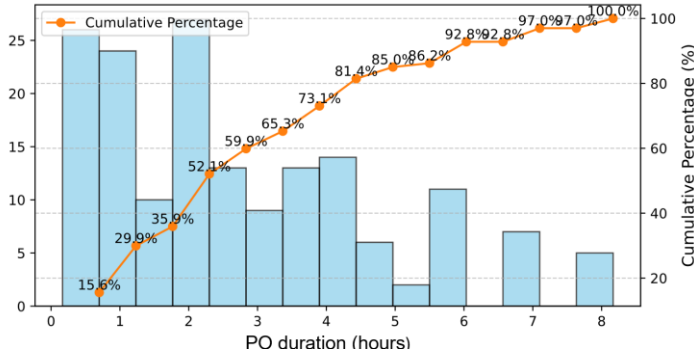
- PO **dataset** is manually collected from **Fluvius** on daily basis (on going)
- Data from **June 2024 to October 2024** is used to generate the PO profiles



Collected historical PO data distribution on (a) Duration and Impact users, (b) Duration and start time

CASE STUDY

- 2,000 profiles were generated, 200 of them contain at least one PO.
- 163 profiles were chosen for these POs overlaps the HVAC schedule.

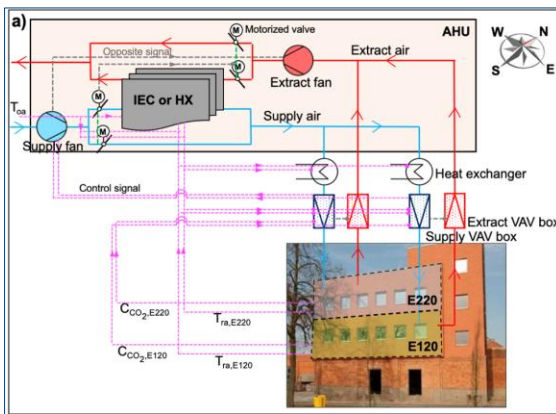


Histogram of the selected PO profiles (total 163)

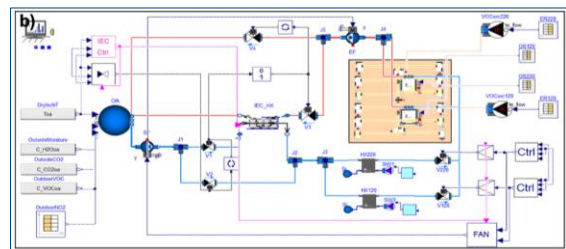
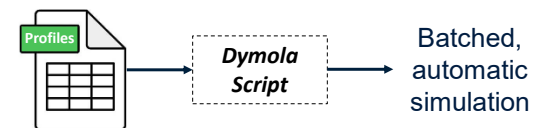
- POs with duration **less than 3 hours** takes up **60%**;
- POs with duration **less than 6 hours** takes up **92%**;
- PO that last **over 8 hours** is **extremely rare** based on the collected data.

CASE STUDY

- The dymola model is validated by previous works [1].

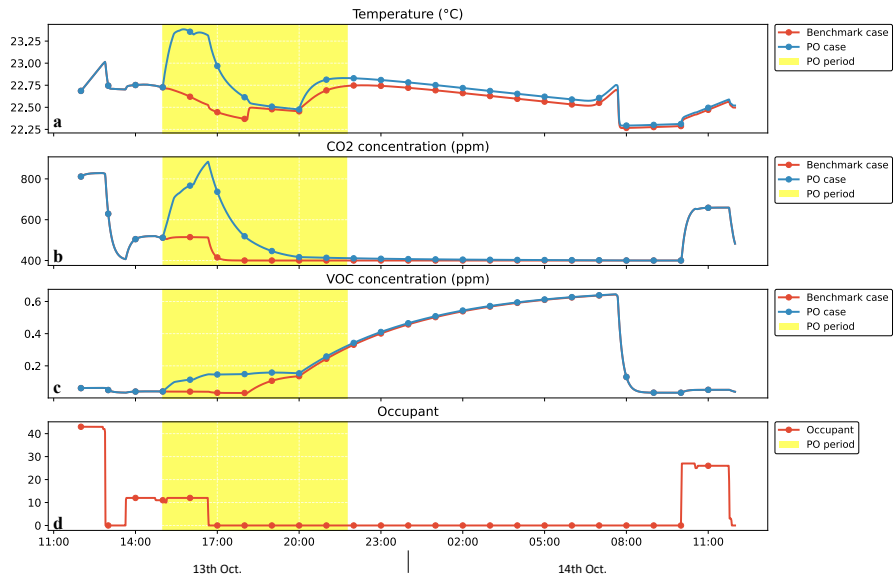


Schematic of the test lecture room building consisting of classrooms E120, E220 and the associated AHU



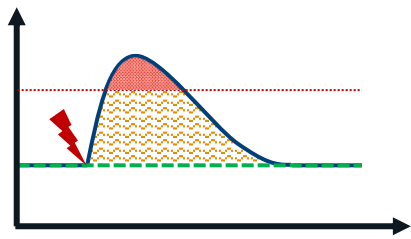
Associated Dymola model for the case study

RESULT



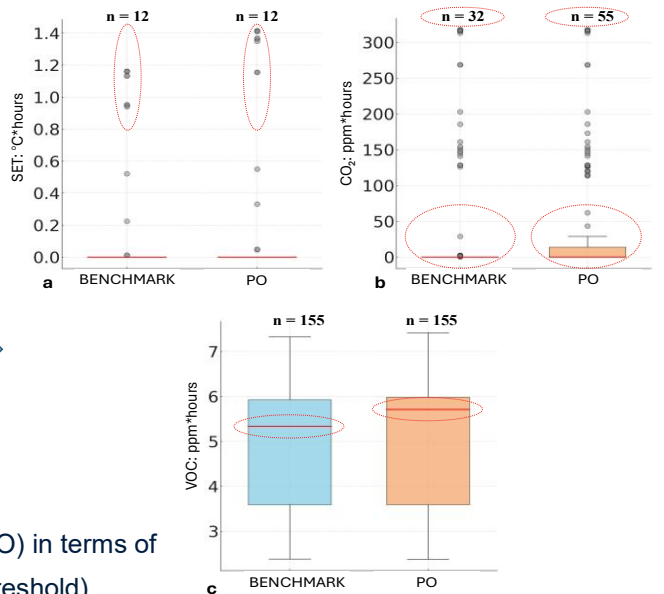
Sample outputs of building performance simulation (Dymola) in terms of (a) Operative Temperature (b) CO₂ concentration (c) VOC concentration and (d) Occupants

RESULT



Threshold:
 SET: 26.7 °C
 CO₂ : 950 ppm
 VOCs: 0.049 ppm

Distribution of the 163 simulation runs (bench+PO) in terms of
 Cumulative performance losses(over threshold)

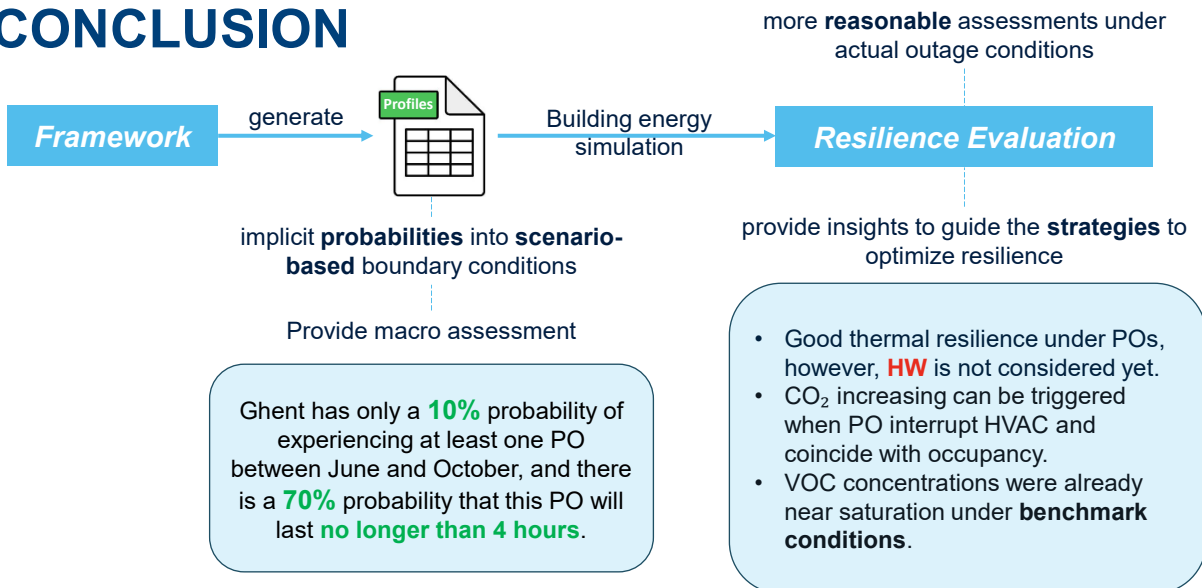


DISCUSSION

- The overall PO risk for this case study is extremely **Low**: only **8%** (June-October).
- IAQ resilience differs in terms of KPI:
 - CO₂ concentration is highly sensitive to PO **timing and duration**: significant rising occurs only when outages are long enough and overlap with occupancy.
 - For VOC, PO acts as a systematic amplifying factor, highlighting the limited effectiveness of ventilation and the need to prioritize **source control**.
- Strong thermal resilience under PO alone, however the **combined impact of HW and PO** still need further research.



CONCLUSION



CONCLUSION

Limitations

- Data acquisition challenge
- How to set Spatial Boundary
- Uncertainty in MCS



Future work

- To include external factors
- To refine and optimize the methodology
- To apply this framework under PO+HW