#### LIFE-CYCLE ASSESSMENT OF AN OFFICE BUILDING: INFLUENCE OF THE STRUCTURAL DESIGN ON THE EMBODIED CARBON EMISSIONS



SUSTAINABLE PLACES 2024

**Global Research and Development** 

06/06/2024 José Humberto M. de Paula Filho<sup>a</sup> <sup>a</sup> ArcelorMittal Global R&D, Esch-sur-Alzette, Luxembourg

 $\frac{\partial f_{i,j}(\vec{x},\vec{c})}{\partial x_i} = \sum_{k \neq i} c_{k,j}$ 

STEFI

## **Presentation Outline:**



- 1. INTRODUCTION
- 2. METHODOLOGY
- 3. RESULTS AND INTERPRETATION
- 4. CONCLUSION

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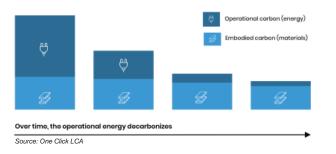
## **INTRODUCTION:**



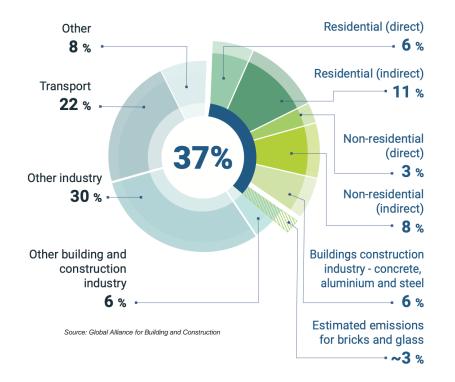
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#### **Environmental emergency**

- 37% of the global share of energy-related CO<sub>2</sub>e emissions were attributed to buildings and the construction sector.
- Efforts have been brought to reduce the operational carbon footprint of buildings by improving their energy efficiency



 Awareness has also been raised on embodied carbon. Emissions from materials and products must be urgently addressed to ensure constructions are optimized as lowcarbon solutions.





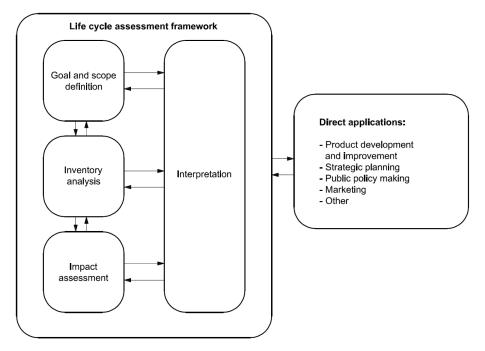
## **METHODOLOGY:**



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#### Life-cycle Assessment (LCA):

- LCA is a scientific and quantitative method for determining and assessing environmentally relevant processes.
- The ISO 14044 Environmental management Life cycle assessment — Requirements and guidelines define the steps that shall be followed for LCA:
  - Goal and Scope;
  - Life Cycle Inventory (LCI)
  - Life Cycle Impact Assessment (LCIA);
  - LCA Interpretation;



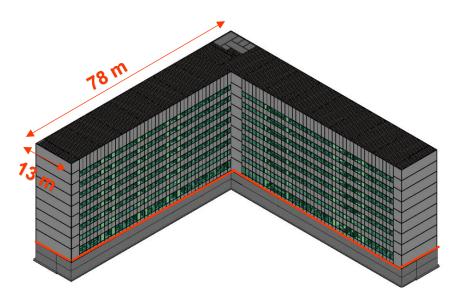
Source: ISO 14040



#### • Goal:

 The goal of this study is to assess the embodied carbon of equivalent structural options, composed of different materials, of a given office building located in Luxembourg for a reference service life of 50 years.

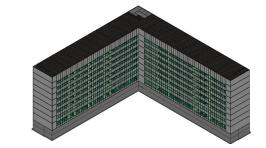
Functional unit	One L shaped office building with a gross floor total area (GFA) of 15552 m <sup>2</sup> located in Luxembourg over a 50-year analysis period
Reference unit	m².
Location	LU
Quantification	Material content as defined by the design office Schroeder & Associés.

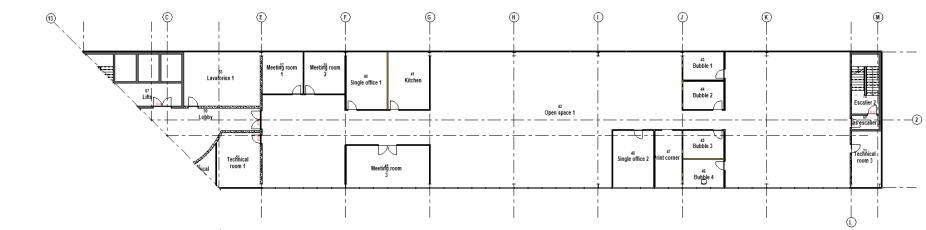


Dimension	Value
Building depth [m]	13.60
Building length [m]	78.15
Superstructure number of levels	R + 8
Infrastructure number of levels	2
Free height on the ground floor [m]	3.5
Free height on the intermediate floors [m]	2.7





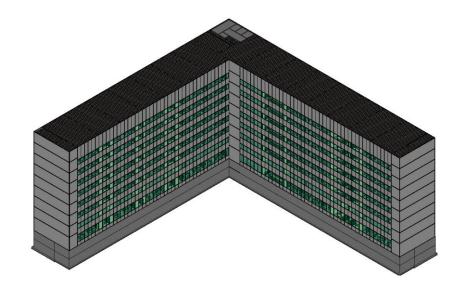






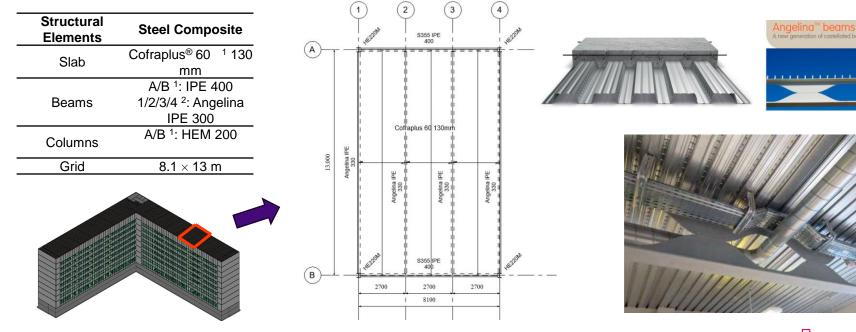
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- Object of assessment (Product description):
  - Three structural alternatives for the office building structure were studied:
    - Composite steel-concrete option;
    - Prefabricated reinforced concrete option;
    - Timber option;
  - The structural design was conducted by the Luxembourgish office Schroeder & Associés;
  - The functional equivalency of these solutions is ensured by adopting the same boundary conditions such as design assumptions, building location, soil condition, safety requirements, and actions on the structure;





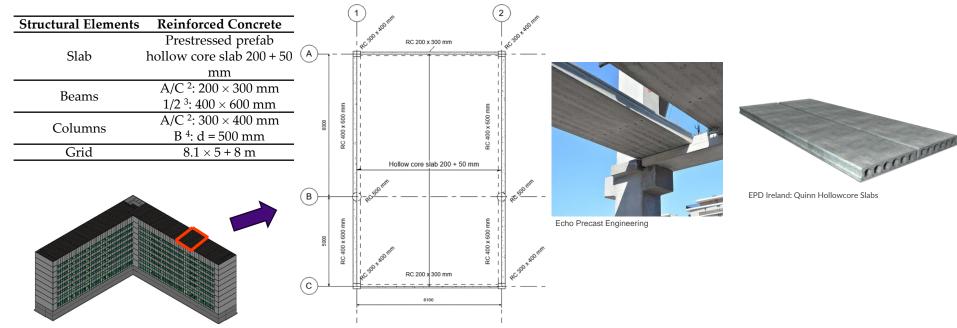
- Object of assessment (Product description):
  - 8.1 m by 13 m (clear span) for the composite steel-concrete option;





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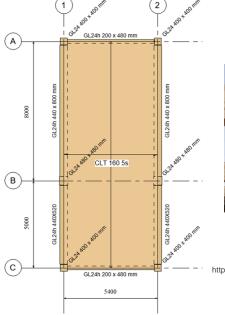
- Object of assessment (Product description):
  - 8.1 m by 5 m + 8 m (with intermediate columns) for the prefabricated reinforced concrete option;





- Object of assessment (Product description):
  - 5.4 m by 5 m + 8 m (with intermediate columns) for the timber options;

Timber			
CLT panel 160 mm			
A/C <sup>2</sup> : 200 × 480 1/2 <sup>3</sup> : 440 × 800			
A/C <sup>2</sup> : 400 × 400 mm B <sup>4</sup> : 480 × 480 mm			
$5.4 \times 5 + 8 \text{ m}$			





LEVER Architecture

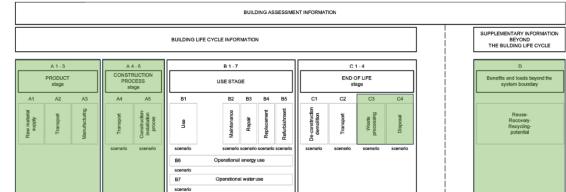
https://oxleytrade.com.au/product/clt-cross-laminated-timber/



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#### System Boundaries:

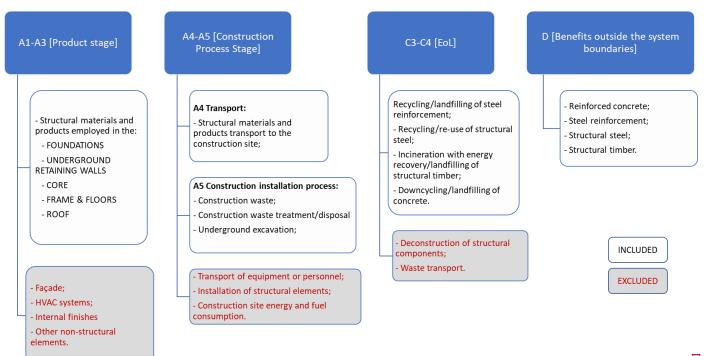
- The EN 15978 set out a common life cycle model for construction works. The life cycle model includes modular definitions for the life cycle stages, allowing each stage to be compared in isolation with other projects.
- Product Stage A1-A3;
- Construction Process Stage A4-A5;
- End of Life C3-C4
- Benefits and loads outside the systems boundaries – D.



Source: EN 15978



System Boundaries:





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#### • Time coverage:

 The study represents a hypothetical initial construction in 2022 and continued use of the office building for 50 years.

#### Technology Coverage:

 Technologies are representative of Luxembourg and Europe for manufacturing the structural materials, and the construction of the office building. Current technologies have been applied to future end-of-life.

#### Geographical Coverage:

 Design assumptions, background data on environmental impacts, and life cycle assessment scenarios were intended to represent an office building construction in Luxembourg. The results of this study are only applicable to the location, the boundary conditions and LCA scenarios considered.

#### Allocation

- Co-product allocation was not necessary for the foreground processes, as there are no coproducts known or considered in the construction of the office building.
- For all background data used in the model, the standard allocation assumptions of the used datasets were maintained.
- Cut-off criteria:
  - No cut-off criteria were applied in this study.



- Selection of Life Cycle Impact Assessment Methodology and Types of Impacts:
  - Global warming potential (GWP) was chosen because of its relevance to climate change. Climate change is of high public and institutional interest and is deemed to be one of the most pressing environmental issues of our times.
  - For GWP, the characterization factors from EC-JRC were applied (EN15804+A1/A2). (Environmental Footprint (EF).
  - In EN 15804+A1, GWP is presented as a single indicator, whereas in EN 15804+A2, GWP-total is divided into three sub-indicators: GWP-Fossil, GWP-Biogenic, and GWP-luluc.

- A partial compatibility is guaranteed by:
  - The global warming potential impacts caused by fossil fuels are remarkably similar between the different standards and characterization methodologies;
  - Biogenic carbon storage (negative carbon impact) is separated from the GWP in EN 15804+A1 data.
  - Then:
    - If and EPD as per EN 15804+A2 is used, then:
       GWP = GWP-fossil
    - If an EPD as per EN 15804+A1 is used, then:
       GWP = GWP biogenic carbon storage.



• Assumptions (scenarios):

- Transport A4

Material	Transport Type	Distance [km]
Structural steel	Trailer combination, 40-ton capacity, 100% fill rate	370
Steel rebars	Trailer combination, 40-ton capacity, 100% fill rate	370
Reinforced concrete	Concrete mixer truck, appr. 8 m <sup>3</sup> , 100% fill rate	60
Timber	Trailer combination, 40-ton capacity, 100% fill rate	220



## Construction A5 (waste)

Material	Percentage %
Structural steel	3.3
Steel rebar	4.85
Reinforced concrete	4
Timber	16.7





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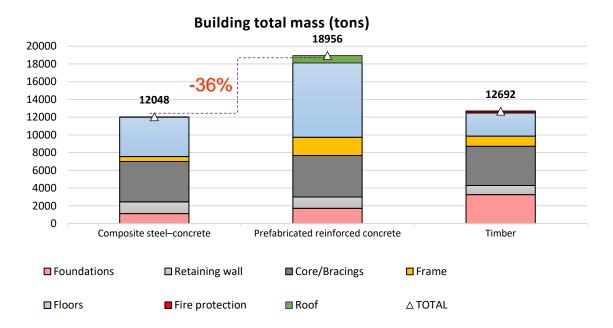
Assumptions:
 EOL C3-C4

Structural Elements	Landfill %	Re-Use %	Recycling %	Downcycling %	Incineration with Energy Recovery %
Structural steel sections	1	11	88	-	-
Hot-dip galvanized carbon steel	2	-	98	-	-
Steel reinforcement bars	10	-	90	-	-
Reinforced concrete	25	-	-	75	-
Timber	-	-	-	-	100



#### Life Cycle Inventory (LCI):

• Building mass:





#### Life Cycle Inventory (LCI):

• Material and Products – Datasets (Environmental Product Declaration):

Product/Material	Dataset name	Compliance system	Year	Geography	Upstream database
Structural steel sections	EPD, XCarb <sup>®</sup> recycled and renewably produced structural steel sections and merchant bars, ArcelorMittal Europe	EN 15804+A1	2023	RER	GaBi
Structural steel sections	EPD, structural steel sections in HISTAR	EN 15804+A1	2017	RER	GaBi
Hot-dip galvanized carbon stee	EPD, XCarb <sup>®</sup> recycled and renewably I produced hot dip galvanized steel with Magnelis <sup>®</sup> coating, ArcelorMittal	EN 15804+A2	2023	RER	GaBi
Steel rebar	EPD, XCarb <sup>®</sup> reinforcing steel in bars, ArcelorMittal Europe	EN 15804+A1	2021	RER	GaBi
Steel rebar	EPD, reinforcing steel in bars, ArceionMittai	EN 15604+A1	2016	RER	Gabi
Ready-mix concrete C30/37	One Click LCA	EN 15804+A2	2022	LU	Ecoinvent
Ready-mix concrete C40/50	One Click LCA	EN 15804+A2	2022	LU	Ecoinvent
Ready-mix concrete C50/60	One Click LCA	EN 15804+A2	2022	LU	Ecoinvent
CLT	EPD, cross-laminated timber (X-Lam), Studiengemeinschaft Holzleimbau e.V.	EN 15804+A1	2017	DE	GaBi
GluLam	EPD, binderholz Glulam—binderholz Bois lamelle-colle BSH—Legno lamellare BSH binderholz—binderholz BSH glulam	EN 15804+A1	2019	DE	GaBi



#### Life Cycle Impact assessment (LCIA):

Calculation matrix & GWP emission factors:						Data Source	Material	FU	A1–A3 GWP [kg CO <sub>2</sub> eq./FU]
						EPD, XCarb <sup>®</sup> recycled and renewably produced structural steel sections and merchant bars, ArcelorMittal Europe	Structural steel sections	kg	0.33
						EPD, structural steel sections in HISTAR grades, ArcelorMittal	Structural steel sections	kg	0.52
Amount of products		Environment per unit of	•	i	Environmental mpact of stage <i>i</i>	EPD, XCarb <sup>®</sup> recycled and renewably produced hot dip galvanized steel with Magnelis <sup>®</sup> coating, ArcelorMittal	Hot-dip galvanized carbon steel	kg	0.794
	EPD data for a1	EPD data	EPD data	EPD data		EPD, XCarb <sup>®</sup> reinforcing steel in bars, ArcelorMittal Europe	Steel rebar	kg	0.30
	stage i	for a <sub>2</sub> stage i	for a <sub>3</sub> stage i	for a <sub>n</sub> stage i		EPD, reinforcing steel in bars, ArcelorMittal	Steel rebar	kg	1.23
a 1 <i>i</i>						One Click LCA	Ready-mix concrete C30/37	m <sup>3</sup>	270.88
a 2 i		GWP a 2 <i>i</i>		GWP a n i	GWP i	One Click LCA	Ready-mix concrete C40/50	m³	355.83
a 3 <i>i</i>	GWP a 1 <i>i</i>	GWP a 27	GWP a 3 <i>i</i>	GWPan7		One Click LCA	Ready-mix concrete C50/60	m <sup>3</sup>	429.00
a n <i>i</i>						EPD, cross-laminated timber (X-Lam), Studiengemeinschaft Holzleimbau e.V.	CLT	m³	187.23
						EPD, binderholz Glulam—binderholz Bois lamelle-colle BSH—Legno	GluLam	m³	205.3

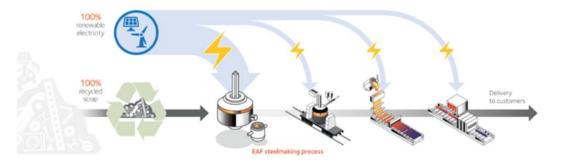
lamellare BSH binderholz—binderholz

BSH glulam



#### Life Cycle Impact assessment (LCIA):

- XCarb<sup>™</sup>:
  - ArcelorMittal's XCarb<sup>™</sup> initiative, which emphasizes recycled and renewably sourced steel, is centered around the Electric Arc Furnace (EAF) route.
  - This approach involves using 100% recycled material combined with 100% renewable electricity.



Material/Product	XCarb <sup>™</sup> A1-A3 kgCO <sub>2</sub> eq./kg	WorldSteel association kgCO₂eq./kg	Reduction %
Hot Rolled Steel Sections	0,33	1,91 (global production)	82,72
Steel reinforcement (rebars)	0,30	1,94 (global production)	84,53
Hot Rolled Coil	0,532	2,28 (global production)*	82,72
Metallic Coated Coil	0,797	-	-



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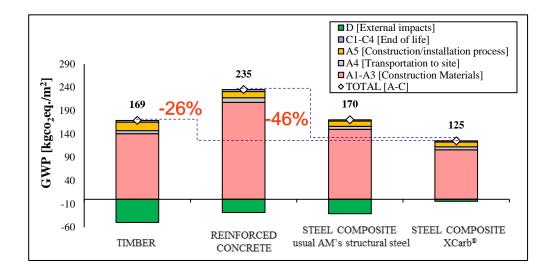
## **RESULTS AND INTERPRETATION:**



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#### **Results and Interpretation:**

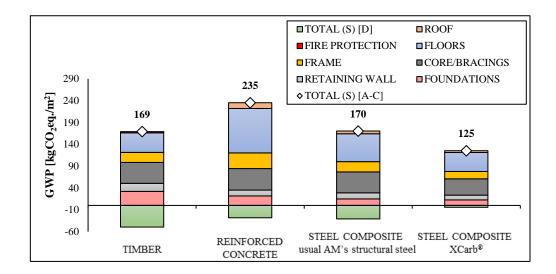
• The results are presented in terms of FU (m<sup>2</sup>):





#### **Results and Interpretation:**

• The results are presented in terms of FU (m<sup>2</sup>):





## **CONCLUSIONS:**



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#### **Conclusions:**

- The whole life cycle GWP is calculated for all structural options. The steel-concrete composite option outperforms the:
  - Reinforced concrete option reducing by 46% when comparing to the XCarb<sup>®</sup> steel-concrete composite option;
  - Timber option reducing by 26% the overall GWP for the XCarb<sup>®</sup> steel-concrete composite option.
  - The steel-concrete composite option profits from having a high content of recycled steel (scrap), up to 100%. In addition, a greater reduction is achieved for the use of XCarb<sup>®</sup> steel since it is produced with 100% renewable electricity.
- The production stage (modules A1-A3) is the major responsible for the CO<sub>2</sub>e emissions.
  - Because of that the reduction of raw material extraction using high-recycled content materials is key to the reduction of embodied carbon.
- Floors are identified as the building part that contributes the most to the overall GWP. Ready-mix concrete is the main responsible for the impacts related to the floors for the steel and reinforced concrete options.





# THANK YOU!!

a manager and a second

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# **S-LCA OF RECYCLED PVC PRODUCTS** CIRCUSTAIN PROJECT

### Nirvana A. Marting V., Thomas Schaubroeck

**Post-doctoral researcher** 

September 25<sup>th</sup>, 2024



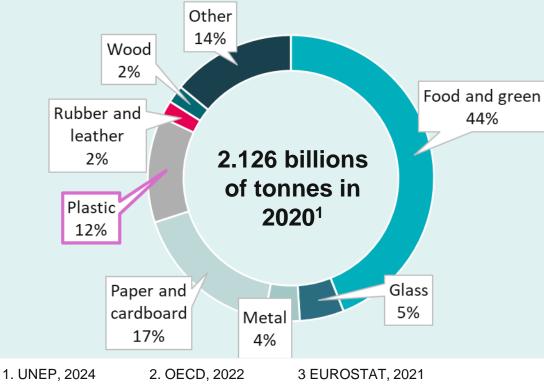


# CONTENT

- ▶ Motivation: PVC value chain
- ► CIRCUSTAIN project: Objectives and scope
- Methodology: Social Life Cycle Assessment
- Preliminary results
- Outlook



# SOLID WASTE PRODUCED GLOBALLY



Global plastic waste accounted for 353 Mt in 2019<sup>2</sup>.

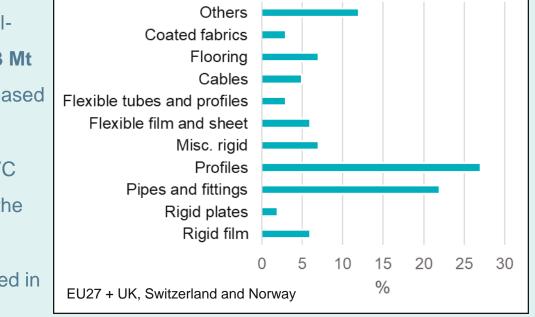
Approx. 16 Mt of plastic **waste** were generated in the EU in 2021<sup>3</sup>.



# **PVC FIGURES**

- Production: In the EU in 2022 fossilbased PVC represented 9.1% or 5.3 Mt of a total of 58.7 Mt plastics (fossil-based + recycled)<sup>4</sup>.
- Waste: 2.4 Mt of post-consumer PVC waste in the EU, 44% generated in the construction sector<sup>5</sup>.
- 70% of PVC is used for products used in the building sector<sup>6</sup>.

#### PVC applications in the EU<sup>7</sup>



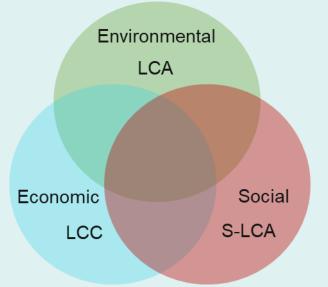
# **PVC FIGURES**

- ► The plastic processing sector employs 338000 employees □ 5.3% of the manufacturing sector in Germany<sup>8</sup>.
- ► Around 41% of companies in this sector contain <50 employees.
- ▶ Only 1% of companies >1000 employees<sup>8</sup>.
- ▶ In 2019: 73% male, 27% female<sup>8</sup>.
- ▶ 92% full-time, 8% part-time<sup>8</sup>.



# **CIRCUSTAIN PROJECT**

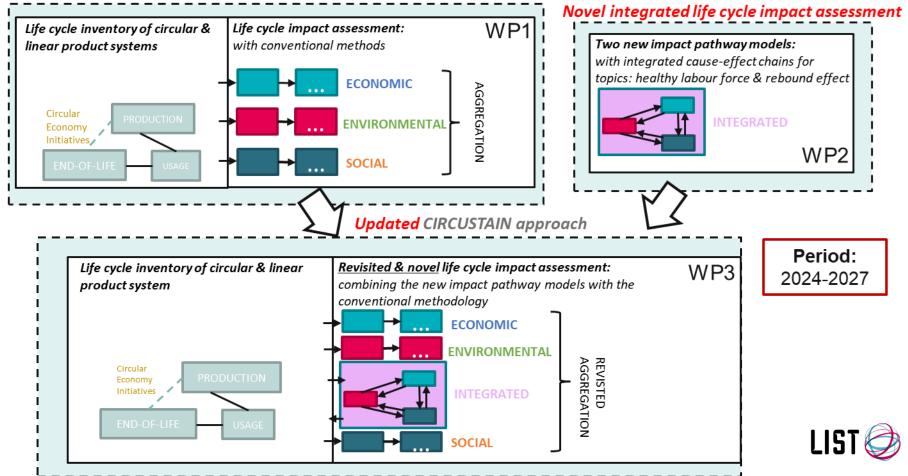
- ► Circular economy (CE) strategies, like plastic recycling
- Sustainability of CE practices is not guaranteed better than conventional ones.
- LCSA (social, economic and environmental)



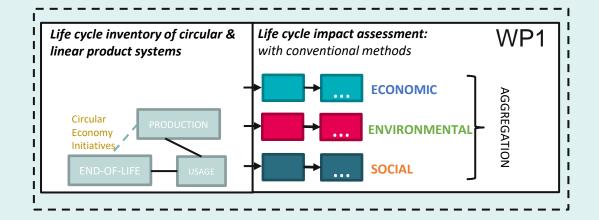


#### CIRCUSTAIN: Integrated sustainability assessment across all three sustainability pillars of circular economy initiatives

Conventional Life cycle sustainability assessment of circular vs. linear economy

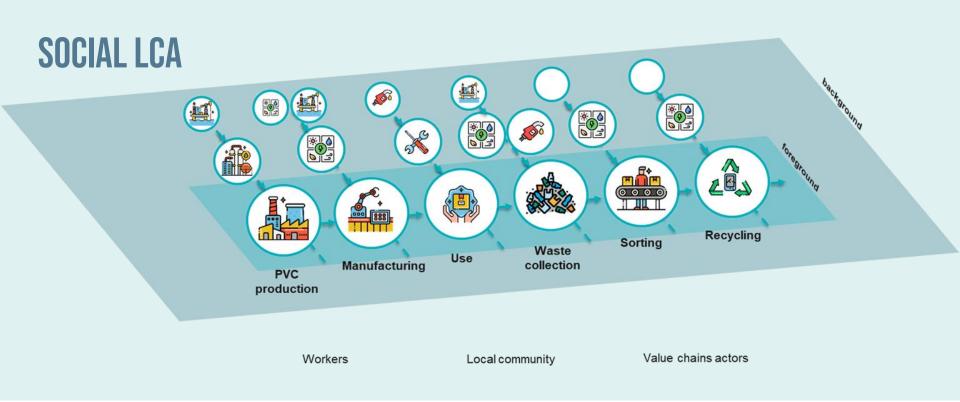


# **OBJECTIVE**



To quantify the **social issues, focusing on labour aspects**, alongside economic and environmental impacts of identified circular versus linear product systems using conventional LCSA.







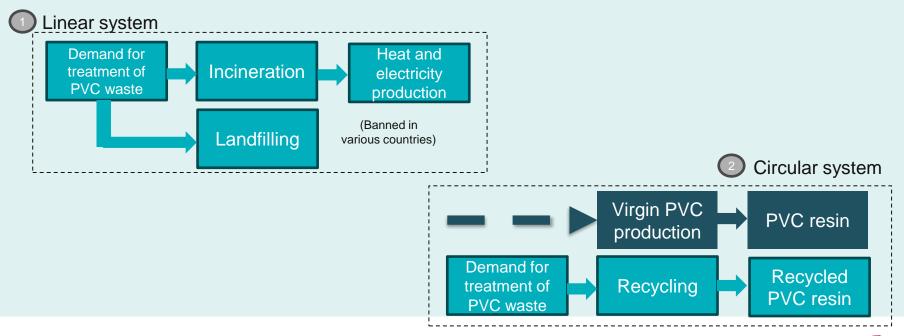
# **SOCIAL LCA FRAMEWORK**



Complete framework in the UNEP Guidelines for S-LCA of products and organizations<sup>9</sup>, 2020



# CASE STUDY: PRODUCT SYSTEM





# **CASE STUDY: SCOPE**

- **Geographical delimitation:** PVC production and recycling in Germany and Luxembourg
- **Stakeholder covered:** Workers, local community
- ► Impact categories addressed: job creation potential and health and safety
- Sectors in the foreground system: construction (installation and unmounting of PVC products) and manufacturing (plastics, chemicals production)



# CASE STUDY: METHODOLOGY

- > Job Creation
  - Input-Output Tables
    - ► PSILCA<sup>10</sup>
    - ▶ EXIOBASE v3 monetary<sup>11</sup>

- Health and safety
  - Exposure to chemical substances
  - Rate of fatal and non-fatal accidents



# **INPUT-OUTPUT TABLES**

 Input–output analysis is an analytical framework that provides data of the structure of the economy based on the interdependence of industries<sup>10</sup> presented in input–output tables.

The input–output table of a country contains information on monetary transactions between industries in a specific location. It shows the flows of products that industries purchase (inputs) and sell (outputs) <sup>10</sup>.

 Industries are related to each other because they buy products from other industries to produce their own products and in turn, provide products to other industries, which use them in their own production<sup>10</sup>.

	Country 1 – sector 1	Country 1 – sector 2	Country 2 – sector 1	Country 2 – sector 2
Country 1 – sector 1				
Country 1 – sector 2				
Country 2 – sector 1				
Country 2 – sector 2				



# CASE STUDY: METHODOLOGY

- ► PSILCA
  - Based on EORA MRIO
  - ► 189 countries
  - It contains worker-hours by sector and country
  - Aim of the database: to conduct Social Risk Assessment

- ► EXIOBASE
  - ► MRIO
  - ▶ 49 countries + regions
  - Worker-hours and employment accounts by sector and country



# **WORKER HOURS ESTIMATION**

PSILCA estimated worker-hours based on EORA following:

Worker hours  $[h] = \frac{\text{Unit labor costs}}{\text{Mean hourly labor costs per employee [USD/h]}}$ 

Unit labor costs =  $\frac{\text{Compensation of employees by sector and country } [€]}{\text{Gross output by sector and country } [€]}$ 

EXIOBASE contains total employment and worker-hours accounts

	Country 1 – sector 1	Country 1 – sector 2	Country 2 – sector 1	Country 2 – sector 2
Compensation of employees				
Total employment				
Vulnerable employment				



## **INVENTORY WORKER HOURS – FOREGROUND SYSTEM - PRELIMINARY**

Process units	Total wh per tonne PVC	Unit	Linear				
PVC production	2.68	wh/PVC tonne	Circular				
Profile production	4.50	wh/PVC tonne					
Windows assembly	58.38	wh/PVC tonne	0 0.2 0.4 0.6 0.8 1 Jobs per tonne PVC				
Montage	224.72	wh/PVC tonne	Jobs per tonne PVC				
Use in B&I			Virgin production and manufacturing				
Unmounting	337.08	wh/PVC tonne	Collection, recycling and production				
Sorting	1.00	wh/PVC tonne	End-of-life				
Mechanical recycling							
Profile production	4.50	wh/PVC tonne	<ul> <li>Manual activities require more hours per functional</li> </ul>				
Windows assembly	58.38	wh/PVC tonne	unit.				
Montage	224.72	wh/PVC tonne	EXIOBASE: Windows assembly and PVC production				
Use in B&I			<ul> <li>Costs variation</li> </ul>				
Unmounting	337.08	wh/PVC tonne	<ul> <li>Mechanical recycling</li> </ul>				
Sorting	1.00	wh/PVC tonne	► Technologies				



# WORKER HOURS IN THE LIFE CYCLE<sup>12</sup>

- Direct jobs.- Jobs provided by a project
- **Direct induced jobs**.- Jobs provided by a supplier and created to serve the project
- **Indirect induced jobs**.- Jobs provided by the supplier of a supplier to develop an input for the project
- Secondary jobs.- Jobs created incentivise consumption creating in turn other jobs

A = aij = Xij/Xj

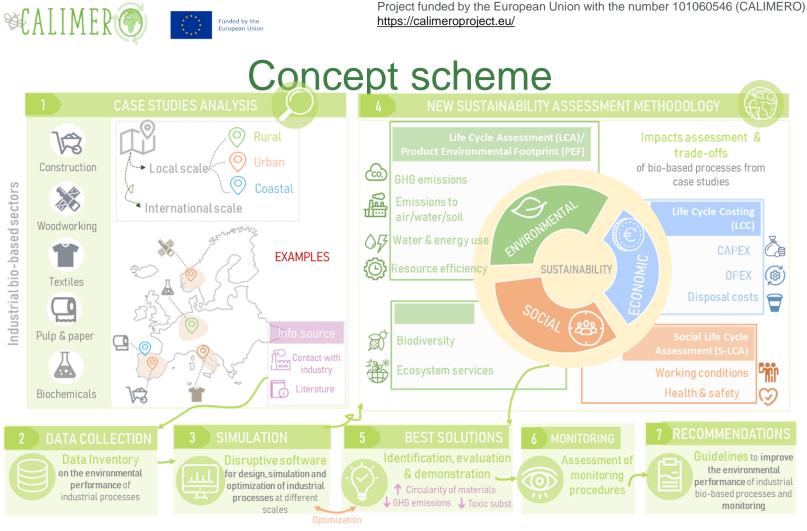
- A: Matrix of technical coefficients
- aij: Technical coefficient of a sector by total sector output
- Xij: Quantity of monetary units from sector (i) necessary to the sector (j)
- Xj: Total monetary output of sector (j)

 $f = S (I-A)^{-1}$ 

- f: direct and indirect worker hours to produced one monetary unit
- S: social matrix in hours per monetary unit
- I: identity
- A: technical coefficients matrix





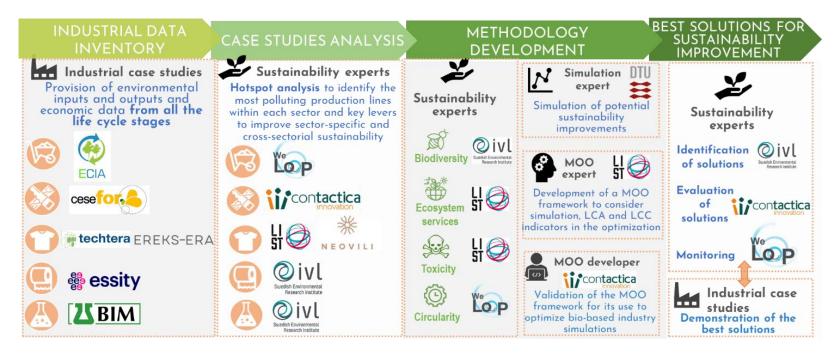






Funded by the European Union Project funded by the European Union with the number 101060546 (CALIMERO) <u>https://calimeroproject.eu/</u>

## Implementation





# **DATABASE COMPARISON (EXAMPLE)**

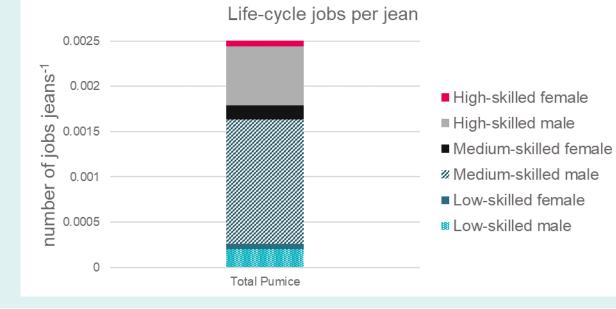


#### **Databases comparison**

► 30-35%



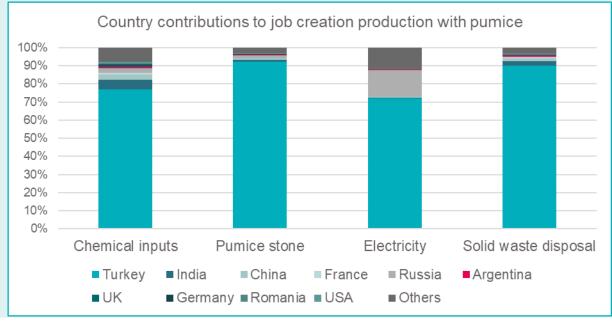
# **SKILL-LEVELS (EXAMPLE)**



#### ► Skill-level

LIST 🥏

# **REGIONALIZATION (EXAMPLE)**



- ► PSILCA
- Job creation by region



## **OUTLOOK**

➤ Data quality

- Health 
   Human toxicity of different chemicals workers are exposed to along the value chain using USEtox.
- > Physical hazards  $\square$  Rate of accidents in the plastics and construction sectors.



[1] United Nations Environment Programme (2024). Global Waste Management Outlook 2024: Beyond an age of waste – Turning rubbish into a resource. Nairobi. https://wedocs.unep.org/20.500.11822/44939 URL: https://www.unep.org/resources/global-waste-management-outlook-2024. Accessed 12/09/2024

[2] OECD (2022), Global Plastics Outlook: Economic Drivers, Environmental Impacts and Policy Options, OECD Publishing, Paris, https://doi.org/10.1787/de747aef-en.

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- > https://www.flaticon.com/free-icons/conveyor-belt" title="conveyor belt icons">Conveyor belt icons created by Flat Icons Flaticon
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# thank you

#### contact nirvana.marting@list.lu













# PREMISES project: impact of predictive maintenance and loT on the SRI

Igor PEREVOZCHIKOV, R2M Solution France

Sustainable Places 2024 25 September 2024 Luxembourg, Luxembourg



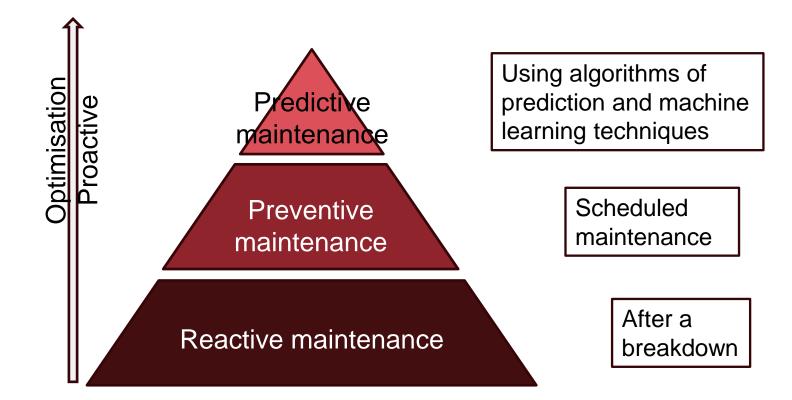


# Why the predictive maintenance?

breakdown of the b	o reduce udget of enance? How to manage building in more smarter way?
3 <sup>rd</sup> floor is not more 9 <sup>th</sup> heat working is not more We pay higher cost The cost for	not producing on quantity

# What is the predictive maintenance?

Predictive maintenance uses data analysis to identify operational anomalies and potential equipment defects, enabling timely repairs before failures occur. It is often based on sensor and meter data.



PREMISES



# Challenges of predictive maintenance

Predictive maintenance uses data analysis to identify operational anomalies and potential equipment defects, enabling timely repairs before failures occur. It is often based on sensor and meter data.





# Predictive maintenance for energy optimization

- Most deviations involve overconsumption.
- Anticipating deviations automatically increases the energy performance of installations.
- Contextualizing data is the only way to interpret it.



G R2M Solution



# Objective, innovations, KPIs

PREMISES reimagines smart building management by introducing a groundbreaking approach to HVAC predictive maintenance and energy optimization. Traditional systems offer data without context, while PREMISES introduces decentralized control, seamlessly integrating **self-learning AI algorithms** with **IoT** devices. This approach promotes cooperation among building equipment, enhancing energy efficiency, reducing costs, and improving comfort. Initial focus will be on **industrial buildings**. PREMISES targets a shift from TRL4 to TRL6 through lab testing in a small building.

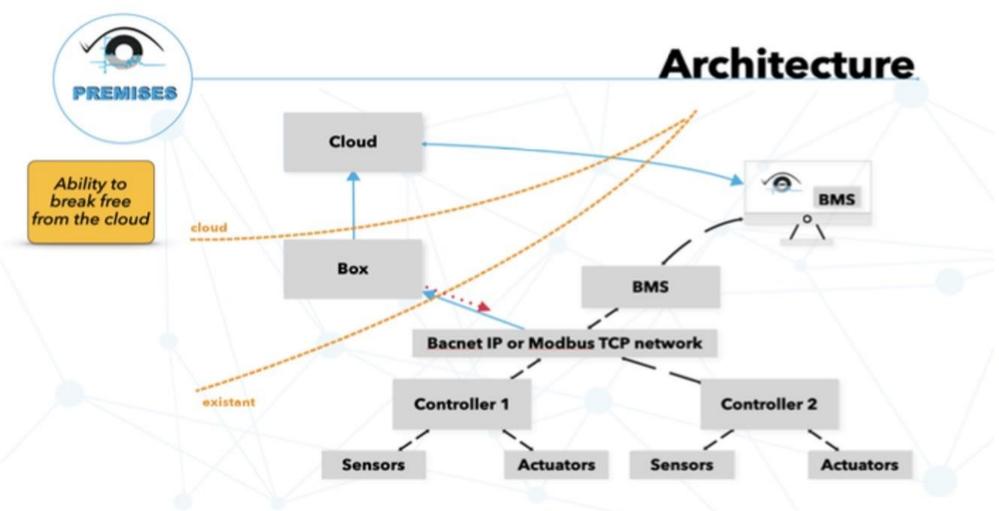
Duration: 9 months
 Budget: 50k€
 Start date: 1st April 2024
 End date : 31st December 2024

#### Key innovations:

- Low installation costs
- Efficient integration with existing systems
- Al-driven predictive maintenance
- ROI of under 1.5 years.
- SRI score enhancement

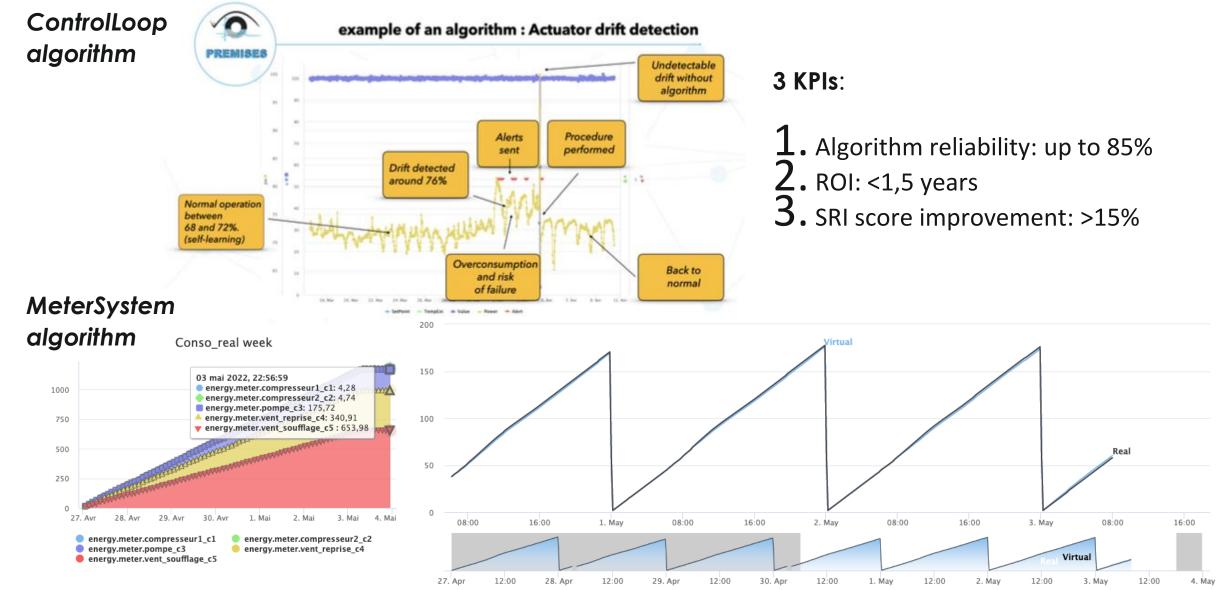


# Architecture of PREMISES solution





# Algorithms and KPIs





# Demo building

Location: South-East of France Area of building: 1100 m2 Age of building: 2019 Use of building: office + manufacture + warehouse Number of floors: 2 + basement Number of users: 35











# Project workplan

## Technical requirements, installation and implementation:

- $_{\rm \circ}~$  Selection and installation of the required IoT devices
- Setup of Home Assistant and Windows with the Oris BMS on the server
- Connection of all equipment via MQTT, Zigbee and Modbus-TCP communication protocol
- Implementation of the AI-driven algorithms

Lab testing:

Al-driven algorithms setup

SRI analysis:

- SRI assessment of pilot building realized
- SRI results analysis, recommendations and simulation performed KPI's assessment.



# Smart Readiness Indicator (SRI) initial score

	Total SRI score		Energy performance and operation			32.4% 29.0%		
	22	2.28%	E R	nergy flexibility (including demanesponse)	d	5.4%		
				7 IMPACTS				
		ÀÀ		$\mathbb{R}$	<b>P</b>			I
Er	nergy efficiency	Energy flexibility	Comfort	Convenience	Health	Maintenance	Information to occupants	
	39%	5%	31%	29%	19%	26%	37%	
Heating								34%
DHW								47%
Cooling								29%
Ventilation								12%
Lighting								23%
Dynamic en	velope							2%
Electricity								56%
eVehicle								0%
M&C								20%



# Smart Readiness Indicator (SRI) improvements

SRI score improvements recommended:

# 1) Equip the building with IoT sensors for measurement of ambient parameters (T, RH, IAQ sensors) and implement a BMS of classes B or A in order to:

- 1. Centralize monitoring and control of all building technical systems (heating, cooling, ventilation, DHW, lighting and electrical outlets).
- 2. Implement building management based on real-time data from local weather station to be installed.
- 3. Collect next-day local weather forecasts to prepare the predictive control of building systems.

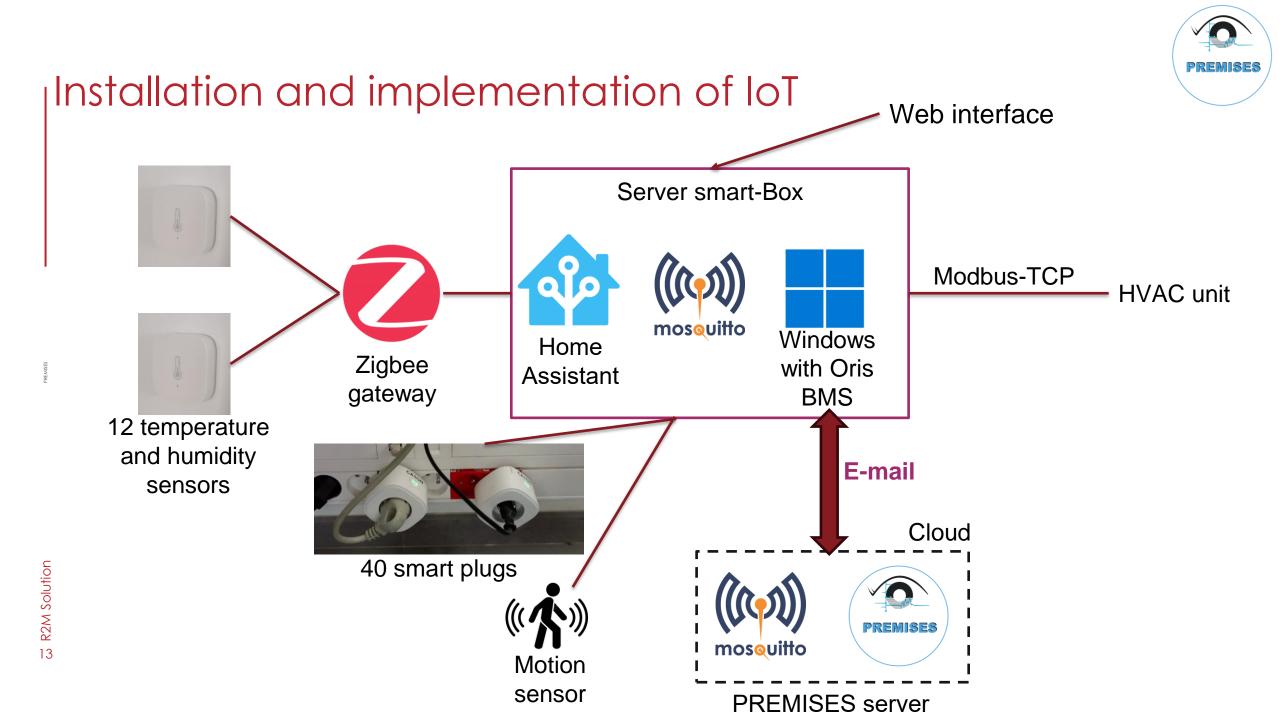
#### 2) Implement a PREMISES solution for predictive maintenance of HVAC systems.

3) Provide flexibility capabilities by connecting BMS supervision server to electrical grid signals (RTE, Enedis) via an API allowing to adjust HVAC operation based on grid status.

#### **Optional Measures:**

- Install renewable energy production systems (PV panels, solar thermal, mini wind turbines) with storage.
- Connect these systems to grid commands (RTE, Enedis) via an API.

REA

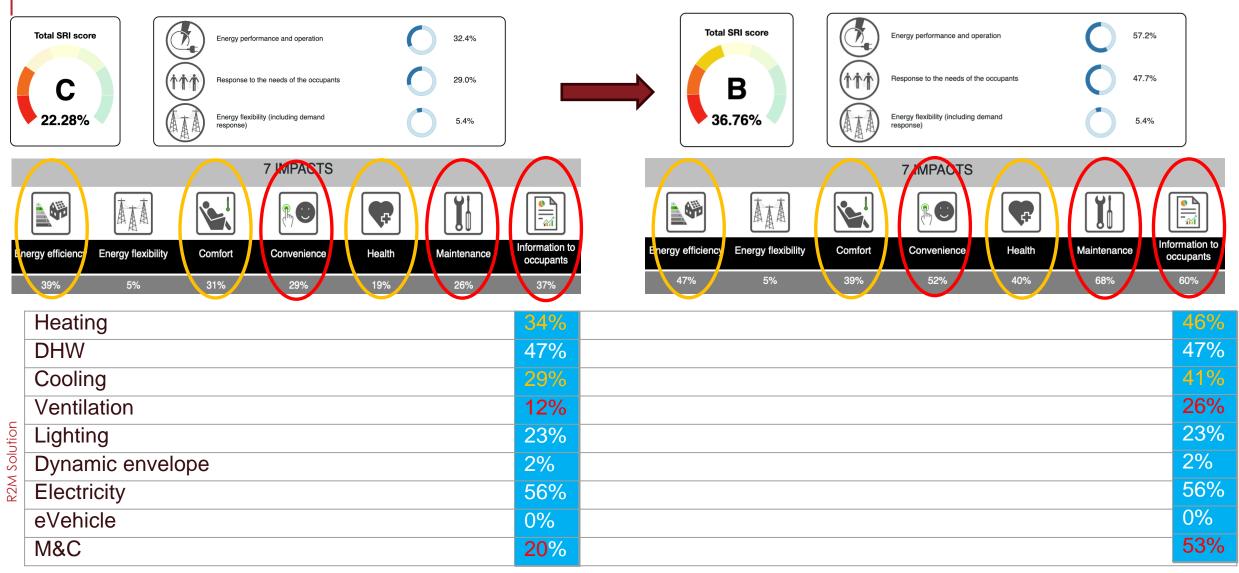


### Smart Readiness Indicator (SRI) analysis



SRI of the building in its current state with scores by main functionality and detailed scores by domain and impact criterion

Simulated SRI of the OSMOSE building with recommended intelligence improvement measures



# Main outcomes



The PREMISES predictive maintenance associated with IoT measurement devices installed in a nonresidential building without TABS impacts mainly **detecting faults of TBS** SRI service and **reporting of performances** services associated to **Heating**, **Cooling**, **Ventilation** and **Monitoring&Control** domains. The impact of PREMISES solution is equal to **14,48**%.

The technical domain of *Monitoring&Control* is the mostly impacted by proposed measures. Its associated SRI score raised up by **33**%. This is followed by *Ventilation* domain whose score is raised up by **14**%.

The SRI criteria the more impacted by improvements proposed is the *Maintenance* criteria, raised by 42%. It is directly related to the predictive maintenance functionality of the PREMISES solution. It is followed by *Information to occupants* and *Convenience* criterias, raised by 23%.

The optional measure related to connection of technical building systems (heating and cooling) to signals from grid in order to provide flexibility control based on grid signals has a big impact on the SRI score, +11,63%.





# PREMISES project

# THANK YOU!

Igor PEREVOZCHIKOV, R2M Solution France

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Content

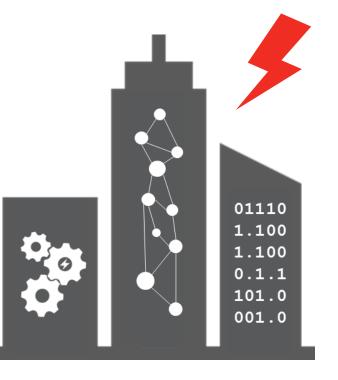


- Background and objective
- Residual generating FDD
- Hybrid AI methodology for FDD
- Case studies
- Summary

•

#### 3

- High implementation and parametrization effort required for FDD methods
- Improving the transferability and quality of FDD through a hybrid AI method based on a rule-based system combined with a machine-learning based residual generation



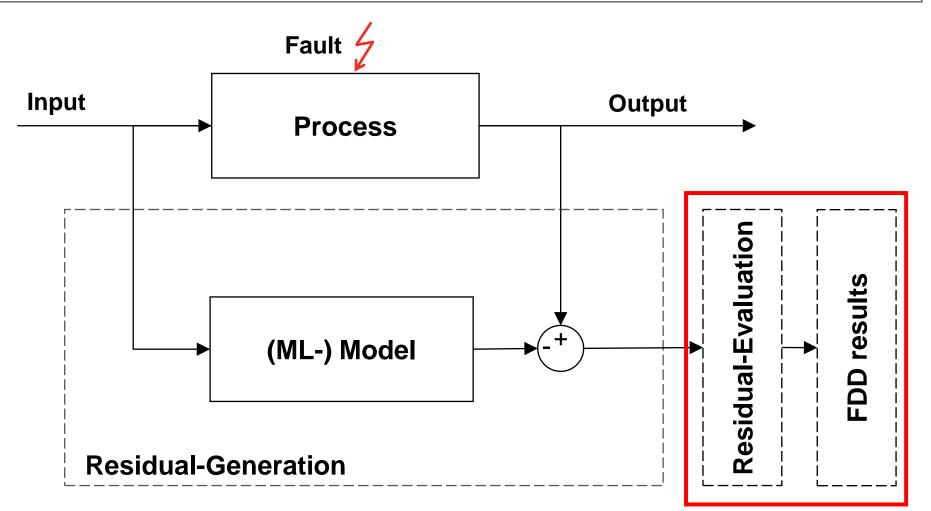


Automated fault detection and diagnosis (FDD)

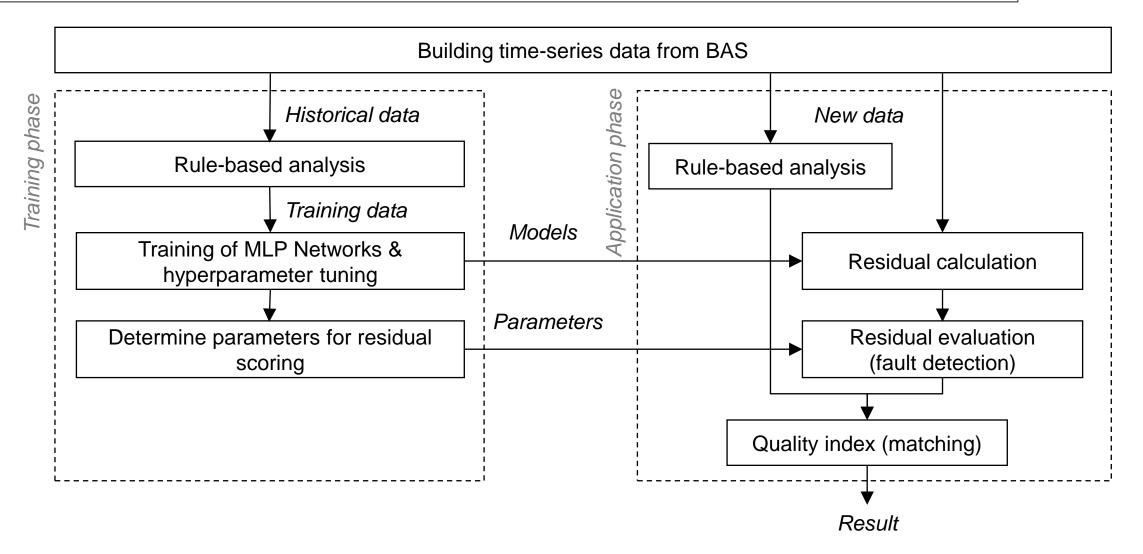


can save up to 20% energy









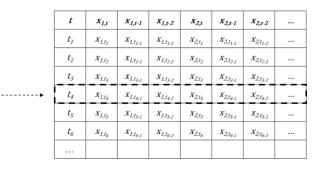


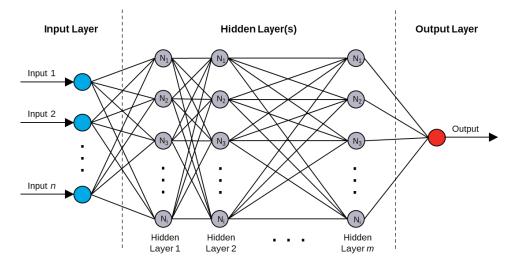
- Features
  - Extensive set of "if-then rules" for HVAC systems
  - Automated discovery of data points and plant topologies based on unified semantics
- Advantages
  - Simple and comprehensible implementation
  - Immediate diagnosis
  - High specificity if parameterized correctly
- Drawbacks
  - Rich in parameters high parametrization efforts
  - Low sensitivity (no detection of new faults)

TH He	ome Assets Data-Manager Configure	Rules Configure CON	IETH AI Log Book	🗾 Fraunhof	fer ISE
BA3135					- 10
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AHUExhaustAirCO2ConcentrationIsHigh	Exhaust air CO2 concentration is	{'exha_co2_max': 900.0}	33.0	2022-11-08 19:05:00+00:00	2022-11-14 23:55:00+00:
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**MLP Networks & hyperparameter tuning** 

- Transformation of the input data into a sliding time window (2h)
- Creation of separate MLP network models for each estimated variable (n inputs, 1 output)
  - Regression for continuous variables
  - Classification for discrete variables
- Bayesian hyperparameter optimization (e.g. hidden layers = 3-6, neurons per layer = 24-552)







 $X_1$ 

 $X_{1,t_1}$ 

 $X_{1,t_2}$ 

 $X_{1,t_3}$ 

 $X_{1,t_4}$ 

 $X_{1,t_{5}}$ 

 $X_{1,t_6}$ 

t,

 $t_3$ 

 $t_5$ 

 $t_6$ 

 $\vec{x}_w(t_i)$ 

 $X_2$ 

 $X_{2,t_1}$ 

 $X_{2,t_2}$ 

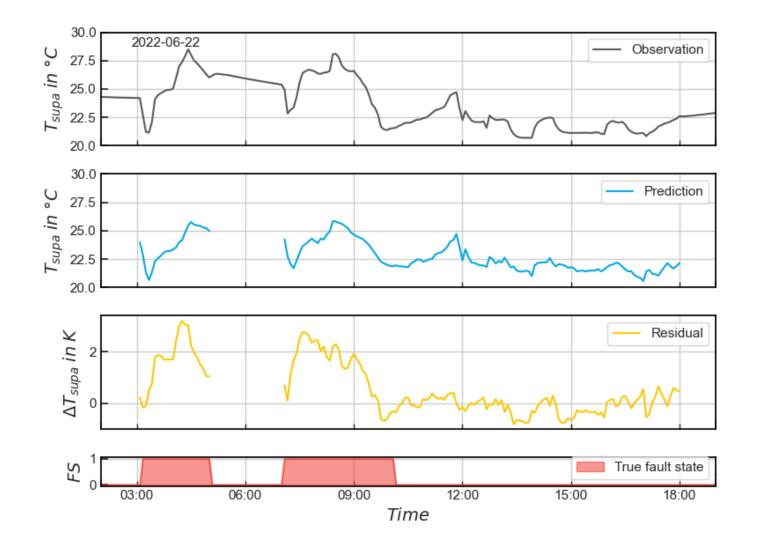
 $X_{2,t_3}$ 

 $X_{2,t_{5}}$ 

 $X_{2,t_6}$ 

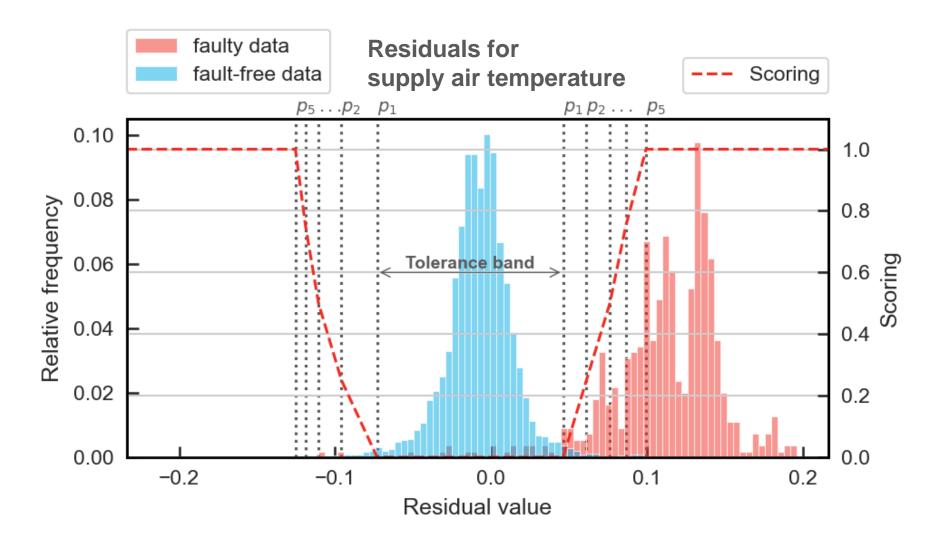
# **Residual evaluation**





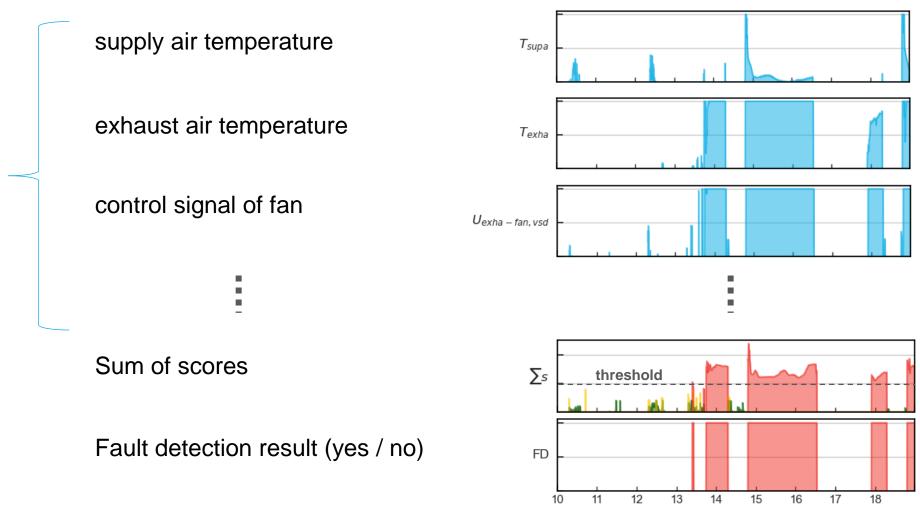
- Time series example of residual generation for supply air temperature
- Fault shown in the period: Heating without heat recovery







Single scores



Oct





• 9 different air handling units (AHUs) from 3 buildings (2850 m<sup>3</sup>/h - 6000 m<sup>3</sup>/h)

Primary school Hohen Neuendorf (PSHN)



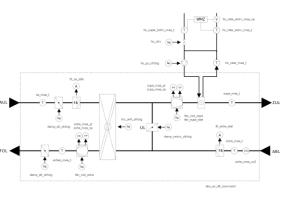
- Combined natural & mechanical ventilation
- Supplying classrooms (FLA, FLB) and a sports hall (SPO)
- Supply and exhaust air systems, plate heat exchanger & heating coil

Maison du Savoir (MSA)



- University of Luxembourg
- Supplying lecture rooms (AUD2, AUD7, AUD8)
- Supply and exhaust air systems, heat recovery wheel, heating coil and adiabatic cooling

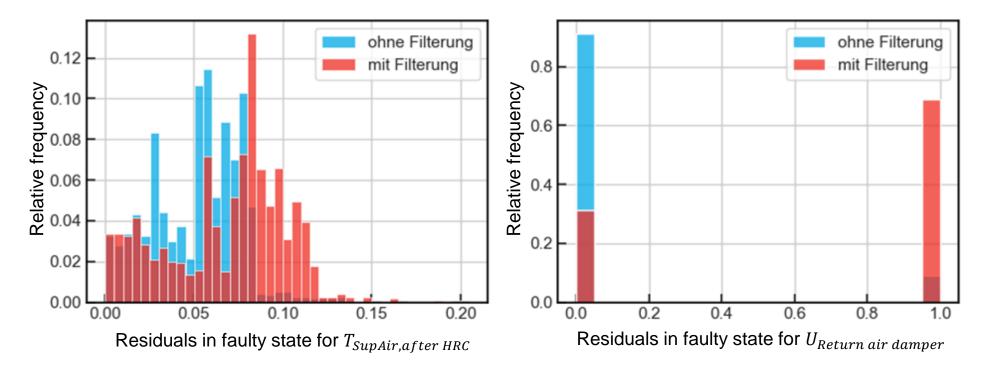
Anonymised boarding school building (ISC)



- Supplying a theater hall (AUD2) ,laboratories (AUD8) and accommodation areas (AUD17)
- Supply and exhaust air systems, heat recovery wheel, heating coil, adiabatic cooling for AUD2



#### Residual-Modulus with and without filtering of faulty states in the training data

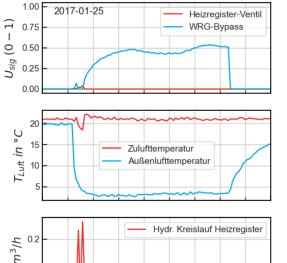


Rule-based pre-filtering of the training data increases the quality of the residuals in the event of a fault

# **FD time series examples**



HC valve leackage (FLB, PSHN)



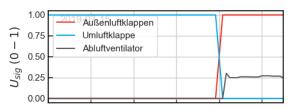
Uhrzeit

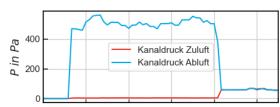
 Volume flow with closed heating coil valve signal indicates a leakage

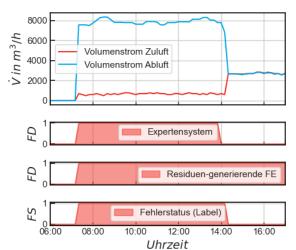
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Limited data availability prevents the fault from being detected by the rules (supply air temperature upstream of the heating coil is missing)





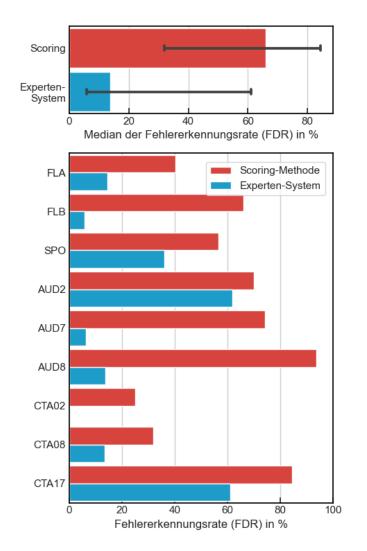




- High duct pressure with closed dampers indicate fan operation (although the control signal = 0)
- Consistency of both methods shows high confidence in the result

## **Fault detection performance**





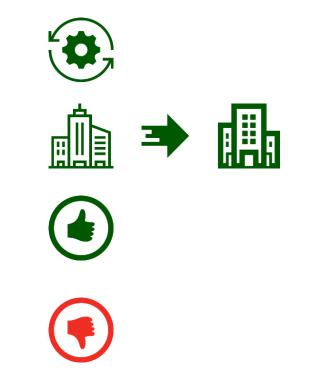
#### Residual generating approach versus expert system

- Higher sensitivity of the residual generating method
- In case of bad training data or changing boundary conditions, risk of high FARs exists
- Detection of novel faults for which no rules are implemented, or rules can't be implemented due to data availability
- Combined application increases the FD performance



## A Hybrid AI methodology for FDD in buildings based on rules & ML-techniques

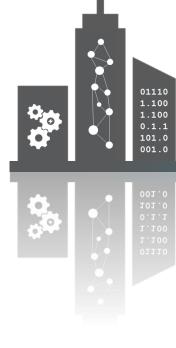
- Highly automated / low implementation effort
- Easily transferable to new systems
- Increased FD-Performance
- Limitation: Requires re-training if boundary conditions change (e.g. set points)











# Thank you for your attention!

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