



Robust and reliable general management tool for performance and durability improvement of fuel cell stationary units

## Degradation-aware energy management of fuel cell-based VPPs

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RUBY WORKSHOP, LUCERNE (CH)

5 JULY 2022

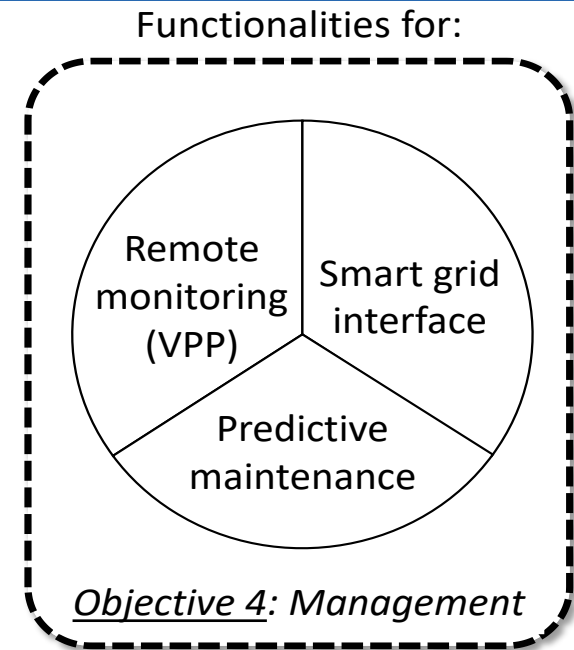
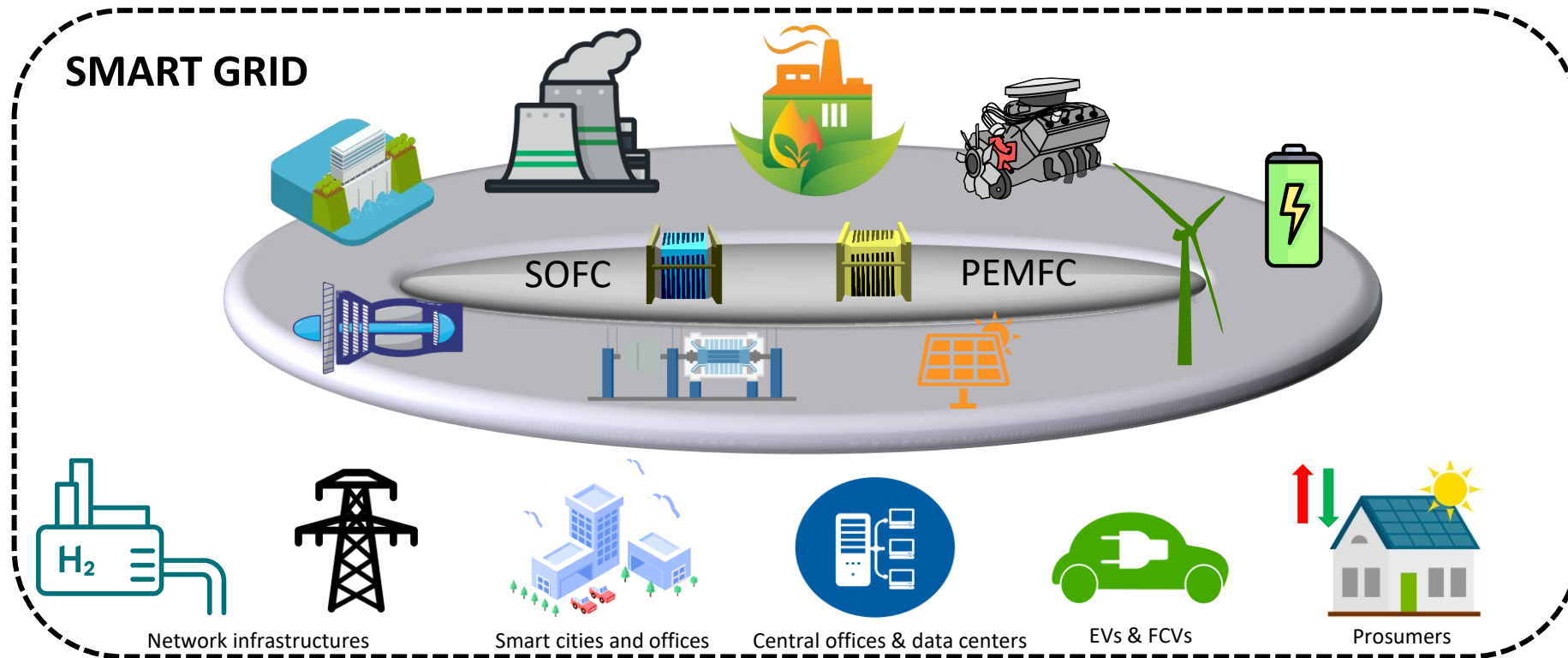
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# Scenario – Energy Technologies & Uses



RUBY aims at improving **FCs** performance and durability (>40000 h) through an **MDPC** algorithm

**VPPs** are ready-made energy options to be suitably embedded in **Smart Grids**

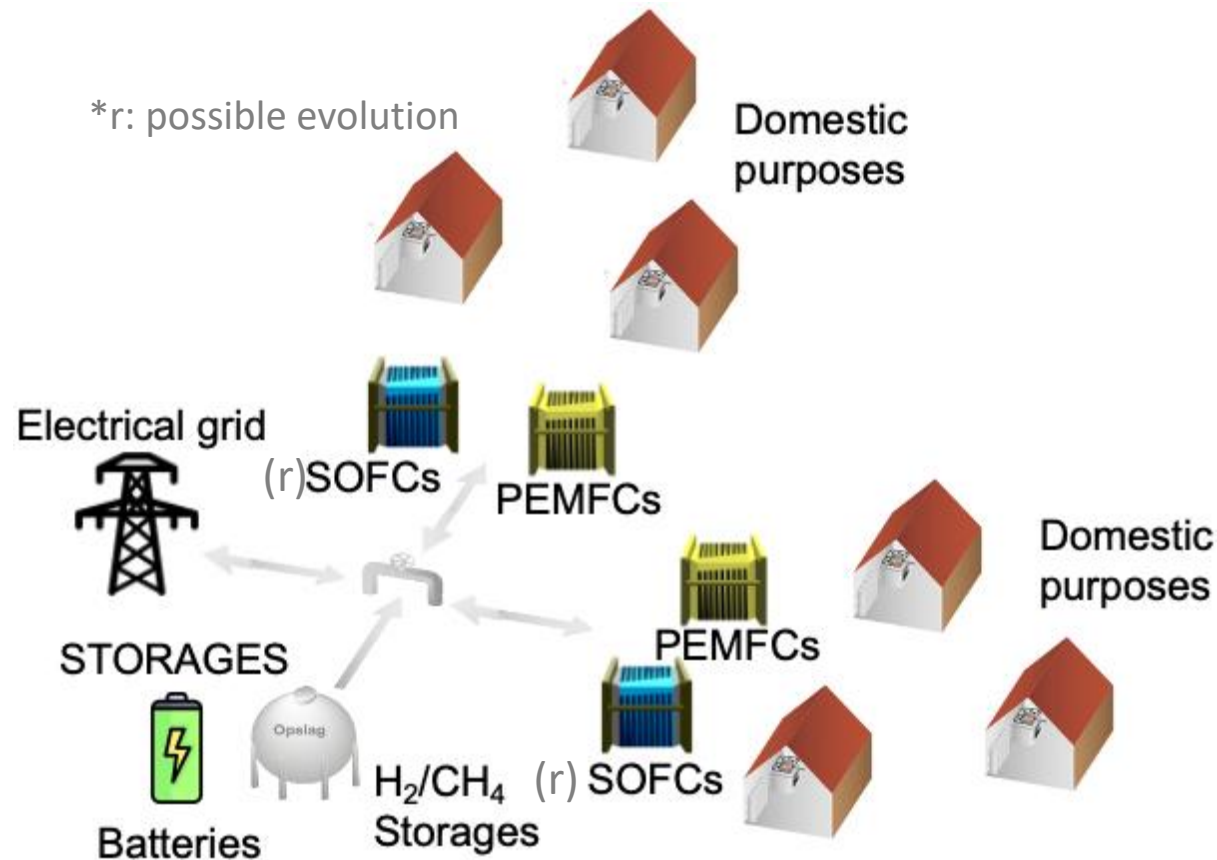
# What is a Virtual power plant (VPP)?

- A **Virtual Power Plant** is a cloud-based combination of different distributed energy systems (with innovative & traditional technologies).
- The integration of **VPPs** in the energy-market can reduce the **risks of energy shortages**, improve the overall **efficiency of the network** and enhance the **flexibility of the the grid** with respect to load and RES fluctuations.

In such environments, systems like **fuel cells, electrolyzers, reversible cells** and **RES** are perfectly suitable

- FC/Electrolysers/Reversible cells can further support the transition to more distributed energy generation paradigm.

## Practical case-study: residential complex\*

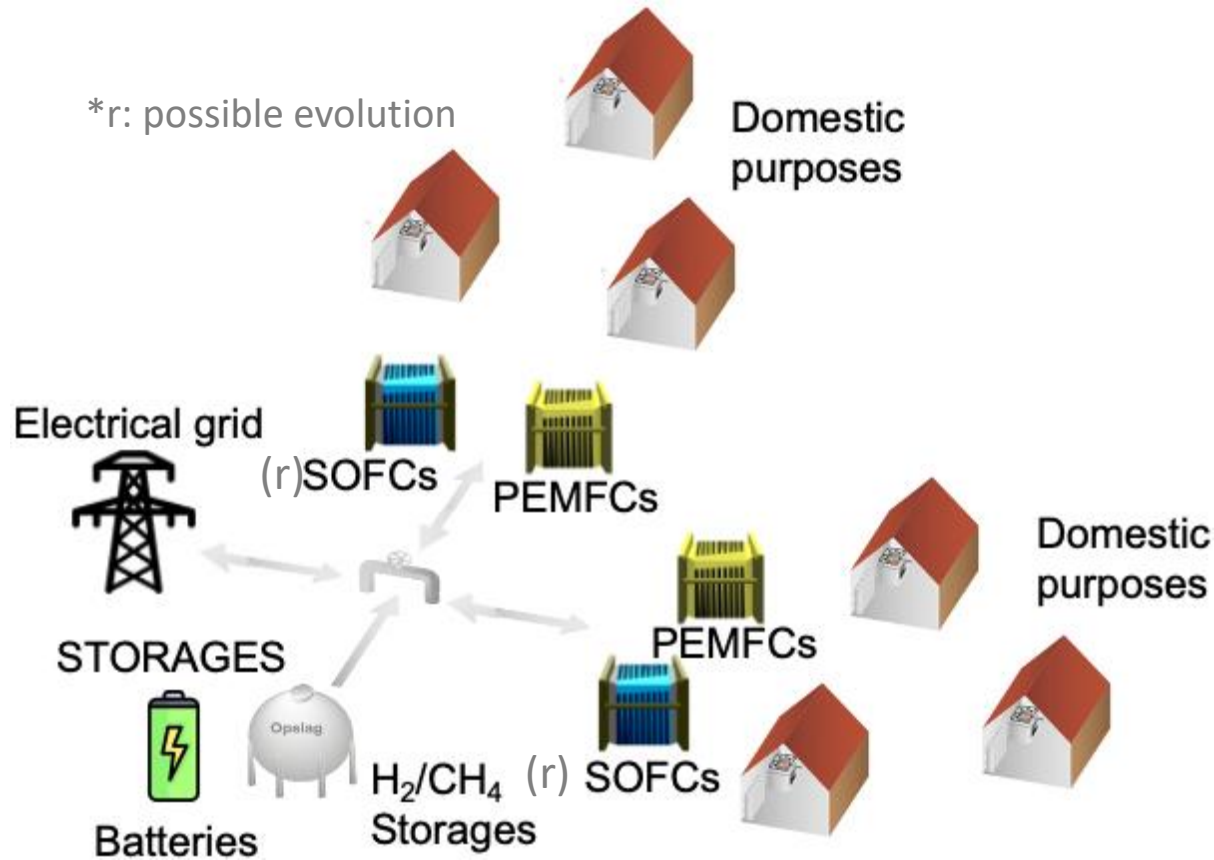


### Configuration:

- Grid-connected virtual power plant
- Both SOFCs and PEMs as energy devices systems
- Electrical storage by means of batteries as well as hydrogen/methane
- Electric and thermal loads coming from the overall residential complex (6 apartments)

In such a case the SOFC and PEM devices can be seen as VPP themselves

\* M. Califano et al. – Optimal heat and power management of a reversible solid oxide cell based microgrid for effective technoeconomic hydrogen consumption and storage. Applied Energy 319 (2022) 119268



The **MDPC** algorithm serves at:

- FCs monitoring and diagnostics
- FCs management taking advantage of both SOFCs and PEMFCs strengths
- Suitably interfacing with the smart grid requests → Optimal energy management
- FCs scheduled maintenance

Potential combination of two competitive technologies (PEM & SOFC) to:

- different or unexpected power request transients
- reliably meeting peak power demand

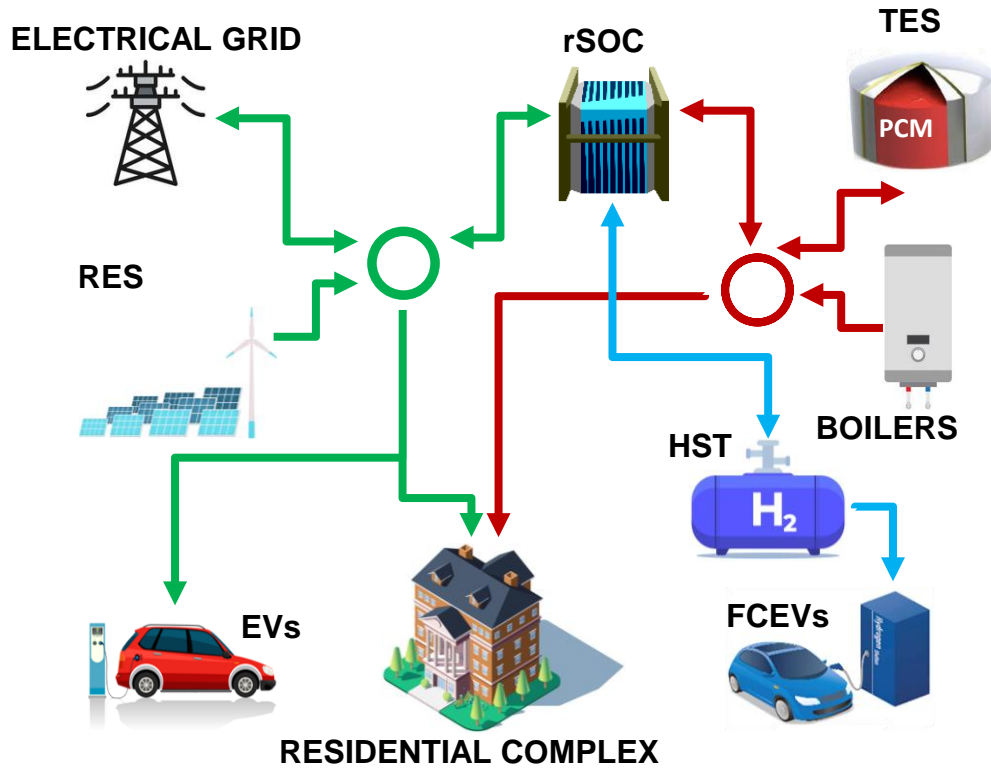
# Case study – Example of microgrid

## An all-in-one solution

- Renewables → Electricity
- rSOC + Hydrogen storage tank → Hydrogen
- Thermal energy storage → Thermal energy

## Test Case Data

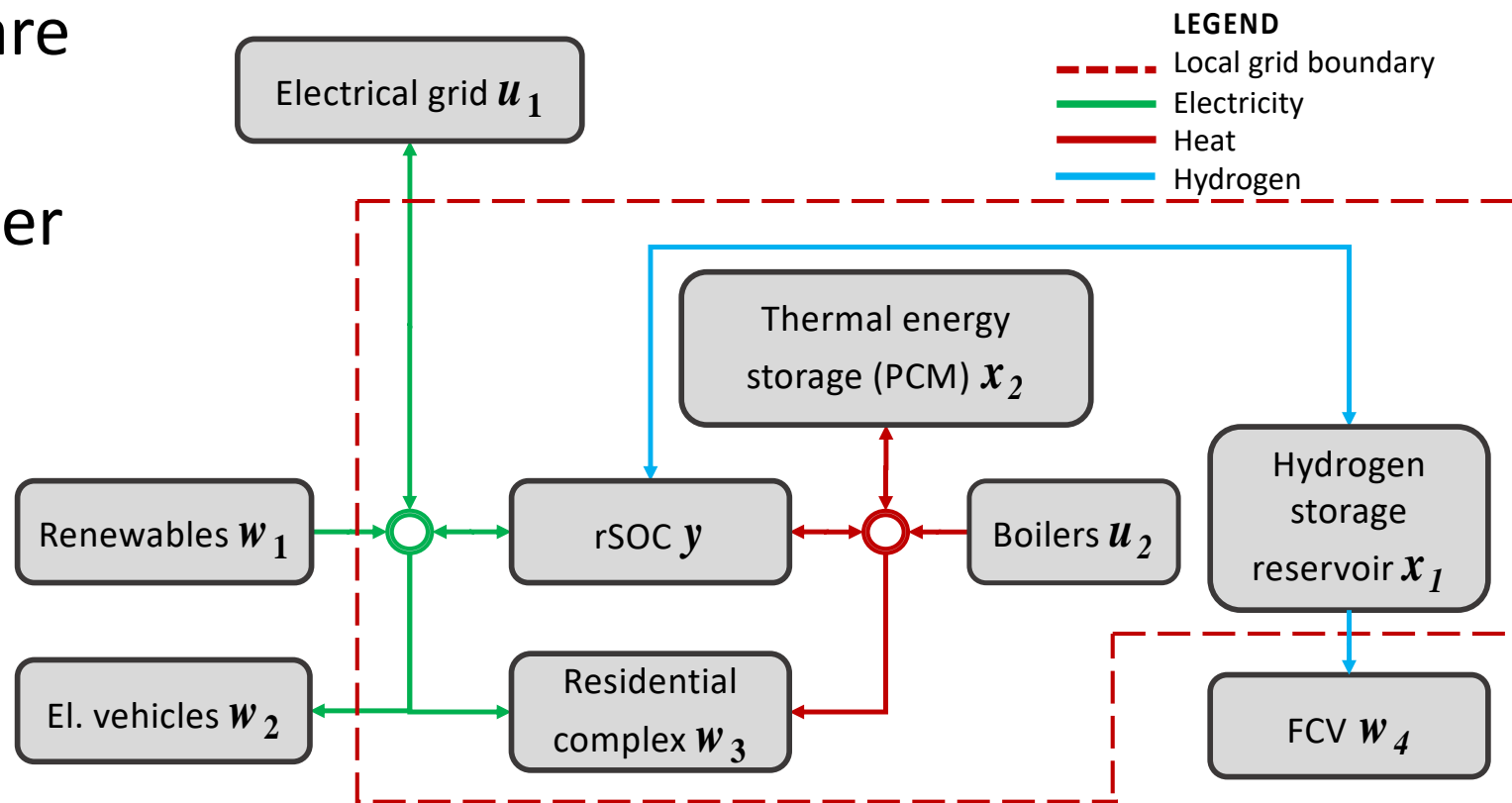
- rSOC = 15 kW
- PV = 30 kW
- Wind turbines = 41 kW
- H<sub>2</sub>ST capacity = 136 kg
- TES thermal capacity = 51 kWh



# Optimal energy management DP

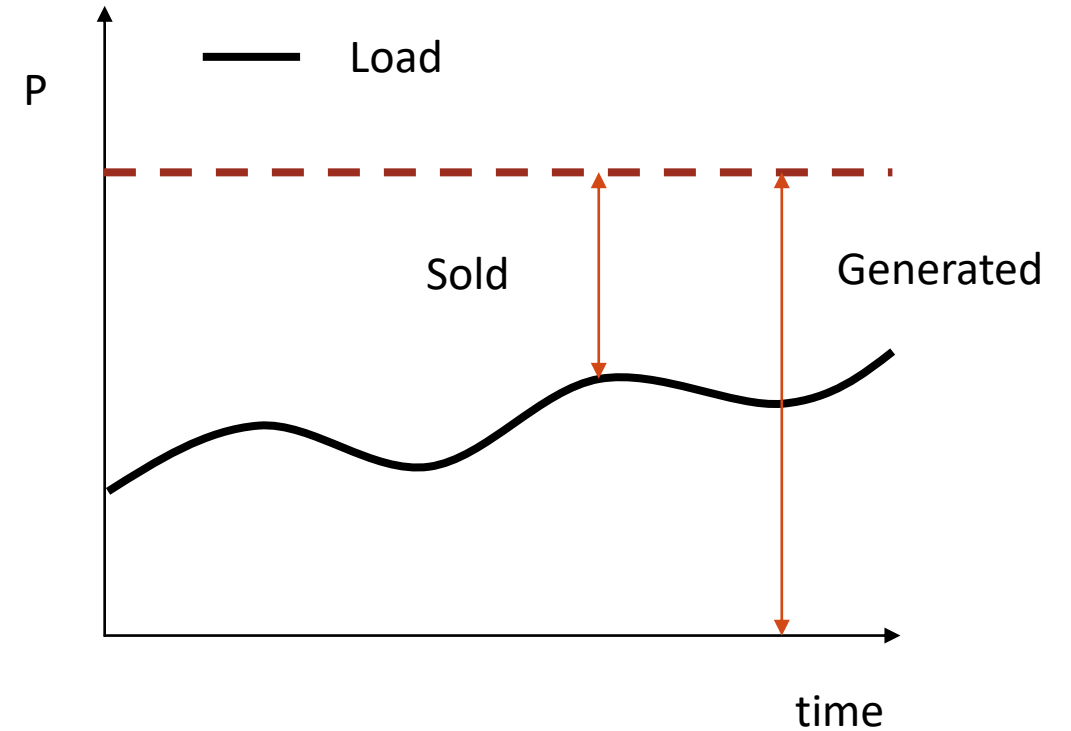
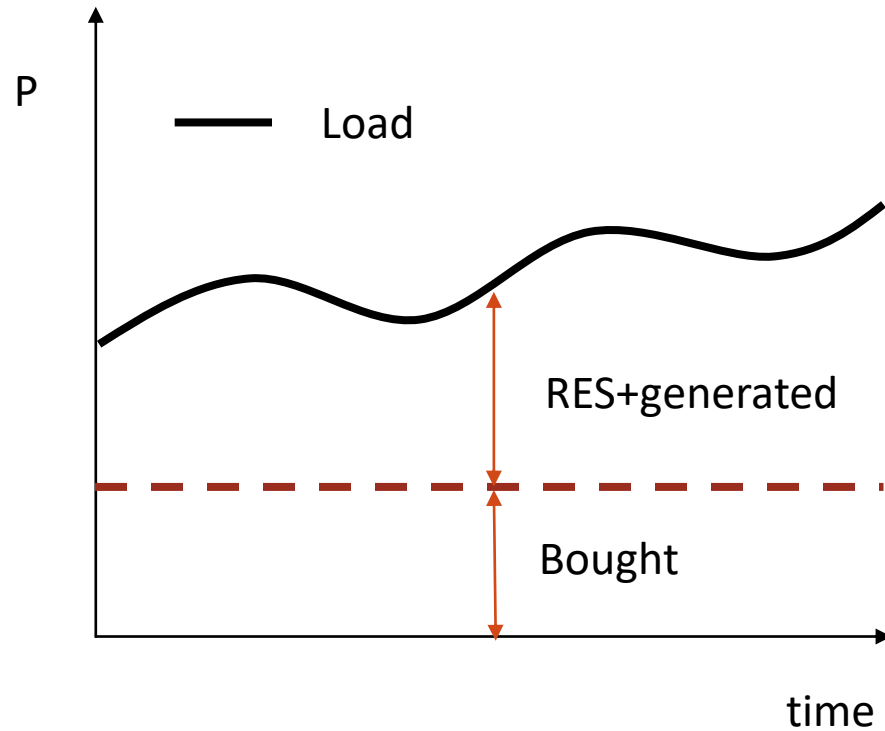
## System variables

- Loads and production (RES) are known a priori ( $w_i$ )
- **Control variables** are El. Power from the grid and Thermal energy from boilers ( $u_i$ )
- Two **state variables** parallel optimization for both H<sub>2</sub>ST and TES state of charge ( $x_i$ )
- Output variable ( $y$ ) is the rSOC electrical power



# Management concept (prosumer)

## Example of Electrical Energy (SOFC mode)



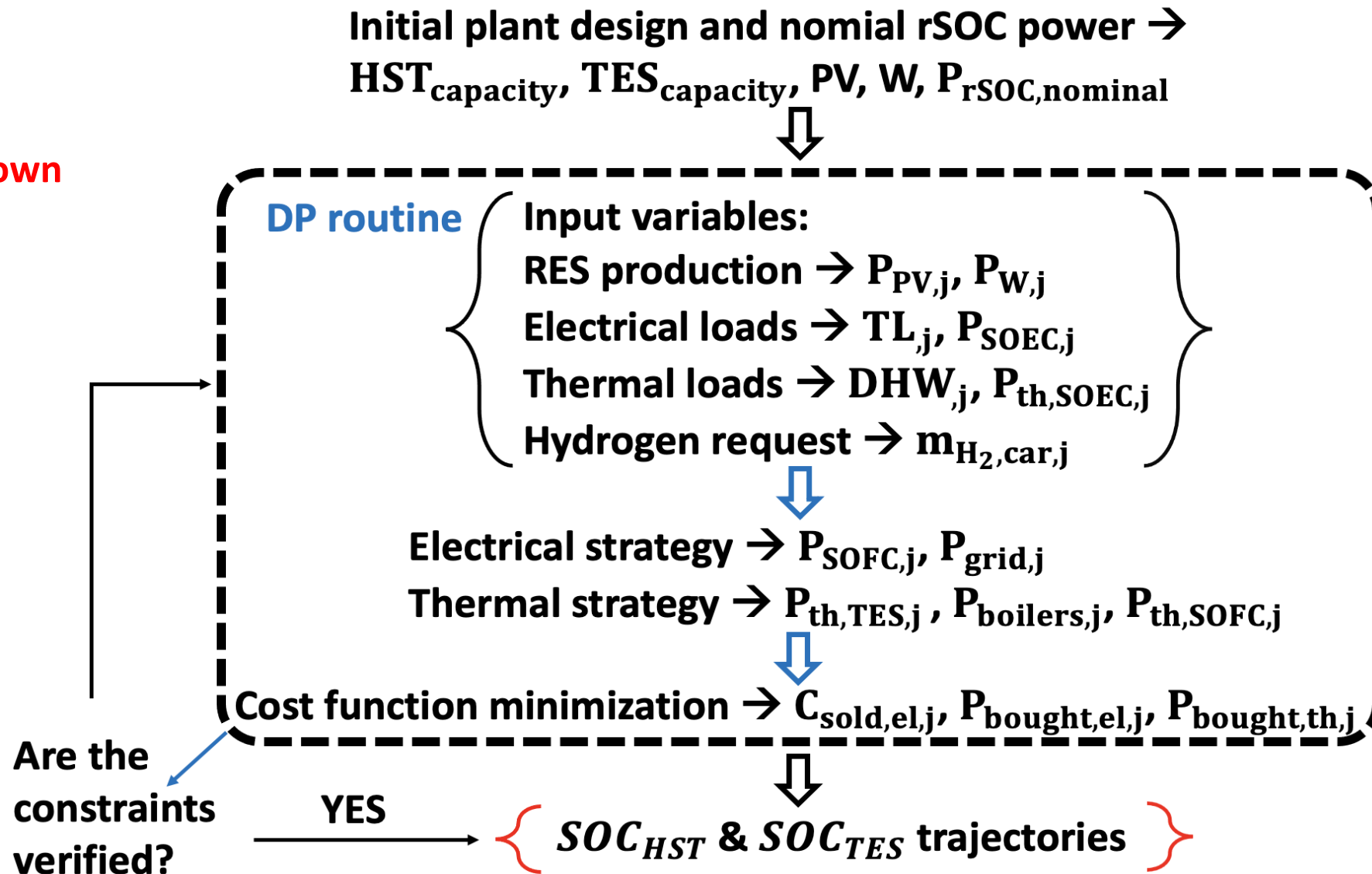
Potentially applicable to thermal energy



# DP solution scheme

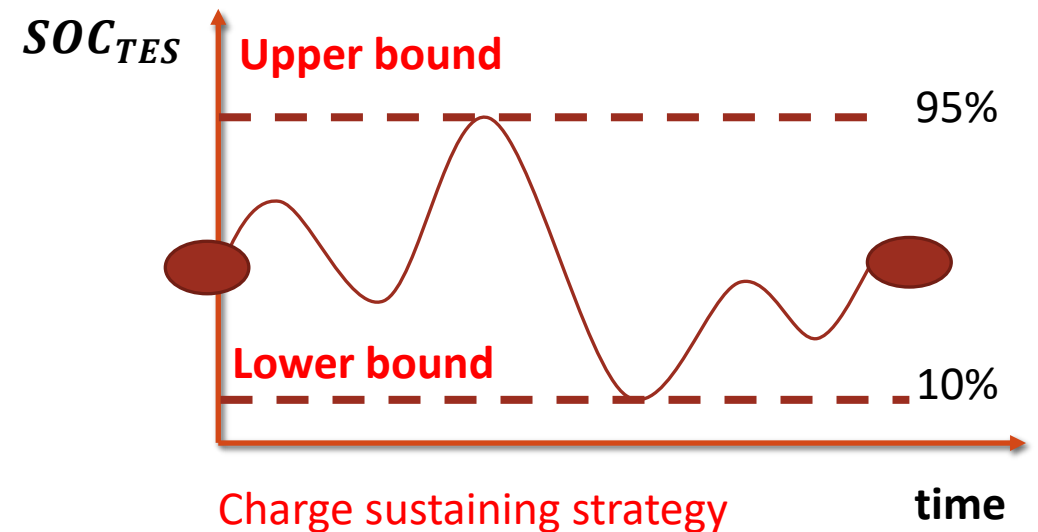
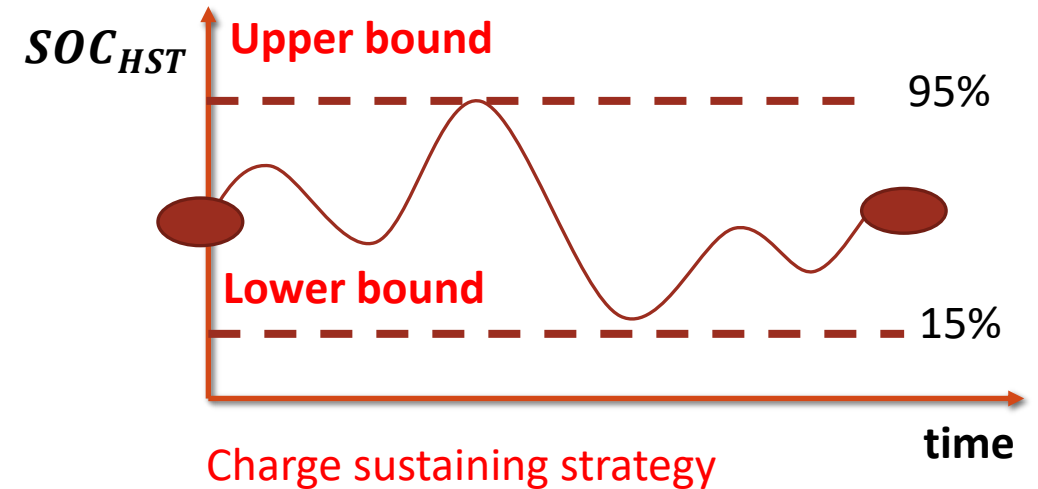
Load time profiles are known

$$\Delta t = 1 \text{ h}$$



# DP routine - State Variables

- The **first** state variable of the DP optimization is the Hydrogen Storage Tank (HST) state of charge  $\rightarrow SOC_{HST}$
- The **second** state variable of the DP optimization is the Thermal Energy Storage (TES) state of charge  $\rightarrow SOC_{TES}$
- They must be in a region identified by the upper bound and the lower bound
- The **initial and final values** must be the same



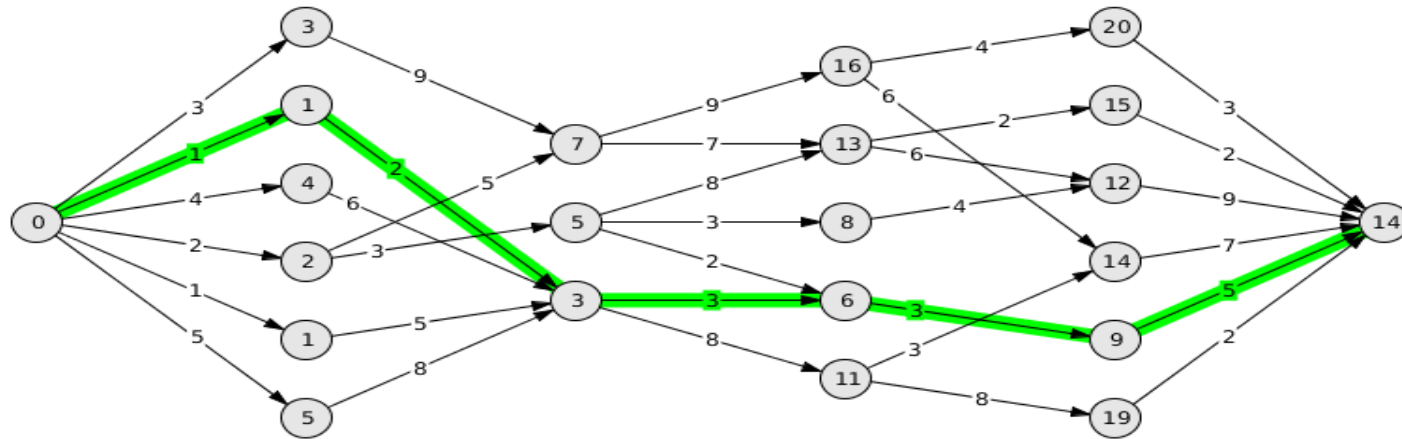
# DP routine - Cost function

**DP** minimizes the performance index that is calculated through the cost function:

$$J = \int_0^t (P_{bought} \cdot C_{bought} - P_{sold} \cdot C_{sold}) dt$$

$P_{bought}$  includes both the electric and thermal power (gas), instead  $P_{sold}$  includes only the electric power

Different values of the control variables return different state variables paths (green and black) leading to different values of  $J$ .



1-D example

**DP** provides as output the path that has the minimum value of the performance index  $J$

# DP Routine - Algorithm Scheme 1/2

**Input variables:** electric load (complex),  $H_2$  demand from FCHVs, PV and wind power production and nominal  $P_{rsoc}$

$$P_{rsoc} = (P_{el} - P_{PV} - P_W) - P_{grid}$$

1° control variable

$$m_{h2,tank} = (m_{SOEC} - m_{SOFC})$$

1° state variable

$$SOC_{HST,i+1} = SOC_{HST,i} + \frac{m_{h2,tank,i} - m_{h2,car,i}}{tank\ size}$$

# DP Routine - Algorithm Scheme 2/2

**Input variable:** thermal load (DHW), thermal power provided by rSOC, thermal power used by rSOC, nominal TES energy

$$P_{th, TES} = P_{boilers, th} + P_{th, rSOC} - DHW$$

2° control variable

*TES charging*

$$E_{th, in} = P_{th, TES} * \sqrt{RTE_{TES}} * \Delta t$$

*TES discharging*

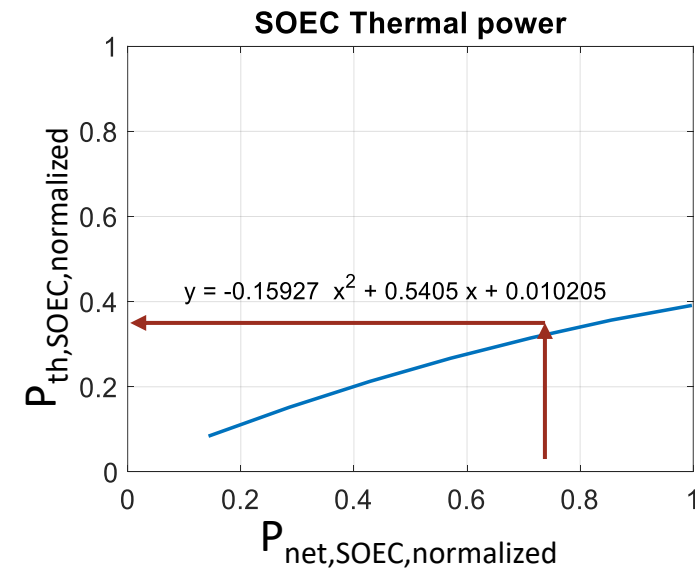
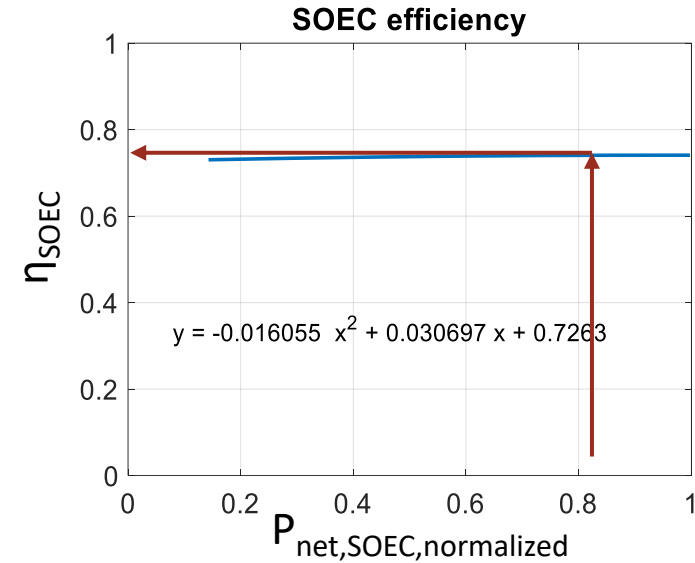
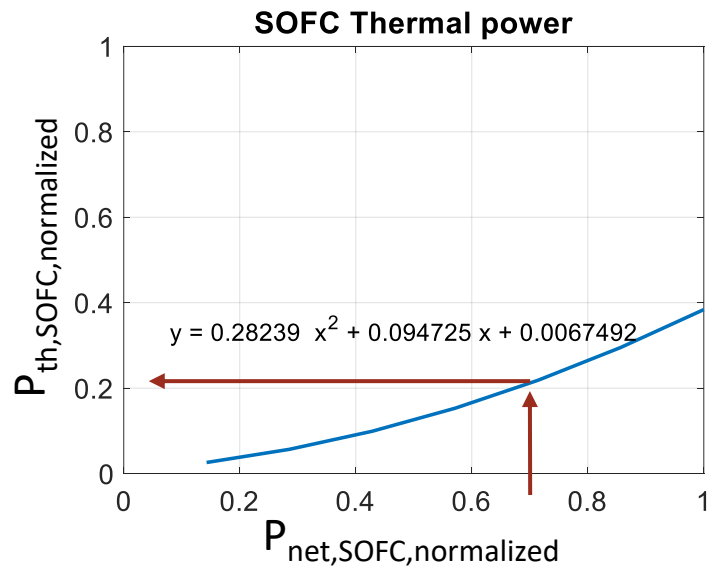
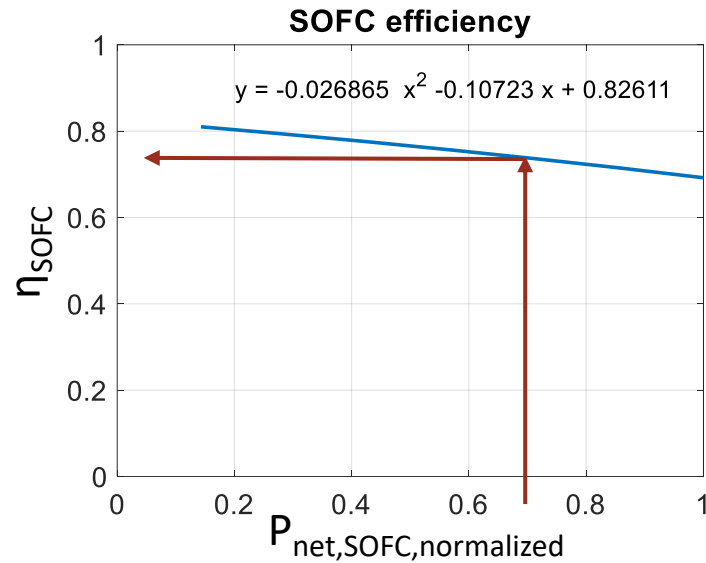
$$E_{th, out} = P_{th, TES} * \Delta t / \sqrt{RTE_{TES}}$$

2° state variable

$$SOC_{TES, i+1} = SOC_{TES, i} + \frac{E_{th, in, tank, i} - E_{th, out, tank, i}}{tank\ size}$$

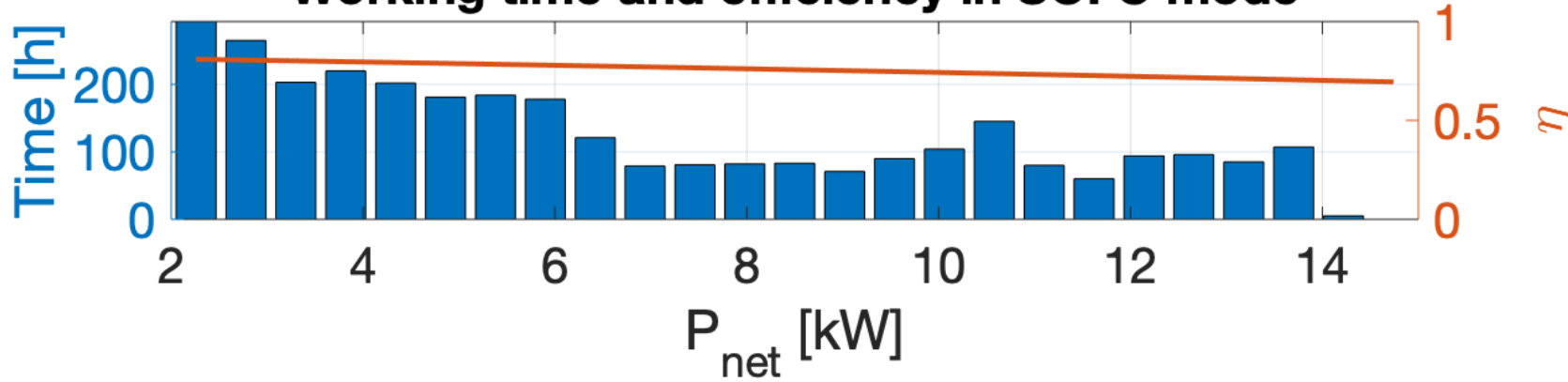
# rSOC modelling (maps\*)

\* M. Scaltritti, Master Thesis Politecnico di Milano (2016)



# Working time of rSOC in both modes

### Working time and efficiency in SOFC mode



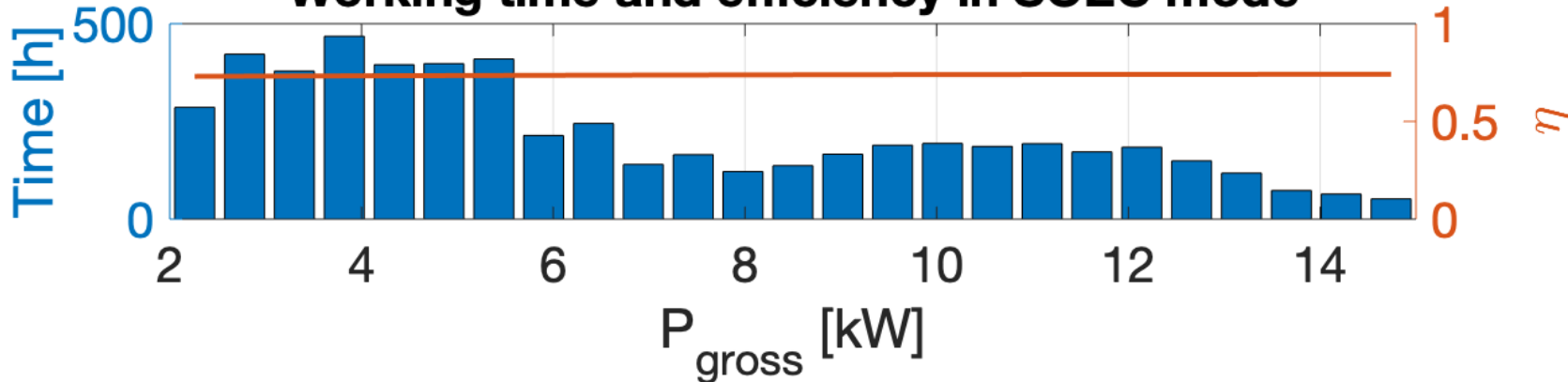
**rSOC 15 kW**

DP results\*

SOFC=3760 hours

SOEC=5000 hours

### Working time and efficiency in SOEC mode



The rSOC works 100% of the time

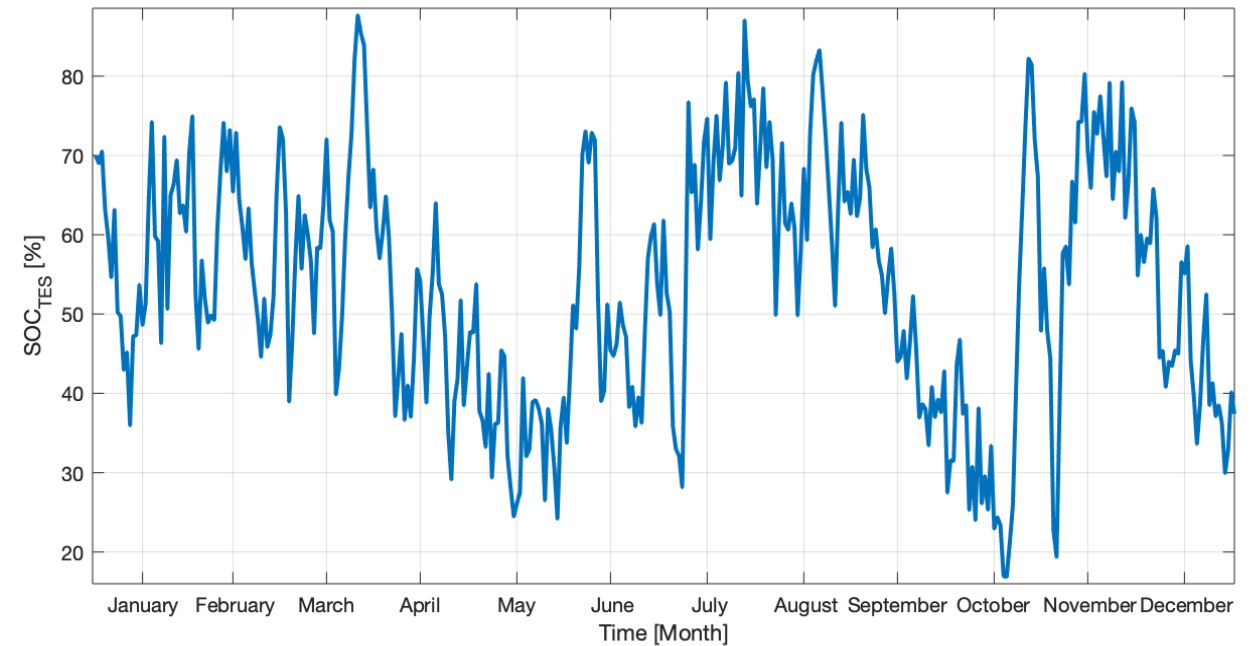
\* M. Califano et al. – Optimal heat and power management of a reversible solid oxide cell based microgrid for effective technoeconomic hydrogen consumption and storage. Applied Energy 319 (2022) 119268

## rSOC 15 kW

1° state variable ->  $SOC_{HST}$



2° state variable ->  $SOC_{TES}$

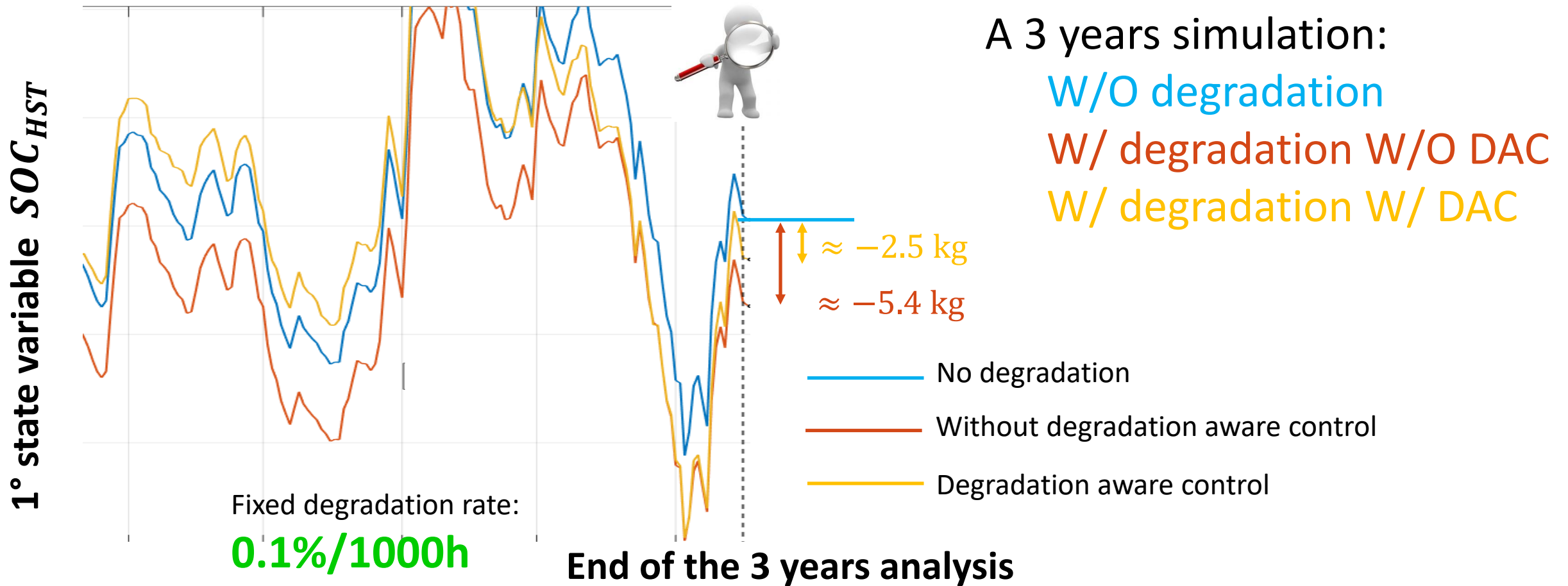


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# Degradation-Aware Control w/ MDPC

The use of a Prognostic tool (MDPC) for degradation prediction allows to optimize the control leading to a lower H2 consumption (i.e.  $x_I = \text{SOC}_{\text{HST}}$ )

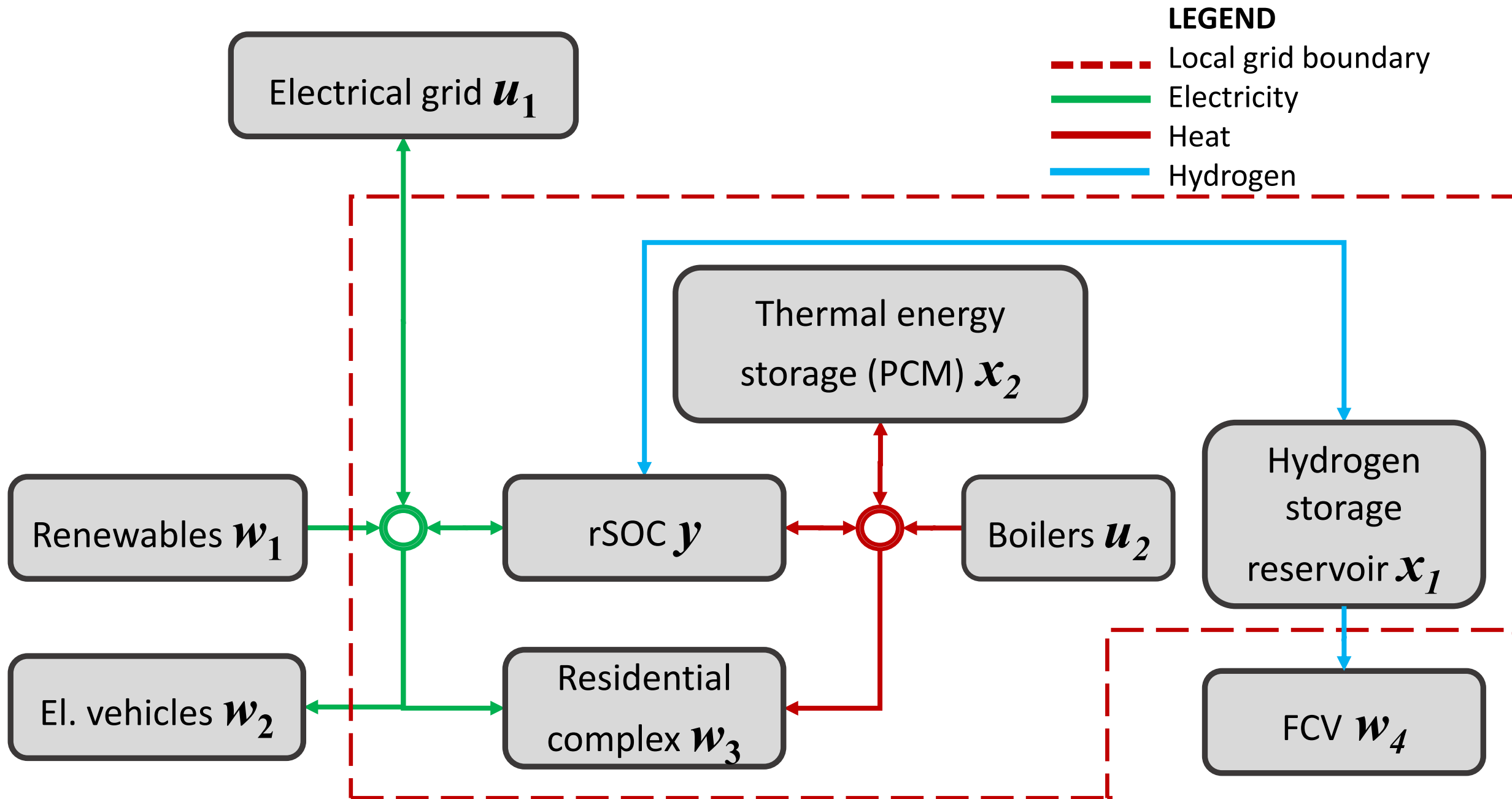


# Thanks for your attention

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This project has received funding from the Fuel Cells and Hydrogen 2 Joint Undertaking (JU) under grant agreement No 875047. The JU receives support from the European Union's Horizon 2020 research and innovation programme and Italy, France, Denmark, Slovenia, Finland, Germany, Switzerland



# Benefits of MDPC tool & VPP

RUBY outcomes will improve

- Remote monitoring
- Smart grid management &
- Virtual Power Plant-based management
- Maintenance (predictive)

Outline of the presentation

- VPP and application to a

