

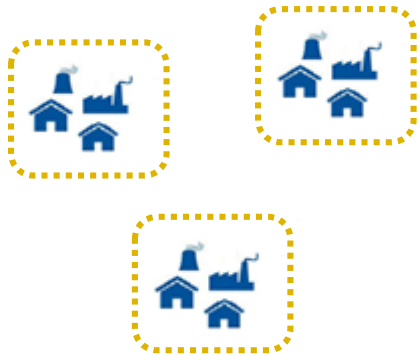


Efficacité énergétique pour les smart grids grâce à l'optimisation et à la prédiction

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The emergence of smart grids

Production islands
operated by
cities/industries



CENTRALISATION for
availability service



DECENTRALISATION for
« territory » services

Enablers:
- Renewable energies
- Internet of Things

Need for advanced
and decentralized
energy management
solutions



1900

2000

Energy management context

Energy vs. Power Management:

	Power Management	Energy Management
Scope	Stability	Energy efficiency
Time scale	ms ~ s	mins ~ hours

Energy management objectives:

- Minimize energy costs
- Reduce CO₂ emissions
- Mitigate power outages
- Improve the Quality of Service

Model Predictive Control

Principle & Ingredients

At each decision instant: Anticipate the future to find the optimal control.



Two energy management examples

- PV farm with storage



→ Flexibility for grid support

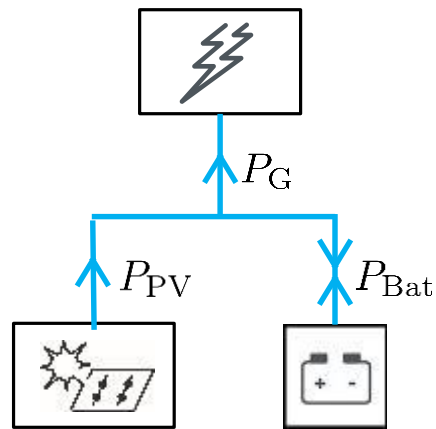
- Learning Grid by Grenoble



→ Flexibility for energy self-consumption

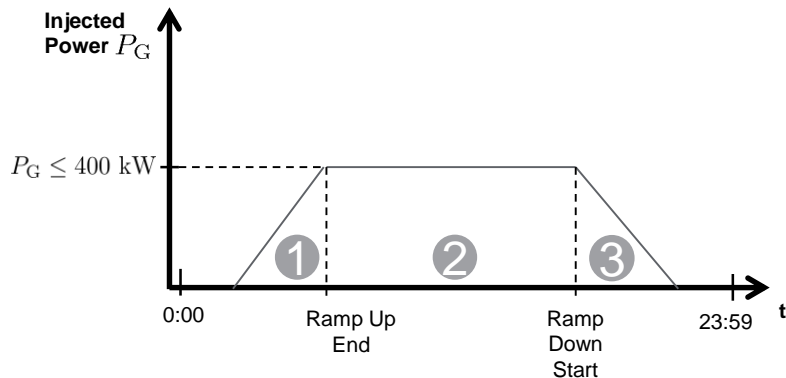
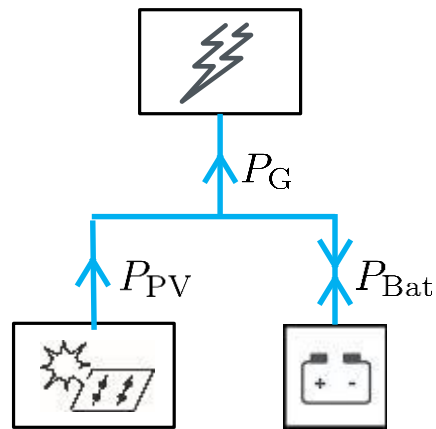
Energy context examples

- Example 1: **Langa Solar, Corsica**
- PV plant revenue maximization under regulatory constraints



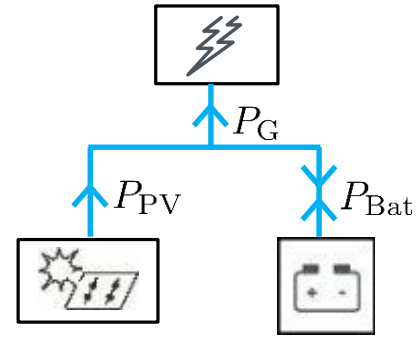
Energy context examples

- Example 1: **Langa Solar, Corsica**
- PV plant revenue maximization under regulatory constraints
- Context
 - Declare day-ahead injection profile P_G
 - Respect trapezoidal profile shape
 - Penalty if realized $P_G \neq$ declared P_G



Energy context examples

- Example 1: **Langa Solar, Corsica**
- Control problem formulation (Mixed-Integer-Linear-Programming)



$$\text{Maximize}_{P_G, P_{\text{Bat}}, \text{SoC}} \int_{t=0}^{24} P_G(t)$$

$$\text{s.t.} \quad P_G(t) = P_{\text{PV}}(t) - P_{\text{Bat}}(t)$$

$$\text{SoC}(t+1) = \begin{cases} \text{SoC}(t) + P_{\text{Bat}} \cdot \Delta t \cdot \eta_{\text{charge}} & \text{if } P_{\text{Bat}} \geq 0 \\ \text{SoC}(t) - P_{\text{Bat}} \cdot \Delta t \cdot \eta_{\text{discharge}} & \text{if } P_{\text{Bat}} < 0 \end{cases}$$

$$0 \leq \text{SoC}(t) \leq \text{SoC}_{\text{max}}$$

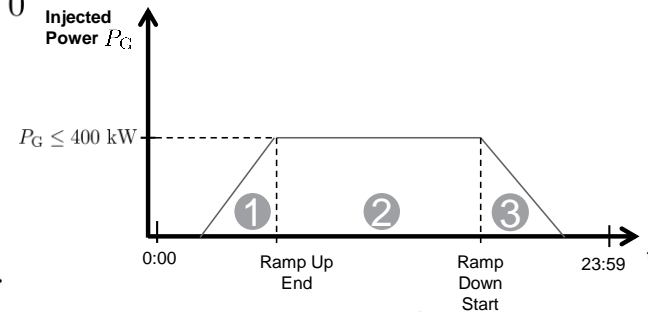
$$-P_{\text{Bat,max}} \leq P_{\text{Bat}}(t) \leq P_{\text{Bat,max}}$$

$$0 \leq P_G(t) \leq P_{\text{G,max}}$$

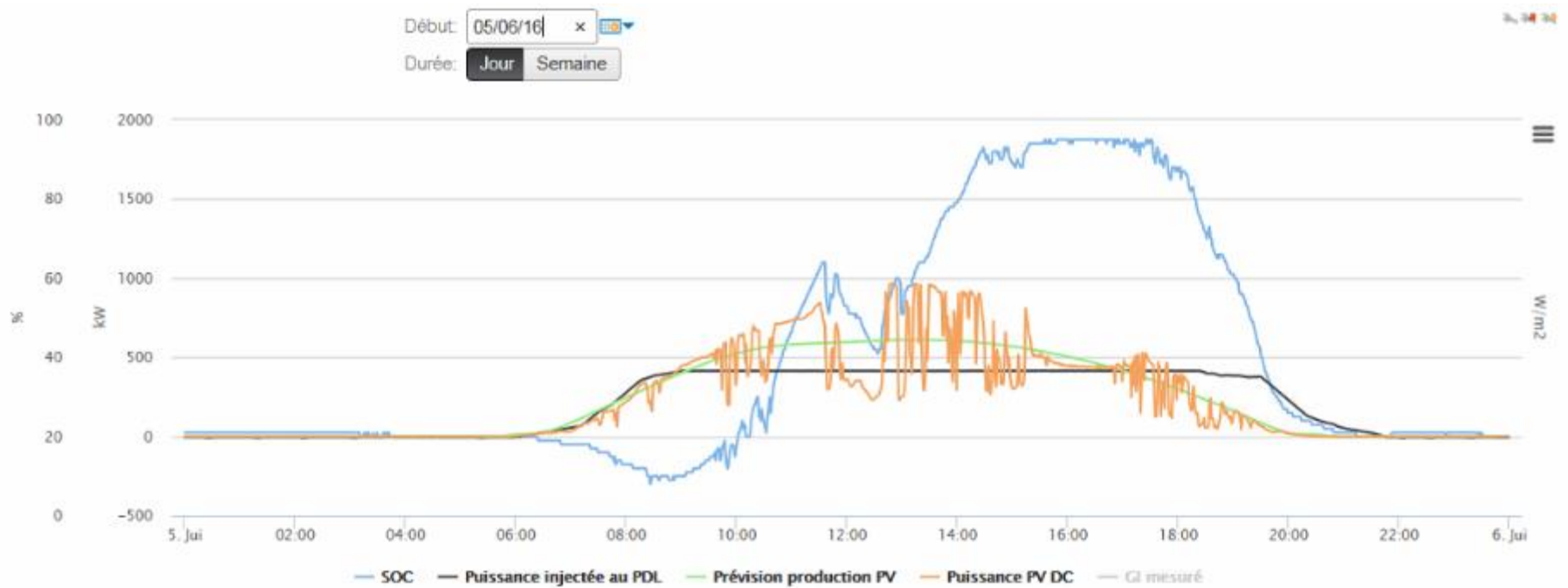
$$|P_G(t+1) - P_G(t)| \leq 6 \text{ kW/min}$$

Several logical constraints to distinguish between the three phases.

SoC	State of Charge
η	battery efficiency
Δt	time step



Screenshot from the Langa Solar monitoring system



Energy context examples

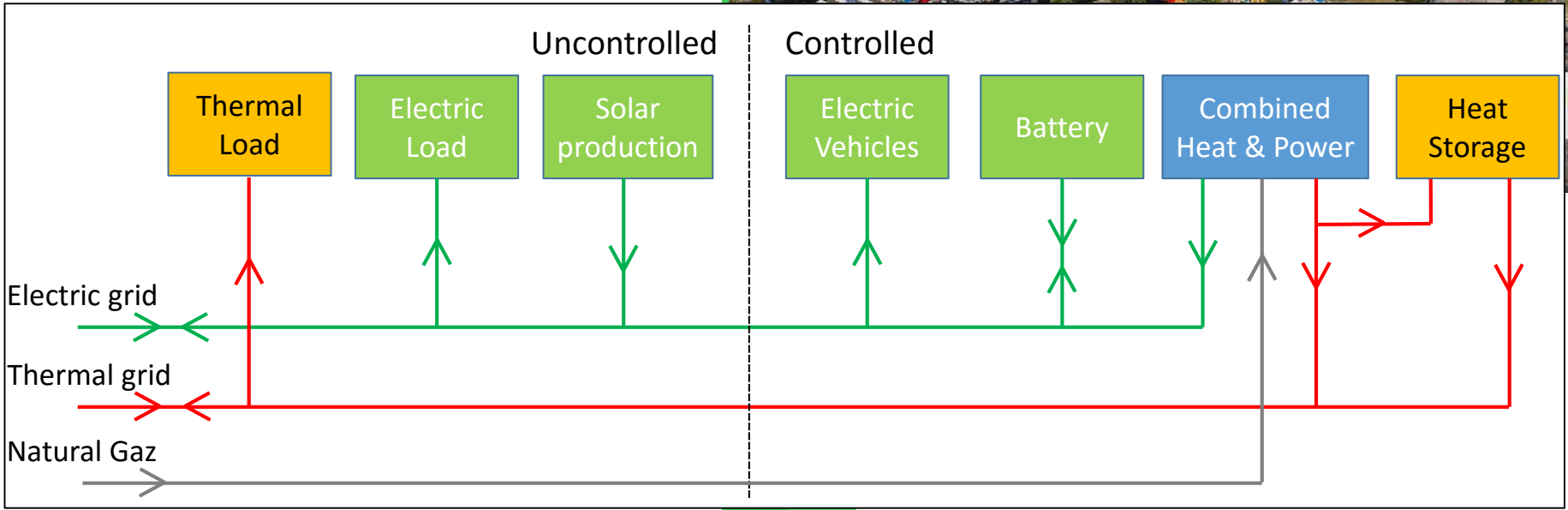
- Ex. 2: **Learning Grid by Grenoble**
- Multi-energy campus
 - Electricity- and District heating network
 - Batteries
 - Solar panels
 - CHP (Combined heat and power plant)
 - EV charging station



- Use flexibilities for local purposes
 - Maximize local usage of renewables
 - Minimize energy costs (based on variable energy tariffs)

Energy context examples

- Ex. 2: Learning Grid by Grenoble



Conclusion

- **Model Predictive Control** is a powerful method for **energy management** applications



14 partners, 10 M€



ARROWHEAD

80 partners, 68 M€

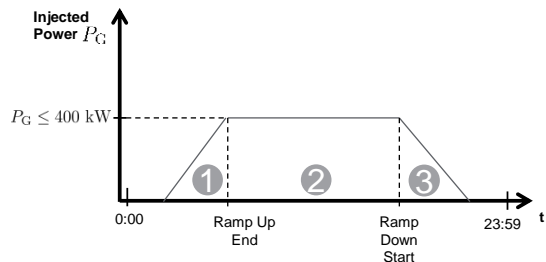
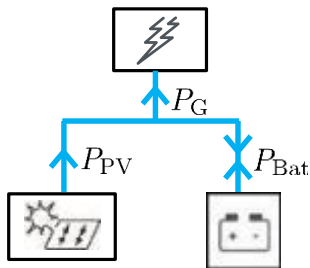


- **Reliable forecasts** are a crucial prerequisite to obtain good performances

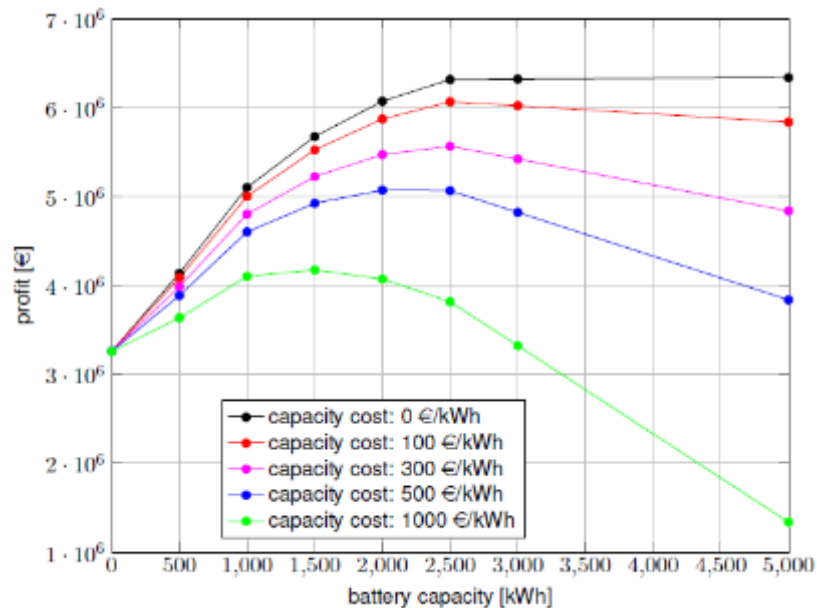
Conclusion

Challenges:

- **Optimal sizing** of smart grid components must **integrate advanced control solutions**



- In **multi-owner use cases**, the **economic model for flexibility** will have a strong impact on the control/optimization model



Thank you!

Life Is On

