

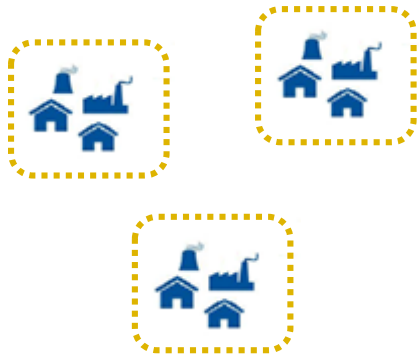


# Efficacité énergétique pour les smart grids grâce à l'optimisation et à la prédiction

Peter Pflaum

# The emergence of smart grids

Production islands  
operated by  
cities/industries



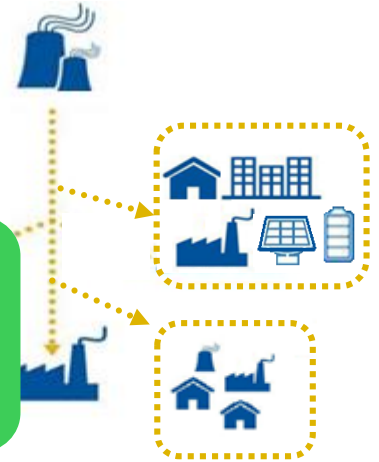
CENTRALISATION for  
availability service



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:

	<b>Power Management</b>	<b>Energy Management</b>
Scope	Stability	Energy efficiency
Time scale	ms ~ s	mins ~ hours

## Energy management objectives:

- Minimize energy costs
- Reduce CO<sub>2</sub> 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.



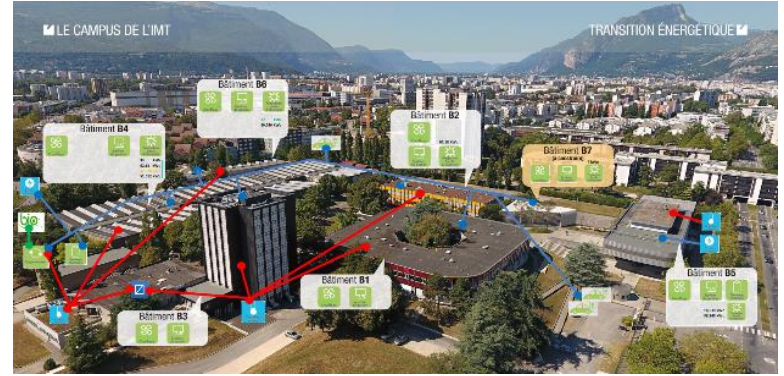
# Two energy management examples

- PV farm with storage



→ Flexibility for grid support

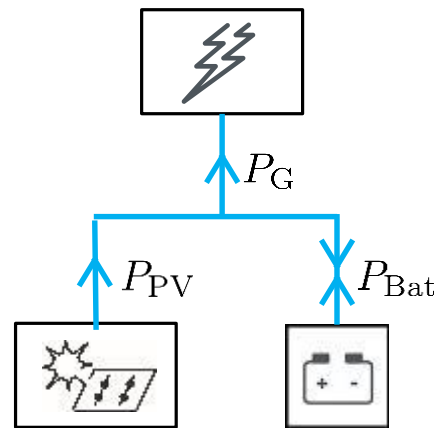
- Learning Grid by Grenoble



→ Flexibility for energy self-consumption

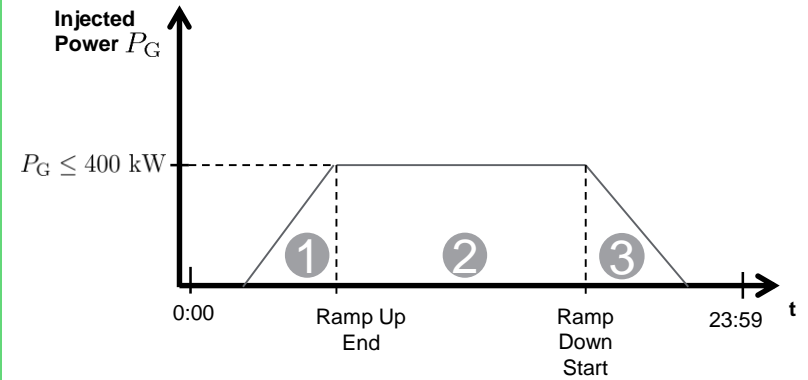
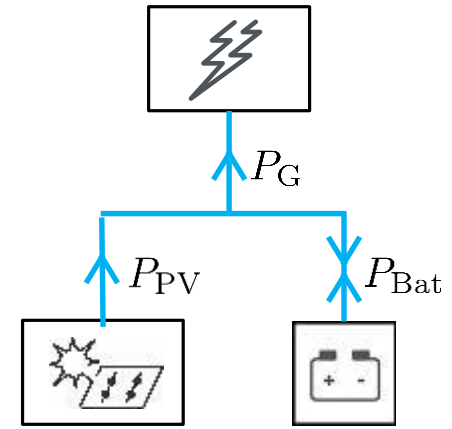
# Energy context examples

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



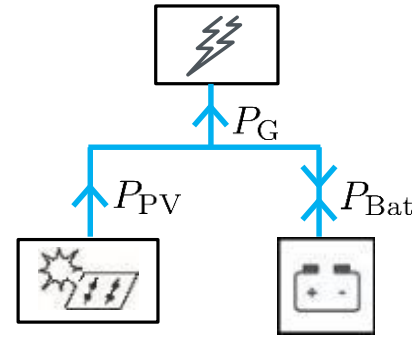
# Energy context examples

- 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)



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

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

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

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

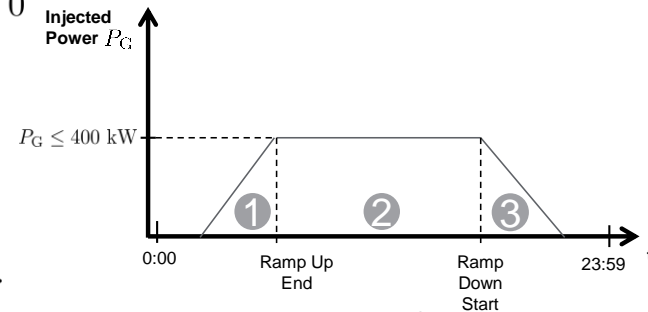
$$-P_{\text{Bat,max}} \leq P_{\text{Bat}}(t) \leq P_{\text{Bat,max}}$$

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

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

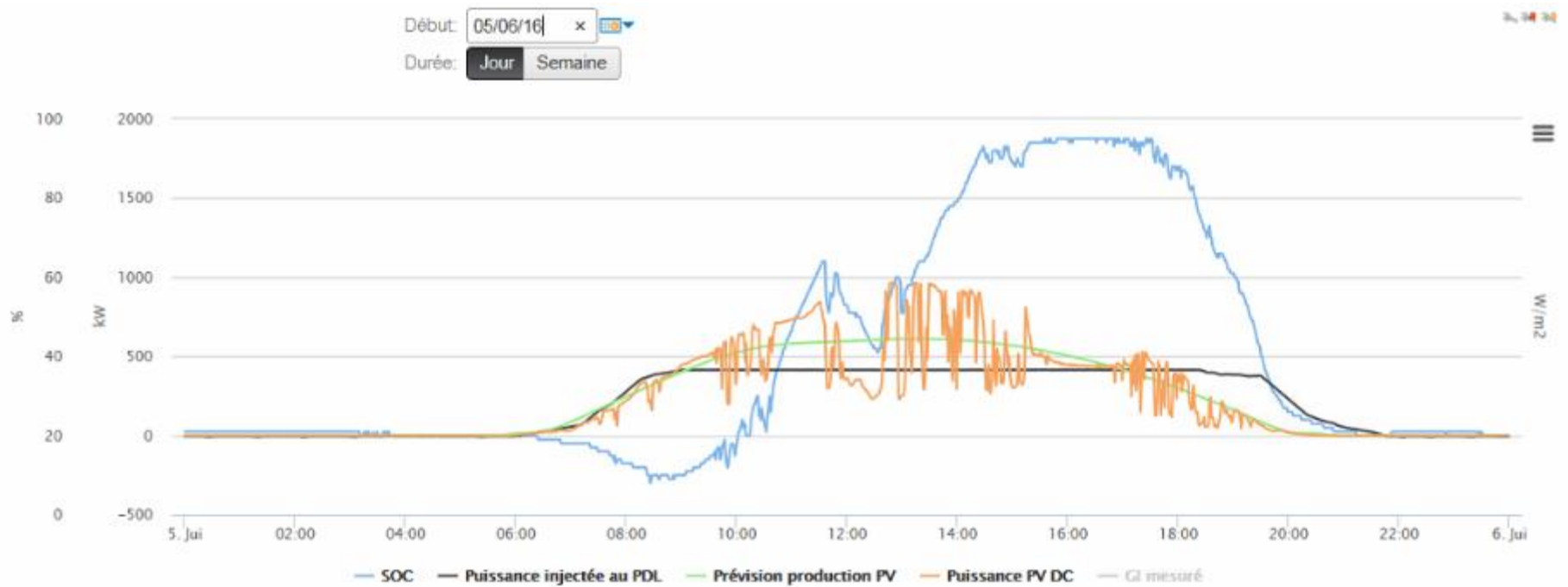
Several logical constraints to distinguish between the three phases.

$\text{SoC}$	State of Charge
$\eta$	battery efficiency
$\Delta t$	time step





# Screenshot from the Langa Solar monitoring system



# Energy context examples

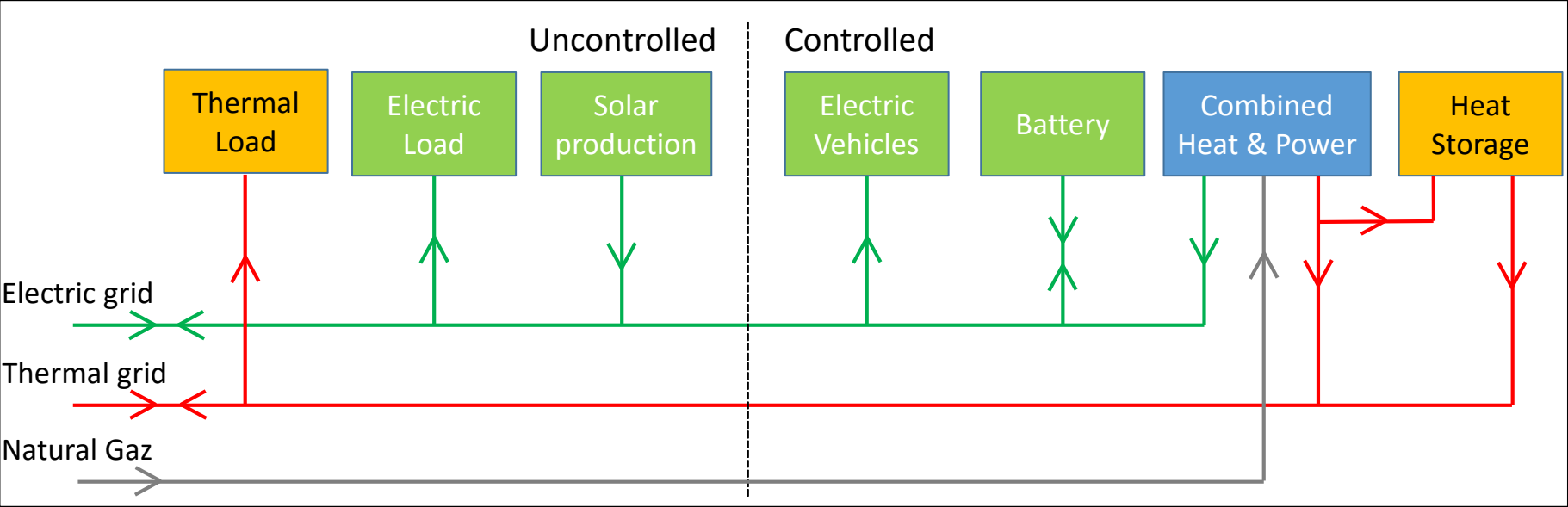
- Ex. 2: **Learning Grid by Grenoble**
- Multi-energy campus
  - Electricity- and District heating network
  - Batteries
  - Solar panels
  - CHP (Combined heat and power plant)
  - EV charging station



- 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€

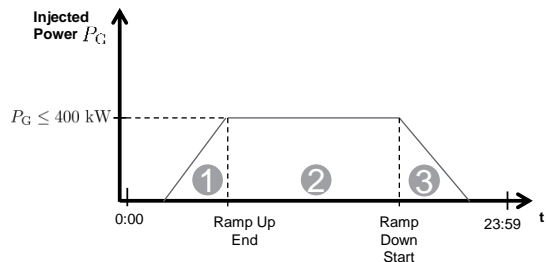
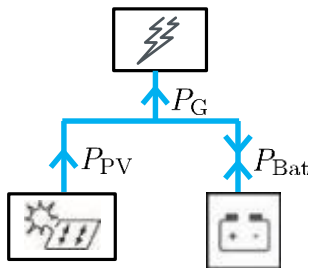


- **Reliable forecasts** are a crucial prerequisite to obtain good performances

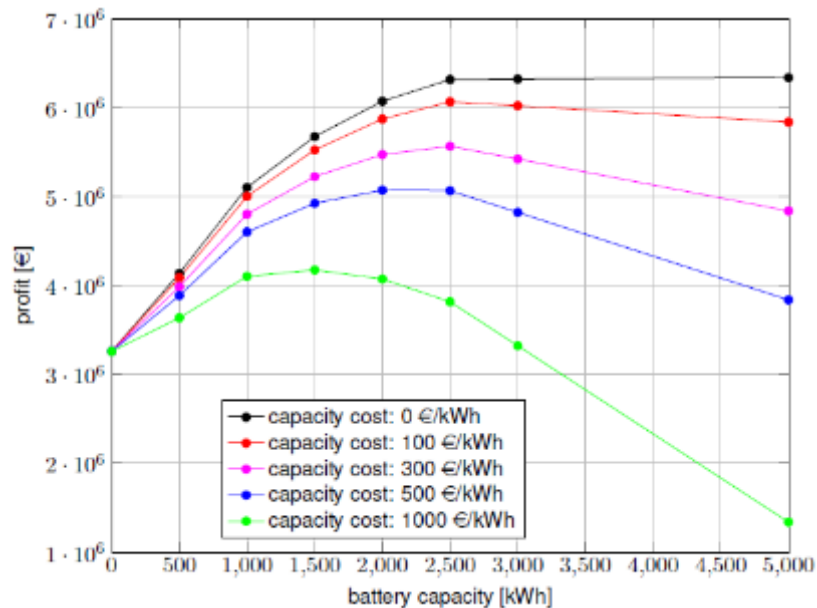
# Conclusion

## 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

**Schneider**  
Electric