IMITATIVE LEARNING FOR ONLINE PLANNING IN MICROGRIDS

INCENTIVES

- Load (e.g. house) proximity ;
- Drop in the cost of PV panels.



LINEAR DYNAMICS

$$\forall t \in \{1 \dots T\}, \sigma \in \Sigma, T \in N:$$

- $a_t^{\sigma,+}, a_t^{\sigma,-} \rightarrow \text{Storage system } \sigma \text{ actions };$
- $s_t^{\sigma} = s_{(t-1)}^{\sigma} + a_{t-1}^{-,\sigma} + a_{(t-1)}^{+,\sigma}, s_0^{\sigma} = 0 \rightarrow$ Storage system σ state ;
- $F_t = d_t \sum_{\sigma \in \Sigma} \left(\frac{a_t^{+,\sigma}}{\eta^{\sigma}} + \eta^{\sigma} a_t^{-,\sigma} \right) \rightarrow$ Power cut. $d_t = prod_t - cons_t$ is the net demand. η^{σ} is storage system σ efficiency.

Objective function (operational costs)

Levelized Cost of Energy : $LEC = \frac{\sum_{t=1}^{T} \frac{-\sum_{\psi \in \Psi} k_t^{\psi} F_t^{\psi}}{(1+r)^{y'}} + I_0}{\sum_{y=1}^{n} \frac{\epsilon_y}{(1+r)^y}} \text{ where }$

- $I_0 \rightarrow$ Initial investment cost ;
- $k_t^{\psi} F_t^{\psi} \to \text{Cost of consumption not met for}$ load $\psi \in \Psi$ at time $t \in \{1 \dots T\}$.

Linear programming Minimization of objective function with contraints related to the dynamics \rightarrow Optimization of planning strategies given a complete scenario.

SAMY AITTAHAR, VINCENT FRANÇOIS-LAVET, STEFAN LODEWEYCKX, DAMIEN ERNST AND RAPHAËL FONTENEAU UNIVERSITY OF LIÈGE, DEPARTMENT OF ELECTRICAL ENGINEERING AND COMPUTER SCIENCE

FULLY OFF-GRID MICROGRID















IMITATIVE LEARNING

 Compute optimal sequences of actions from production and consumption scenarios (lin-



• Build smart online planning agent with op-

• Benchmarking of others machine learning

con-

• Transfer learning (i.e. adaptation of an ex-