

# A Demand-Response Framework in Balance Groups through Direct Battery-Storage Control

**G. Chasparis**

M. Pichler

T. Natschläger

Software Competence Center Hagenberg GmbH, Austria

Naples, Italy  
26th June 2019

# Demand Response in Residential Buildings



- **Recently, we observe**

- A constant increase in the number of battery storage systems.
- A need for larger absorption rates of renewable energy

- **We need:**

- to incentivize users to provide “*flexibility*”
- to effectively *coordinate* flexibility extraction

# Demand-Response Approaches

- **Commitment-based approaches**

- The users commit to *reduce load* during peak-hours
- The operator *distributes* the desired aggregated demand to the users  
(e.g., Ruiz et al (2009), Chen et al. (2014))

# Demand-Response Approaches

- **Commitment-based approaches**

- The users commit to *reduce load* during peak-hours
- The operator *distributes* the desired aggregated demand to the users  
(e.g., Ruiz et al (2009), Chen et al. (2014))

- **Incentive-based approaches:**

- The users provide *preferences* over availability and cost functions
- The operator computes optimal flexibility extraction that *minimize cost*  
(e.g., Herter (2007), Triki & Violi (2009), Xu et al (2016))

# Demand-Response Approaches

- **Commitment-based approaches**

- The users commit to *reduce load* during peak-hours
- The operator *distributes* the desired aggregated demand to the users  
(e.g., Ruiz et al (2009), Chen et al. (2014))

- **Incentive-based approaches:**

- The users provide *preferences* over availability and cost functions
- The operator computes optimal flexibility extraction that *minimize cost*  
(e.g., Herter (2007), Triki & Violi (2009), Xu et al (2016))

- **Combination of the two (*demand-response aggregation*)**

- The operator *directly* extracts flexibility when necessary
- In return, the operator offers to the owners a compensation  
(e.g., Parvania et al (2013), Iria et al (2017), Nan et al (2018))

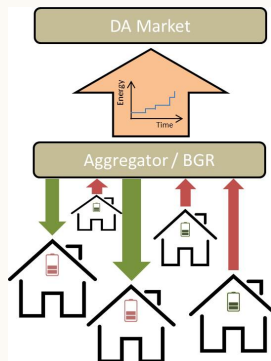
## Demand-Response Aggregation: *Requirements & Objective*

- *Features/requirements:*

- optimizing *flexibility* extraction
- optimizing over a *future time horizon*
- optimizing over *multiple households*

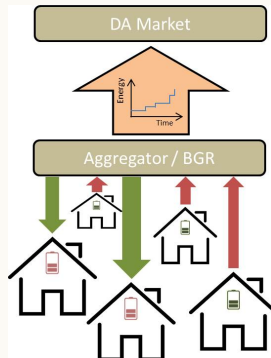
- *Objective*

- Respond to *load reduction/increase goals*
- Participate in a *wholesale* electricity spot-market (e.g., Day-ahead, Intra-day)



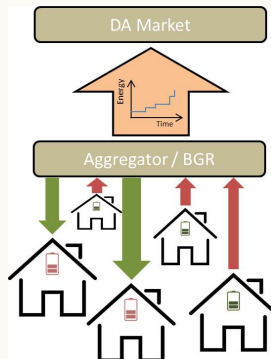
# Demand-Response Aggregation: *State-of-the-art & Challenges*

- So far, *analysis*
  - is primarily restricted to *single* households  
(e.g., Mohsenian-Rad (2016), He et al (2016), Jiang & Power (2016))
  - includes *detailed modeling* of the battery and cycle costs  
(e.g., He et al (2016))
  - has taken into account *uncertainty* and *imperfect forecasts*  
(e.g., Jiang & Power (2016))
- *Challenges*,
  - *computationally efficient* methods for a coordinated response of a pool of battery-storage systems
  - *dynamic-programming scheduling problem* with equilibrium constraints



## Setup and Goal of this paper

- *Austrian market organization*
  - *balance-group* (BG) organization
  - every prosumer is part of a BG
  - feed-in and off-take are *cleared* in 15min
- *Incentives for strategic behavior*
  - (O1) *Market participation*: The BG may exchange electricity in a spot-market
  - (O2) *Imbalance optimization*: Reduce current imbalances to zero, or generate imbalances
- *Goal* of this paper:
  - *formalize* these two (instantaneous) optimization problems
  - *provide building blocks* for a large scale *dynamic* optimization

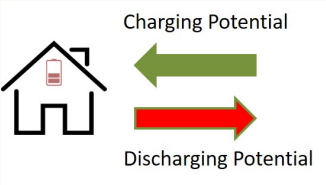


# Outline

- 1 Introduction
- 2 Energy Potential
- 3 Activation Costs
- 4 Centralized Optimal Activation
- 5 Summary & Future Work

# Outline

- 1 Introduction
- 2 Energy Potential**
- 3 Activation Costs
- 4 Centralized Optimal Activation
- 5 Summary & Future Work

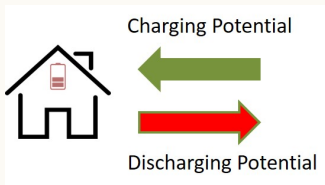


- **Charging (power) potential:**

$$V_{c,i}(t) \doteq \overline{P_{g,i}}(t) - P_{g,i}(t) \geq 0$$

- $\overline{P_{g,i}}(t)$  : maximum (positive) power from the grid
- $P_{g,i}(t)$  : power from the grid under baseline operation

## Energy Potential



## Energy Potential

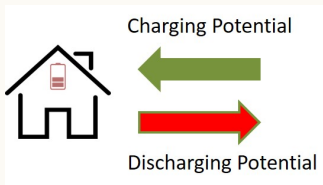
Energy potential or flexibility of a household  $i$ , for a time interval  $\Delta T$ , is defined as the amount of energy that can be “**charged**” to or “**discharged**” from the household.

- **Discharging (power) potential:**

$$V_{d,i}(t) \doteq \underline{P}_{g,i}(t) - P_{g,i}(t) \leq 0$$

- $\underline{P}_{g,i}(t)$  : maximum (negative) power to the grid
- $P_{g,i}(t)$  : power from the grid under normal operation

## Energy Potential



## Energy Potential

Energy potential or flexibility of a household  $i$ , for a time interval  $\Delta T$ , is defined as the amount of energy that can be “*charged*” to or “*discharged*” from the household.

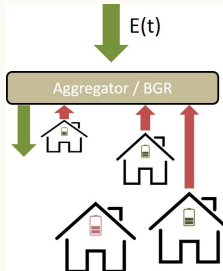
- **Example:** When the battery is currently *charging* with rate  $c_i(t) \geq 0$ , the charging (energy) potential is:

$$V_{c,i}(t) = \underbrace{c_{\max,i} \cdot \min\{\Delta T_{c,i}^*(t), \Delta T\}}_{\text{maximum charging energy}} - \underbrace{c_i(t) \cdot \min\{\Delta T_{c,i}(t), \Delta T\}}_{\text{current charging energy}}$$

# Outline

- 1 Introduction
- 2 Energy Potential
- 3 Activation Costs**
- 4 Centralized Optimal Activation
- 5 Summary & Future Work

# Opportunity Costs



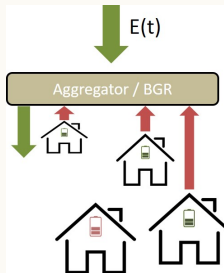
## Opportunity Costs

Independently of the way that the BRP is currently utilizing the battery, the opportunity costs of activation can be defined as follows:

$$C_{\text{act},i}(P'_{g,i}(t)) = \underbrace{U_{\text{base},i}^*(t)}_{\text{optimal baseline operation}} - \underbrace{U_{\text{BRP},i}(P'_{g,i}(t))}_{\text{BRP intervention}}$$

under the selected by the BRP power exchange with the grid.

# Opportunity Costs (cont.)



## Opportunity Costs: *Revenues under Baseline Operation*

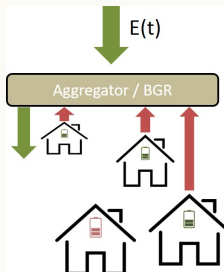
What would have been the revenue when the user maximizes its own utility:

$$U_{\text{base},i}^*(t) = \max_{P_{g,i} \in [\underline{P}_{g,i}, \overline{P}_{g,i}]} U_{\text{base},i}(P_{g,i})$$

where

$$U_{\text{base},i}^*(P_{g,i}) \doteq U_{\text{e.sell},i}(P_{g,i}) + U_{\text{b.store},i}(P_{g,i}) - C_{\text{e.buy},i}(P_{g,i}) - C_{\text{b.loss},i}(P_{g,i}) - C_{\text{b.wear},i}(t)(P_{g,i}).$$

# Opportunity Costs (cont.)

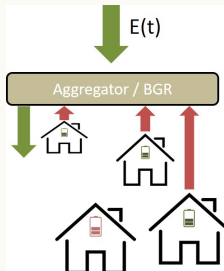


## Opportunity Costs: *Revenues under BRP intervention*

What is the revenue of the household when the BRP intervenes:

$$U_{\text{BRP},i}(P'_{g,i}) = U_{\text{e.sell},i}(P'_{g,i}) + U_{\text{b.store},i}(P'_{g,i}) - C_{\text{e.buy},i}(P'_{g,i}) - C_{\text{b.loss},i}(P'_{g,i}) - C_{\text{b.wear},i}(P'_{g,i}),$$

# Generic Activation Costs



## Generic Activation Costs

The user submits a linear activation curve that values its own preferences over the use of the battery. For example, when the BRP wants to establish a positive imbalance:

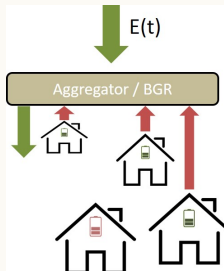
$$C_{\text{act},i}(t) = \alpha_i(t) \cdot V_{c,i}(t) \cdot \beta_{c,i}(t)$$

under the selected by the BRP power exchange with the grid, for some positive constant  $\beta_i(t)$  and activation parameter  $\alpha_i(t) \in [0, 1]$ .

# Outline

- 1 Introduction
- 2 Energy Potential
- 3 Activation Costs
- 4 Centralized Optimal Activation**
- 5 Summary & Future Work

# Generic Activation Costs



## Market Participation

We need to compute the optimal subset of participants and their schedules to generate a specific commitment  $E(t) > 0$  at time interval  $t$ . In the case of generic costs,

$$\begin{array}{ll} \min & \mathbf{z}_c(t)^T \mathbf{a} \\ \text{s.t.} & V_c(t)^T \mathbf{a} = E(t) \\ \text{var.} & \mathbf{a} \in [0, 1]^N \end{array}$$

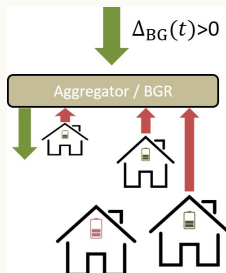
# Market Participation Optimization (cont.)

## Optimal activation for $E(t) > 0$

```

1: procedure OPTIMALACTIVATION( $V_c, \mathbf{z}_c, E(t)$ )
2:   order participants  $i = 1, 2, \dots, N$  as follows
                                      $\beta_{c,1} \leq \beta_{c,2} \leq \dots \leq \beta_{c,N}$ 
3:   for  $i = 1, 2, \dots, N$  do
4:     if  $i \leq k^*$  then
5:        $\alpha_i^* = 1$ 
6:     else
7:       if  $i = k^* + 1$  then
8:          $\alpha_i^* = (E(t) - \sum_{j=1}^{i-1} V_{c,j}) / V_{c,i}$ 
9:       else
10:         $\alpha_i^* = 0$ 
11:   return  $\mathbf{a}^*$ 
  
```

# Imbalance Optimization



## Market Participation

We need to compute the optimal subset of participants to minimize a current imbalance, say  $\Delta_{BG} > 0$

$$\begin{aligned} \min \quad & -\mathbf{z}_d(t)^T \mathbf{a} + \lambda_{imb}(t) (\Delta_{BG}(t) + V_{d,i}(t)^T \mathbf{a}) \\ \text{var.} \quad & \mathbf{a} \in [0, 1]^N \end{aligned}$$

# Outline

- 1 Introduction
- 2 Energy Potential
- 3 Activation Costs
- 4 Centralized Optimal Activation
- 5 Summary & Future Work**

# Summary

## Summary:

- We described a framework for *activation optimization* in balance groups.
- Essential building blocks:
  - energy potential calculation
  - opportunity/activation costs
- *Recall*: This is an optimization that applies to single interval  $\Delta T$ .

## Future Work:

- Day-ahead optimization for spot-market participation
- Upcoming paper in DACH+ Energy Informatics 2019

# Snap-shot of Day-ahead Optimization

