

C-DAX Support & Validation for EV Charging Services





C-DAX is funded by the European Union's Seventh Framework Programme (FP7-ICT-2011-8) under grant agreement n° 318708

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Smart EV charging overview

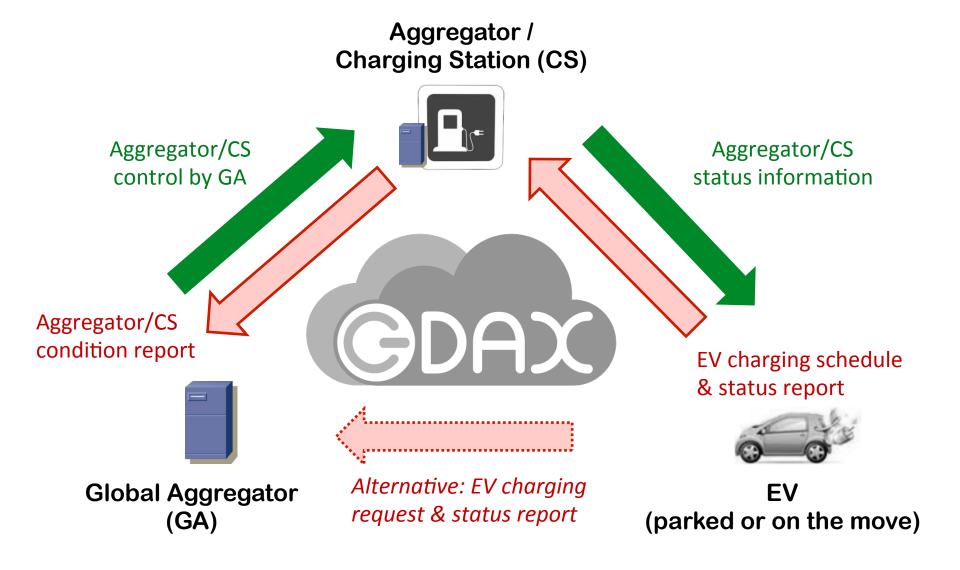
- Use case objectives
 - Design an efficient C-DAX platform for EV charging services
 - For the grid: Efficient demand response control, Energy load balancing
 - For the EVs: Optimised service performance (e.g., reduced charging waiting time)
- Charging scenarios with EV grid communication:
 - Parked: Smarter charging for optimal grid operations and user experience
 - On-the-move: Ensure optimal selection of a charging station
- Main actors
 - Aggregator: global aggregator (GA) and/or at charging station level
 - EVs either in parking mode or on the move







Smart EV charging overview









Why is C-DAX suitable?

Flexible support of different communication modes

Mode	Scenario Example		
1-to-1	GA controls aggregators/CSs for information publication		
M-to-1	Aggregator/CS condition update to GA		
M-to-1 / 1-to-M	Aggregator/CS information publication to EVs		

Security functionality

- C-DAX key management to defend against malicious parties
- Where necessary, encryption techniques should be used (e.g., shared privacy sensitive information such as EV charging preferences)

Scalable and flexible platform

To add and remove clients: Simply join/leave the relevant topics







Part I: Parked EV charging

Demand Response for parked EVs Outline

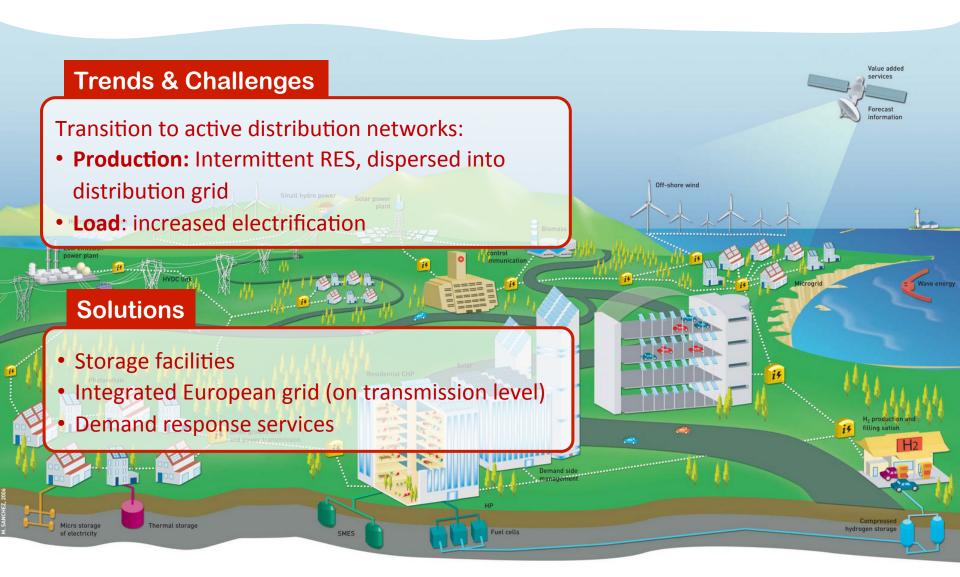
- Introduction to demand response and smart charging
- Example use case for smart charging
- Smart charging
 - System overview
 - Information flows
 - Deployment on C-DAX
 - Performance analysis
- Conclusions







Introduction: Demand Response (1)









Introduction: Demand Response (2)

Algorithms focus on distributed and hybrid approaches

- Scalability: computational, memory, and communication benefits
 - Computational: Computational complexity is spread over users (i.e., many small and simple models versus one large and complex model)
 - Memory: Reduced model complexity (e.g., constraints, decision variables)
 - Communication: Localize communication intensive parts
- Reliability: avoid single point-of-failure, limited to one way communication, ...
- Privacy: users no longer need to share privacy sensitive information and hand over control to a third party

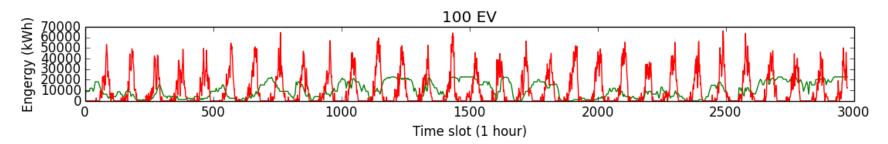






Example scenario: Wind balancing

- Aim: demand-supply balancing by exploiting flexibility of EVs
 ... instead of grid storage, additional generating power, etc.
 - Supply: Power generated by wind turbine
 - Demand: EV charging
- Exploiting flexibility
 - Always respect user constraints (e.g., fully charged when departing)
 - Time shifting: Control action = turning charger on/off
 - Demand shaping: Adapt charging power (in [0, P_{max}])



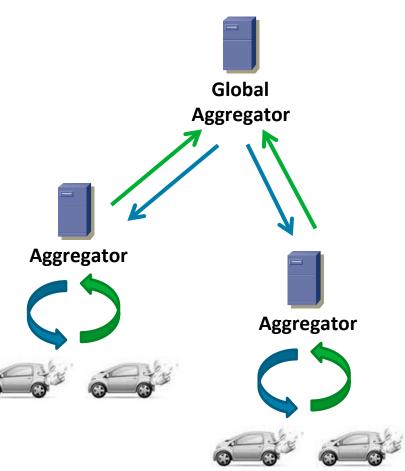






System Overview

- Information sharing
- Incentive (control) signal



Global Aggregator

- Manage system objective (e.g., balancing)
- Process flexibility info
- Create target profiles

Aggregator

- Manage user & system objectives
- Collect, aggregate, and forward flexibility info.
- Negotiate charging plans
- Manage sessions.
- Localize communication and information sharing

Electric vehicles

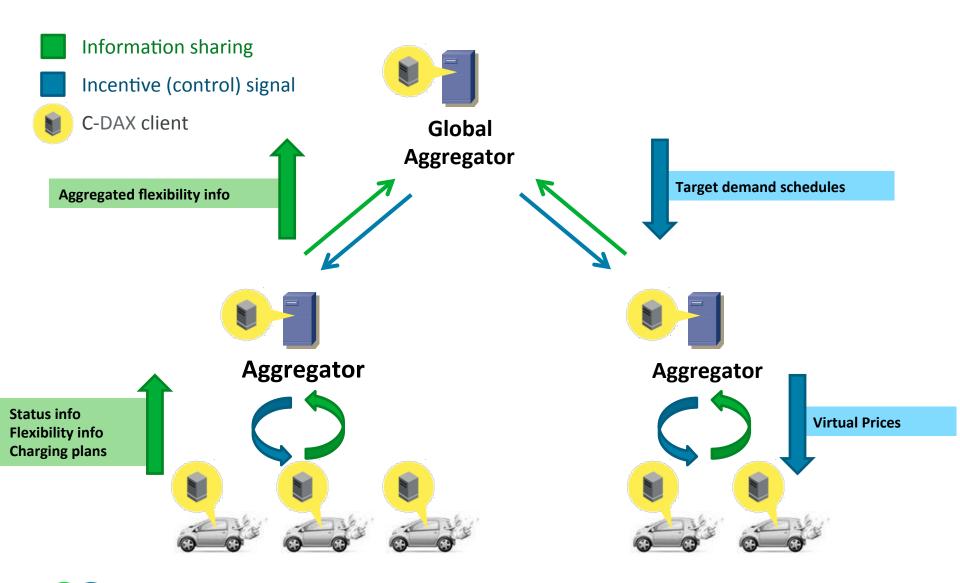
- Provide flexibility info
- Negotiate and execute charging plans







Information Flows

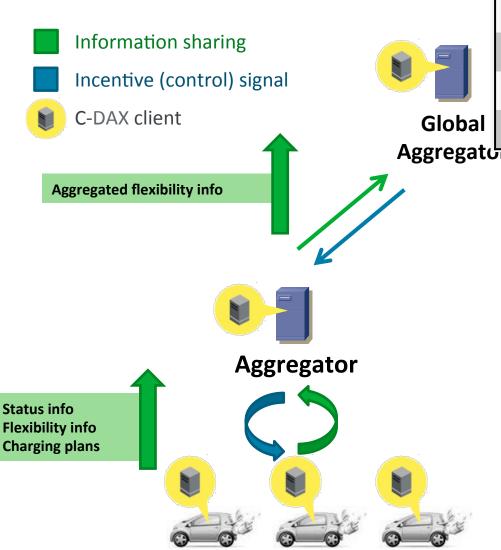






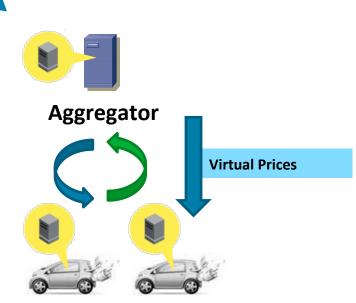


Information Flows



Topic	Mode	Pub	Sub
Target sched.	1-to-1	GA	Aggr.
Aggregator flex.	M-to-1	Aggr.	GA
Virtual prices	1-to-M	Aggr.	EV
Charging plans	M-to-1	EV	Aggr.
Flexibility info	M-to-1	EV	Aggr.
71			

Target demand schedules

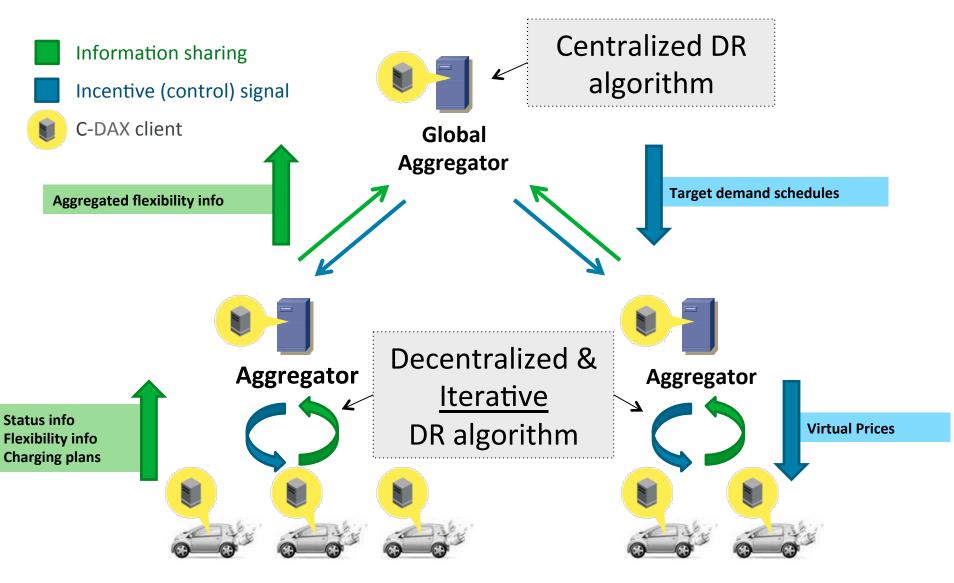








Information Flows

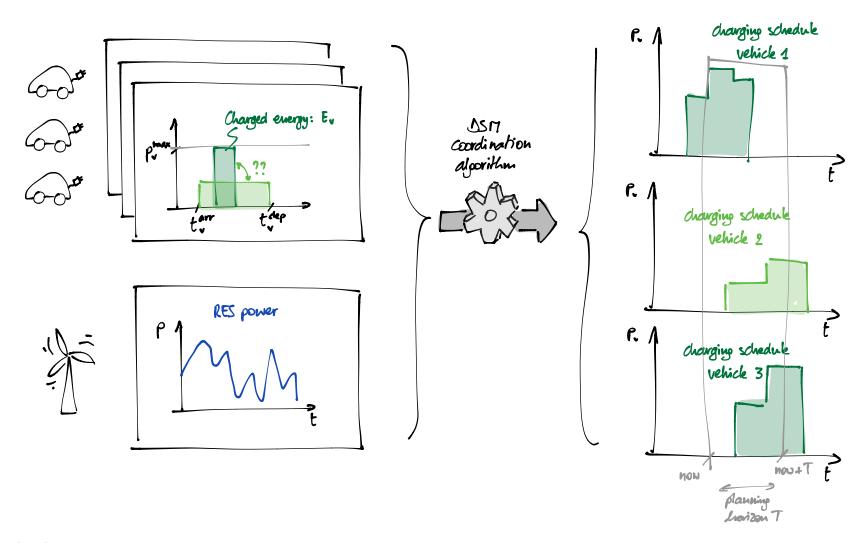








RES balancing: Problem Statement

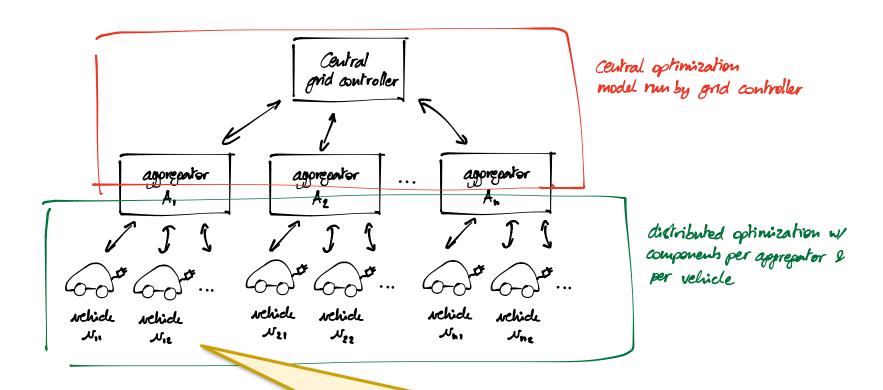








System overview



Multiple **types of users**, e.g.:

- Fast: wants to be charged as quickly as possible
- Flex: doesn't care, as long as charging happens







Global Aggregator optimization model

Minimize "black"
$$\min_{\{\mathbf{x_a}, \mathbf{b}, \mathbf{d}\}} \sum_{t \in \{1, \dots, T\}} (\beta_t \ b_t + \delta_t \ d_t) + \sum_{\mathbf{x_a}, \mathbf{b}, \mathbf{d}} \sum_{t \in \{1, \dots, T-1\}} \alpha \in A$$
 Avoid "discarding" supply from RES

subject to:

$$\sum_{a \in A} x_{a,t} + d_t = g_t + b_t \qquad t \in \{1, \dots, T\} \qquad (3)$$

$$\sum_{t'=1}^{t} x_{a,t'} \ge e_{a,t} \qquad t \in \{1, \dots, T-1\}, a \in A \qquad (4)$$

$$\sum_{t'=1}^{T} x_{a,t} = e_{a,T} \qquad a \in A \qquad (5)$$

$$p_{a,t}^{\min} \le x_{a,t} \le p_{a,t}^{\max} \qquad t \in \{1, \dots, T\} \qquad (6)$$

$$b_t \ge 0, \quad d_t \ge 0 \qquad t \in \{1, \dots, T\} \qquad (7)$$





Aggregator optimization model – central version

Minimize "black" supply (i.e., non-RES)
$$\min_{\mathbf{x_v}, \mathbf{b_a}, \mathbf{d_a}} \beta \cdot \mathbf{b_a} + \delta \cdot \mathbf{d_a} + \max_{\mathbf{v}} \sum_{\mathbf{v} \in V_a} \mathbf{p_v} \cdot (\mathbf{p_v^{max} - x_v})$$
 (9)

 $\mathbf{b_a} \geq \mathbf{0}, \quad \mathbf{d_a} \geq \mathbf{0}$

subject to:

$$\sum_{v \in V_a} \mathbf{x_v} + \mathbf{d_a} = \mathbf{g_a} + \mathbf{b_a}$$

$$\mathbf{1} \cdot \mathbf{x_v} = e_v$$

$$\mathbf{p_v^{\min}} \le \mathbf{x_v} \le \mathbf{p_v^{\max}}$$

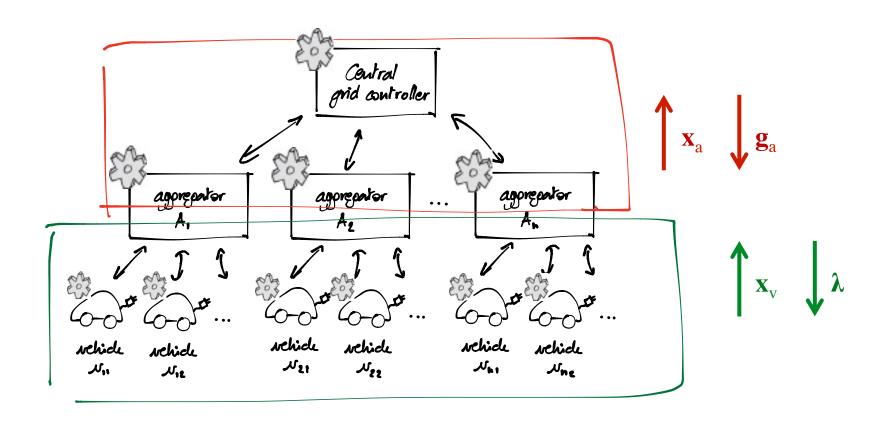
$$v \in V_a$$





(13)

Communication over C-DAX platform









Sample scenario

- T = 31 days divided in 15 minute time slots
- 1 Global Aggregator, 2 Aggregators
- 100 EVs at home charging sessions
 - 3.68 kW maximum charging power
 - 10 kWh batteries
 - Time shifting and/or demand shaping
 - Different flexibility profiles tailored to user preferences
 - Sessions randomly assigned to aggregator
- Power from wind turbine: Scaled to match total demand over T

2 types of users:

- Fast: charge as quickly as possible
- Flex: only deadline to meet

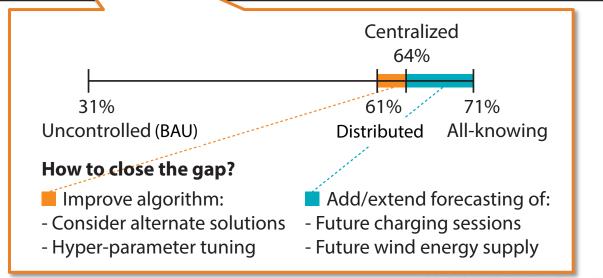






Sample results: RES contribution to EV charging

User mix	% of load supplied by RES				
Fast-Flex	Distributed	Centralized	BAU	Upper bound	
100% – 0%	48.85%	62.65%)	<u> </u>	
75% - 25%	56.57%	63.51%			
50% - 50%	61.73%	64.13%	39.49%	71.17%	
25% - 75%	63.76%	64.87%			
0% - 100%	63.97%	28%	J	J	









Thanks. Any questions?

