Intelligent Energy Systems in CITIES and Solar Power Forecasting





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Henrik Madsen, Technical University of Denmark COST WIRE and CITIES Workshop, DTU Risø, June 2014

DTU Compute Department of Applied Mathematics and Computer Science



Wind integration in Denmark

• Notice – wind only:

Key figures for wind power*

	2013	2012
Wind power generation	11.1 billion kWh	10.3 billion kWh
Electricity consumption (including loss in the electricity grid)	33.5 billion kWh	34.1 billion kWh
Wind power share of electricity consumption the entire year	33.2%	30.1%
Wind power share of electricity consumption in December	54.8%	33.5%
Wind power capacity at the end of the year	4,792 MW	4,166 MW
Energy content of the wind	Approx. 93% of a standard year	Approx. 102% of a standard year

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Solar district heating in Denmark





Concepts in CITIES



Integration based on *IT* solutions and forecasting leading to methods for operation and planning for future energy systems



Example: Storage by Energy Systems Integration



- Operational (simplified) models for integration, optimization and control
- (Virtual) storage principles:
 - Buildings provide storage up to, say, 5-10 hours ahead
 - District heating systems can provide storage up to 1-2 days ahead
 - Gas systems can provide seasonal storage





Scientific Objectives

To establish forecasts and ITC solutions for design and operation of integrated electrical, thermal, fuel pathways at all scales (budget 10 mill. Euros)





Control/Forecasting Principles



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Control/Opt. Principles



Day Ahead:

- Stoch. Programming based on eg. Scenarios
- Cost: Related to the market (one or two levels)
- Operational optimization also for the grid

Direct Control:

- Actuator: Power
- Cost: eg. MV, LQG, EMPC, ... (a single large problem)
- Two-way communication
- Models for DERs are needed
- Constraints for the DERs (calls for state est.)
- Contracts are complicated

Indirect Control:

- Actuator: Price
- Cost: GPC, LQG at high level, VaR-alike
- Cost: E-MPC at low (DER) level, ...
- One-way communication
- Models for DERs are not needed
- Simple 'contracts'



Direct vs Indirect Control

Level	Direct Control (DC)	Indirect Control (IC)
III	$\min_{x,u} \sum_{k=0}^{N} \sum_{j=1}^{J} \phi_j(x_{j,k}, u_{j,k})$	$ \min_{\hat{z}, p} \sum_{k=0}^{N} \phi(\hat{z}_k, p_k) $ s.t. $\hat{z}_{k+1} = f(p_k) $
IV	$\downarrow_{u_1} \dots \downarrow_{u_J} \uparrow_{x_1} \dots \uparrow_{x_J}$ s.t. $x_{j,k+1} = f_j(x_{j,k}, u_{j,k}) \forall j \in J$	$\min_{u} \sum_{k=0}^{N} \phi_j(p_k, u_k) \forall j \in J$ s.t. $x_{k+1} = f_j(x_k, u_k)$

Table 1: Comparison between direct (DC) and indirect (IC) control methods. (DC) In direct control the optimization is globally solved at level III. Consequently the optimal control signals u_j are sent to all the J DER units at level IV. (IC) In indirect control the optimization at level III computes the optimal prices p which are sent to the J-units at level IV. Hence the J DERs optimize their own energy consumption taking into account p as the actual price of energy.



Forecast requirements



Day Ahead:

- Forecasts of loads
- Forecast of Grid Capacity (using eg. DLR)
- Forecasts of production (eg. Wind and Solar)

Direct Control: .

- Forecasts of states of DERs
- Forecasts of load

Indirect Control:

- Forecasts of prices
- Forecasts of load



Which type of forecast to use?

- Point forecasts
- Conditional mean and covariances
- Conditional quantiles
- Conditional scenarios
- Conditional densities
- Stochastic differential equations



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Ongoing CITIES projects using MET forecasts

- Temperature control in houses (Grundfos, ENFOR)
- HVAC systems (Grundfos)
- Supermarket cooling (Danfoss, UCD)
- Consumption in family houses (TI, ENFOR, ...)
- District heating networks (Cowi, ENFOR, Rambøll, DFF-EDB)
- Combined Heat and Power plants (Dong Energy)
- Heat Pumps in District Heating networks (HOFOR, Cowi, ENFOR)
- Rainfall Run-off Systems (DHI and Rambøll)
- Wastewater treatment plants (Krüger)
- 🧿



Example

Solar Power Forecasting in CITIES





Solar Power Forecating





- Grid connected PV-systems mainly installed on rooftops
- Average of output from 21 PV systems in Brædstrup





- Based on MET forecasts and online readings of output
- Two-step method:
 - -1) Transformation to atmospheric transmittance with statistically clear sky (see above),
 - -2) A dynamic model + adaptive quantile regression.



Example (quantile forecasts - up to 36h ahead)



Adaptive correction method





Adaptive correction method



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Adaptive correction method (correction function)





Adaptive correction method





Use of MET forecasts for:

- Electricity Load
- Wind Power Production
- Solar Power Production
- Gas Load
- District Heating Load
- Price forecasts
- Urban Meteorologi
- Grid operations

- Important aspects:
 - Use the same MET service(s)
 - Adaptivity
 - Corrections
 - Full probabilistic
 - 'Correlation' (auto- and cross correlation)



Case study

Electrical Heating of Buildings; Control of Load by Price





Price responsivity

Flexibility is activated by adjusting the temperature reference (setpoint)



- **Standardized price** is the % of change from a price reference, computed as a mean of past prices with exponentially decaying weights.
- **Occupancy mode** contains a price sensitivity with its related comfort boundaries. 3 different modes of the household are identified (work, home, night)



Two data sources



Olympic Pensinsula project

- 27 houses during one year
- Flexible appliances: HVAC, cloth dryers and water boilers
- 5-min prices, 15-min consumption
- **Objective:** limit max consumption

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Simulation framework

- Modular design
- Runge-Kutta solver (diff. equations)
- Scalable (linear computation time)
- Variable sampling rate CITIES Centre for IT Intelligent Energy Systems

Aggregation (over 20 houses)



Identify price response



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Forecasted Response on Price Step Change

Simulated

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Olympic Peninsula



Adaptive control setup

As the systems changes over time





Control performance

With a price penality avoiding its divergence

- Considerable reduction in max consumption
- Mean daily consumption shift





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