



Model Predictive Control of Heat Supply to Greenhouses

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 $f(x+\Delta x) = \sum_{i=0}^{\infty} \frac{(\Delta x)^{i}}{i!} f^{(i)}(x)$

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Outline





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- Model for Greenhouses
- Online Predictions
- Model Predictive Control
- Simulation vs. Prediction based Control
- Flexibility in DH Systems





Models for Greenhouses





Lumped models



Fig. 1. Serial energy fluxes in a greenhouse with n nodes. Each node is assumed to be spatially uniform with a constant temperature and heat capacity







Table 1 Statistic of the models considered

Model no.	No. of nodes	No. of parameter	log(L)	SBC
1	1	6	22 0 58	- 44 0 55
2	2	10	34 680	- 69 258
3	3	14	34833	- 69 523
4	4	16	34 836	- 69 509
5	4	18	34851	- 69 519

L = likelihood function; SBC = Schwartz's Bayesian Criterion.

Lumped models for greenhouses Conclusions – so far

A lumped parameter models with 3 nodes is adequate for describing the heat dynamics of the greenhouse

The solar radiation enters the first node in the model

This node seems to represent the air temperature and 'outer' surfaces of other objects in the greenhouse

The next node interacts with the air temperature and represents the plants, the soil in the pots, the inner part of the bench and a few centimeters of the ground

- The third node represents the deeper part of the soil
- Nonlinear models taking the wind speed and humidity into account must be formulated







Forecasting and Control





Predictive Control





Heat load – 96 hour forecasts (Sønderborg DH system)



Horisont [timer]



Stoch. Models for the DH Network (simplified)





Set-point selection Use of uncertainties

User supply temperature



CITIES Consortium Meeting 2015 Control of Greenhouses Ambient air temperature



Optimal Control



$$\begin{split} \min_{u_k} J(\Gamma_k, \Lambda_k, \omega_k; k, u_k) \\ &= E_k[(y_k - y_k^0)^T \Gamma_k (y_k - y_k^0) \\ &+ u_k^T \Lambda_k u_k + 2\omega_k^T u_k] \;, \end{split}$$

If we assuming that the output can be predicted by a piecewise linear model, we obtain easily

$$u_k = -\left[H_k^T \Gamma_k H_k + \Lambda_k\right]^{-1} \left[H_k^T \Gamma_k^T \beta_k + \omega_k\right] \; .$$

This defines the optimal supply temperature for each of the greenhouses. The resulting supply tempeturature is then found as the maximum of all.





Та

CITIES Control of Supply Temperatures



FJERNVARMEN | 5 2010

Styring af temperatur rummer kæmpe sparepotentiale

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Conclusions:

- Control using simulation gives up to 10 pct reduction of heat loss.
- Control using forecasting and measurements gives up to 20 pct reduction of heat loss



Models and Optimal Control



Control of Greenhouses



Conclusion on Control

- Use simulation based control when:
 - No data is available from the DH net
 - A new layout of the DH system is selected
- Use prediction based control when:
 - Data is available online (so they can be used for control and forecasting)
 - Meteorological forecasts are used for improved control
 - For adaptive self calibrating control setups







Flexibility in DH systems





Interface with power system









- Lumped parameter models for greenhouses
- Load forecasting in DH systems which takes advantage of Meteorological forecasts
- Controllers for minimizing the supply temperature in DH networks
- Principles for flexibility in DH systems are described
- Interface with power system is outlined





Thanks for your time...

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