

Fusion of Data and Physics-Based Modeling; State-of-the-Art and Applications

| White | Grey | Black |
|---|---------|-----------------------|
| Physics | Data | Data |
| Deterministic equations | | Stochastic models |
| Detailed submodels | | Simple models |
| Parameter calibration often problematic | | Parameter calibration |
| Parameters~physics | Physics | Parameters~black |

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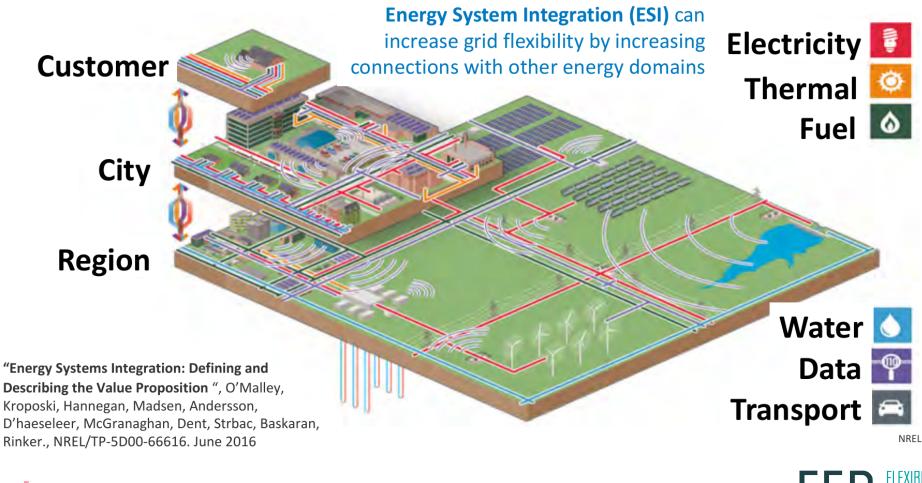








Energy Systems Integration





IEA Wind Task 43 on Digitalization, May 2020

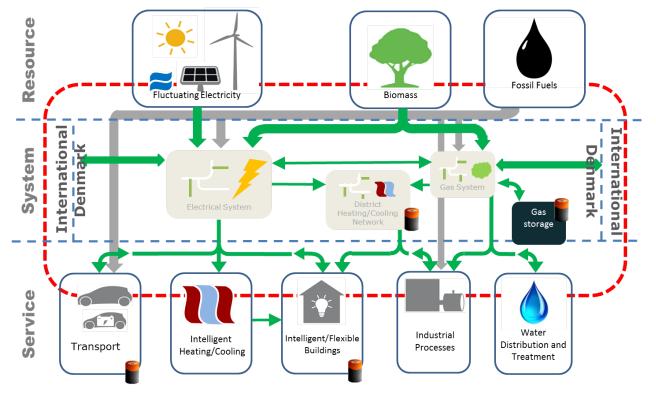
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Energy System Models for Real Time Applications and Data Assimilation



Grey-box models are simplified models for the individual components facilitating system integration and use of sensor data

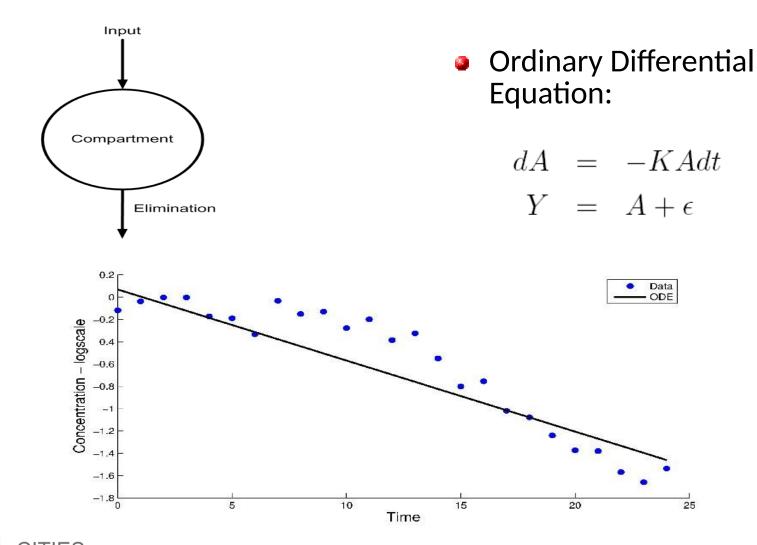




Traditional Dynamical Model



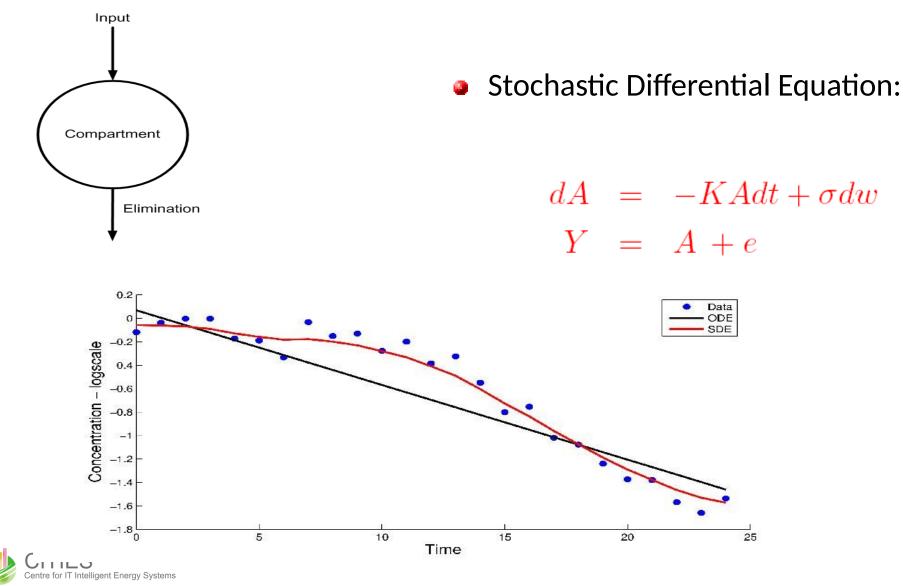
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Stochastic Dynamical Model

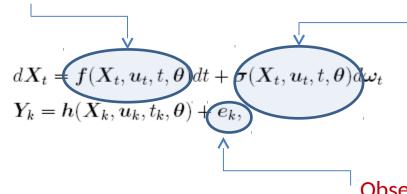




m ETU

Grey box or HMM model on state space form

Drift term



Diffusion term

System equation Observation equation

Observation noise

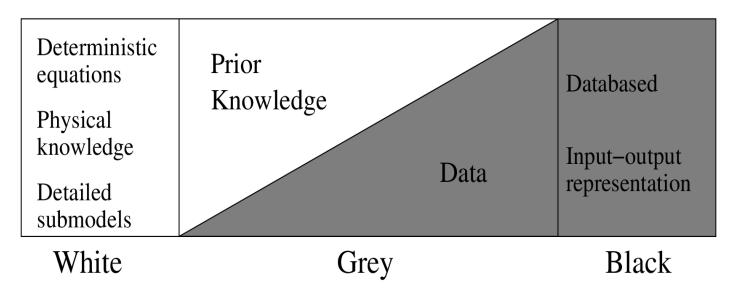
Notation:

- X_t : State variables
- u_t : Input variables
- θ : Parameters
- Y_k : Output variables
- t: Time
- ω_t : Standard Wiener process
- e_k : White noise process with N(0, S)

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Grey-box modeling concept





- Combines prior physical knowledge with information in data
- Equations and parameters are physically interpretable



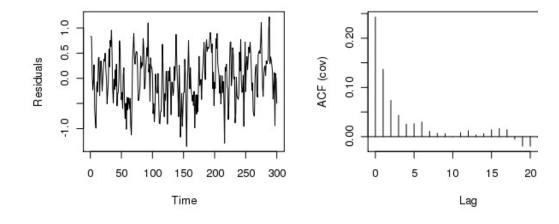


Grey-Box models are well suited for ...

- One-step forecasts
- Software state estimation (software sensors)
- K-step forecasts
- Simulations
- Control
- <u>م</u>
- Provides a framework for pinpointing model deficiencies – like:
 - Time-tracking of unexplained variations in e.g. parameters
 - Missing (differential) equations
 - Missing functional relations
 - Lack of proper description of the uncertainty

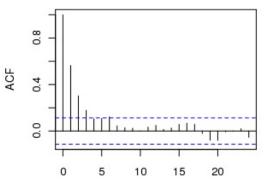


Identification of the needed number of states / differential equations (example)



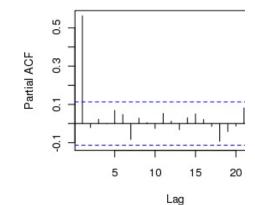
Series Residuals





Lag







The software ...



- CTSM-R Continuous Time Stochastic Modelling in R
- Download from http://ctsm.info
- User's and Math Guides are available
- For more information:

Email to

info@ctsm.info





Example on



Blade manufacturing (molding) ...

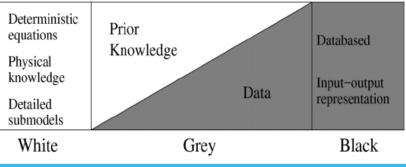






Monitoring and Modelling Framework

- Sensors
 - It would require a lot of sensors to get the full picture
- Computational Fluid Dynamics
 - Solving multidimensional PDEs in real-time is computationally heavy
- Greybox Models
 - We combine knowledge from physical models with sensor data to further develop our models and to explain discrepancies





chnical University Denmark **SIEMENS** Gamesa





Simulation of Flow-Front Progression

• We combine Darcy's law with the equation for conservation of mass

$$q = -\frac{\kappa H}{\mu} \nabla p$$
$$\dot{h} + \nabla \cdot q = 0$$
$$\dot{h} = \frac{dh}{dp} \dot{p} = \nabla \cdot \left(\frac{\kappa H}{\mu} \nabla p\right)$$

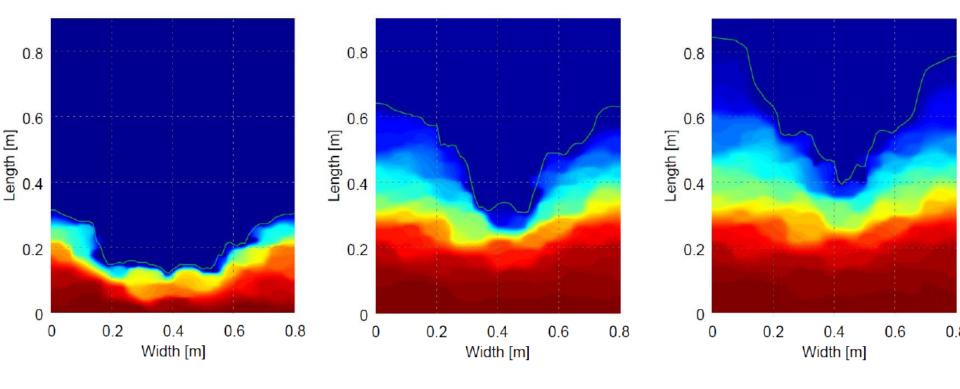
$$q = q(x, y, t), \nabla = (\partial_x, \partial_y)$$

$$h = \min(H, \frac{p}{\rho g})$$





Simulation of Flow-Front Progression



| DTU Technical University of Berimark | SIEMENS Gamesa | 12-08-2019 | 12 |
|---|----------------|------------|----|
| | | | |





Example on Prob. Wind Power Forecasting









DTU

Prob. Wind Power Forecasting using SDEs

Wind dynamics given by:

$$dX_t = \left(\left(1 - e^{-X_t} \right) \left(\rho_x \dot{p}_t + R_t \right) + \theta_x \left(p_t \mu_x - X_t \right) \right) dt + \sigma_x X_t^{0.5} dW_{x,t}$$

$$dR_t = -\theta_r R_t dt + \sigma_r dW_{r,t}$$

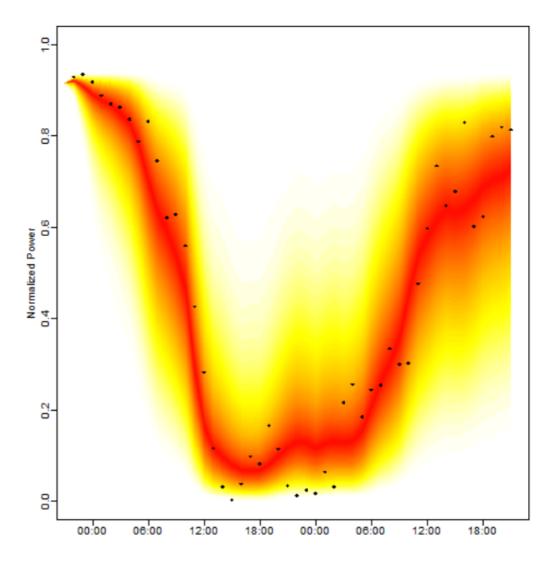
$$Y_{1,k} = X_{t_k} + \epsilon_{1,k}$$

Wind to power dynamics given by:

$$\begin{split} dQ_t &= (S_t - \theta_q Q_t) dt + \sigma_q dW_{q,t} \\ dS_t &= -\theta_s S_t dt + \sigma_s dW_{s,t} \\ Y_{2,k} &= (0.5 + 0.5 \tanh(5(X_{t_k} - \gamma_1))) (0.5 - 0.5 \tanh(\gamma_2(X_{t_k} - \gamma_3))) \\ & \frac{\zeta_3}{1 + e^{-\zeta_1(X_{t_k} - \zeta_2 + Q_{t_k})}} + \epsilon_{2,k} \end{split}$$

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Predictive density of production in percent out of rated power for the Klim wind farm:









DFNMARK

Examples on advanced topics ... (Solar power forecasting)



DTU Compute Institut for Matematik og Computer Science







- A solar power plant with a nominal output of 151 MW.
- Measurements of 91 inverters every second for one year.
- ► We consider a cutout of 5 by 14 inverters for modeling.





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IEA Wind Task 43 on Digitalization, May 2020

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Stochastic Partial Differential Equation

- Normalize the parameters with the spatial distance in appropriate way.
- Parameters become grid-invariant.
- Can be interpreted as a stochastic partial differential equation.

The dynamical model interpretation:

$$dU(x,t) = \bar{v}\theta\nabla U(x,t)dt + \sigma dW(x,t),$$

with the deterministic part $dU(x, t) = \bar{v}\theta\nabla U(x, t)dt$ being a uni-directional wave equation.





SPDE Model Performance



| _ | Auto- Regressive | Model |
|----------------------------|---------------------|---------|
| $CRPS_5$ | 0.00262 | 0.00131 |
| $CRPS_{20}$ | 0.00982 | 0.00666 |
| CRPS_{60} | 0.02886 | 0.02455 |
| $CRPS_{120}$ | 0.04883 | 0.04675 |





Innovation Fund Denmark





Center Denmark

National Digitalization/AI Hub for Smart Energy Systems and Integration of Wind/Solar Power









Connect networks and data for a green world

Danmarks nationale Center

Fremme den grønne omstilling. Samle og bygge bro, mellem forskning, teknologi, natur og formidling, på tværs af interesseorganisationer, virksomheder, skoler og universiteter.

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Thanks for your time ...

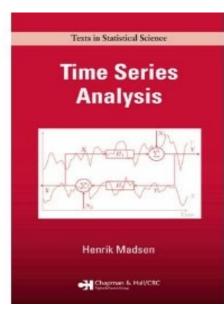


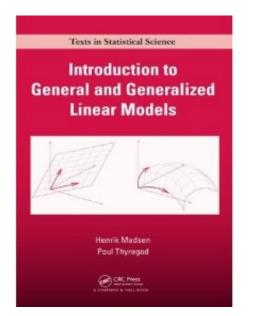






Some 'randomly picked' books on modeling and renewable integration





International Series in Operations Research & Management Science

Juan M. Morales - Antonio J. Conejo Henrik Madsen - Pierre Pinson Marco Zugno

Integrating Renewables in Electricity Markets

Operational Problems



2 Springer





Digitalization Hub - Center Denmark



- A digitalization hub for data intelligent operation of integrated energy systems (electricity, thermal, gas, water)
- A national hub for <u>unlocking the flexibility</u> potential for large scale integration of fluctuating renewable energy
- Tests on framework conditions have to be <u>representative</u> and <u>scaling</u> is important
- The new national smart energy hub is <u>Center Denmark</u> (10.000 m2 facilities for Research, Education, Development and Testing plus Dissemination)
 - The <u>Societal objective</u> is to establish a realistic and concrete pathway to a fossil-free society
- The S<u>cientific objective</u> is to establish methodologies and solutions for the future intelligent and integrated energy system using digitalization and a smart energy hub
- The <u>Commercial perspective</u> is to being able to idenfy and test solutions which can form the background for commercial success stories. We believe that this setup has the unique characteristics for being the ultimate smart energy hub for test and demonstration of future smart energy solutions

