

### Ensemble based updating of distributed, physically based, urban drainage models

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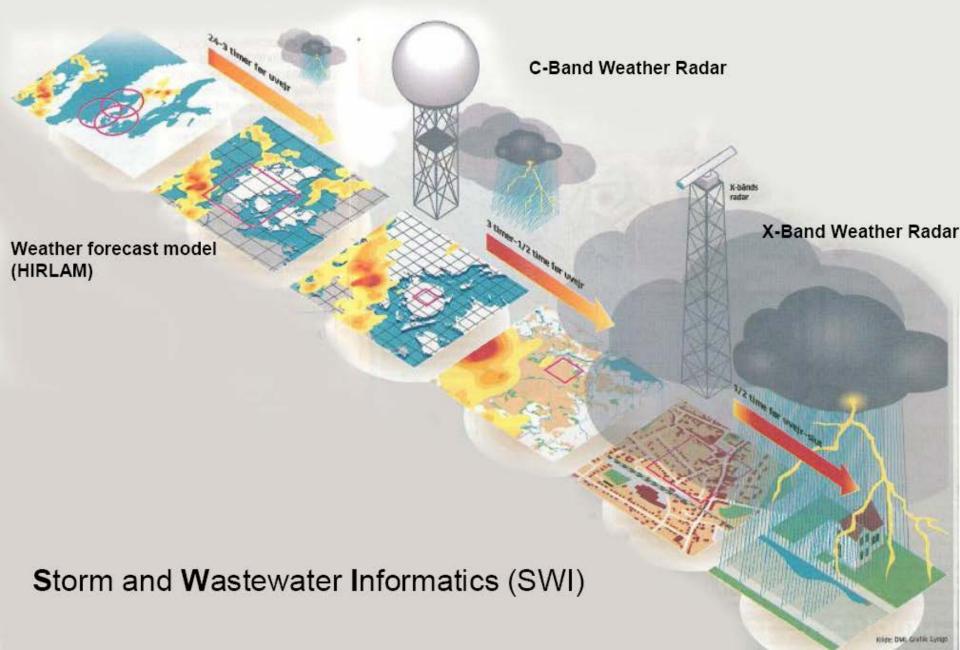
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Intelligence Inside



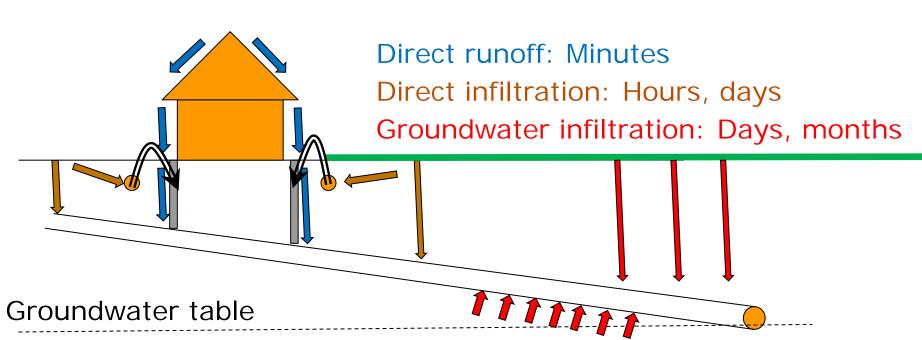
### Outline

- Urban drainage systems: Background
- Physically based, distributed urban drainage models and what they can do
- EnKF issues
- Small example



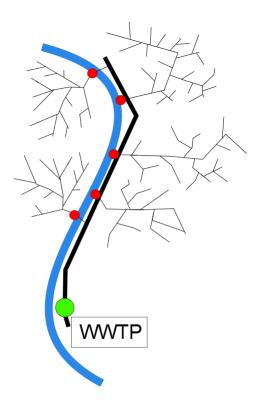
### **Generating Urban runoff**







### **Urban drainage Systems**





# **Urban drainage Systems** WWTP

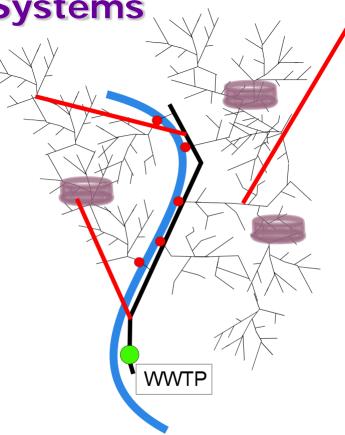


# **Urban drainage Systems** WWTP

### Urban drainage Systems

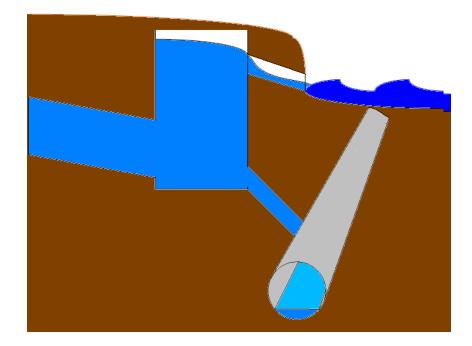
- It's a mess
- Suboptimal for most events
- Control is needed







### **Overflow structure**





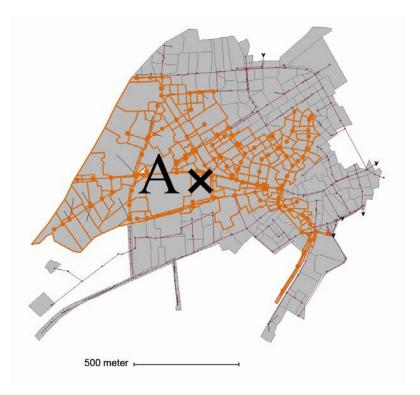
### Differences to other hydrological system

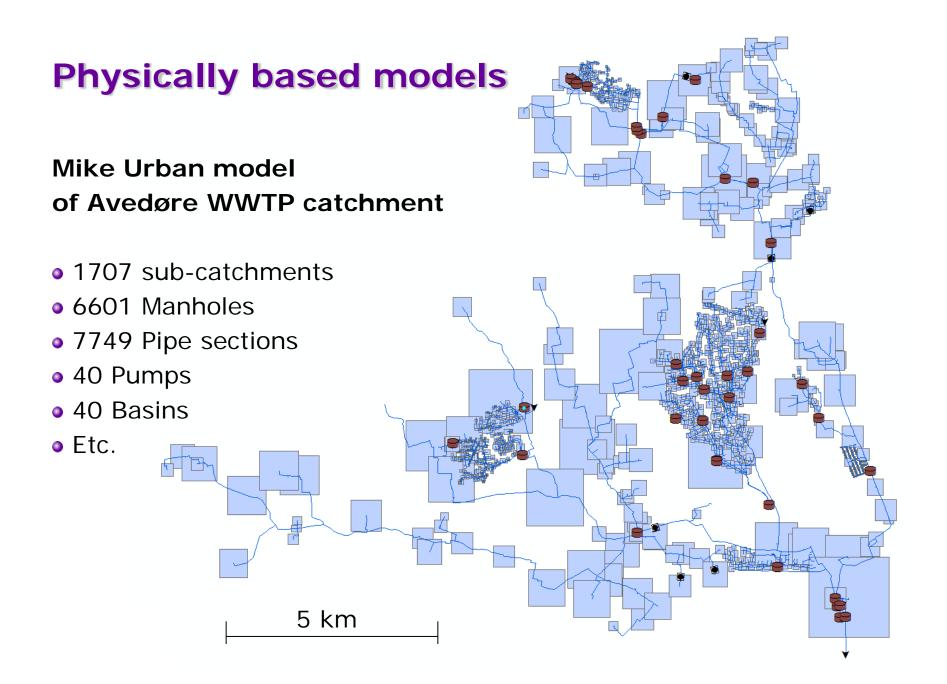
- Fast response times
- Closed conduits -> max. capacity
- Overflows: Water disappears out of system
- Real time control change hydraulic behaviour in seconds

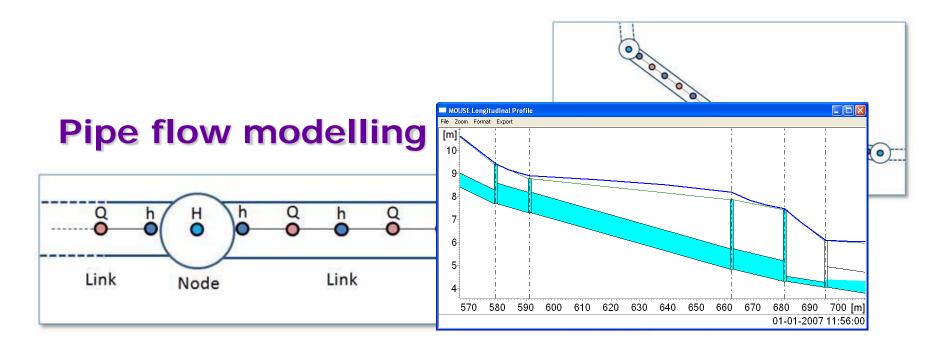


# Distributed, physically based, urban drainage models

- Mixture of models: Runoff + Hydrodynamic + Water Quality
- Developed mainly for design purposes
- Can be build purely from physical data without calibration







• Full 1D St. Venant equations Conservation of mass:

$$\frac{\partial Q}{\partial x} + \frac{\partial A}{\partial t} = 0$$

Conservation of momentum:

$$\frac{\partial Q}{\partial t} + \frac{\partial \left(\alpha \frac{Q^2}{A}\right)}{\partial x} + gA \frac{\partial y}{\partial x} + gAI_f = gAI_0$$

### Multi purpose model

#### • Dimensioning of new system elements

Max frequency of water on terrain, basements etc.

### Calculate yearly overflow and pollution loads

Documentation to authorities

#### • Develop and **optimize control** strategies

Both PID and model predictive control using simple models

### Are NOT used as online models



# Purpose of online model

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## Real time health risk assesment

70

København Lyngbyvej 16 aug. 2010

Photo: Bente Schou

## **Error detection!**

0

Human



### Model assisted real time control

#### • Aim:

- reduce flooding
- reduce or redirect overflow
- Reduce cost of electricity consumption





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### **Currently not used online because**

- Computational cost hardly run real time
- Very uncertain rain input
- No efficient update algorithm



# Ensemble based updating

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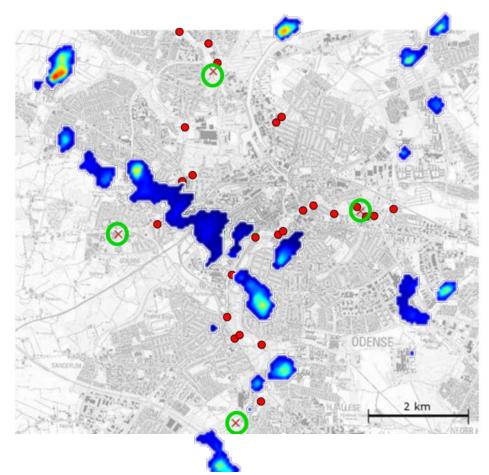
### **Dominating error source: Rain estimates**

 Gauge: Accurate in small area
-> huge ensemble required to represent spatial variability

#### Radar

Very inaccurate short term rain depth

But spatial information -> reduced ensemble size





# Q Update

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### **Q** changes makes no lasting change

 Q update do not change the volume of water in an area and thus the change is only local and temporary.

Full 1D St. Venant equations

Conservation of mass:

$$\frac{\partial Q}{\partial x} + \frac{\partial A}{\partial t} = 0$$

Conservation of momentum:

$$\frac{\partial Q}{\partial t} + \frac{\partial \left(\alpha \frac{Q^2}{A}\right)}{\partial x} + gA\frac{\partial y}{\partial x} + gAI_f = gAI_0$$



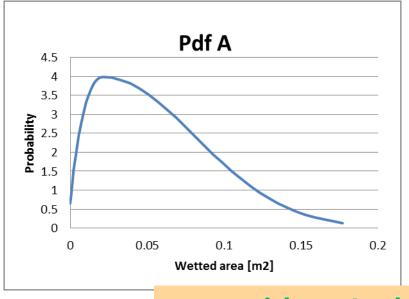
# H update

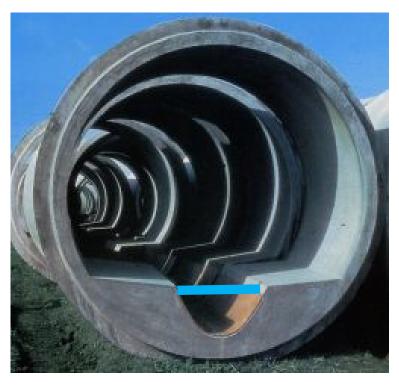
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### **Overestimated observation variance**

 Change in volume per change in h is depth dependent.





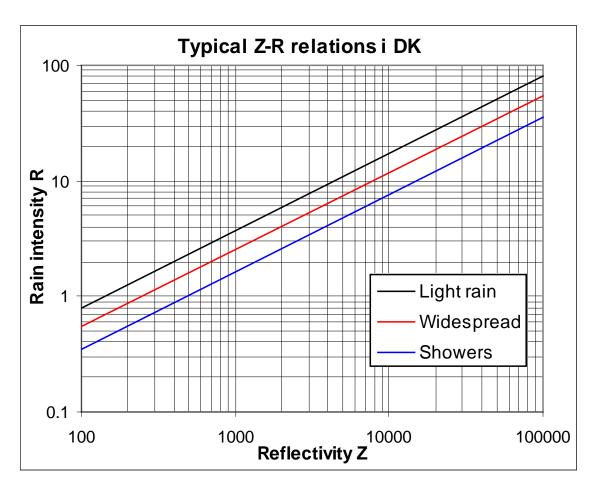
### -> avoid perturbated observations



# EnKF Example using constructed radar data

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### **Radar data**

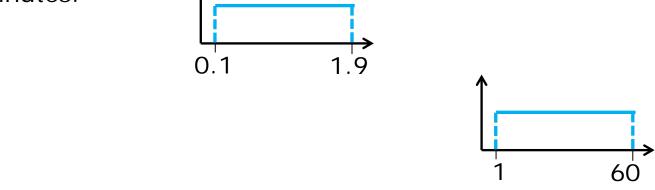


### **Radar rain perturbation**

- Assuming radar rain estimates
  - Assuming factorial error (Wrong guess at Marshall Palmers)
  - Assuming constant error factor in intervals

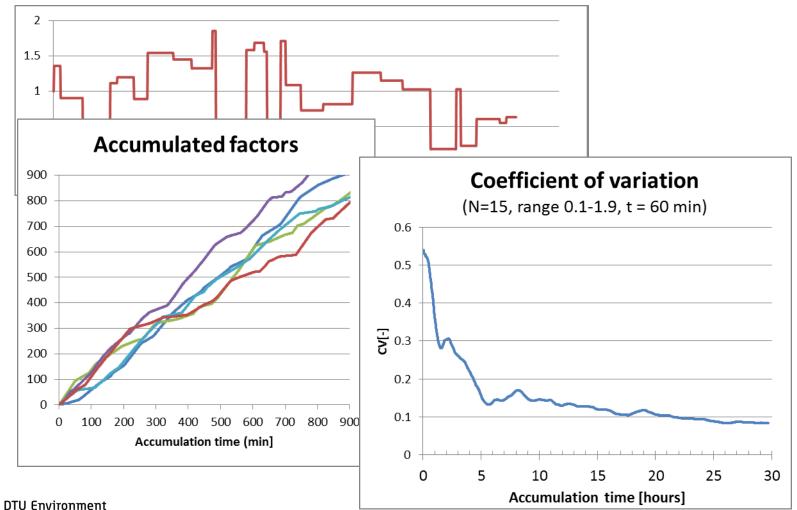
### *Estimated rain* = *actual rain* \* *f*

*f* is drawn from uniform distribution *unif*(0.1, 1.9) every t minutes.



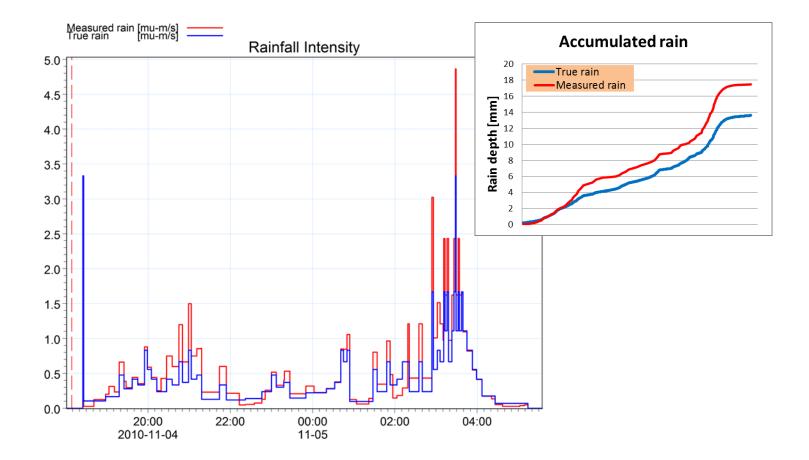
t is drawn from *unif*(1, 60) every t minutes.

### **Realization of f**

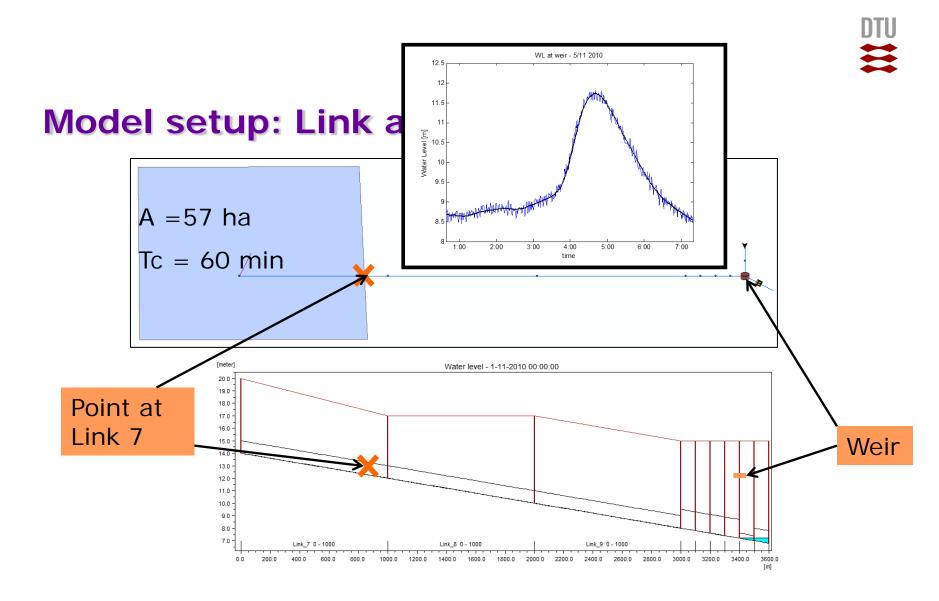


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### **Constructed radar estimate**

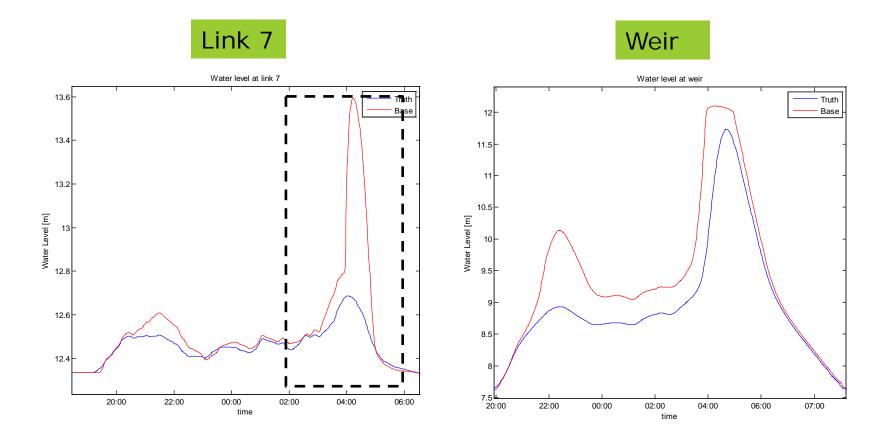


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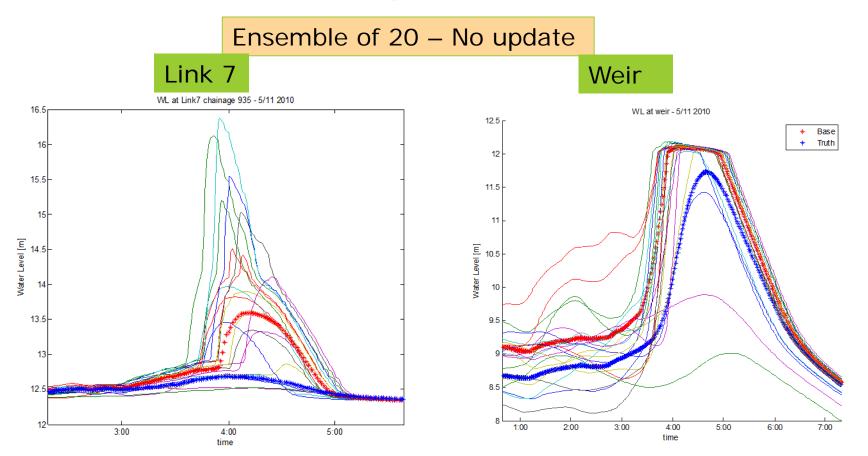


### Situation without update



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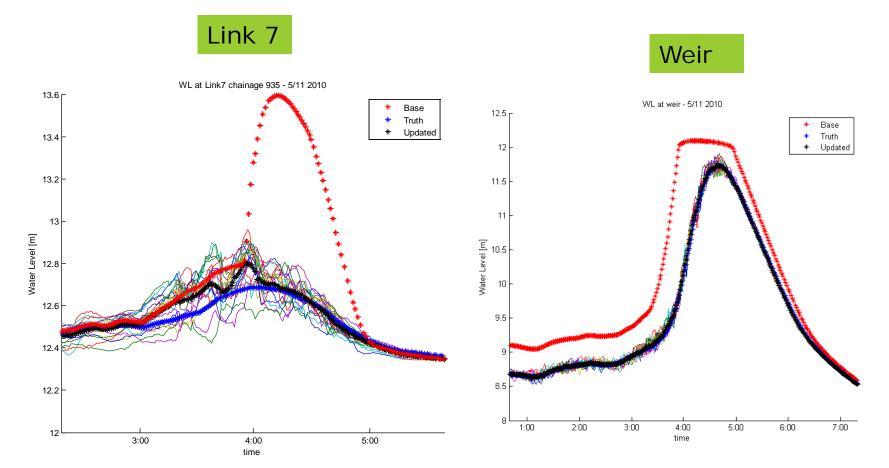
### Situation without update



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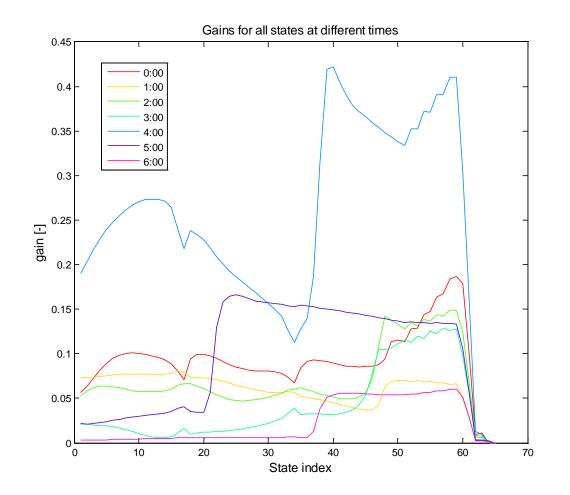
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### When updating using EnKF

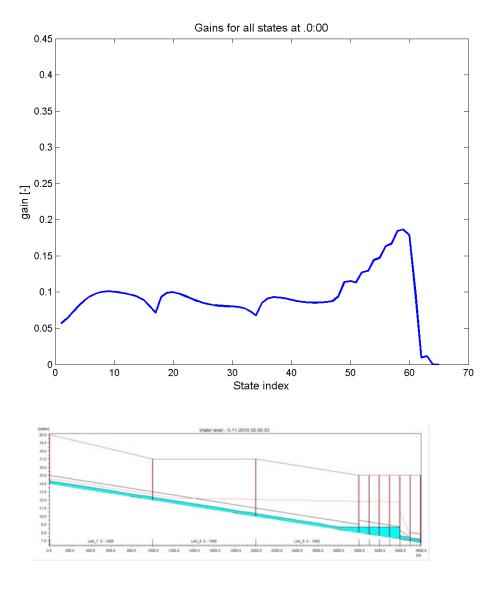


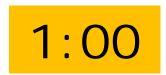
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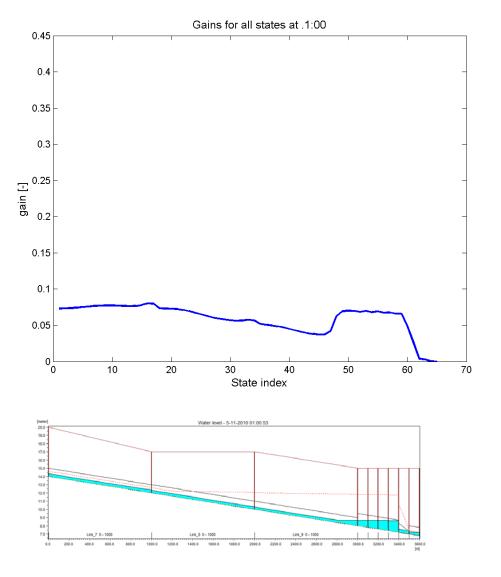
### **Gain and backwater**



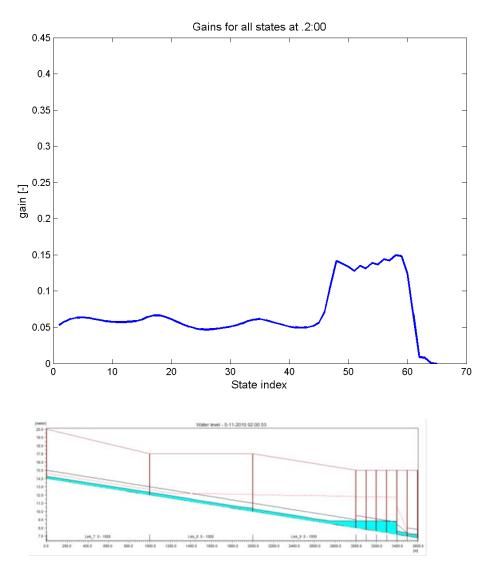




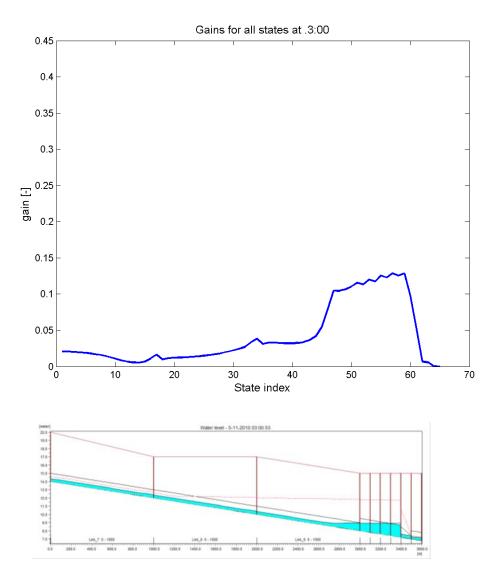






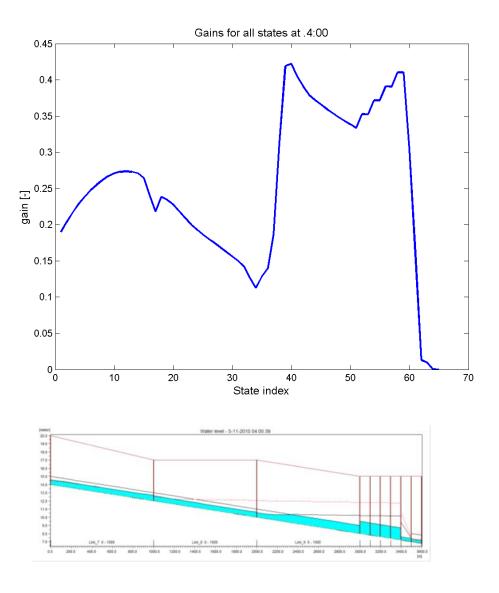




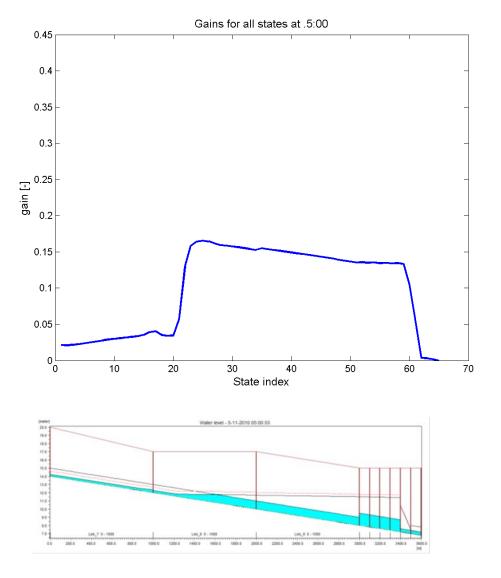


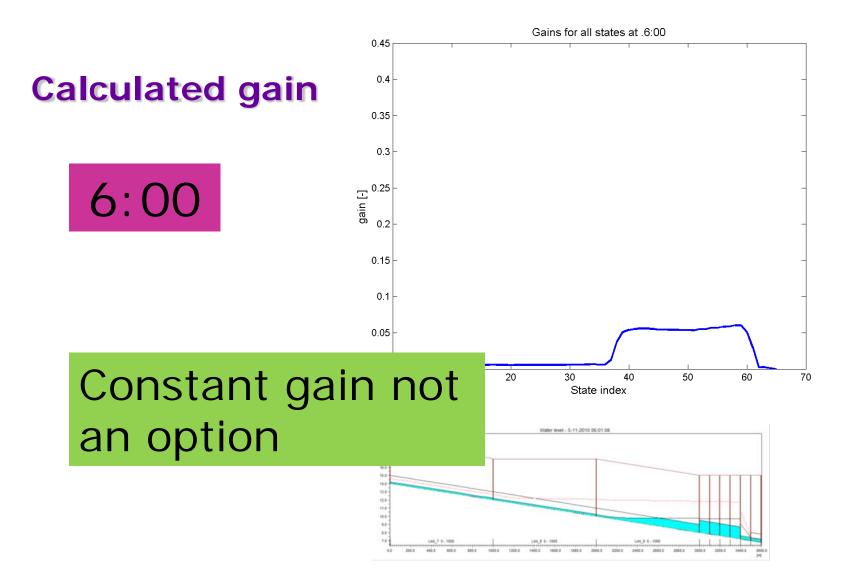
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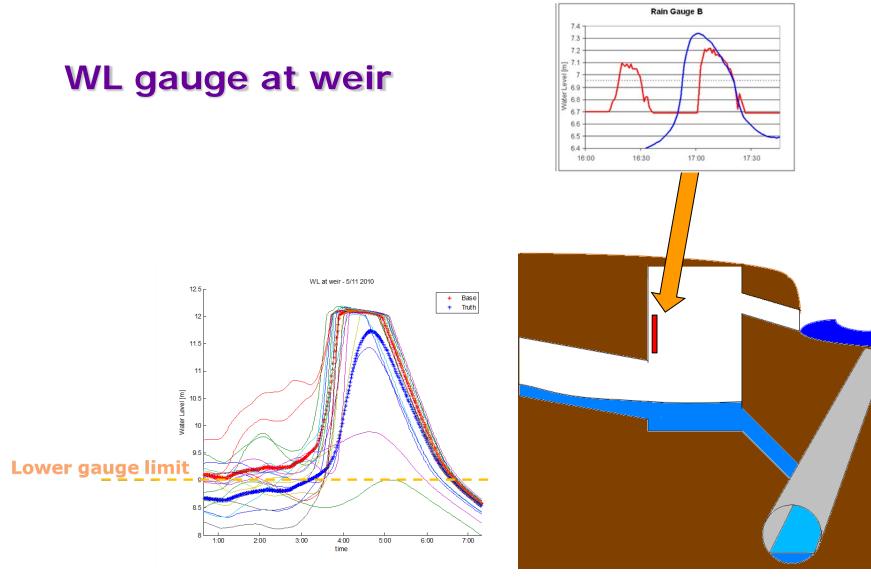




## Non measurements

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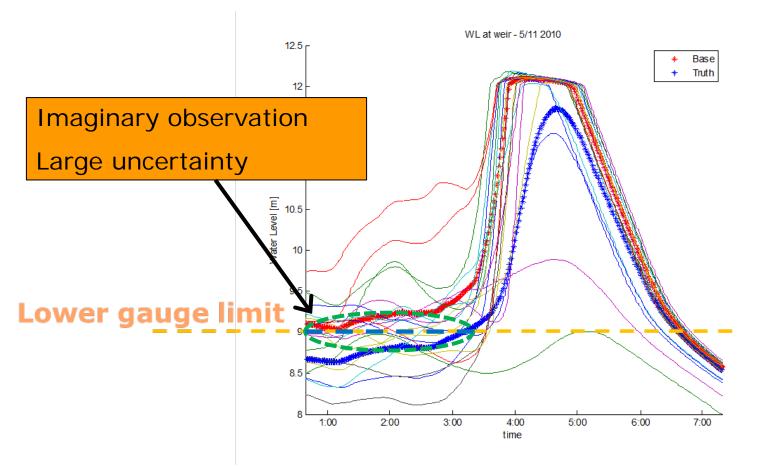




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### Partial ensemble updating



## Partial DEnKF when no data

Modified from [Sakov, 2008]<sup>1)</sup>:

 $dy = gaugeLimit - Hx^{f}$   $x^{f} = mean(X^{f})$   $A^{f} = X^{f} - [x^{f}, ..., x^{f}]$   $A^{a} = A^{f} - \frac{1}{2}KHA^{f}B$   $B = Diagonal matrix. 1 where HA_{i} > = dy, otherwise 0$ 

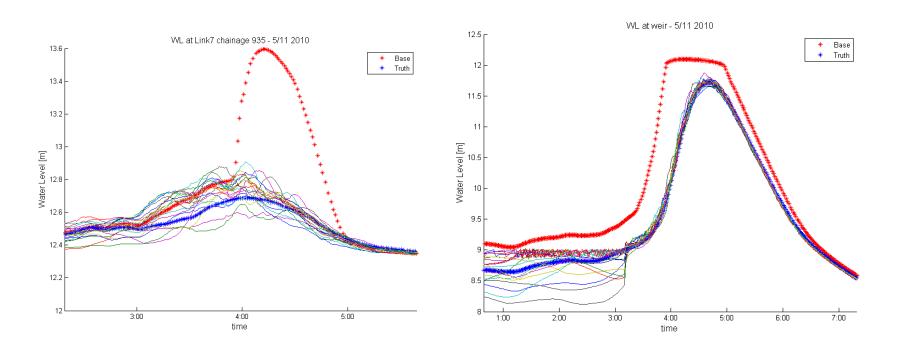
 $X^{a} = A^{a} + [x^{f}, \dots, x^{f}]$  $x^{a} = median(X^{a})$ 

DTU Environment Department of Environmental Enginer 1) Sakov, P., & Oke, P. R. (2008). A deterministic formulation of the ensemble Kalman filter: an alternative to ensemble square root filters. *Tellus A*, *60*(2), 361–371.



## Partial DEnKF when no data

### Example



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### Summary

- Static gain not sufficient
- Radar data is almost a requirement for EnKF
- Ensemble spread can be reduced in periods without measurement
- Probably best to avoid perturbed observations



# **Questions** ?

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