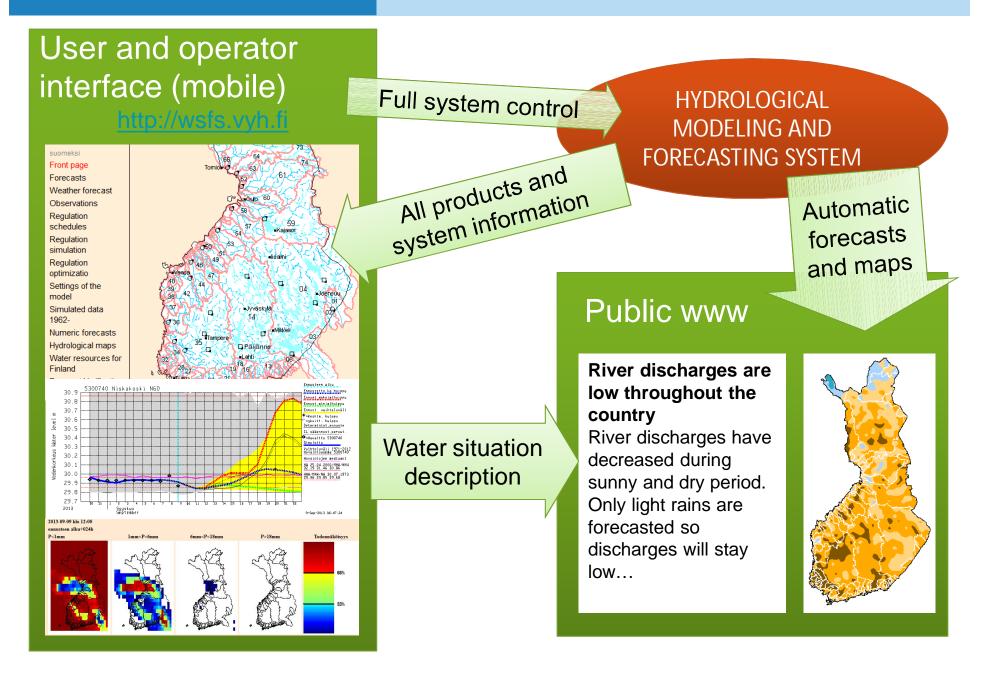
Modeling in aquatic environment Lecture 10 Data assimilation and data fusion in models

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Hydrological Modeling and Forecasting System (4/4)



Weather radar observations and nowcasts (1/1)

Weather radar and LAPS in WSFS

Weather radar

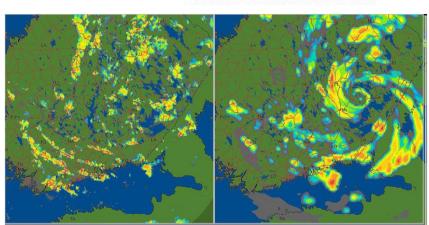
- Hourly radar data in 2x2km grid
- Used for 2 days in model
- Underestimates large rainfalls
 - Corrected manually against rain gauges

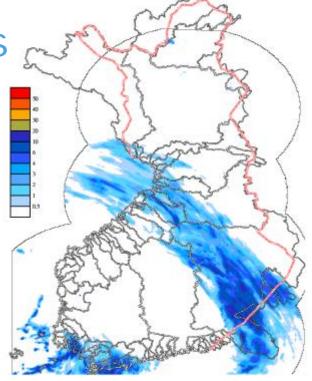
Radar nowcasting

- 3hrs in Southern Finland
- Ensamble of 50 members

LAPS

- Combines information from weather radar, automatic real-time rain gauges, road weather measurements, ...
- Will probably be used in WSFS as a weather radar replacement in near future



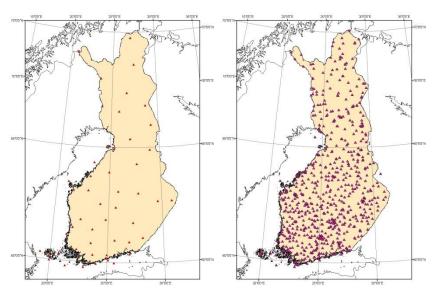


Cumulative precipitation sum (mm) UTC 201909020000-201909020800

Realtime and historical hydrometeorological and nutrients observations (1/4)

Observation networks

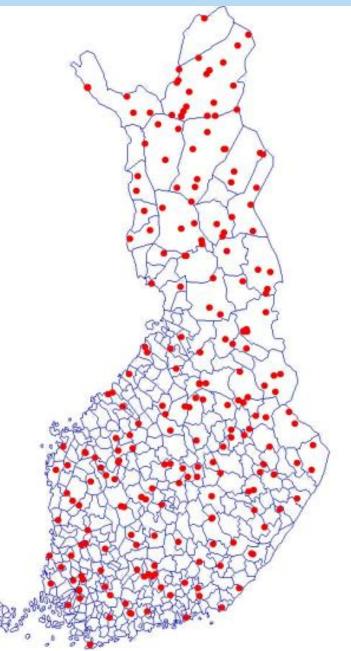
- Automatic realtime precipitation 1-hour measurements from 100 stations
- 380 discharge stations
 - 220 with daily measurements
 - 160 external stations with usually daily measurements
- 660 water level stations
 - 400 with daily measurements
 - 260 external stations with usually daily measurements
- Snow courses
 - 140 montly measurements
- Water quality measurements



Synoptic weather stations, 50 daily measurements Other weather stations, 200 daily measurements Realtime and historical hydrometeorological and nutrients observations (2/4)

Measuring the snow water equivalent

- Snow water equivalent is measured by snow course measurements
 - About 140 snow courses in Finland (less than what is shown at the map)
- Areal snow water equivalents are calculated for approximately 110 areas
- Snow courses are 2-4 km long routes through various terrains
 - 80 depth measurements
 - 8 manual weightings
- Measurements are made once or twice a month

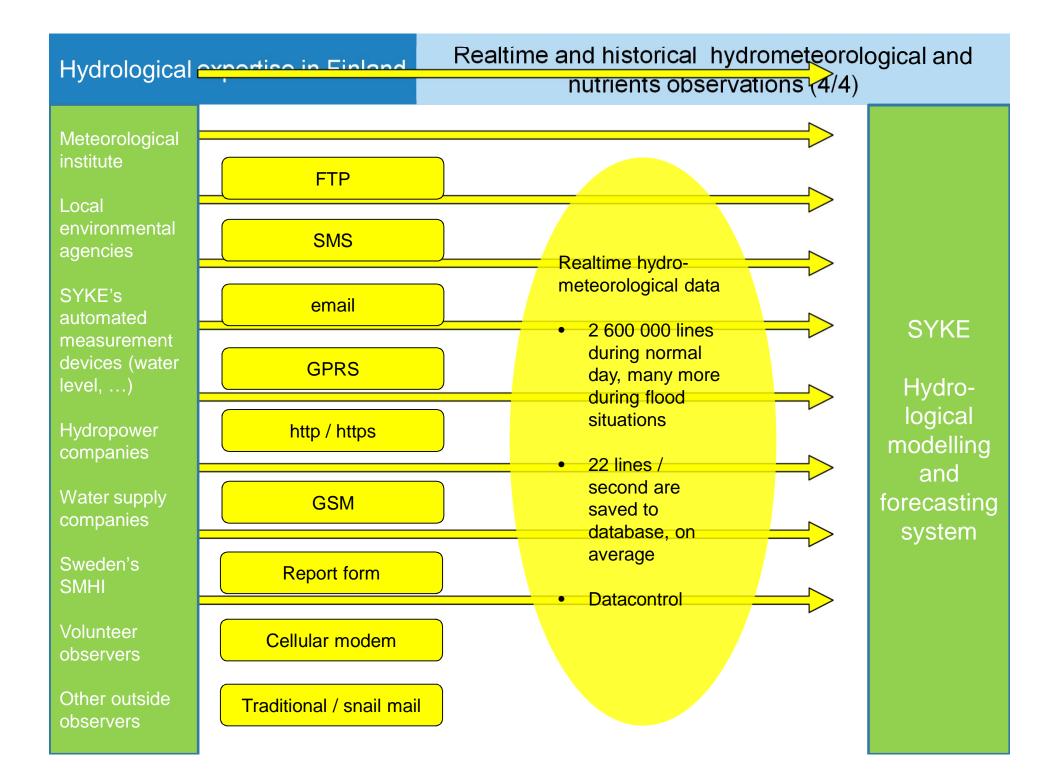


Hydrological expertise in Finland

Realtime and historical hydrometeorological and nutrients observations (3/4)

Measuring the snow water equivalent





Hydrological expertise in Finland

Data assimilation and control (1/6)

Manual measurements

- Manual river discharge measurement

- Manual precipitation observations

- Snow line measurements Automatic observations - Automatic river water level observation - Automatic

precipitation observations

- Snow depth from automatic stations

Derived results / algorithmic data - Weather radar precipitation - Wind corrected precipitation observations

- "Near-bycomplemented" precipitation observations

- Satellite value for snow water equivalent

- Interpolated snow water equivalent

Simulations

- Areal precipitation

- Runoff, discharge, water level, soil moisture, ...

- Areal precipitation corrected to match water balance

- Snow water equivalent, snow depth, ground frost depth

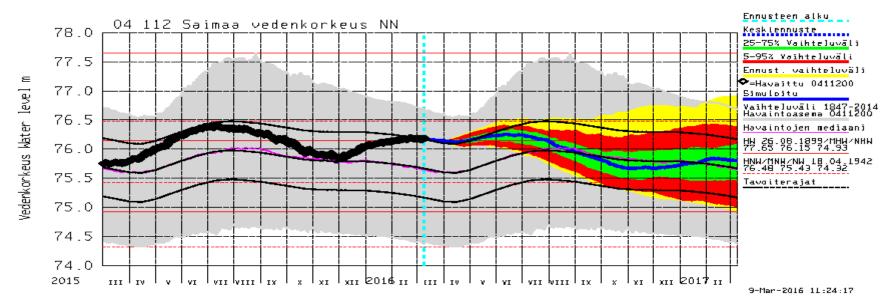
- Ice thickness

Calculated data

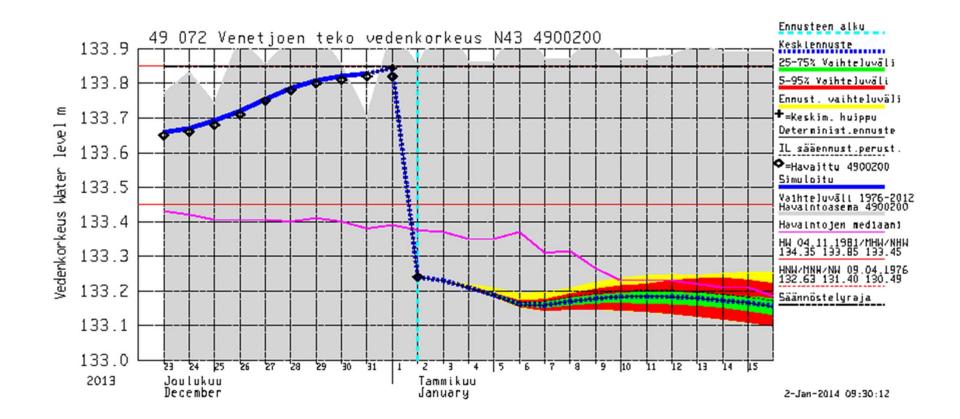
Observed data

Data assimilation algorithm

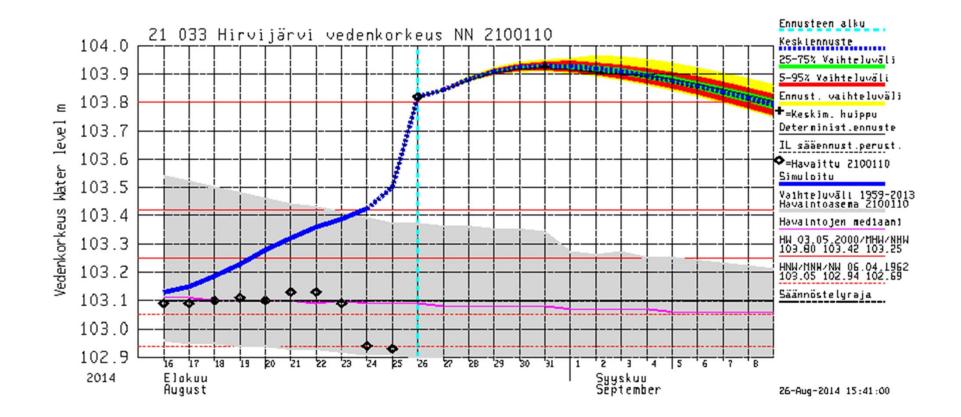
- To estimate the state of the hydrological system today
- Assimilation observations of:
 - discharge and water levels (over 400 stations)
 - snow water equivalent (over 150 stations)
 - SnowCoverArea satellite data
 - flood cover area (experimental)
- Corrects inputs of the model (daily precipitation and temperature)
- Simulation is corrected to agree with observations on a 1-2 year long period backward



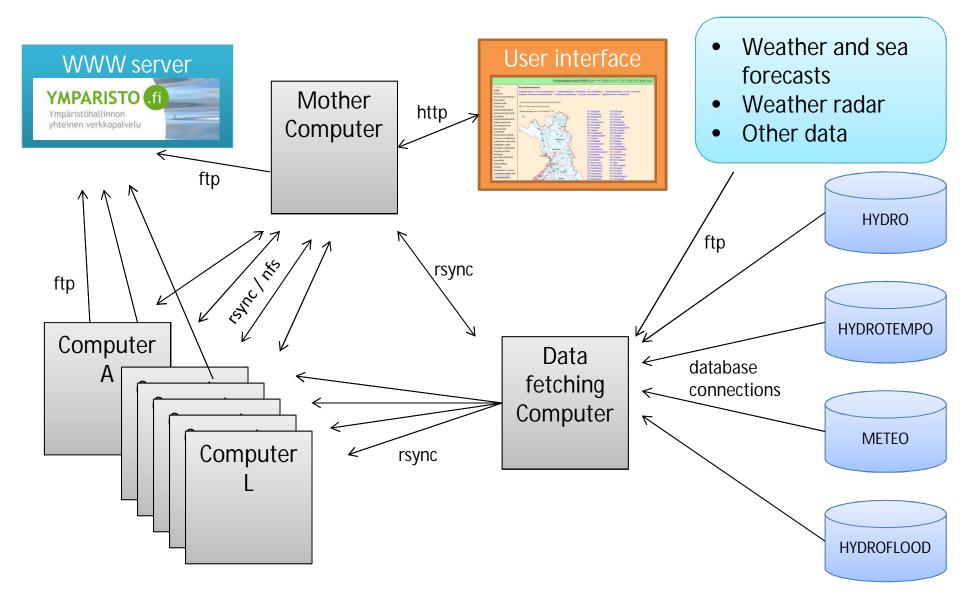
Example: forecast gone wrong when data is not filtered



Example: forecast gone wrong when data is not filtered



Data transfers in Watershed Simulation and Forecasting System



Data Control State Machine

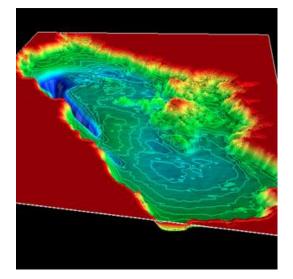
State: UNKNOWN (at the beginning)	State: OK	State: DISCARD	State: SUSPICION
U_AO (accept observ.) → U_FO	O_AO (accept observ.) → O_FO	D_AO (accept observ.) → D_FO	S_AO (accept observ.) → S_FO
U_FO (fetch observ.)	O_FO (fetch observ.)	D_FO (fetch observ.)	S_FO (fetch observ.)
→ U_RF	→ O_RF	→ D_RF	→ S_RF
U_RF (read flag) • ACCEPTABL. \rightarrow U_AO • REVISED \rightarrow O_AO • SUSPICIOUS \rightarrow U_AO • NULL \rightarrow O_AO	O_RF (read flag) • ACCEPTABLE \rightarrow O_AO • REVISED \rightarrow O_AO • SUSPICIOUS \rightarrow O_TR • NULL \rightarrow O_TR	D_RF (read flag) • ACCEPTABLE \rightarrow D_AO • REVISED \rightarrow O_ AO • SUSPICIOUS \rightarrow D_TR • NULL \rightarrow D_TR	S_RF (read flag) • ACCEPTABLE \rightarrow S_AO • REVISED \rightarrow O_ AO • SUSPICIOUS \rightarrow S_TR • NULL \rightarrow S_TR
	O_TR (test rejection)	D_TR (test rejection)	S_TR (test rejection)
	• PASS → O_TS	• PASS \rightarrow D_TS	• PASS → S_TS
	• FAIL → D_FO	• FAIL \rightarrow D_FO	• FAIL → D_FO
	O_TS (test suspicion)	D_TS (test suspicion)	S_TS (test suspicion)
	• PASS → O_AO	• PASS → O_AO	• PASS \rightarrow O_AO
	• FAIL → S_AO	• FAIL → S_AO	• FAIL \rightarrow S_AO

Wind data assimilation to Coherens (Shuku&Suito)

Local wind fields strongly impact on water current fields (Suito et al., 2014)

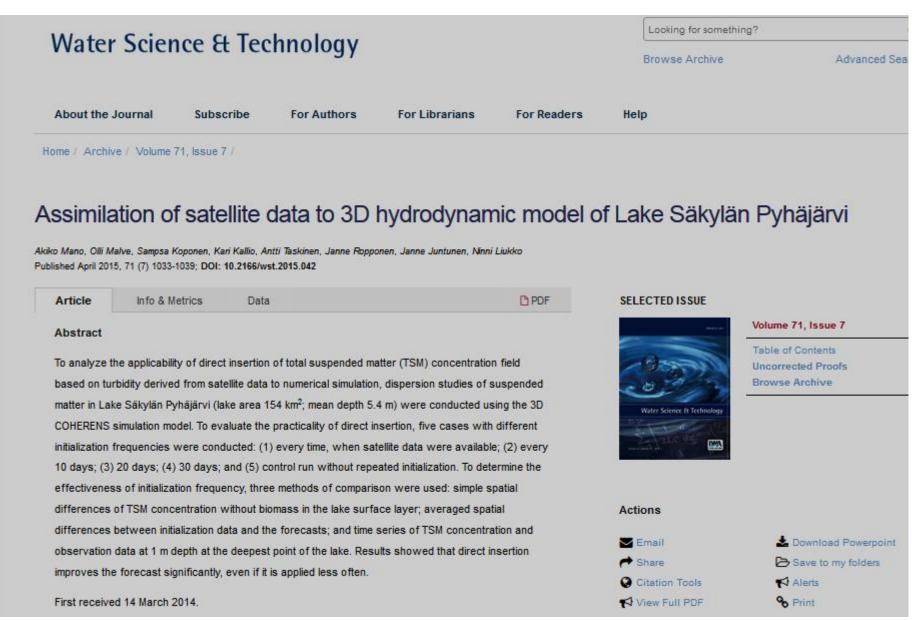
Coupled simulations between air flow and water flow are recommended

- Time consuming
- Complex (Confusing)



(http://www.ems.okayama-u.ac.jp/suito/)

How should we deal with the dilemma?



Data and methods

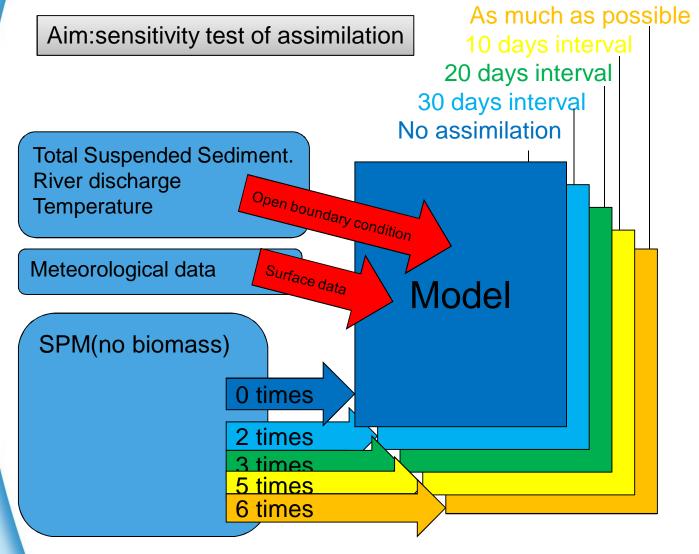
- Models
 - COHERENS V2(Luyten, 2011)
- Open boundary conditions at 3 river mouths:
 - River discharge (observation extracted from Hertta data base of the Finnish Environmental Administration)
 - Temperature (model results provided by VEMALA*)
 - Total suspended sediment (observation interpolated by linear function)
- Surface data
 - Meteorological data such as wind speed and direction, air temperature, humidity, cloud coverage and air pressure. (observation provided by Finnish Meteorological Institute)

*VEMALA: the water quality component of the Watershed Simulation and Forecasting System (Vehviläinen B et al., 2005) of the Finnish Environment Institute. This system simulates variables such as the transport of total phosphorus and nitrogen and suspended solids in land area, rivers and lakes. (Huttunen I. et al., 2008).

- Huttunen I, Huttunen M, Tattari S, Vehviläinen B. 2008. Large scale phosphorus load modelling in Finland. In Northern Hydrology and its Global Role, Volume 2, Sveinsson ÓGB, Garðarsson SM, Gunnlaugsdóttir S (eds). XXV Nordic Hydrological Conference 2008. NHP Report No. 50. Icelandic Hydrological Committee: Reykjavik; 548-556.
- Vehviläinen B, Huttunen M, Huttunen I. 2005. Hydrological forecasting and real time monitoring in Finland: The watershed simulation and forecasting system (WSFS). In Innovation, Advances and Implementation of Flood Forecasting Technology, Conference Papers, Tromsø, Norway, 17–19 October 2005.



Objective



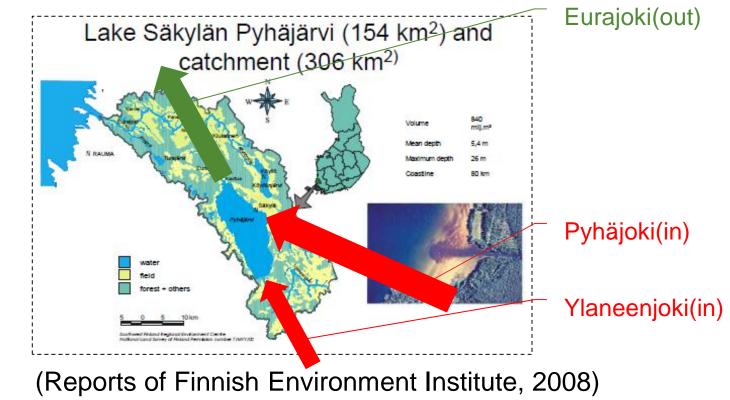
Tested cases and initialization timing

To evaluate the practicality of the assimilation (direct insertion) of the model, five model runs (1–5) were performed using different initialization frequencies: (1) every time, when satellite data were available; (2) every 10 days; (3) 20 days; (4) 30 days; and (5) the control run without repeated initialization (except for the initialization on 16 May 2009). The initialization timings were at noon on: (1) June 1, 8, 18, 21 and 26 and July 6; (2) June 1, 8, 18 and 26 and July 6; (3) June 1 and 21 and July 6; and (4) June 1 and July 6.

SYKE

Target area

Study area



Reports of Finnish Environment Institute 15/2008, 73 p. URN:ISBN: 987-952-11-3125-7 ISBN: 987-952-11-3125-7 (PDF). Vehviläinen B, Huttunen M, Huttunen I. 2005. Hydrological forecasting and real time monitoring in Finland: The watershed simulation and forecasting system (WSFS). In Innovation, Advances and Implementation of Flood Forecasting Technology, Conference Papers, Tromsø, Norway, 17–19 October 2005.

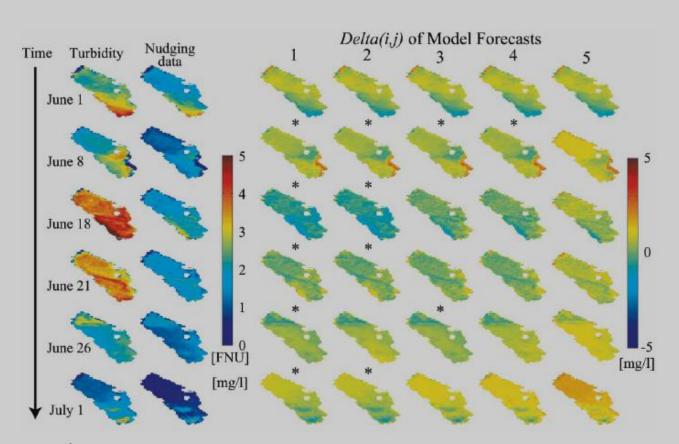
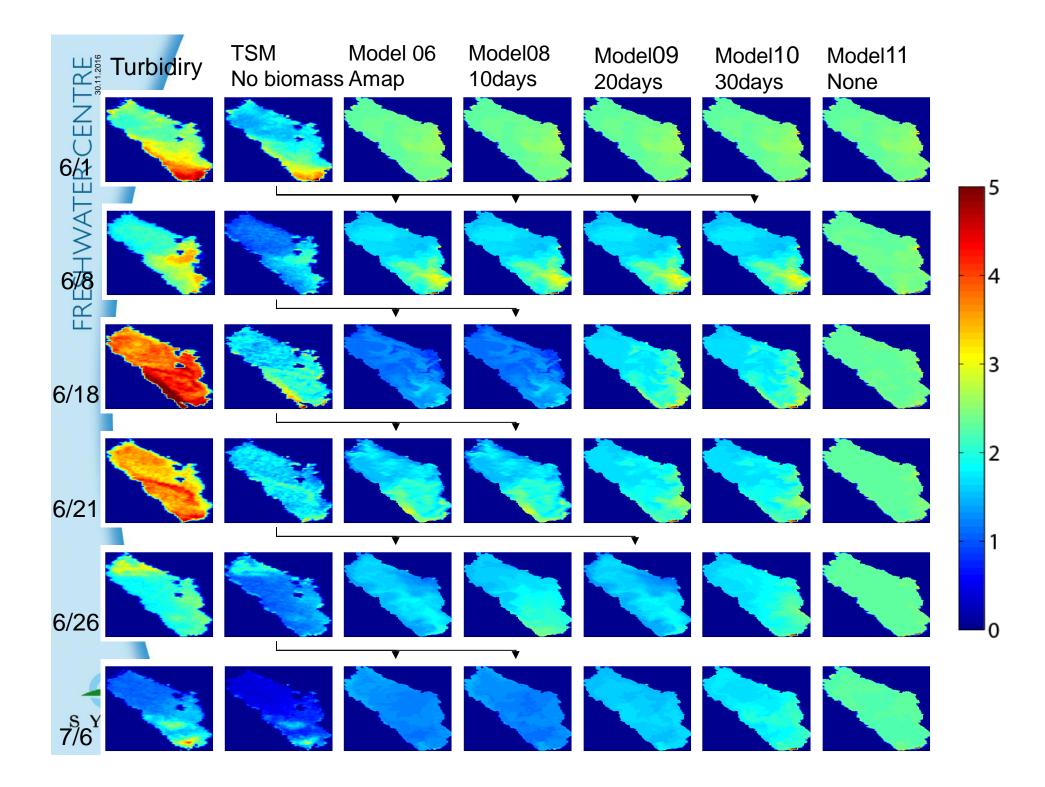
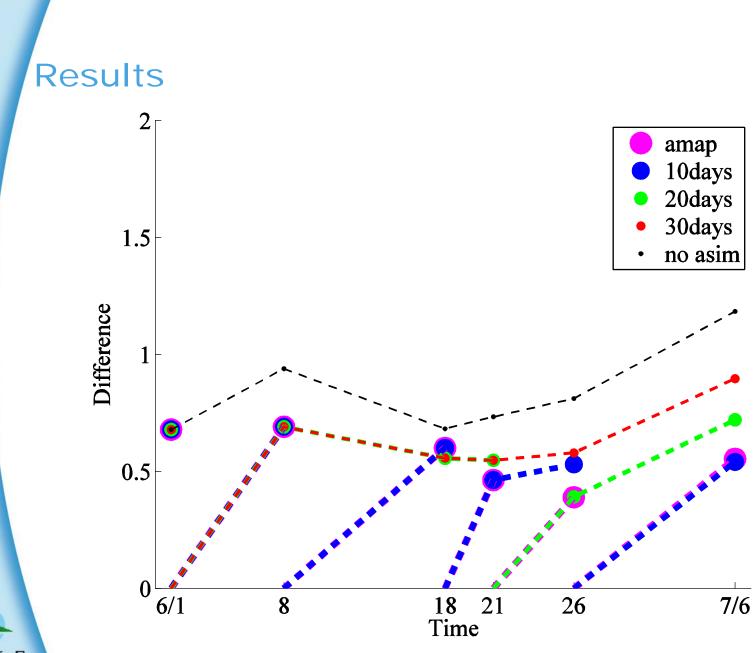


Figure 2 Comparison of turbidity from satellite data [FNU], TSM concentration without biomass (nudging data regarded as 'true' value) [mg/l] and *delta(i,j*) of model forecasts at the lake surface. The six rows represent dates from June 1 to July 6. The columns are turbidity derived from satellite data [FNU], converted TSM concentration without biomass [mg/l] (nudging data), and five cases of *delta(i,j*) taken with Equation (3) from left to right: every time that satellite data were available (1); every 10 days (2); 20 days (3); 30 days (4); and control run without repeated initialization (5). The initialization timings are shown with asterisk (*) symbols below each figure of *delta(i,j*).







SYKE

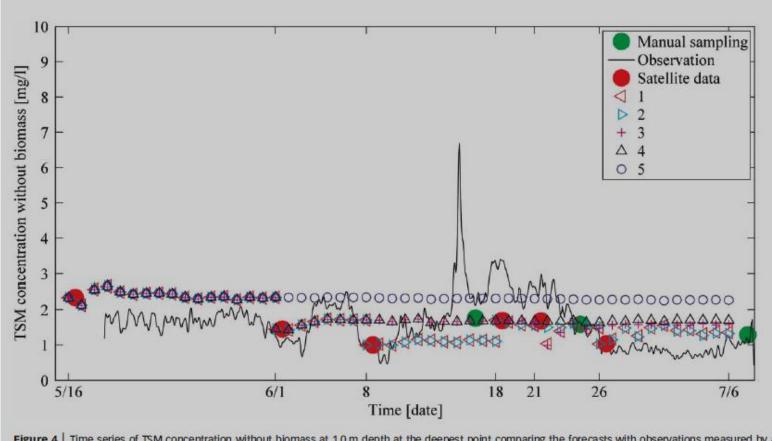


Figure 4 Time series of TSM concentration without biomass at 1.0 m depth at the deepest point comparing the forecasts with observations measured by automatic station and by manual sampling. Values were averaged from nine grids around the point. Green dots represent manual sampling converted from turbidity. The black line represents the automatic station converted from monitored turbidity using Equation (1), the same equation used to obtain the nudging data. Red dots are from satellite data (nudging data). The full colour version of this figure is available online at http://www.iwaponline.com/wst/toc.htm.

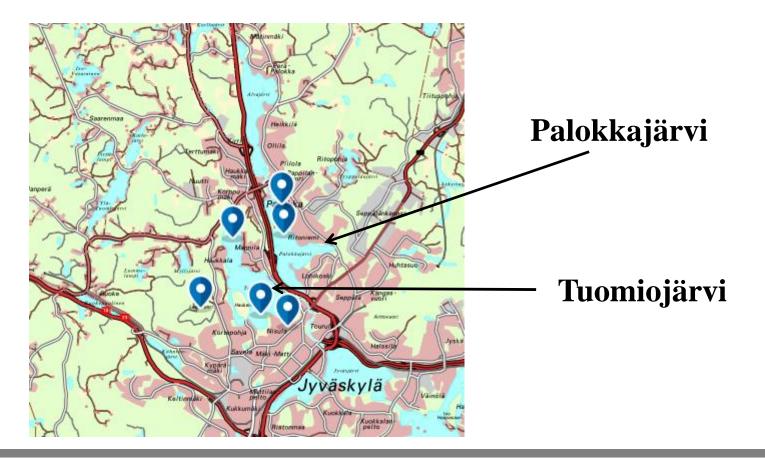
This paper demonstrates that the predictive performance of a coupled circulation-sediment model can be notably improved by incorporating satellite data, even if the used sediment model and assimilation method are simple and its parameterization of sediment transport is not exhaustively calibrated, as only the sinking velocity is calibrated at a single observation point. This is also true when direct insertion is applied less often, such as the interval of 30 days. In particular, the 10-day period shows values of Forecast RMSE increases that are smaller when compared to situations where no simulation is used at all.

For usability, until the necessary modules are implemented and until the demanding calibrations have been done, direct insertion can obtain more realistic forecasts of lake water quality. Future studies should be performed carefully to examine considerations of the possibilities of using satellitebased bathymetrical data and detailed sedimentation processes. Such studies can enrich our experience in using more sophisticated data assimilation methods.

Observation Station

Meteorological data

 \rightarrow wind direction, wind speed, temperature, humidity



Observation Station



Temporal Modeling

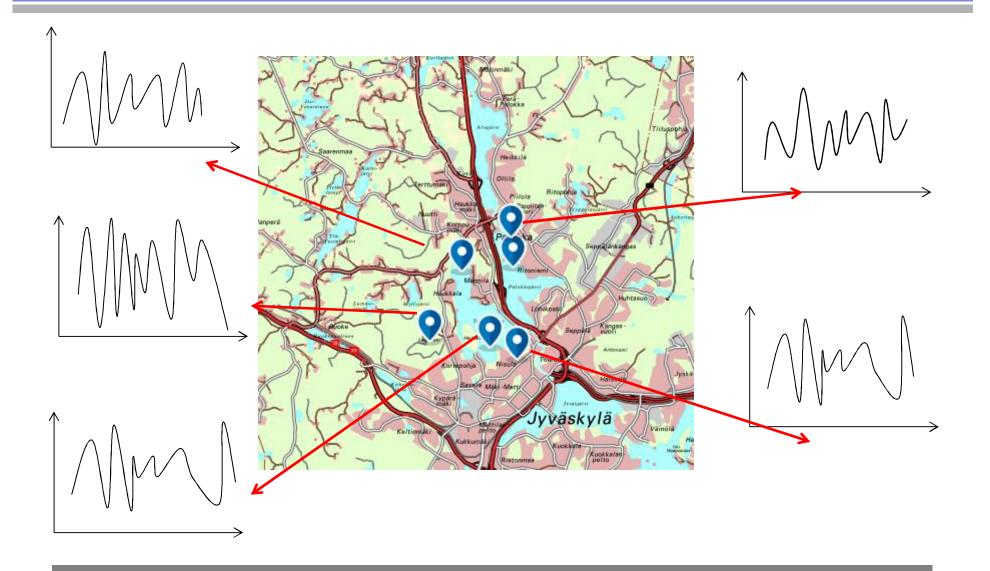
$$y_n = \sum_{i=1}^m a_i y_{n-i} + v_n$$

- $\int Y_n : n^{\text{th}} \text{ time-series data}$
 - a_i : *i*th AR coefficient
 - v_n : white noise
 - m: autoregressive order

<Yule-Walker method>

$$\begin{bmatrix} \hat{C}_{0} & \hat{C}_{1} & \cdots & \hat{C}_{m-1} \\ \hat{C}_{1} & \hat{C}_{0} & \cdots & \hat{C}_{m-2} \\ \vdots & \vdots & \ddots & \vdots \\ \hat{C}_{m-1} & \hat{C}_{m-2} & \cdots & \hat{C}_{0} \end{bmatrix} \begin{bmatrix} a_{1} \\ a_{2} \\ \vdots \\ a_{m} \end{bmatrix} = \begin{bmatrix} \hat{C}_{1} \\ \hat{C}_{2} \\ \vdots \\ \hat{C}_{m} \end{bmatrix} \quad \begin{bmatrix} \hat{C}_{i} & : \text{ auto-correlation function} \\ a_{i} & : \text{AR coefficient} \end{bmatrix}$$

Spatial Modeling



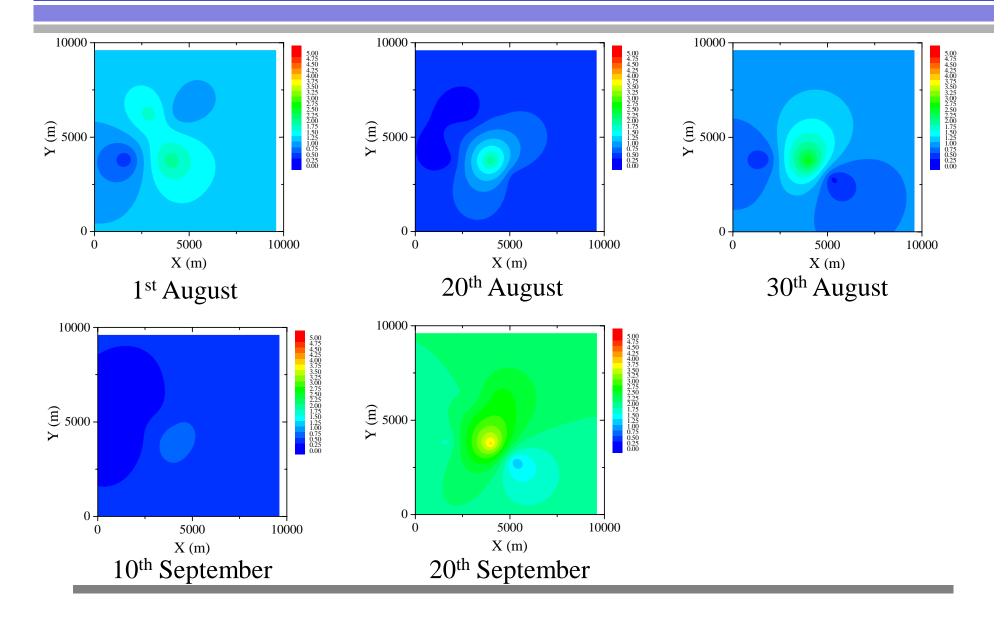
Ordinary Kriging

System equation

$$\begin{bmatrix} \gamma(y_1 - y_1) & \gamma(y_1 - y_2) & \cdots & \gamma(y_1 - y_n) & 1 \\ \gamma(y_2 - y_1) & \gamma(y_2 - y_2) & \cdots & \gamma(y_2 - y_n) & 1 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ \gamma(y_n - y_1) & \gamma(y_n - y_2) & \cdots & \gamma(y_n - y_n) & 1 \\ 1 & 1 & 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \\ \mu \end{bmatrix} = \begin{bmatrix} \gamma(y_1 - y_0) \\ \gamma(y_2 - y_0) \\ \vdots \\ \gamma(y_n - y_0) \\ 1 \end{bmatrix}$$

- γ : Variogram (auto correlation function)
- y_0 : Unknown value (wind speed and wind direction)
- $y_{i(1 \sim n)}$: Known value (wind speed and wind direction)
 - μ : Mean value
 - W_i : Weight

Interpolated data



Estimated local wind field

