

# Towards a combined meteorological-hydrological forecasting system

## Assimilation of in-situ and satellite snow data for hydrological forecasting in Sweden - a hydropower case study

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# Met-Hyd collaboration for surface DA at SMHI

Land surface model (parts of) – SURFEX.

Data assimilation method – EnKF.

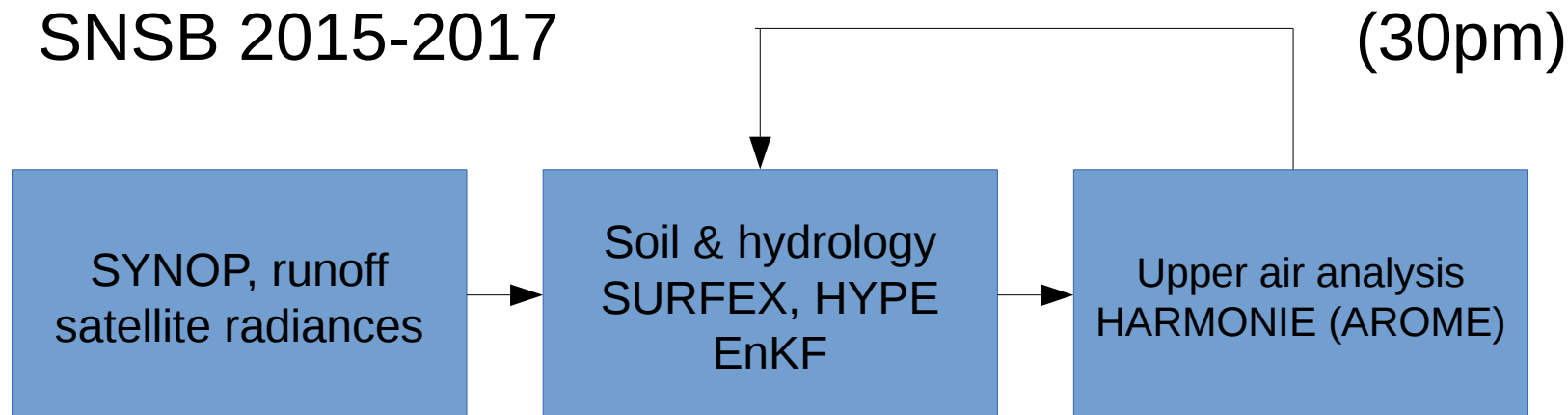
Observations – towards raw radiances / backscatter.

Work together in research projects:

- SNSB: DA for satellite-based measurements of the hydrosphere
- EU H2020: IMproving PRedictions and managment of hydrological Extremes

Collaboration with NILU, Météo France and HIRLAM partners.

SNSB 2015-2017



Assimilation of satellite-based measurements of the hydrosphere  
 - towards a combined meteorological-hydrological forecasting system



AMSR2 (GCOM-W1)



MIRAS (SMOS)

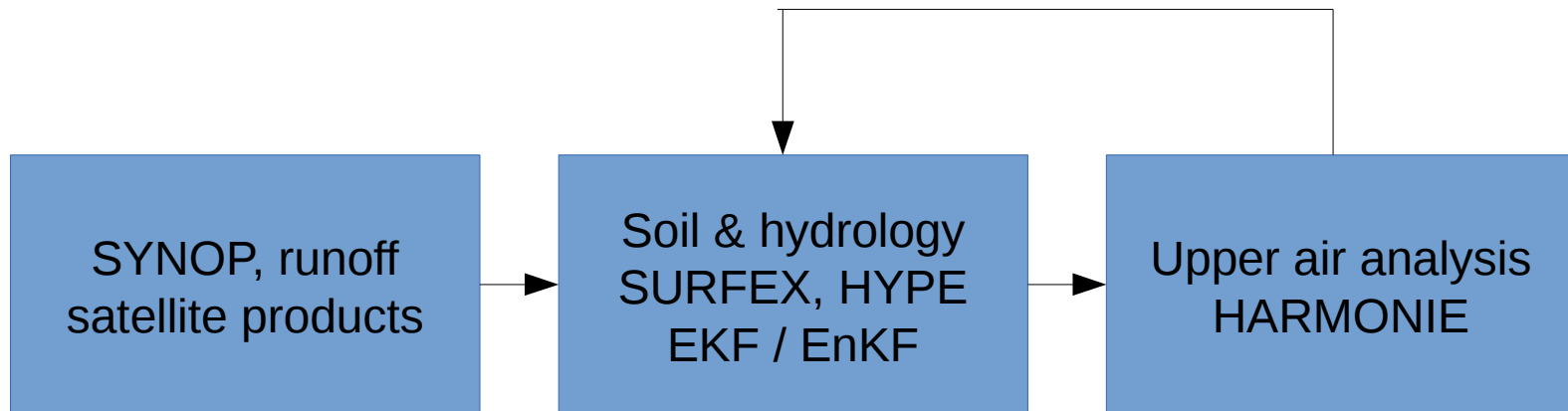


SAR (Sentinel-1)

## EU IMPREX 2015-2018

(43pm)

IMproving PRedictions and management of hydrological Extremes



Soil moisture from ASCAT, SMOS and AMSR2.

Snow water equivalent from H-SAF / GLOBSNOW / NASA (SSMIS, AMSR2)

# Harmonie MetCoOp

AROME cycle 38h1.b2

2.5 km, 750 x 960 grid points, 65 levels

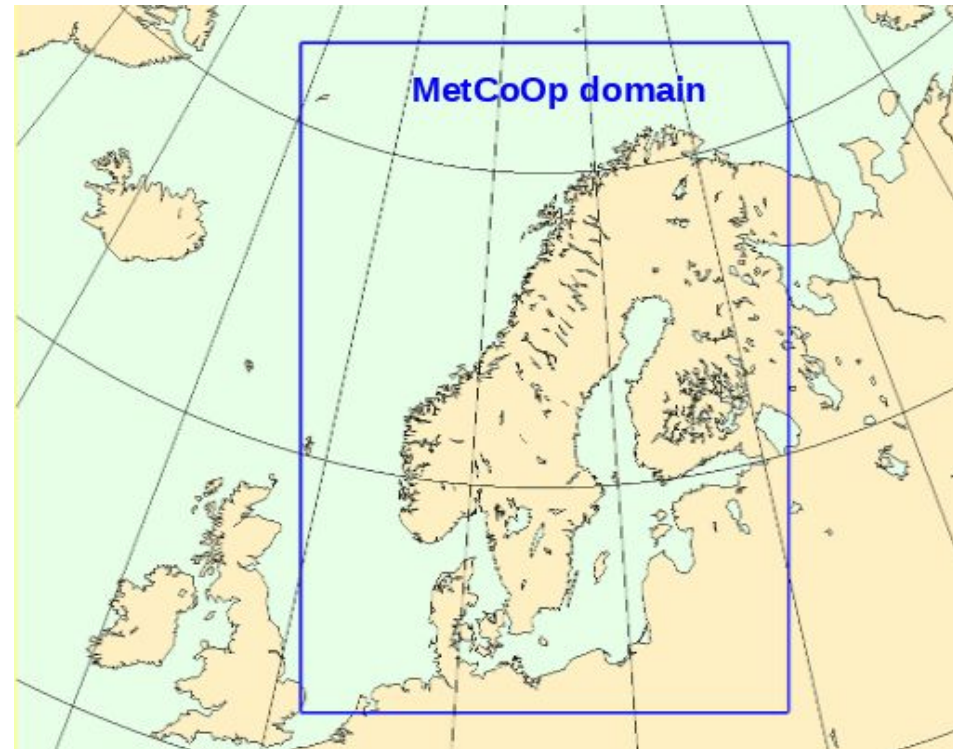
3D-Var, fc +60 hours

8 an + 8 fc per day @ 00,03,...,21

Obs: SYNOP, Aircraft, Buoy, Temp, GPS,  
AMSU-A, AMSU-B/MHS, IASI  
Radar, ASCAT winds.

ECMWF boundaries

Surface DA: CANARI-OI\_main



# SURFEX

Version 8

- Stable OpenMP implementation.
- ISBA-ES “Explicit snow” with multi layer snow packs.

Two patches (low and high veg) to match CMEM observation operator.

Couple with hydrological model

- Extend HYPE surface model with SURFEX.
- Introduce SURFEX river routing via OASIS.
- Interface with topological databases (e.g. Hydro1k).

## Remote sensing data and observation operators

Sentinel-1/SAR-C: wet snow, snow extent, (dry snow?)

- S1A\_EW\_GRDM\_1SDH
- Extra Wide swath mode VV+VH and HH+HV, ca 25 x 80 m
- MEMLS3&a by Proksch et al. (2015)

GCOM-W1/AMSR2

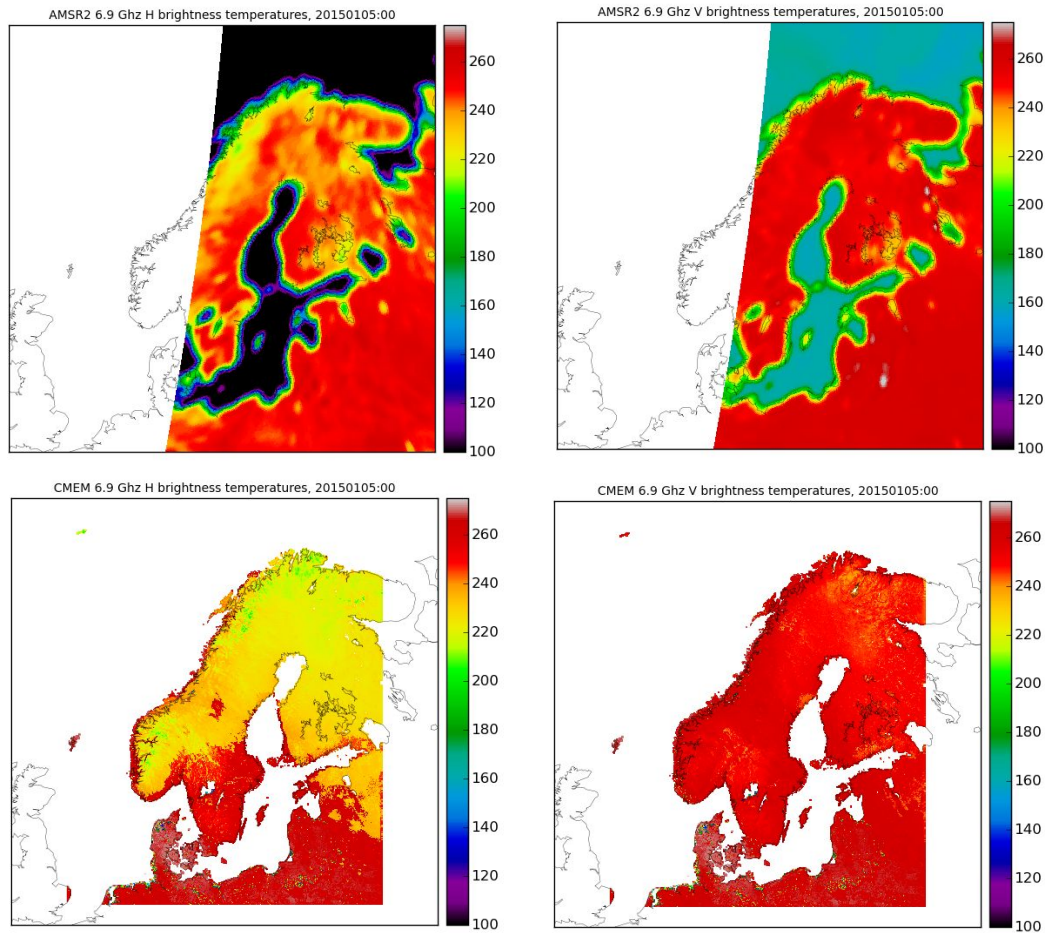
soilm (7 Ghz), deep (10, 19 GHz), moderate (37 GHz), shallow snow (89 GHz)

- L1SGRTBR
- Level 1R V,H, ca 40 x 60 km
- Community Microwave Emission Modelling Platform (CMEM): 1 – 20 Ghz
- FASTEM + RTTOV?

SMOS/MIRAS, L band 1.4 GHz: soil moisture

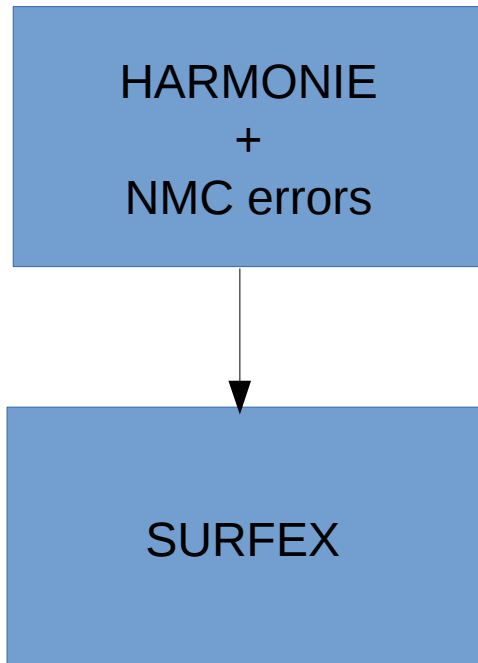
- MIR\_BWSD1C
- Level 1C Browse Brightness Temperatures, dual (or full) polarization, ca 50 x 50 km, ISEA 4-9 hexagonal grid.
- CMEM + FASTEM (water)?

# First technical test: AMSR2 level 1C, 6.9 GHz





## Spatially correlated errors in the forcing



$$x_k = fc_{48} - fc_{24} \quad (m \times 1)$$

$$X = [x_1, \dots, x_n] \quad (m \times n)$$

$$B = X X^T \quad (m \times m)$$

$$B_s = X^T X = E_s D_s E_s^T \quad (n \times n)$$

$$X E_s (X E_s)^T = X X^T = B$$

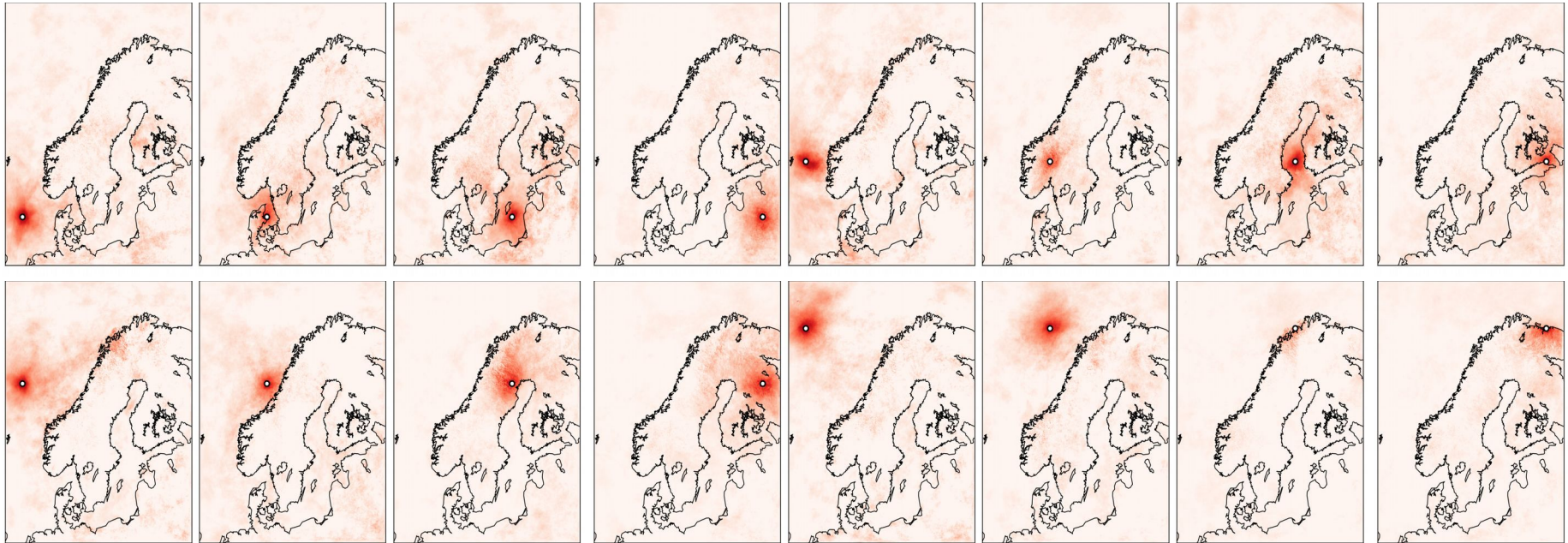
$$= E_l D_s E_l^T, \quad E_l = X \hat{E}_s \quad (m \times n)$$

$$z = E_l D_s^{1/2} e, \quad e \in N(0, I)$$

$$E\{z z^T\} = E_l D_s E_l^T = B$$

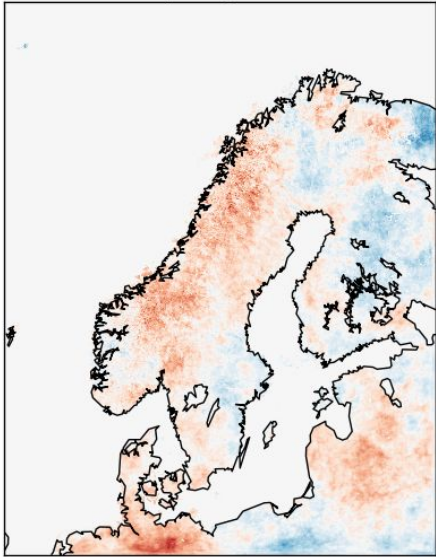
$$m \times n \approx npar * 500\,000 \times n$$

# 16 columns of B (t2m), fc48-fc24: 20140401-1231

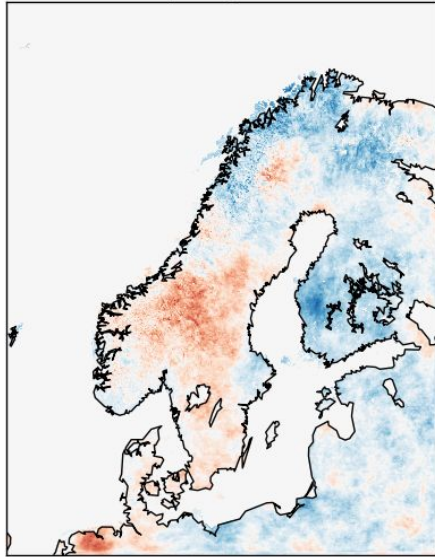


## Four examples of spatially correlated t2m errors (z)

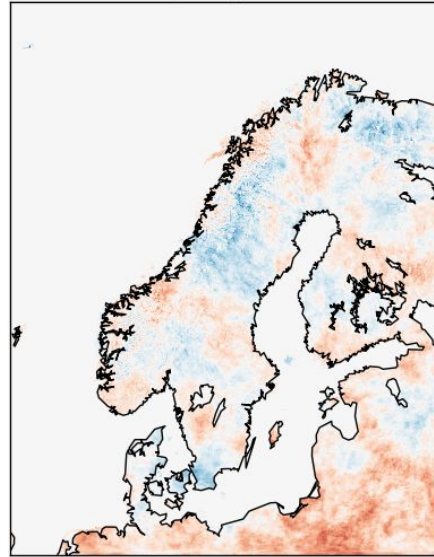
One realization:spatially correlated t2m error



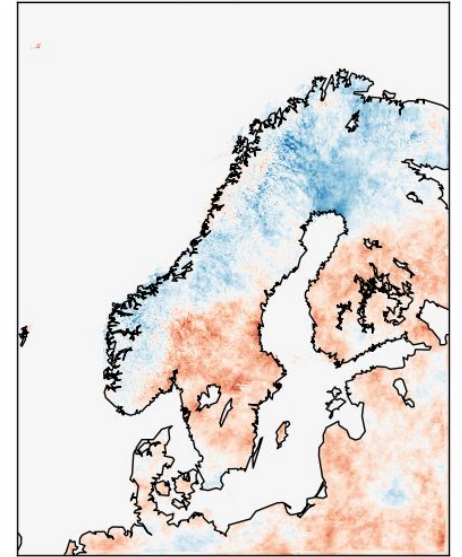
One realization:spatially correlated t2m error



One realization:spatially correlated t2m error

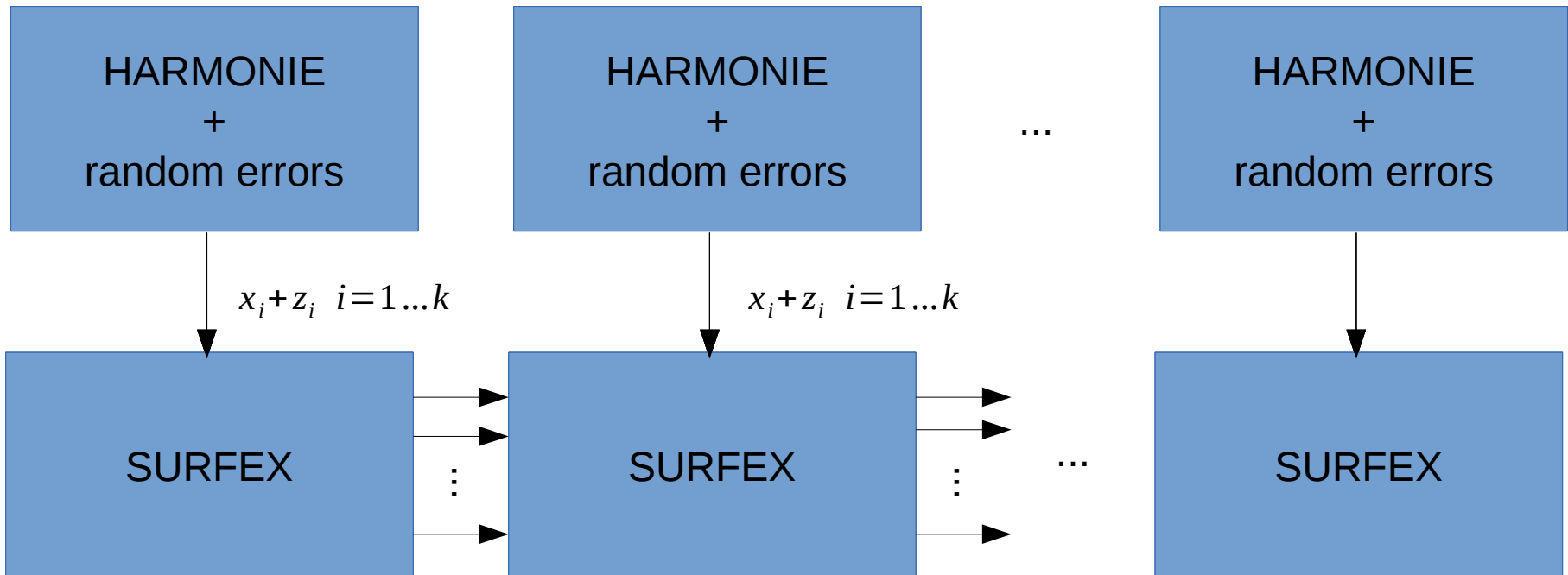


One realization:spatially correlated t2m error



These are not fc48-fc24 differences, but samples drawn using  $z = E_l D_s^{1/2} e$ ,  $e \in N(0, I)$

## Creating an initial SURFEX ensemble (spin-up)



How many members (k) are called for?

How long time do we need to run to spin up a sufficiently rich ensemble?

# Questions

NWP

Alternative (EnKF or En2DVar) DA for t2m and rh2m?

EnKF

Ensemble Kalman Filter for SURFEX – from N 1D to 1 ND. Cycling members?

How to introduce systematic perturbations? Time shifts?

Need for adding horizontally/vertically correlated errors to the SURFEX state?

How to assimilate runoff observations – long time window EnKF?

Observations and observers (obop)

How to make Sentinel-1 data fit HARMONIE scale – work with pdf:s?

How to make HARMONIE fit scale of SMOS/AMSR2 – footprint/antenna func?

Water emissivity at SMOS freq, FASTEM? RTTOV for AMSR2 37 and 89 GHz?

# Assimilation of in-situ and satellite snow data for hydrological forecasting in Sweden - a hydropower case study

## EO and in-situ snow data in Sweden

### Forcing data

- P, T interpolated (4x4 km<sup>2</sup>)
- Elevation EU-DEM (25x25 m<sup>2</sup>)

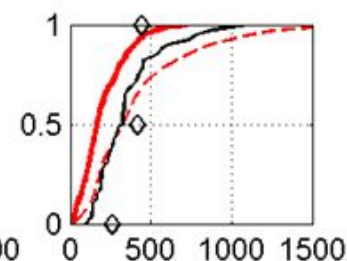
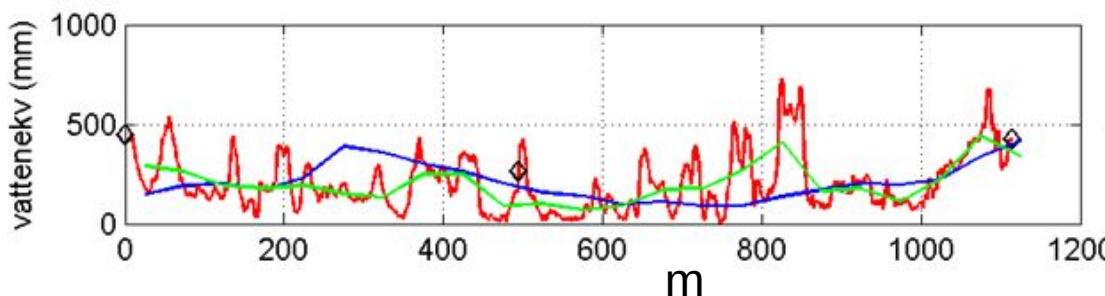
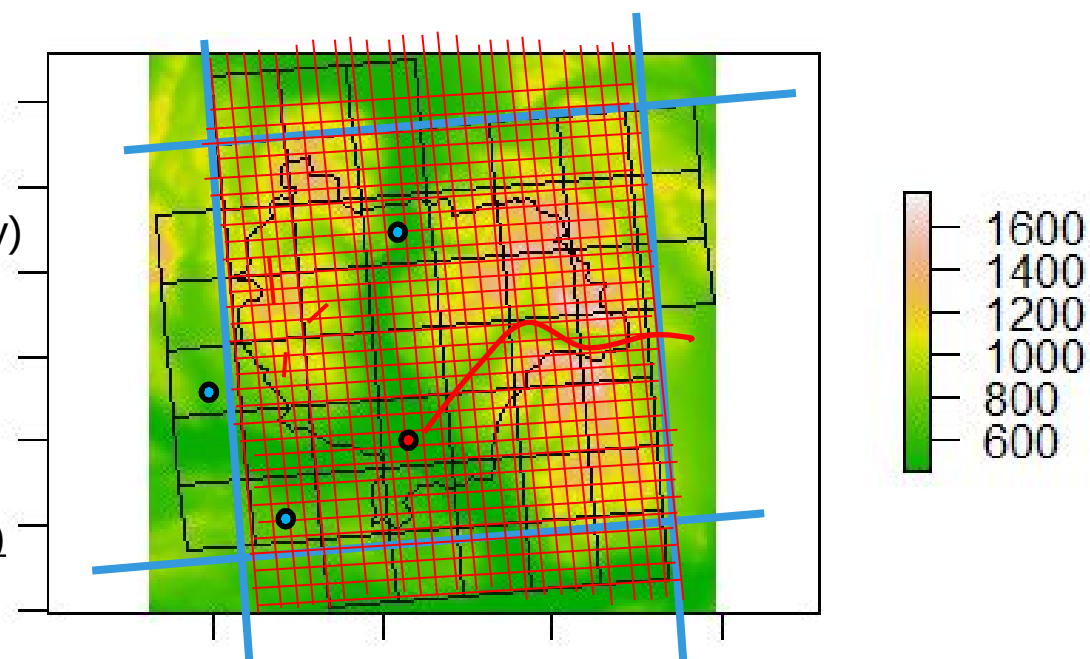
### Snow data

- SMHI snow depth stations (point, daily)
- Hydropower companies:
  - SWE point data (bi-weekly)
  - Snow courses (once per year)

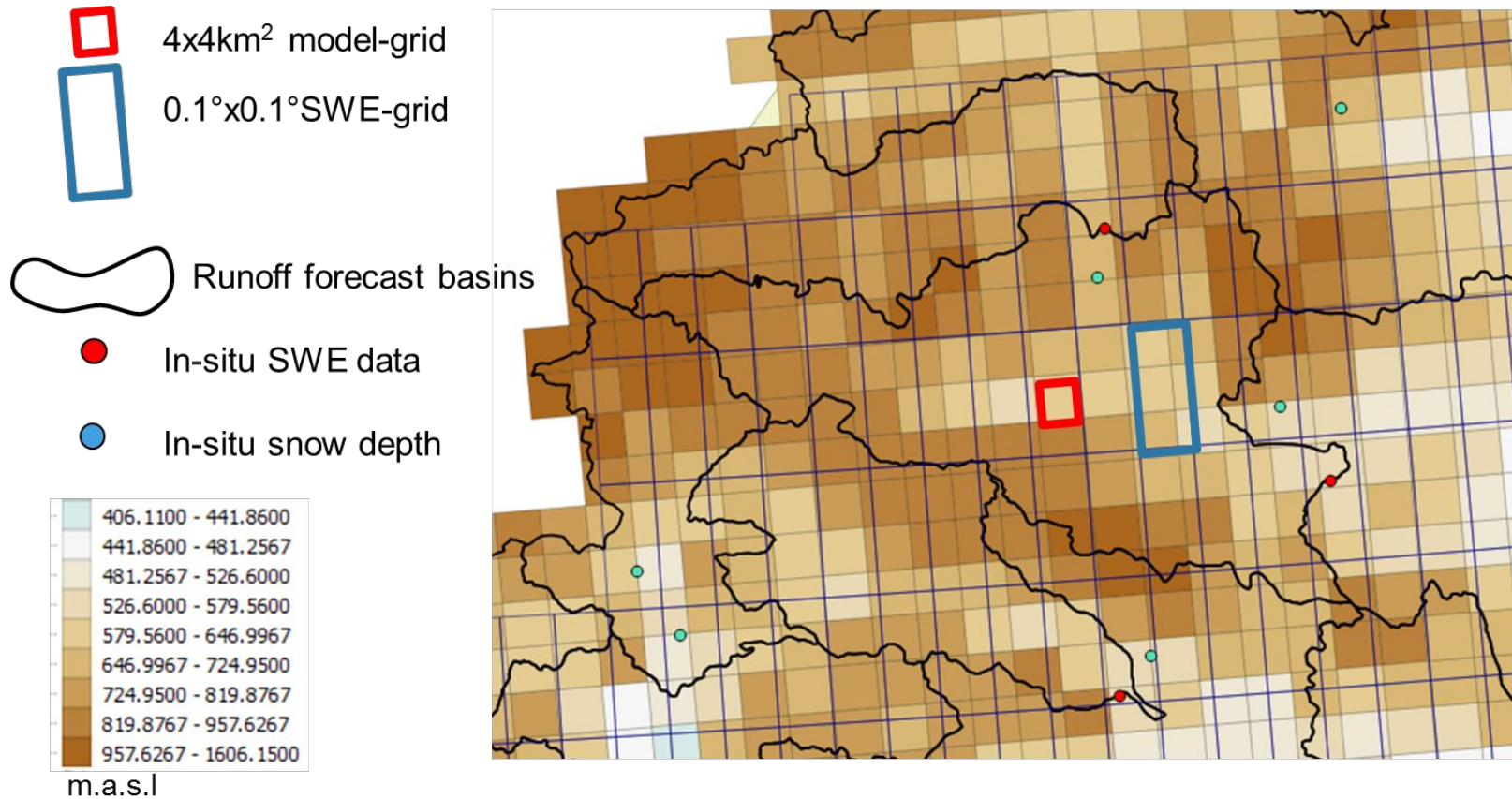
### Satellitdata (GlobSnow, CryoLand, etc)

Fractional snow cover 1x1 km<sup>2</sup>

Snow water equivalent 25x25 km<sup>2</sup>

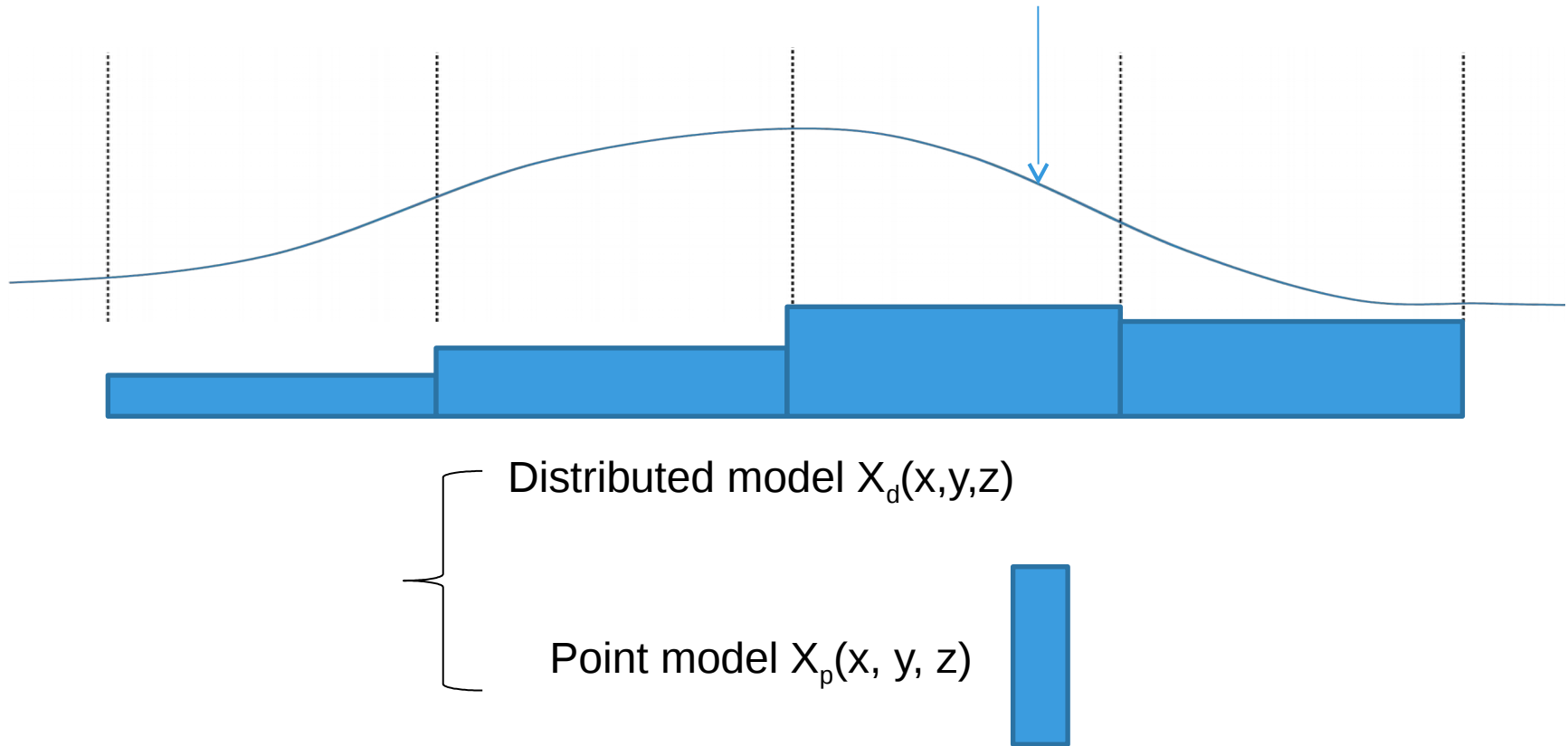


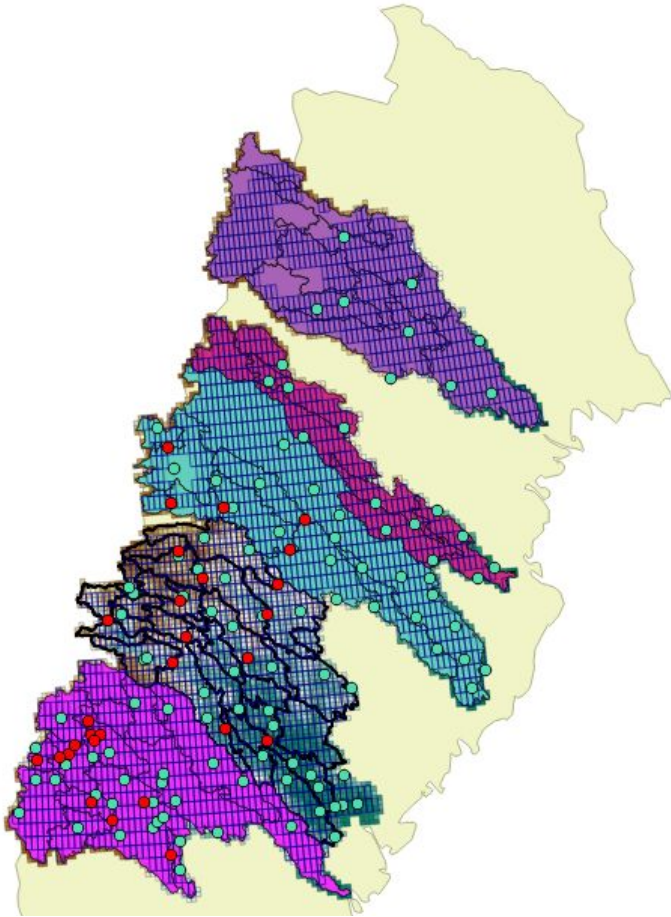
## Distributed hydrological model (HBV-type)





## Assimilation of point observations in a distributed model





## HOPE model application

- Simple HBV type of snow/soil model on the 4x4 km<sup>2</sup> PTHBV grid
- Up to 160 SLC classes for different snow accumulation/melt regimes
- Runoff is aggregated for the VRF forecast areas in Lule river, Skellefte river, Ume river, Ångerman river and Indals river.

## Snow data

- **SMHI** snow depth stations
- **VRF/VF/SVF** snow water equivalent point data at regulation dams
- CryoLand SWE and Snow cover (satellite)
- Local runoff estimated for each forecast area

## Two options for snow data integration:

### 1) Evaluation of Model and Data agreement

**KGE** = agreement in terms of **mean value**, **standard deviation** and **correlation**

### 2) Model state-updating using EnKF (Evensen, 1994)

$$X^a = X^p + K(Y - HX^p) \quad K = \frac{C_{XY}}{C_{YY} + R}$$

**All model states (non only snow) updated** as a function of covariance between the states and the 'innovation' in the snow variables

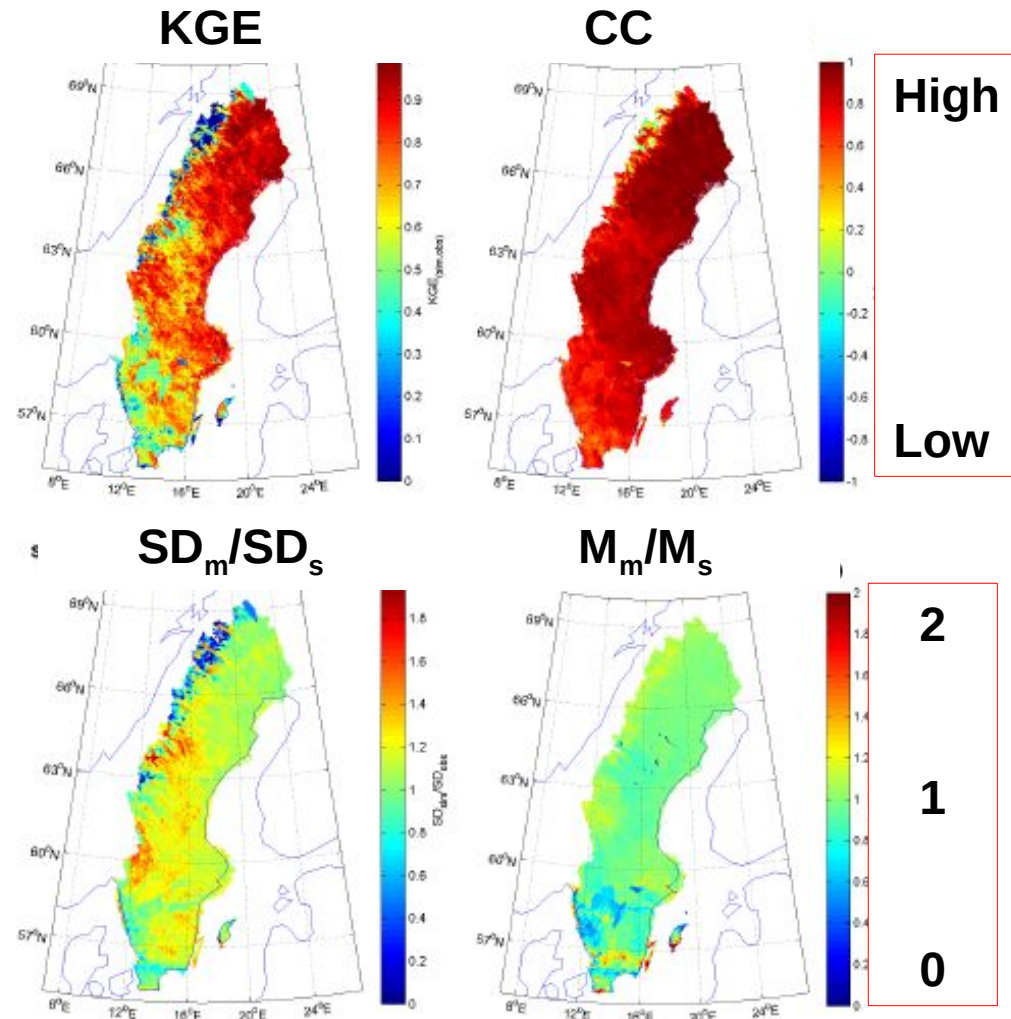
**Important that model and data do not disagree too much!**

# Model and data comparison – FSC

optical product ENVEO/SYKE

Pan-European

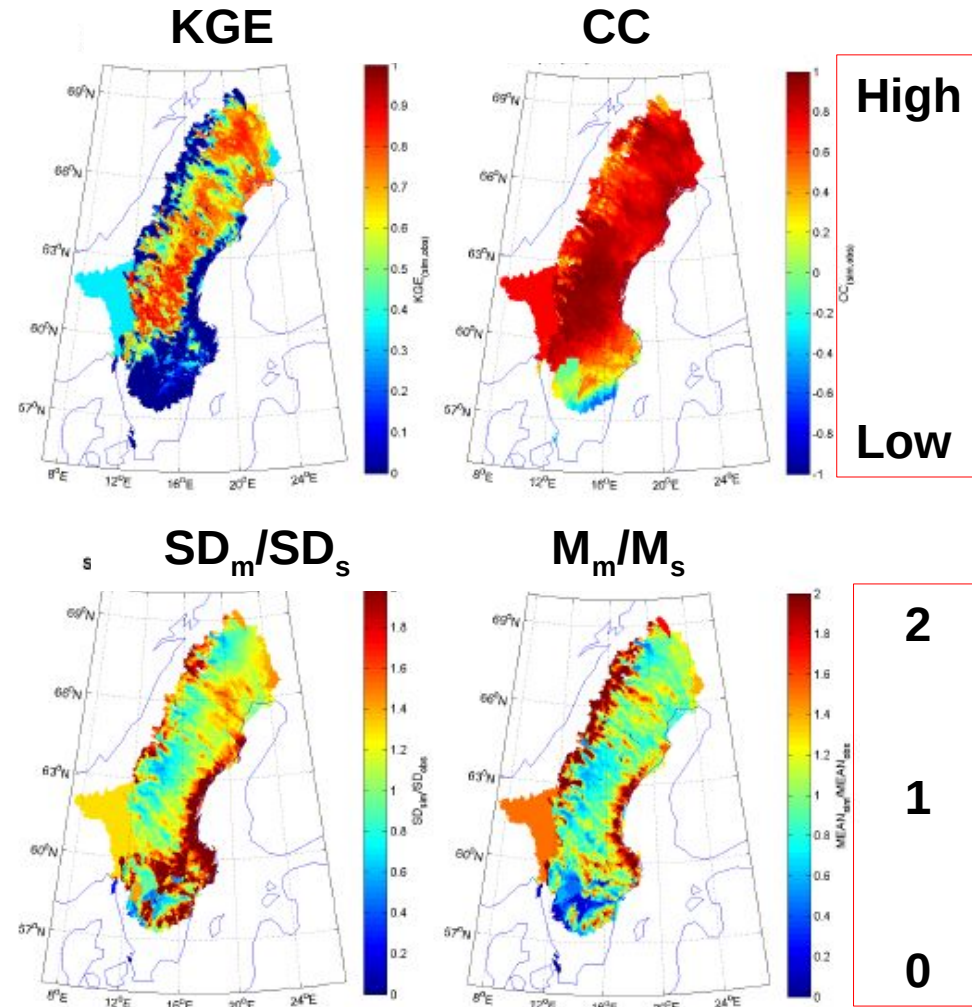
- In general a very good agreement between model and satellite data throughout Sweden
- However, the temporal variability is different in the most alpine part of the mountains in northern Sweden
- Transmissivity model is well-adapted to boreal forests.



# CryoLand SWE vs S-HYPE modellen

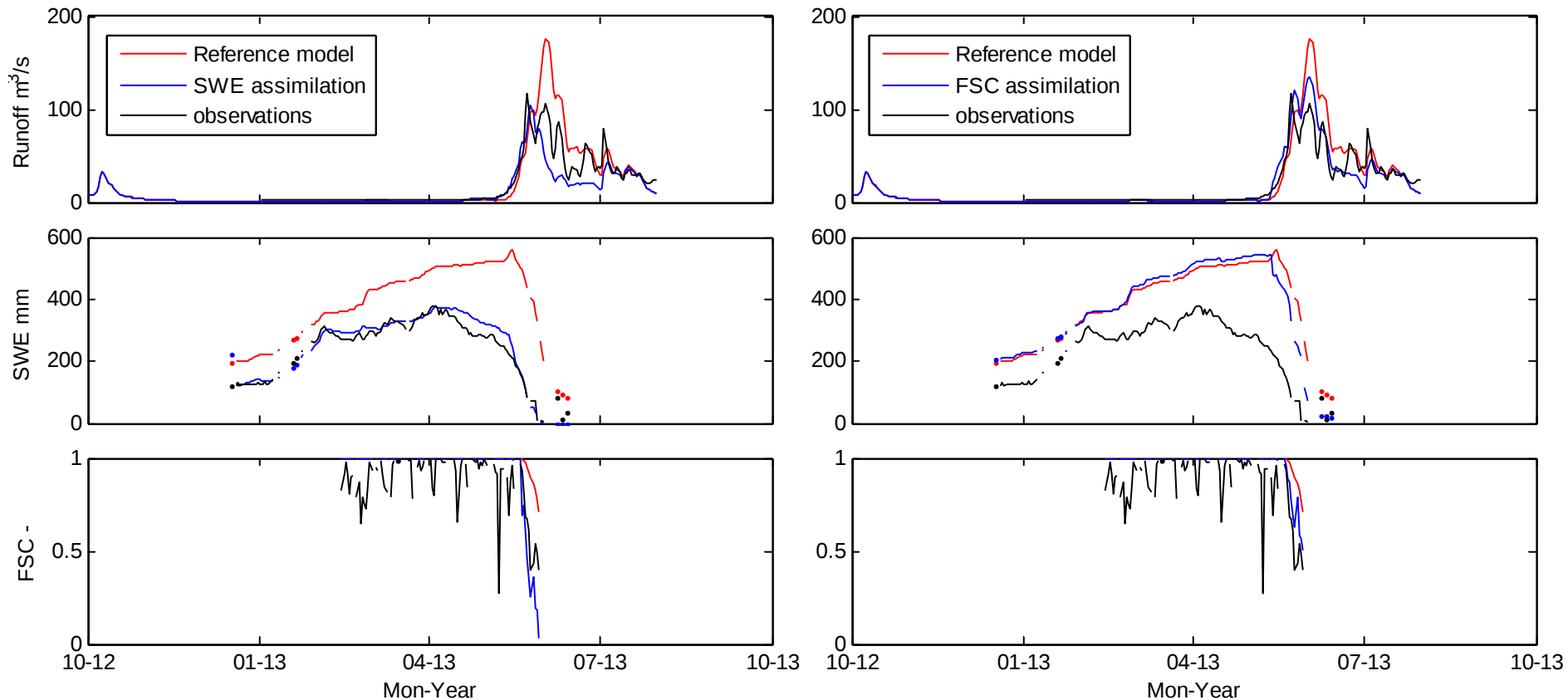
Pan-European SWE product (FMI)

- **Good agreement** in central part of middle and northern Sweden:
  - Forests
  - Non-mountain areas
  
- **Correlation is high** (except for the south)
  
- **Variability and Mean value differs:**
  - In the south (little snow and lakes)
  - along the east coast
  - western mountain range
  
- **Problem for the satellite or model?**
  - Mountains, surface water, coastal areas, spatial distribution of snow



## Good exmple: Abiskojokki, northern Sweden.

Both SWE and FSC data improve stream flow simulations



## Spring flood forecast assimilation experiments

- 5 test areas
- 6 year snow observations (2010-2015)
- Initialization for forecast start dates 15/2, 15/4 och 15/6 assimilating:
  - Local runoff (Q)
  - Snow water equivalent (SWE) – in-situ
  - Snow depth (SD) – in situ
  - Snow fraction area (FSC) – from CryoLand

		Area	Snow water equivalent at forecast start (mean 2010-2015)			
Område	Älv	km <sup>2</sup>	15/2	15/4	15/6	15/6 (% av 15/4)
Tjaktjajaure	Luleälven	2256	376	520	258	50%
Riebnesjaure	Skellefteälven	976	290	440	147	33%
Överuman	Umeälven	653	402	663	236	36%
Kultsjön	Ångermanälven	1095	357	498	118	24%
Landösjön	Indalsälven	1453	148	182	11	6%

## Results spring melt volume forecasts

- Assimilation until forecast start – forecast using ensemble of historical years
- Assimilation initialization is better in most cases than the reference run
- In-situ SWE and satellite based FSC is the most consistent improvement

### Relativt volymfel (%) (medel av absolut volym fel 2010-2015)

<u>Prognos 15/4-31/7</u>							
Område	Ref	Ens0	EnsQ	EnsSWE	EnsSD	EnsFSC	EnsAll
Tjaktjajaure	74.8	71.2	70.0	<b>39.8</b>	41.6	301.4	46.7
Riebnes	35.2	31.6	41.2	<b>15.3</b>	18.8	28.9	25.5
Överuman	20.8	20.1	22.3	20.1	21.6	<b>12.8</b>	26.9
Kultsjön	10.2	9.3	15.4	16.0	32.9	<b>5.4</b>	42.2
Landösjön	26.4	24.8	40.9	23.7	15.2	<b>14.6</b>	23.2
antal lägst fel				2		3	
antal < Ref		5	1	4	3	4	3



## Comparing forecasts 15/2 15/4 och 15/6

- Snow data assimilation important throughout winter
- Largest improvement later in the winter and through the melt period

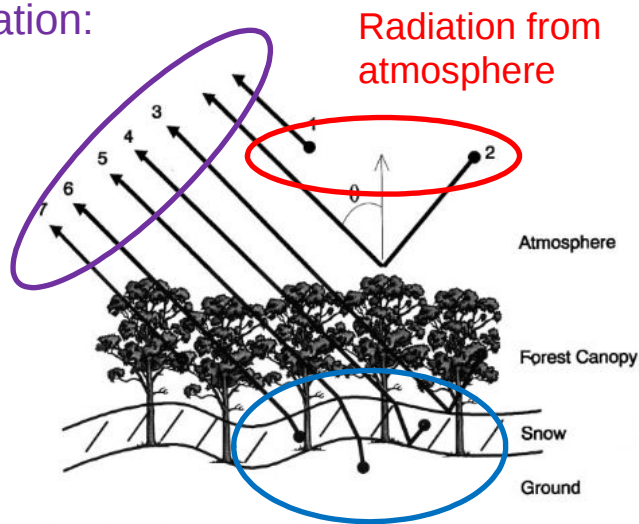
		Ref	Ens0	EnsQ	EnsSWE	EnsSD	EnsFSC	EnsAll
Relativt volymfel (%)	15/2	12.8	12.6	15.1	13.5	15.9	12.6	15.9
	15/4	23.1	21.5	29.9	18.8	22.1	15.4	29.4
	15/6	28.6	26.8	34.6	23.0	23.1	22.8	27.6
Relativ förbättring (%)	15/2		-2	20	-13	19	-26	26
	15/4		-7	37	-48	14	-29	61
	15/6		-6	27	-41	0	-1	17

Next step: assimilate in-situ data, passive microwave data directly in the hydrological model?

## Radiation emission model

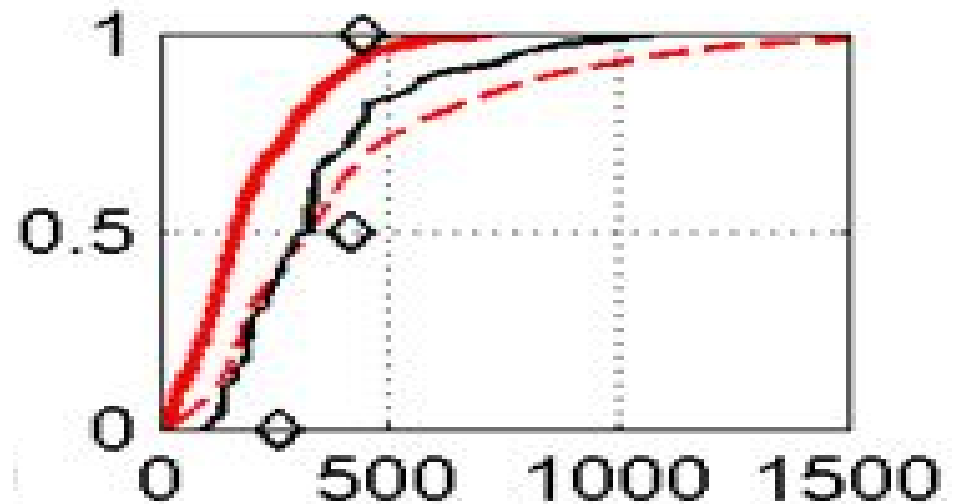
Ex from Pullianen and Hallikainen (2001)

Satellite observed radiation:



Radiation from ground (soil, snow, vegetation)

Spatial distribution of snow  
(from model or from in-situ data)



## Conclusions

- Assimilation of satellite and in-situ snow data reduced snow melt runoff forecast errors with 5-50%.
- In-situ snow water equivalent and satellite based fractional snow cover gave the most consistent improvements when assimilated in the model.
- Further studies will be focused on assimilation of passive microwave data and in-situ data directly in the snow hydrological models.

Tack för uppmärksamheten

