# **Representation of snow in NWP**

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with contributions by

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### Introduction

Snow observations Snow forecast Snow data assimilation COST ES1404







and methodologies. The Action will co-ordinate efforts to address these

Trans-Domain Proposals

Memorandum of Understanding

<u>COST Action ES1404</u> aims at building a better connection between snow measurements and models, between snow observers, researchers and forecasters, for the benefit of various stakeholders and the entire society

#### Aim of the Action:

To enhance the capability of the <u>research community</u> and <u>operational services</u> to provide and exploit <u>quality-</u> <u>assured and comparable</u> regional and global observation-based data on the <u>variability of the state</u> <u>and extent</u> of snow.



snow water equivalent – temperature - density – grain size – albedo .



COST ES1404 suggested working groups

WG1: Physical characterization of snow

WG2: Instrument and method evaluation

<u>WG3:</u> Snow data assimilation and validation methods for NWP and hydrological models

COST ES1404 will <u>not</u> focus on prognostic snow modelling/parametrisations! And we are talking about <u>snow cover</u>, not snowfall!

#### What does the snow cover mean for NWP?



Observed temperature profiles from the level of -1m (soil) to 50m (mast) when there was 0.4m snow on ground in Sodankylä and air cooled 20K during 24h

# The case of Lake Ladoga, January 2012: role of clouds as predicted by NWP



#### http://netfam.fmi.fi/Lake12/HIRLAM\_Ladoga-anim\_hkrp2014.m4v

Kalle Eerola et al., 2014. Impact of partly ice-free Lake Ladoga on temperature and cloudiness in an anticyclonic winter situation – a case study using HIRLAM model. http://tellusa.org .Video prepared by Homa Kheyrollah Pour, 2014

## NWP for emergency warnings



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Local and remote sensing snow observations

SYNOP and climate stations: Ultrasonic or manual snow depth measurements • Represent local conditions

Satellite instruments: Passive microwave sensors - e.g SMSI Coarse resolution wide area snow water equivalent **Optical/NIR - e.g.MODIS**  High resolution snow extent Limited by cloud and light problems Active microwave - e.g. SAR from ESA's Sentinel-1 Very high resolution indication of wet snow Narrow swath – infrequent data

#### Availability of various snow observations over Finland



Remote Sensing of Environment 147 (2014) 65-781,

#### Example of the first snowfall in November 26-28 2012

IMS



(Land-SAF was not available those days)

#### What are the most valuable snow observations for NWP?

#### SYNOP + climate station snow observations, which provide also no-snow information

- Should be more widely available via GTS
- Should include the national group with no-snow information
- NWP models should read correctly the extended SYNOP code

#### Remote sensing observations

- 1) Snow water equivalent by passive microwave sensors
- Snow extent seen by visible and derived from passive and active microwave signals
- 3) Snow wetness indicated by SAR instruments

Dilemma of using satellite data: ready-made products or spatialization + assimilation of the signals within the surface DA of NWP models?

- Satellites with varying instrument specifications come and go – building long-lasting operational systems is difficult
- Products contain assumptions and rely on additional data sources different from those applied in NWP framework
- NWP model may provide up-to date background based on prognostic snow parametrizations – for quality control, for assimilation

e.g. IMS and Globsnow SWE are products, while SAR backscattering from the just launched Sentinel-1 would represent a raw signal

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# Verification of MODIS snow cover maps with web cameras









Hróbjartur Þorsteinsson et al. 2014



#### Prognostic snow schemes available in SURFEX

Single-layer	D95	Douville et al. (1995a,1995b)
Multi-layer	Explicit-Snow (ES)	Boone (2000); Boone and Etchevers (2001)
Multi-layer	Crocus	Brun et al. (1989,1992); Vionnet et al. (2012)

Table 4.1: Summary of the snowpack schemes available in ISBA\*

#### **ISBA + D95** Operational in HARMONIE-SURFEX

Layers in snowpack: One Prognostic variables: SWE, snow density, snow albedo but no separate snow temperature/liquid water content Data assimilation: SWE updated with optimally interpolated snow depth

\* SURFEX SciDoc v.2

#### Prognostic snow schemes available in SURFEX

Single-layer	D95	Douville <i>et al.</i> (1995a,1995b)
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Table 4.1: Summary of the snowpack schemes available in ISBA\*

#### **ISBA + ES** Next operational in HARMONIE-SURFEX?

Layers in snowpack: ca. 3 Prognostic variables: heat content > temperature and liquid water, layer thicknesses and densities Data assimilation: None yet Other features: Possibly to couple MEB

\* SURFEX SciDoc v.2

Multi-Energy Balance (MEB)





MEB is designed to work with

- snow schemes ES (3-L) and CRO (requires separate snow energy balance)
- soil scheme ISBA-DIF (diffusion) with patches (separate forest/grass/bare land)

#### 2D offline experiment – Snow Water Equivalent



#### With MEB:

- Less snow in forested areas in mid winter (10-20 kg m<sup>-2</sup>) due to snow interception
- More snow in forested areas late in winter (20-50 kg m<sup>-2</sup>) due to a combination of radiation and turbulence effects
- The melting is delayed



Difference SWE ISBA-MEB – ISBA Average over 1978-2008 in kg m<sup>-2</sup>

Patrick Samuelsson, 2014

## Explicit snow and Crocus snowpack model



Brun, E., V. Vionnet, A. Boone, B. Decharme, Y. Peings, R. Valette, F. Karbou and S. Morin, Simulation of northern Eurasian local snow depth, mass and density using a detailed snowpack model and meteorological reanalyses, J. Hydrometeor., 14, 203–219, doi: 10.1175/JHM-D-12-012.1, 2013.

# NWP output can be used to drive stand-alone Crocus



#### Data picked from HIRLAM and HARMONIE

Lowest model level variables to be used as atmospheric forcing for SURFEX/CROCUS, wind drift

Snow-related variables for comparison/validation against observations



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### CROCUS on Kistufell (23.257W 66.074N)



HIRLAM forecast (resolution 7 km/65L) temperature, humidity, wind, downward SW and LW radiation and (snow) precipitation were applied to drive CROCUS for the autumn 2013 at Kistufell target point



### CROCUS on Kistufell (23.257W 66.074N)



HARMONIE/AROME forecast (1km/65L) temperature, humidity, wind, downward SW and LW radiation and (snow) precipitation were applied to drive CROCUS for the autumn 2013 at Kistufell target point



### CROCUS on Kistufell (23.257W 66.074N)



The result is different because of the different atmospheric forcing by two weather models

CROCUS could also be driven by observations, but they are seldom sufficiently available







#### How to use advanced snow schemes in NWP?

#### Our aim:

Multilayer prognostic soil + Soil data assimilation + Multilayer prognostic snow - vegetation + Snow data assimilation

The problem:

Multilayer soil and snow schemes and MEB have been developed for climate models without any data assimilation

Solution would require some work:

Soil Scheme	Soil DA	Snow scheme	Snow-veg scheme	Snow DA	Application
Force-	OI/EKF	D95	none	snowOI(Canari)	NWP
restore	+OI ( <u>Canari</u> )	ES	MEB	[[snowOI/VAR/EKF]] +snowOI(Canari)	NWP
			none	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
		CRO	MEB	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
			none	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
	none	D95	none	none	climate
		ES	MEB	none	climate
			none	none	climate
		CRO	MEB	none	climate
			none	none	climate
Multi-	{OI/VAR/EKF}	D95	none	snowOI(Canari)	NWP
layer 	+Ol( <u>Canari</u> )	ES	MEB	[(snowOI/VAR/EKF)] +snowOI(Canari)	NWP
			none	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
		CRO	MEB	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
			none	[{snowOI/VAR/EKF}] +snowOI(Canari)	NWP
	none	D95	none	none	climate
		ES	MEB	none	climate/ NWP?
			none	none	climate
September		CRO	MEB	none	climate
2014 / EK			none	none	climate

Table by Ekaterina Kurzeneva, 2014

## Future NWP model for dedicated applications?



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#### **Operational snow analyses**

Model	Observations	Assimilation	Operational
СМС	SYNOP	OI	1999
ECMWF	SYNOP IMS	Cressman Cressman Ol	1987 2004 2010
HARMONIE	SYNOP	OI	2010
HIRLAM	SYNOP SYNOP Globsnow	Cressman Ol Ol	1995 2004 Experimental
Met Office	IMS	Update	2009

**Richard Essery** 

http://www.ecmwf.int/newsevents/meetings/workshops/2013/Polar\_prediction/Presentations/Essery.pdf

### **Operational CANARI snow analysis** spreads snow observations to model grid in horizontal

Optimal interpolation of snow depth of SYNOP station observations

Snow depth > SWE using assumed snow density

Background error correlations include horizontal and vertical terms\*



\* presentation by Mariken Homleid, ASW13

#### **Operational snow analyses**



SYNOP snow depths and FMI snow pits (from Timo Ryyppö) Hirlam snow analyses (from Laura Rontu) ECMWF snow analyses (from Patricia de Rosnay)

#### **Richard Essery**

http://www.ecmwf.int/newsevents/meetings/workshops/2013/Polar\_prediction/Presentations/Essery.pdf

#### Development of snow data assimilation methods

Assimilation of ground-based snow data requires:

- good background estimate of snow density
- good estimates of observation and model errors (underestimation of model / observation error ratio is worse than overestimation)
- may not require advanced data assimilation techniques

The use of a Kalman Filter will still be beneficial if information can be propagated to unobserved state variables through off-diagonal elements in the gain matrix, either due to <u>correlation between state</u> <u>variables in the model</u> or the use of a <u>complex observation operator</u> such as a <u>microwave emission model</u> or <u>assimilation of radiance data</u>.

#### **Richard Essery**

http://www.ecmwf.int/newsevents/meetings/workshops/2013/Polar\_prediction/Presentations/Essery.pdf

## **Concluding remarks**

Simple snow schemes are used in present NWP models, with snow mass, density, albedo in one layer but advanced multilayer prognostic snow schemes exist

Horizontal interpolation via optimal interpolation is applied to conventional snow depth observations but a lot more remote sensing and local snow cover observations exist

Advanced data assimilation methods will be needed to combine multilayer prognostic snow and soil parametrizations with various types of remote-sensing observations in operational NWP models

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## COST ES1404 suggested working groups

WG1: Physical Characterization of Snow WG2: Instrument and Method Evaluation WG3: Snow data assimilation and validation methods for NWP and hydrological models

# WG1: Physical Characterization of Snow

Challenges for optical remote sensing of snow:

•Pixels contain not only snow

- •Definition of 100% snow cover?
- •Definition of fractional snow cover
- Definition of snow during melting periods
  Spatial and temporal variation of snow properties



Is this 100% snow coverage? What is the coverage, when the trees are also white?

### WG1: Physical Characterization of Snow properties

- **Task 1.1: Identifying and assessing the essential snow variables** (snow grain size, snow depth, snow density, snow covered area, snow temperature, thermal conductivity, albedo, full microstructure, snow impurity concentration)
- **Task 1.2: Physical characterization of essential snow variables:** Relationships between variables; How snow variables are affected by atmospheric thermodynamic and dynamic (wind drift) forcing; Response time scales of different snow variables, ...
- Task 1.3: Snow network optimization, data quality control and homogenization: How much snow variables vary in space during different seasons and over different environments . User needs; Ground-truth with satellitebased observations; Quality control and homogenisation recommendations.
- Task 1.4: Harmonization of snow observations in terms of measured variables: Practical and organizational actions needed for harmonization will be assessed. The potential generated from a harmonized network of snow observations for network operators and data users will be assessed.
- **Task 1.5: WG1 interacts with WG2** as to techniques applied for measurement of essential snow variables and to WG3 as to the physical characterization of modelled snow variables.

## WG2: Instrument and Method Evaluation

There is a strong need to intercompare, standardize and validate the methods in Europe



Snow height map measured over a 4-metre diameter area versus one snow pit (= point).



Snow grain size in 2D grid versus 3D Snow Specific Area (SSA) measurements

## WG2: Instrument and Method Evaluation

- Task 2.1: Review of space-borne and ground-based sensors/instrumentation with estimates of their uncertainties
- Task 2.2: Guidelines for in-situ snow observations and related training (Accuracy of methods and instruments; Error sources; Representativeness of point values; Recommended length and sampling resolution for line measurements).
- **Task 2.3: Spectroradiometry for snow studies**: Making field spectrometer data consistent; Harmonising data processing (e.g. spectral sampling, geo-rectification in case of airborne measurements, filtering techniques for continuous spectra).
- **Task 2.4: Methods to measure snow grain size:** current worldwide development with varying grain size definition. The wealth of measurement techniques requires a thorough assessment and inter comparison.
- Task 2.5: Methods to measure mechanical properties of snow: High relevance for avalanche formation. Field tests for harmonising snow stability assessment across European avalanche services, and for testing snow properties using snow penetrometry (SnowMicroPen).
- **Task 2.6: WG2 interactions with WG1** as to the definition of measured snow variables and with WG3 in terms of observation uncertainties.

# WG3: Snow data assimilation and validation methods for NWP and hydrological models

Which snow observations do we assimilate into Numerical Weather Prediction models and how?

#### PRESENTLY:

- We take from SYNOP stations only snow depth
- We select only snow extent from satellite data
- We convert data to model grid using the method of "Optimal Interpolation"



#### How to assimilate more remote sensing observations?

- Observations: predicted and observed parameters differ!
- Methods: advanced methods to be developed to assimilate satellite retrievals instead of remote sensing snow products!



# WG3: Snow data assimilation and validation methods for NWP and hydrological models

- Task 3.1: Overview assessment of future perspectives as to snow observations in NWP, hydrology and climate studies for the sake of validation and assimilation.
- Task 3.2: Developing methods to update non-observed forecasted physical snow properties (e.g. snow temperature, wetness, density profiles, and mechanical properties) based on the observed ones
- Task 3.3: Advancing assimilation of new and developing satellite observations of snow properties and their combination with conventional insitu snow data.
- Task 3.4: Improving wider use of conventional snow observations in NWP, hydrological and climate models (i.a. observations from HR national networks).
- Task 3.5: Quantifying model and observational errors for data assimilation from results of WG1 and WG2.
- **Task 3.6**: **Remote sensing and in-situ observations fusion techniques** for snow-melt modelling in all weather conditions (esp. under cloudy conditions.

# Near future NWP snow tasks related to COST ES1404

Acquire more and ensure full usage of SYNOP/climate station snow depth observations

Introduce passive microwave SWE observations (Globsnow via Hydro-SAF) into the snow analysis

Research task: Develop advanced data assimilation methods to combine multilayer prognostic snow to various types of remote-sensing observations

# THANK YOU!