

ECMWF Data Assimilation Training course

Land Surface Data Assimilation

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Outline

Part I (Monday 10 March)

- **Introduction**
- Snow analysis
- Screen level parameters analysis

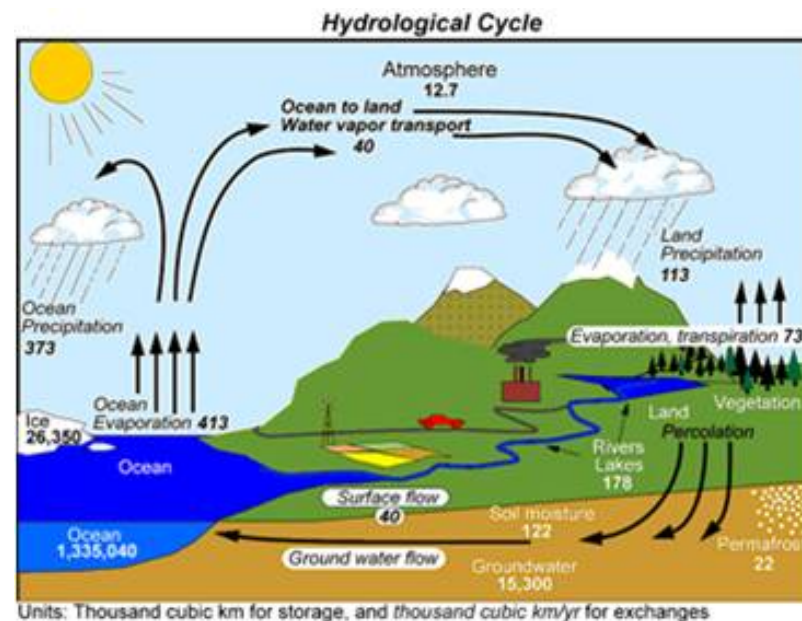
Part II (Tuesday 11 March)

- Soil moisture analysis
- Summary and future plans

Introduction: Land Surface in NWP

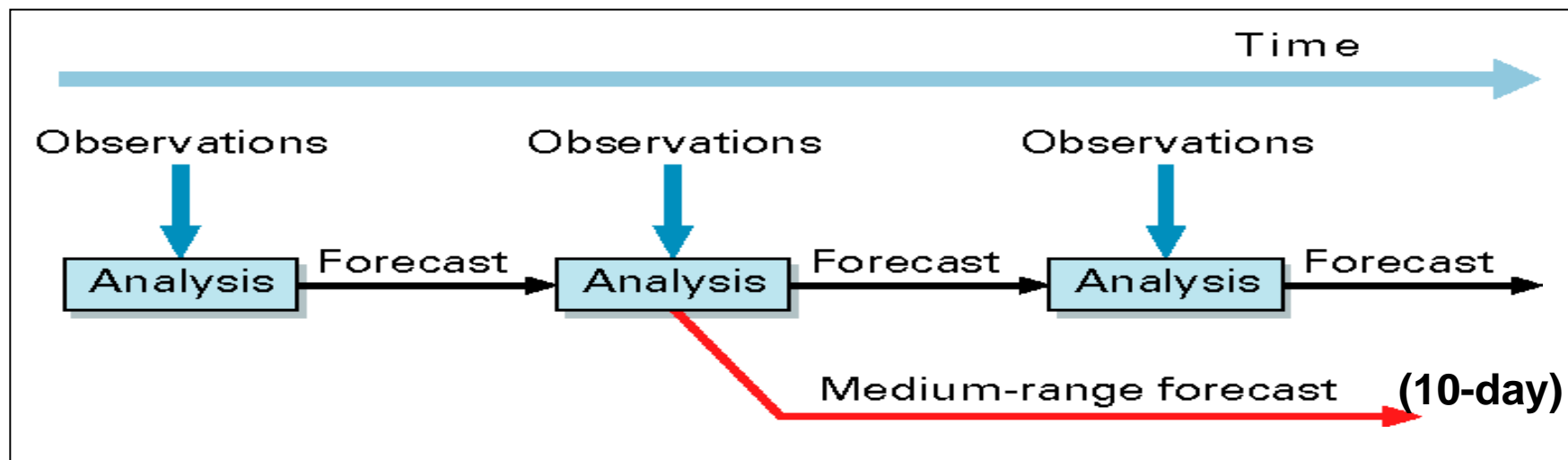
- **Land surfaces:** Boundary conditions at the lowest level of the atmosphere
- **Land surface processes** → Continental hydrological cycle, interaction with the atmosphere on various time and spatial scales, strong heterogeneities
- Crucial for near surface weather conditions, whose high quality forecast is a key objective in NWP
- Land Surface Models (LSMs) prognostic variables include:
 - Soil moisture
 - Soil temperature
 - Snow water equivalent, snow temperature, snow density

- Land surface initialization:
Important for NWP & Seasonal Prediction
(Beljaars et al., Mon. Wea. Rev, 1996, Koster et al., 2004 & 2011)



Trenberth et al. (2007)

Introduction: ECMWF Integrated Forecasting System (IFS) data assimilation system



Data Assimilation System:

Provides best possible accuracy of initial conditions to the forecast model

- 4D-Var for atmosphere
- Land surface data assimilation
- SST and Sea Ice analysis



- Surface and upper air analyses are running separately in parallel
 - Feedbacks provided through the first guess forecast initialised with the analysed fields
- Surface and Atmospheric DA are weakly coupled

Introduction:

Land Surface Data Assimilation (LDAS)

Snow depth analysis

- Approaches: **Cressman** (DWD, ECMWF ERA-I), **2D Optimal Interpolation (OI)** (ECMWF, CMC, JMA)
- Observations: *in situ* snow depth and NOAA/NESDIS IMS Snow Cover

Soil Moisture analysis

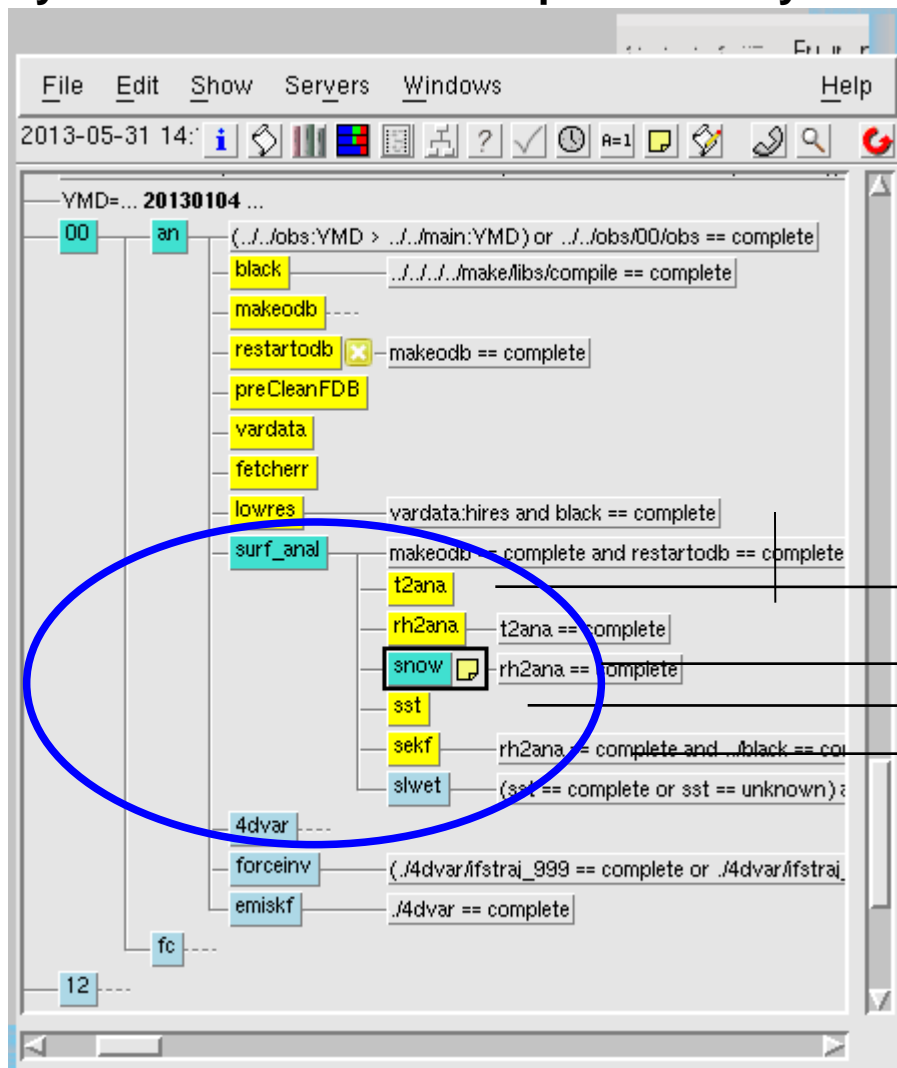
- Approaches:
 - 1D Optimal Interpolation (Météo-France, CMC, ALADIN and HIRLAM)
 - Analytical nudging approach (BoM)
 - Simplified **Extended Kalman Filter (EKF)** (DWD, ECMWF, UKMO)
- Conventional observations: SYNOP data of 2m air relative humidity and air temperature ; **Dedicated 2D OI screen level parameters analysis**
- Satellite data : ASCAT soil moisture (UKMO), SMOS (dvpt ECMWF, UKMO, Env.Canada)

Soil Temperature and Snow temperature also analysed

- 1D OI for the first layer of soil and snow temperature (ECMWF, Météo-France)

Introduction: LDAS tasks organisation

IFS cycle 40r1 is the current operational cycle



SMS: Supervisor Monitor Scheduler

Different tasks performed

Colour code:

- Yellow: task completed
- Green: running
- Blue: in queue
- Red: failed

Screen level parameters

Snow

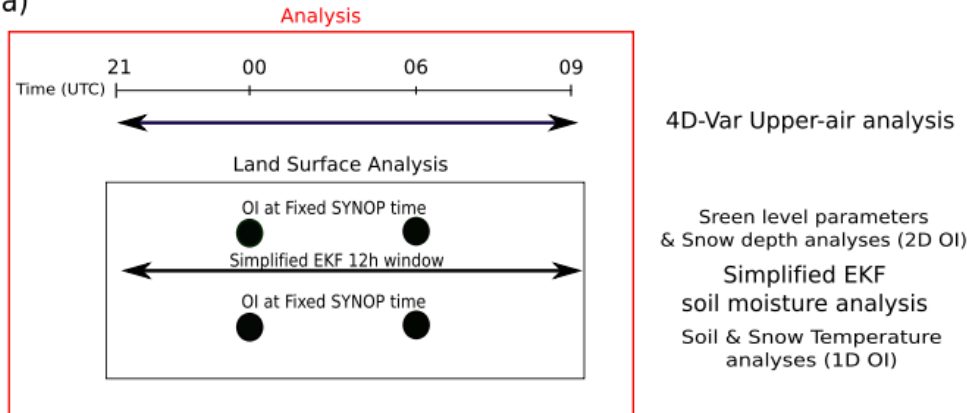
SST and Sea Ice

Soil Moisture
and Temperature

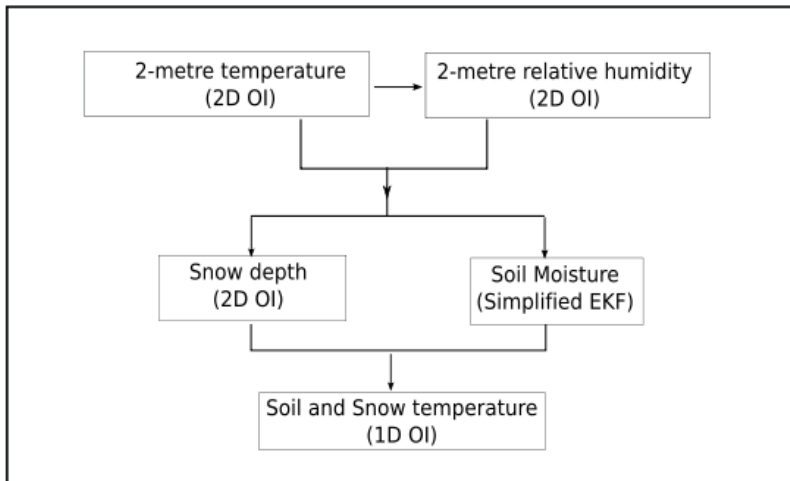
Surface and 4D-Var are running separately and in parallel.

Introduction: LDAS tasks organisation

(a)



(b)



LDAS:

- 2D OI:
Screen-level for T and humidity
Snow depth
- EKF
Soil moisture
- 1D OI:
Snow & soil temperature

Analysed surface fields: used as initial conditions for the next forecast.

- Influence the forecast which will be used as first guess for the next data assimilation window, for both 4D-Var and LDAS
- Feedback surface-atmosphere.

Outline

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- Introduction
- **Snow analysis**
- Screen level parameters analysis

Part II (Tuesday)

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- Summary and future plans



Snow data assimilation

Snow Model: Component of H-TESSSEL

(Balsamo et al., JHM 2009, Dutra et al., 2010)

- Snow depth S (m) (diagnostic)
- Snow water equivalent SWE (m), ie snow mass
- Snow Density ρ_s , between 100 and 400 kg/m³

} Prognostic variables

$$SWE = \frac{S \times \rho_s}{1000} \quad [\text{m}]$$

Observations types used:

- Conventional snow depth data: SYNOP and National networks
- Snow cover extent: NOAA NESDIS/IMS daily product (4km)

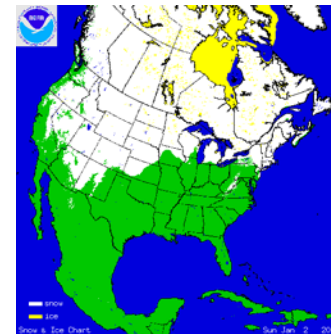
Drusch et al. JAM, 2004 ; de Rosnay et al, SG 2013

de Rosnay et al. Res. Mem. R48.3/PdR/1028 2010,
and Res. Mem. R48.3/PdR/1139 2011

Data Assimilation Approaches:

- Cressman Interpolation in ERA-Interim
- Optimal Interpolation in operations

de Rosnay et al, Survey of Geophysics 2013



NOAA/NESDIS IMS Snow extent data

Interactive Multisensor Snow and Ice Mapping System

- Time sequenced imagery from geostationary satellites
- AVHRR,
- SSM/I
- Station data

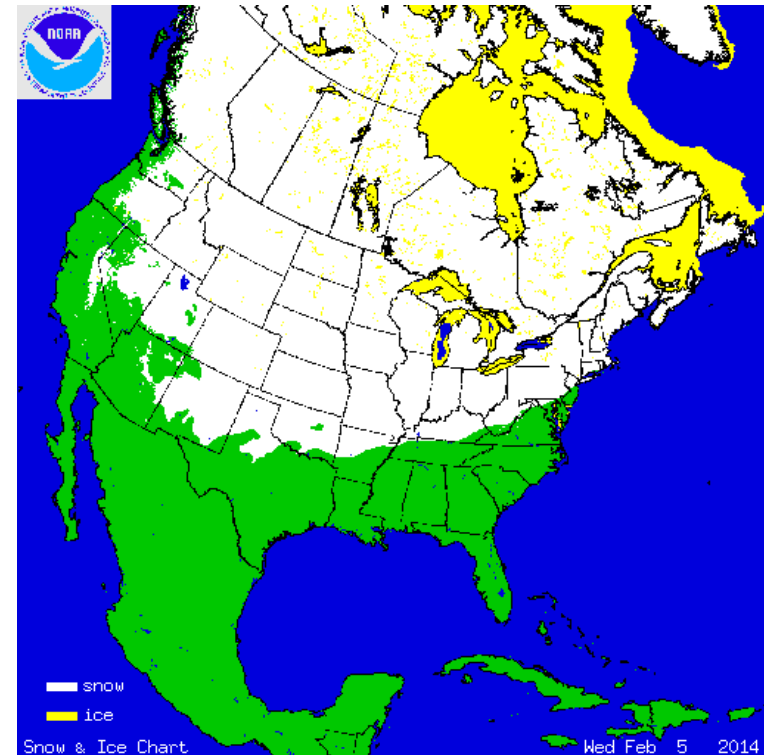
Northern Hemisphere product

- **Daily, no time stamp**
- Polar stereographic projection

Information content: Snow/Snow free

Data used at ECMWF:

- **24km product in Grib**
Used in ERA-Interim (2004-present)
and in operations (2004-2010)
- **4 km product in Ascii**
Revised pre processing
Used in operations (Nov 2010-present)

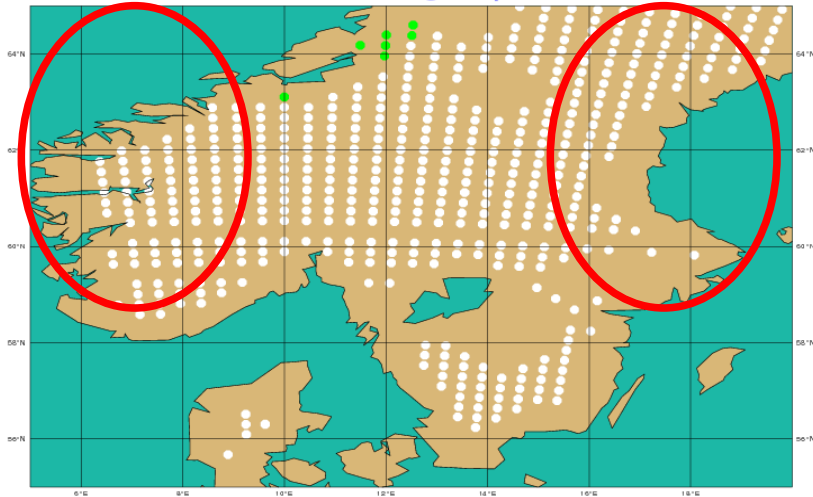


IMS Snow Cover 5 Feb. 2014

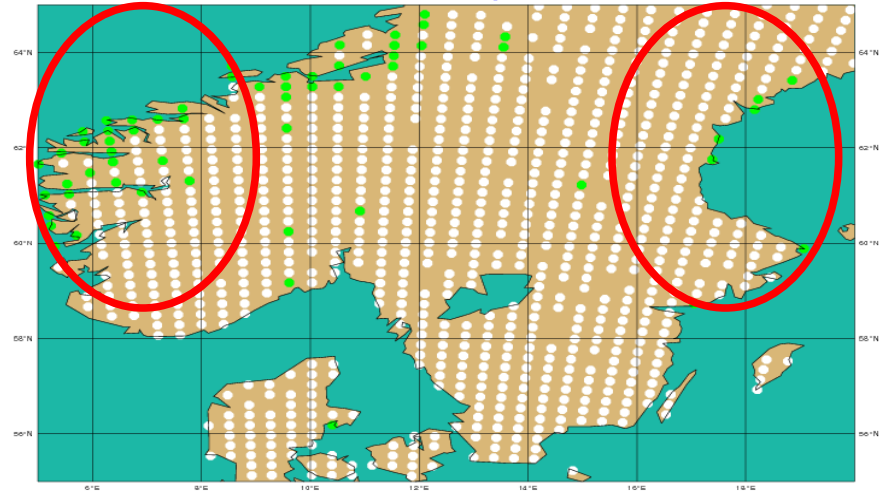
More information at: <http://nsidc.org/data/g02156.html>

Snow Cover 24km vs 4km product

NOAA/NESDIS - 24 km grib product 20091222



NOAA/NESDIS - 4 km product 20091222



IMS Products after pre-processing at ECMWF

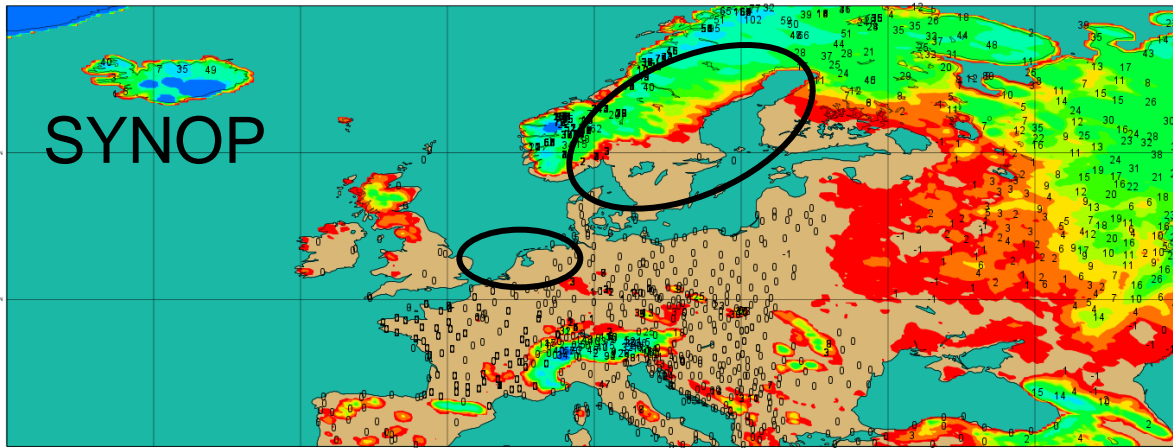
- Coast mask applied in the 24km product (lack of geolocation information in the grib product)
- Data thinning (1/36) of the 4km product -> same data quantity, improved quality

4km product provides more local information than 24km product

→ consistent with the way IMS is used in the data assimilation system

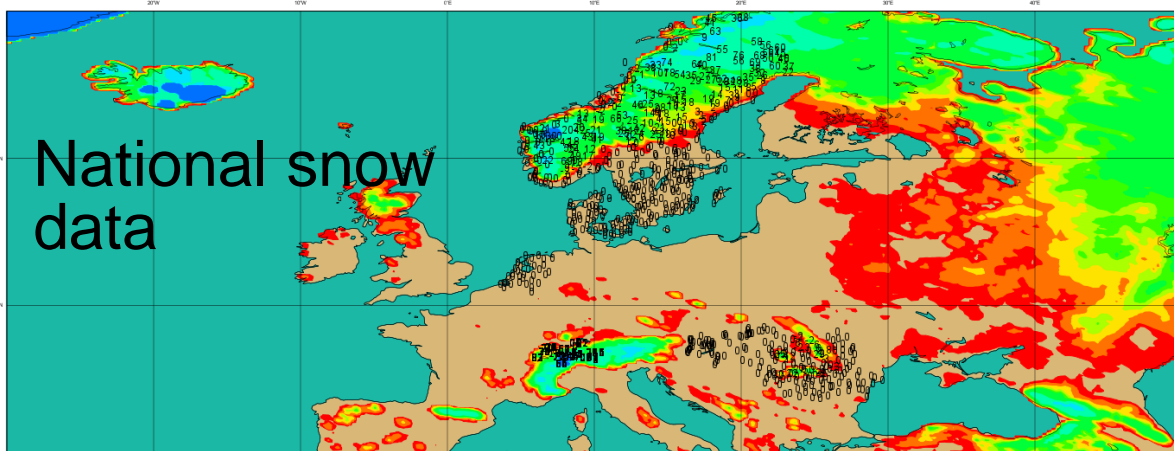
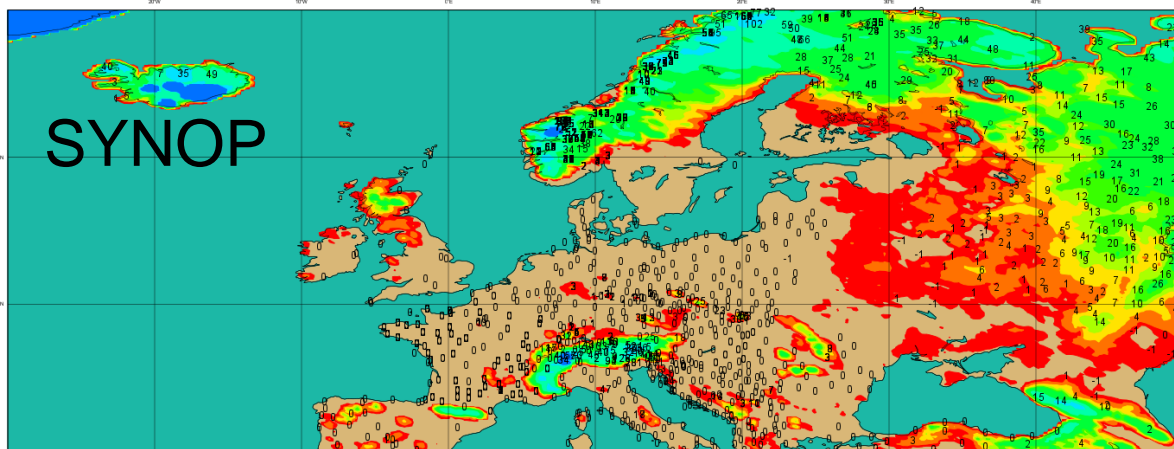
Snow SYNOP

2014 01 01 at 06UTC



Snow SYNOP and National Network data

2014 01 01 at 06UTC



Additional data from national networks from 7 countries:

Sweden (>300), Romania(78), The Netherlands (33), Denmark (43), Hungary (61), Norway (183), Switzerland (332).

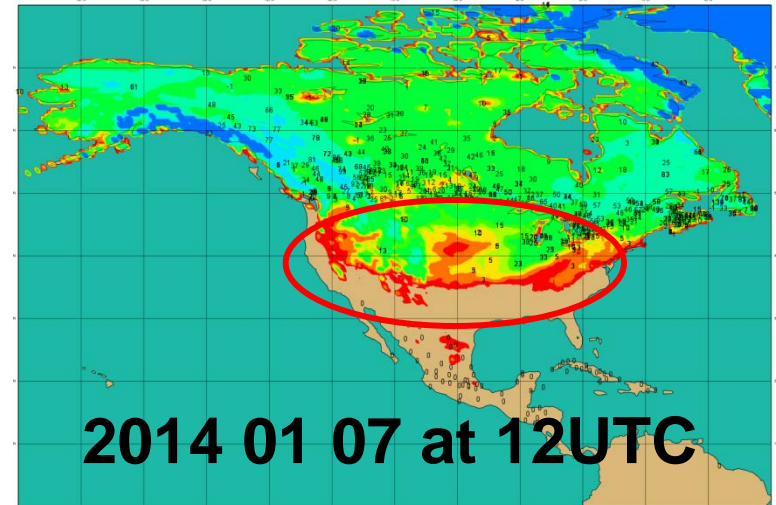
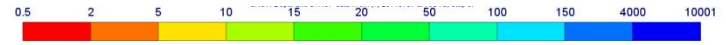
→ Dedicated BUFR

(de Rosnay et al. ECMWF Res. Memo, R48.3/PdR/1139, 2011)

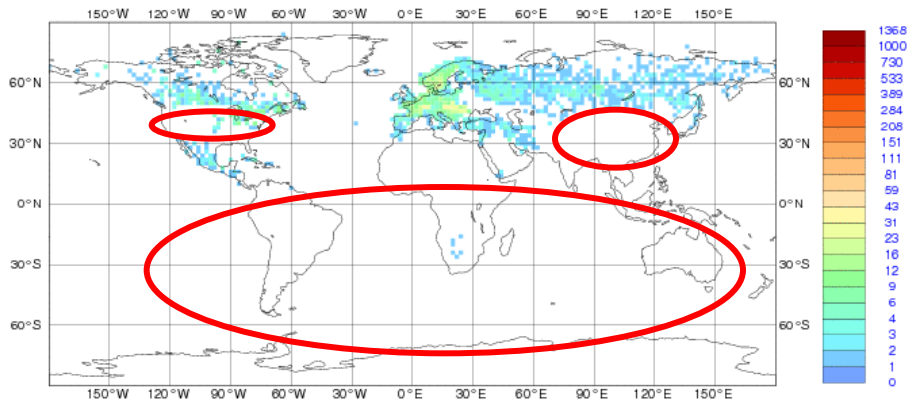
SYNOP Snow depth availability

ECMWF Operational monitoring of SYNOP snow depth: number of observations on 2014 01 04 at 00UTC (21-09 UTC):
observations gap in USA, China and southern hemisphere

<http://www.ecmwf.int/products/forecasts/d/charts/monitoring/conventional/snow/>

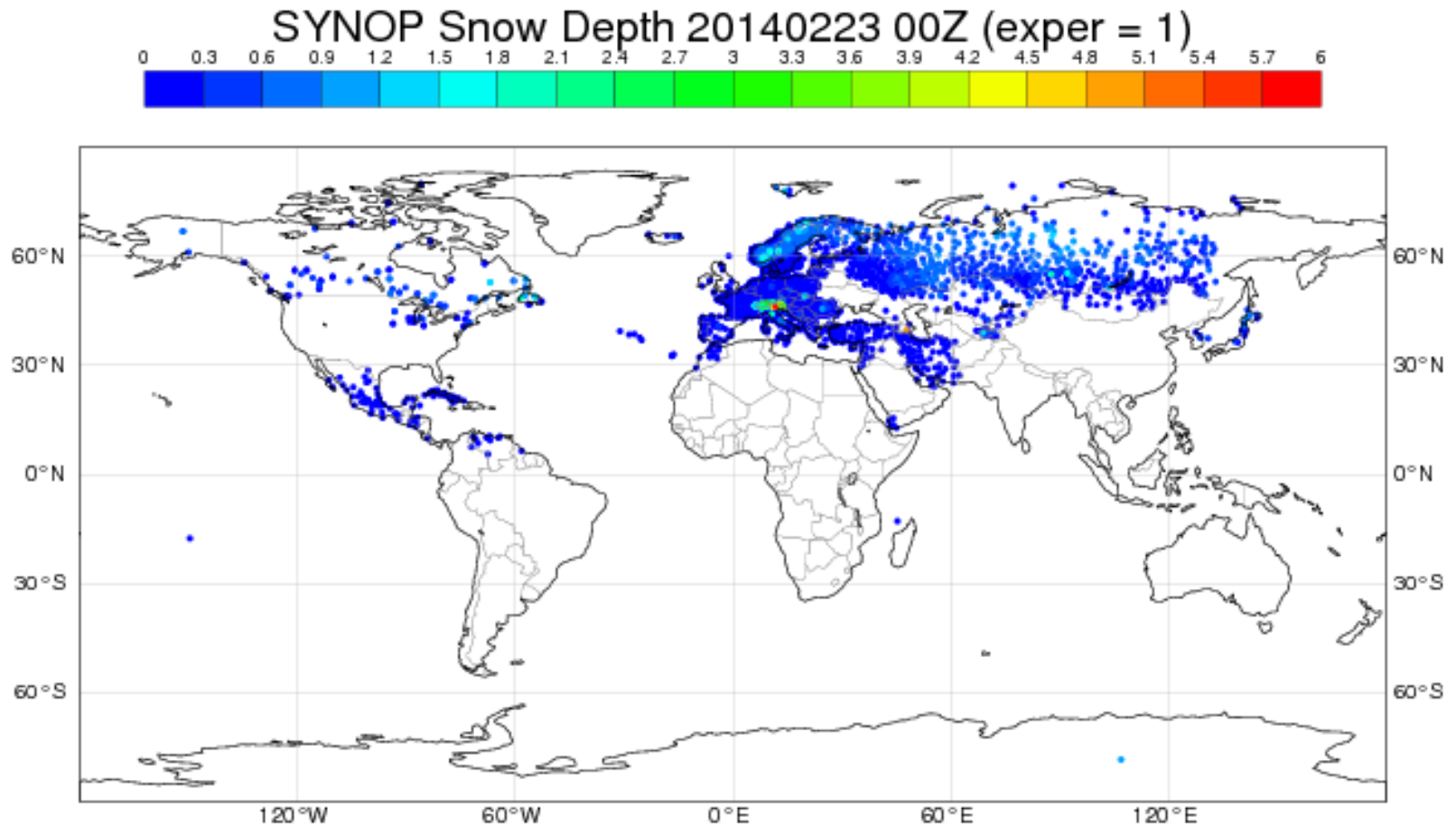


GCW Snow Watch initiative to improve in situ snow depth data access (NRT and rescue), Brun et al 2013



Snow depth observations

Snow depth observations available (>4500 per day in winter time)



Snow Analysis at ECMWF

Pre-Processing:

- SYNOP reports converted into BUFR files.
- IMS converted to BUFR (and orography added)
- SYNOP BUFR data is put into the ODB (Observation Data Base)

Snow depth analysis at 00, 06, 12, 18 UTC :

- **Cressman interpolation:** Operations: 1987-2010
Still used in ERA-Interim
- **Optimal Interpolation (OI):** Operational since November 2010

(de Rosnay et al; SG 2013)

Observation errors:	
BG:	$\sigma_b = 3\text{cm}$
SYNOP	$\sigma_{\text{SYNOP}} = 4\text{cm}$
IMS	$\sigma_{\text{ims}} = 8\text{cm}$

Use NESDIS IMS data in the OI (00 UTC):

NESDIS: \ 1 st Guess:	Snow	No Snow
Snow	x	DA 5cm
No Snow	DA	DA

Snow depth Optimal Interpolation

Used at CMC, JMA, ECMWF

Based on Brasnett, j appl. Meteo. 1999

1. Observed first guess departure ΔS_i are computed from the interpolated background at each observation location i .

2. Analysis increments ΔS_j^a at each model grid point j are calculated from:

$$\Delta S_j^a = \sum_{i=1}^N w_i \times \Delta S_i$$

3. The optimum weights w_i are given for each grid point j by: $(\mathbf{B} + \mathbf{O}) \mathbf{w} = \mathbf{b}$

b : **background error vector** between model grid point j and observation i (dimension of N observations) $b(i) = \sigma_b^2 \cdot \mu(i,j)$

B : correlation coefficient matrix of background field errors between all pairs of observations ($N \times N$ observations)

$B(i_1, i_2) = \sigma_b^2 \times \mu(i_1, i_2)$ with the horizontal correlation coefficients $\mu(i_1, i_2)$ and $\sigma_b = 3\text{cm}$ the standard deviation of background errors.

O : **covariance matrix of the observation error** ($N \times N$ observations):

$$\mathbf{O} = \sigma_o^2 \times \mathbf{I}$$

with σ_o the standard deviation of observation errors (4cm in situ, 8cm IMS)

Snow depth Optimal Interpolation

Used at CMC, JMA, ECMWF

Based on Brasnett, j appl. Meteo. 1999

Correlation coefficients $\mu(i_1, i_2)$ (structure function):

$$\mu(i_1, i_2) = \left(1 + \frac{r_{i_1 i_2}}{L_x}\right) \exp\left(-\left[\frac{r_{i_1 i_2}}{L_x}\right]\right) \cdot \exp\left(-\left[\frac{z_{i_1 i_2}}{L_z}\right]^2\right)$$

Lz; vertical length scale: 800m, **Lx**: horizontal length scale: 55km

r_{i_1, i_2} and Z_{i_1, i_2} the horizontal and vertical distances between points i_1 and i_2

Quality Control: reject observation if $\Delta S_i > \text{Tol} (\sigma_b^2 + \sigma_o^2)^{1/2}$ with $\text{Tol} = 5$

→ Observation rejected if first guess departure larger than 25 cm

Redundancy rejection: use observation reports closest to analysis time

And use a maximum of 50 observations per grid point)

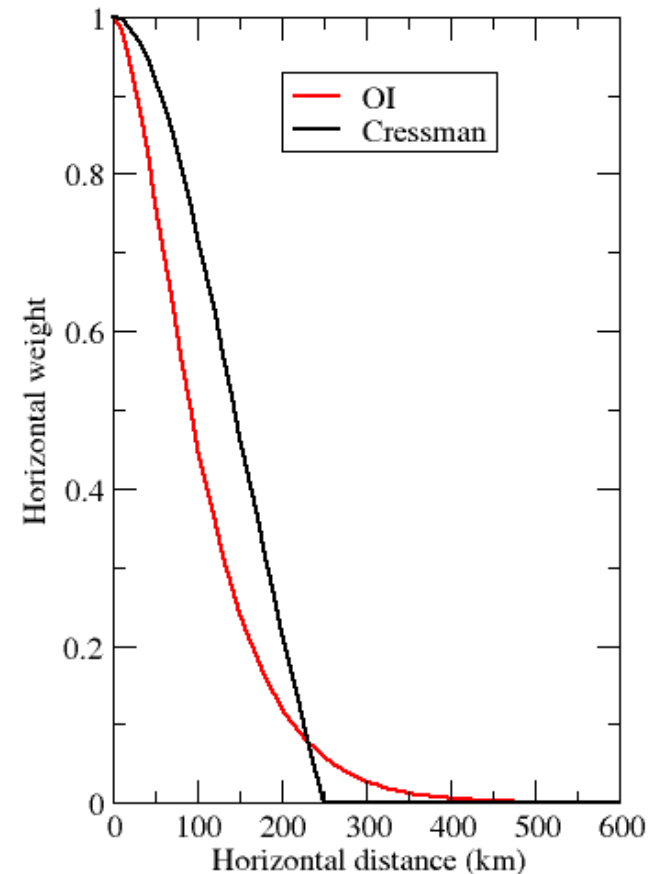
OI vs Cressman

In both cases: $\Delta S_j^a = \sum_{i=1}^N w_i \times \Delta S_i$

Cressman (1959): weights are function of horizontal and vertical distances. Do not account for observations and background errors.

OI: The correlation coefficients of B and b follow a second-order autoregressive horizontal structure and a Gaussian for the vertical elevation differences.

OI has longer tails than Cressman and considers more observations. Model/observation information optimally weighted using error statistics.



Validation data: NWS/COOP

- NWS Cooperative Observer Program
- Independent data relevant for validation
- Used to validate a set of numerical experiments considering different assimilation approaches and IMS snow cover

Numerical Experiments	Bias (cm)	R	RMSE (cm)
Cressman, IMS 24 km	1.1	0.66	18.0
OI, IMS 24 km	- 2.0	0.74	10.1
OI, IMS 4km	- 2.1	0.73	10.3
OI, IMS 4km <1500m	- 1.5	0.74	10.1

- Oper until Nov 2010
- ERA-Interim

- Oper since Nov 2010

Validation against ground data

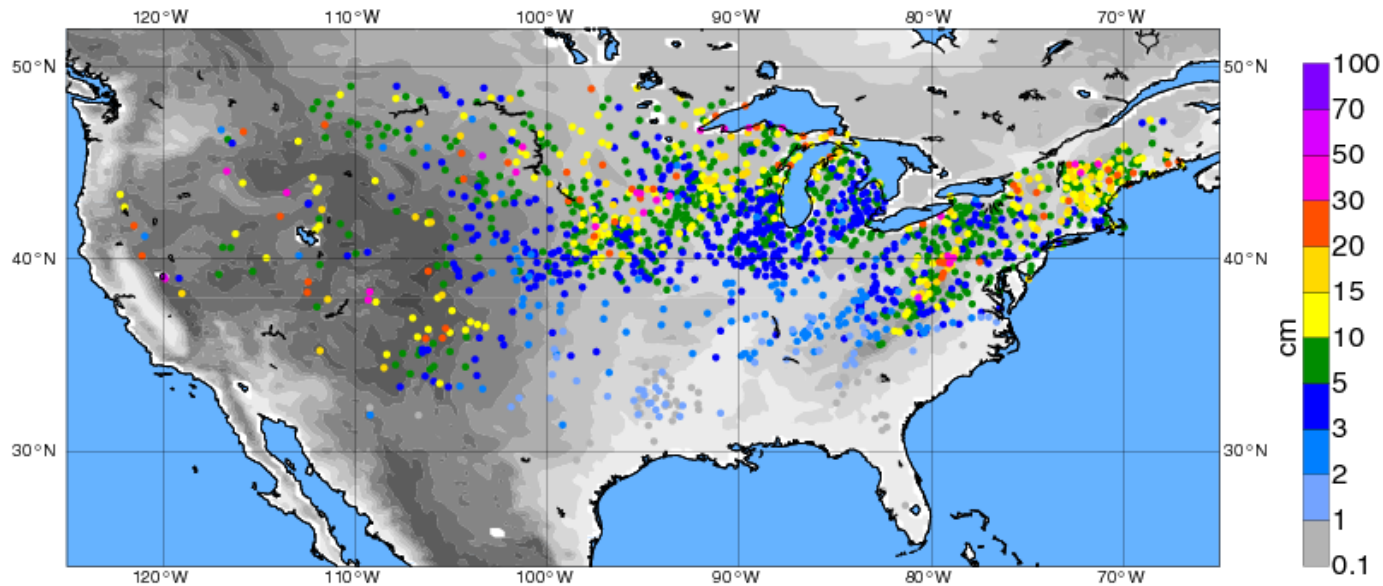
→ Main improvement due to the OI compared to Cressman

Validation data: NWS/COOP

- NWS Cooperative Observer Program
- Independent data relevant for validation
- Used to validate a set of numerical experiments considering different assimilation approaches and IMS snow cover

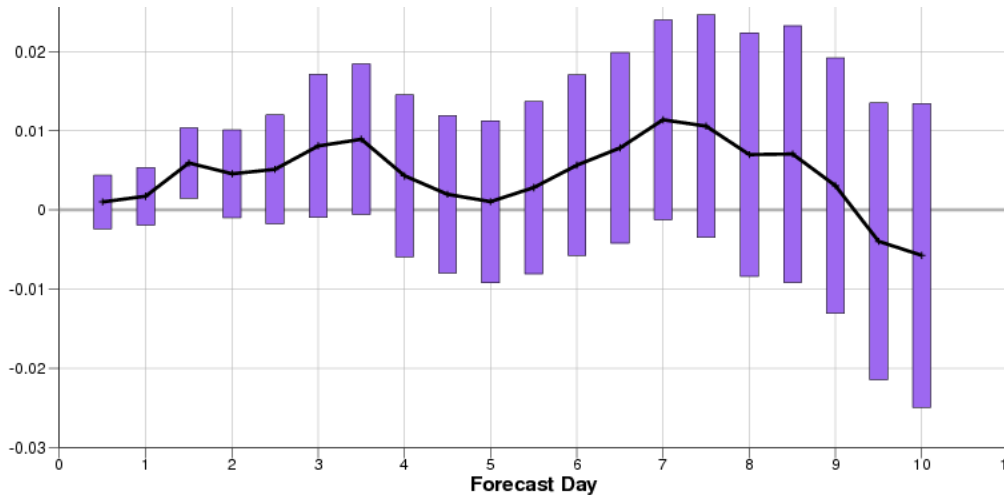
RMSE (cm) for the new snow analysis (OI, IMS 4km except in mountainous areas)

Model-COOP RMSE, Snow Depth, figg, Winter 2010, AN time: 0/6/12/18 (Z)
Mean=10.06 cm (1653pts)



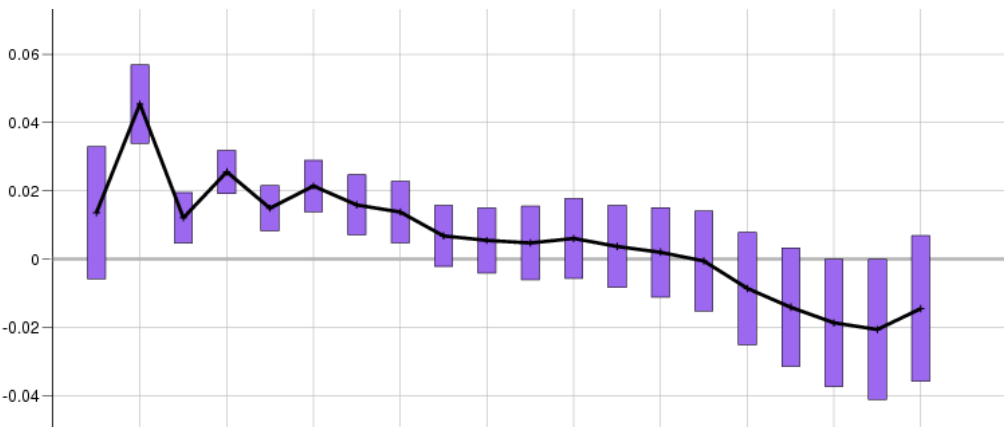
Impact on the Atmospheric Forecasts

**RMS 1000hPa Geopotential height
Northern Hemisphere
DJF 2009-2010**



Top: OI vs Cressman impact
(both use IMS 24km)

Positive means OI improves



Bottom: Overall impact

New OI, IMS 4km
vs Cressman, IMS 24km

Positive means new analysis
improves

Validation with atmospheric forecasts

→ Main improvement due to the IMS 4km and pre-processing

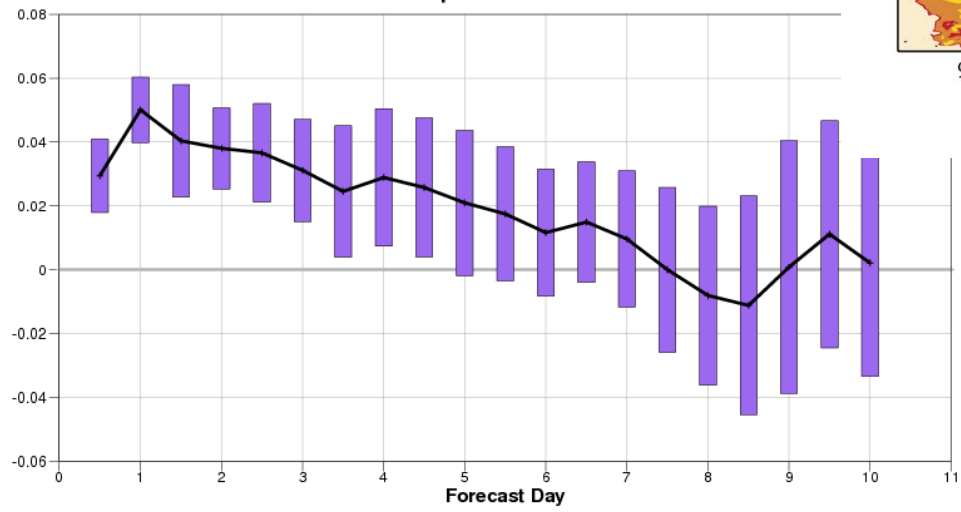
OI snow Analysis in Operations From Nov 2010

Old: Cressman
IMS 24km

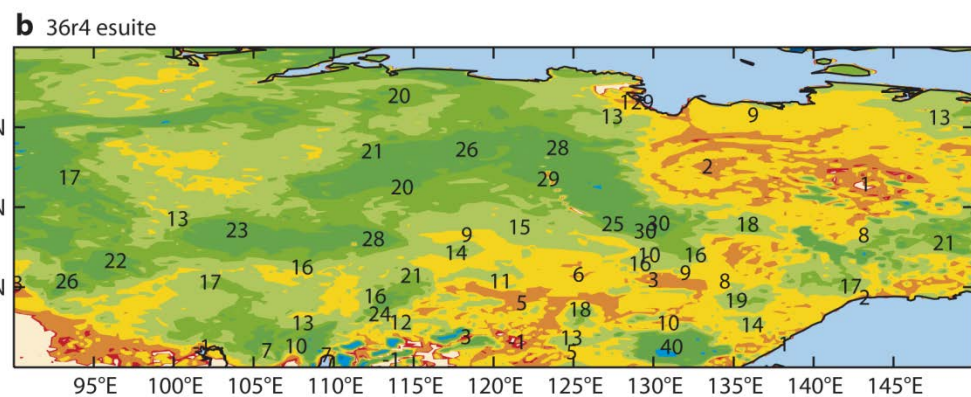
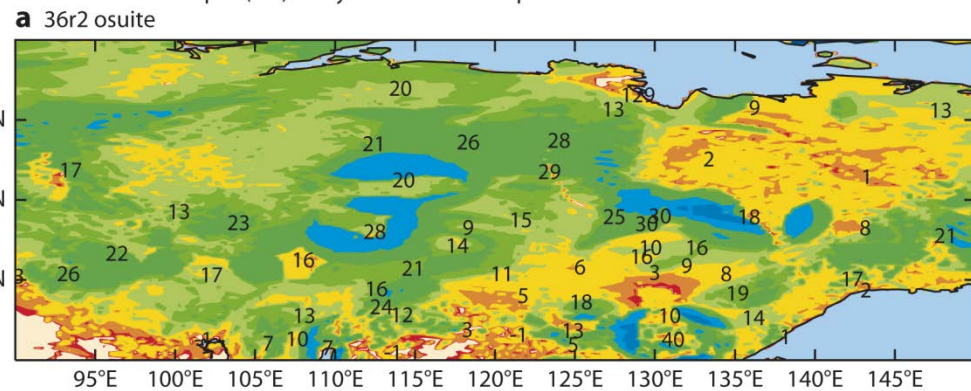
New: OI
IMS 4km & new preprocessing

FC impact (East Asia)
RMSE 500 hPa Geopot H

Confidence: 90%
Population: 90



Snow depth (cm) analysis and SYNOP reports on 30 October 2010 at 00 UTC



New snow analysis improves
both the snow depth patterns
and the atmospheric forecasts

Snow Analysis latest improvements

- 2010: replace Cressman by OI and improved IMS use (4km data and revised preprocessing)
- 2013: further improvement in the ECMWF snow analysis in IFS 40r1:
 - Revised observations error specification for IMS snow cover and assimilation of 5cm of snow instead of direct insertion,
 - Generic snow blacklist,
 - Revised surface analysis code and Observation data base (ODB) feedback
 - New Land surface observations monitoring for conventional and IMS data

<https://software.ecmwf.int/wiki/display/LDAS/Land+Surface+Observations+monitoring>

Snow Analysis latest improvements

Improved use of NESDIS/IMS snow cover data

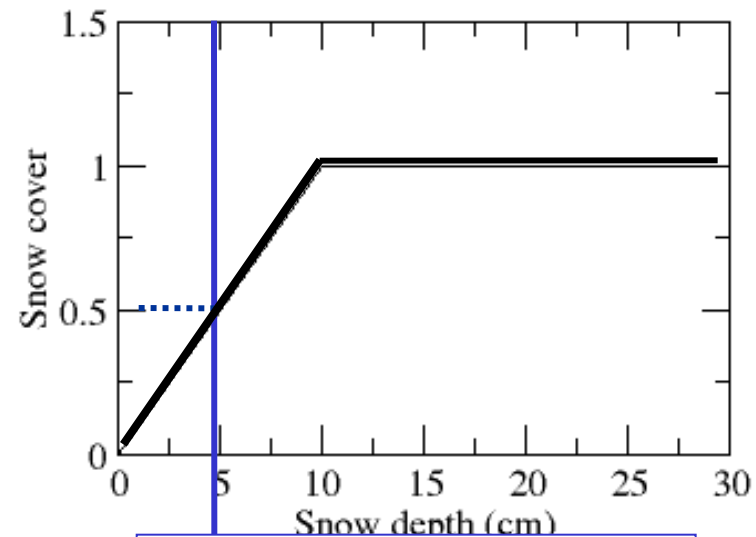
Current version:
Cycle 40r1:

NESDIS \ FG	Snow	No Snow
Snow	x	DA 5cm
No Snow	DA	DA

↓ **OI**

40r1 errors:

BG: $\sigma_b = 3\text{cm}$
 SYNOP $\sigma_{\text{SYNOP}} = 4\text{cm}$
 IMS $\sigma_{\text{ims}} = 8\text{cm}$



40r1:

- Obs assimilated
- SC=1 → SD=5cm

Snow Analysis latest improvements

Improved use of NESDIS/IMS snow cover data

Previous version: IFS Cycle 38r2

NESDIS \ FG	Snow	No Snow
Snow	x	BG: 10cm
No Snow	DA	DA

38r2:

- **BG overwritten**
- **SC=1 → SD=10cm**

OI

Previous cycles errors:
 BG: $\sigma_b = 3\text{cm}$
 OBS: $\sigma_{\text{SYNOPSIS}} = 4\text{cm}$

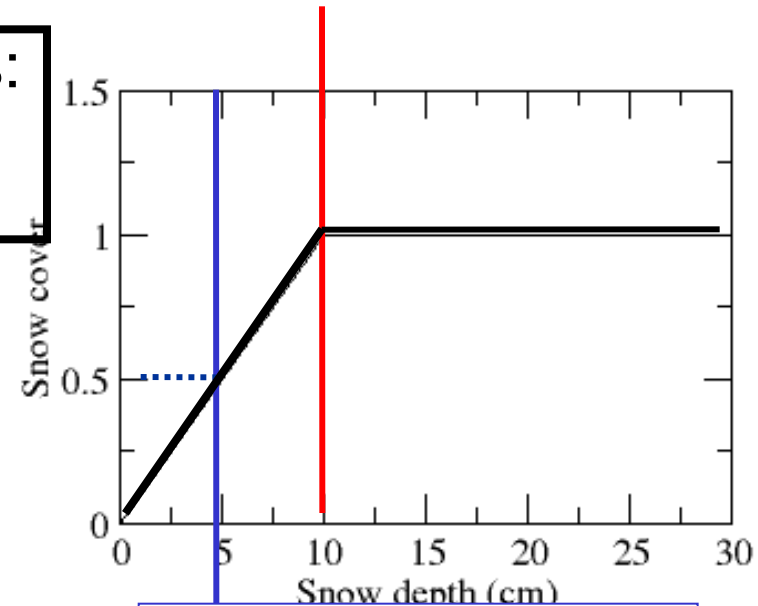
Current version: Cycle 40r1:

NESDIS \ FG	Snow	No Snow
Snow	x	DA 5cm
No Snow	DA	DA

OI

40r1 errors:

BG: $\sigma_b = 3\text{cm}$
 SYNOPSIS $\sigma_{\text{SYNOPSIS}} = 4\text{cm}$
 IMS $\sigma_{\text{ims}} = 8\text{cm}$



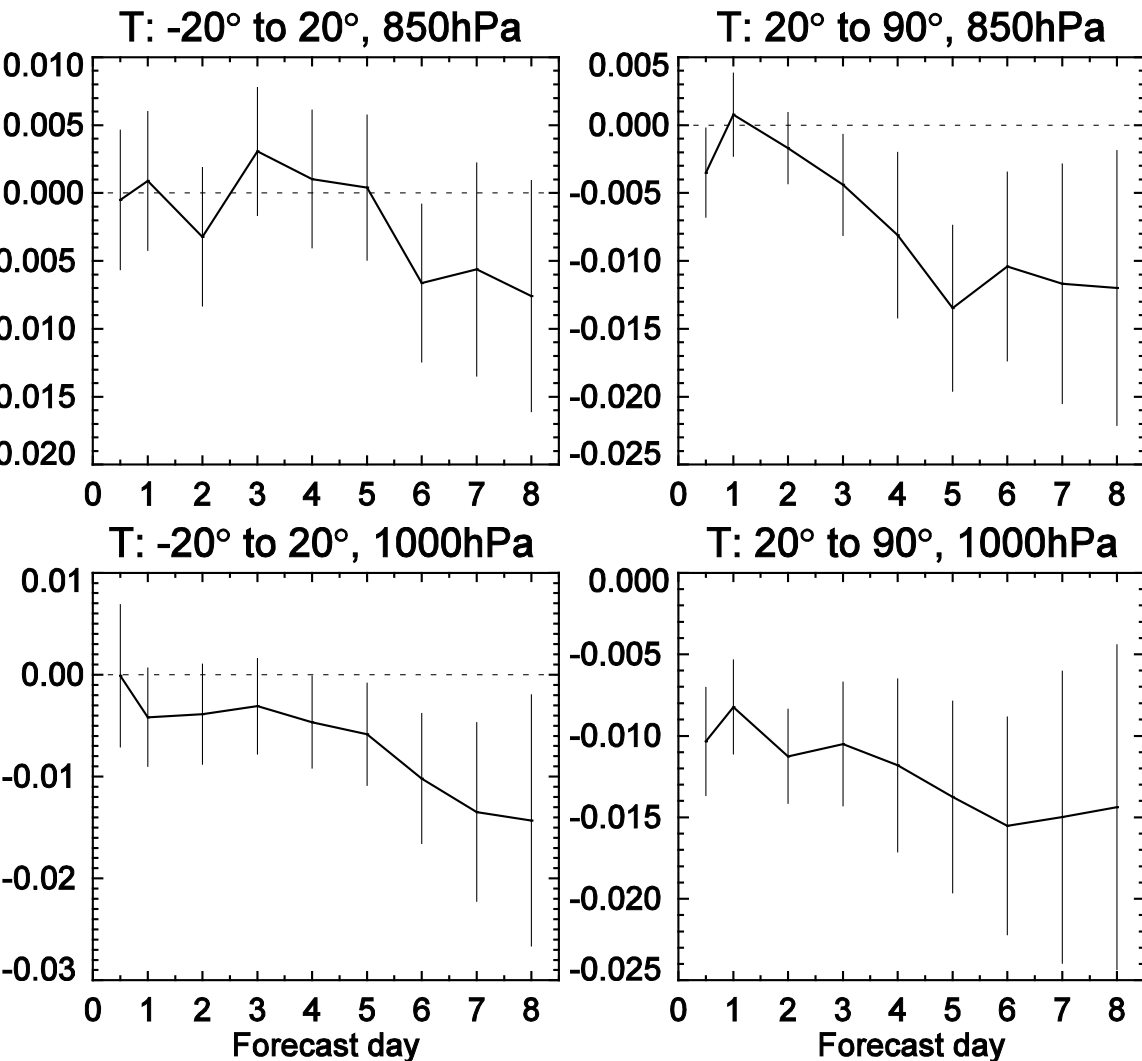
40r1:

- **Obs assimilated**
- **SC=1 → SD=5cm**

Snow analysis latest improvements: Temperature FC verification

Tropics

NH extra-tropics



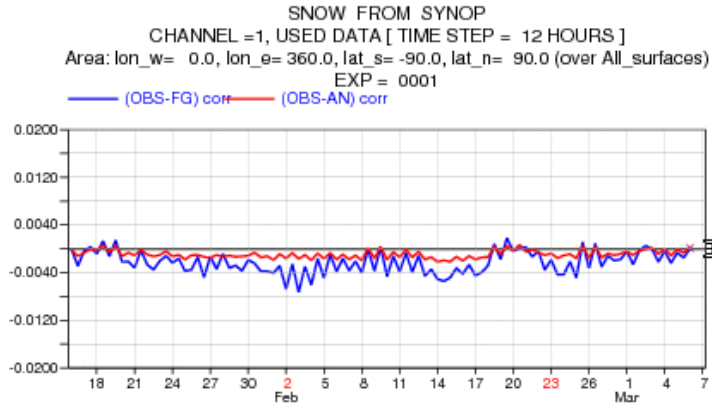
Temp FC RMSE
Verified against own
analysis
(20 Dec 12 – 08 Mar 13)

**40r1-38r2 (current-
previous cycle)**

Improved use of IMS
snow cover → Significant
impact on the
atmosphere and error
reduction in forecasts

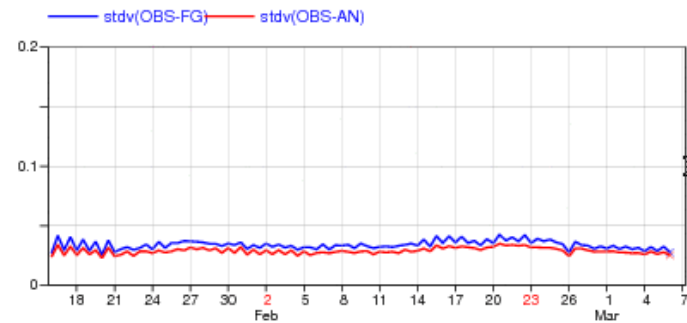
de Rosnay et al. in prep 2014

Snow observations monitoring

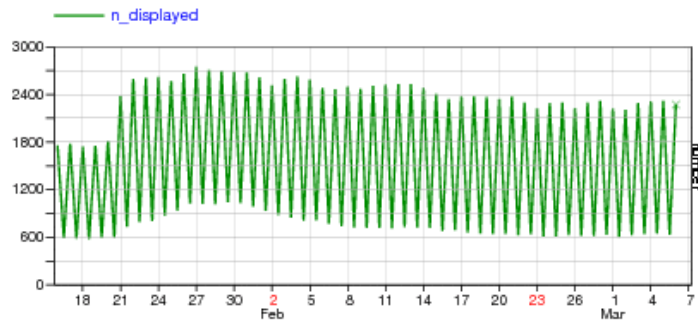


Global first guess departure

Global analysis departure



Standard deviation of departure statistics



Number of in situ observations used:
~600 to 2500 per 12 hours

2014

Summary on Snow Analysis

- Large sensitivity of atmospheric forecasts to snow data assimilation (DA method, observations pre-processing, error specification)
- Current snow analysis based on 2D-OI (CMC, JMC, ECMWF), old approach was based on Cressman (still used in ERA-Interim)
- Importance of in situ snow depth data availability
- Scarce snow depth observations in some areas → European initiative (new BUFR for additional snow data) – action to extend it to WMO Member States
- Snow cover data used (NOAA/NESDIS IMS product)
- **No** use of Snow Water Equivalent product in NWP
- Future investigations on using satellite radiances

Outline

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- Snow analysis
- **Screen level parameters analysis**

Part II (Tuesday)

- Soil moisture analysis
- Summary and future plans

Screen Level parameters analysis

- Screen level variables: 2m Air Temperature (T2m) and air Relative humidity (RH2m), both diagnostic variables.
- Analysis based on an Optimal Interpolation using SYNOP observations, every six hours: 00UTC, 06UTC, 12UTC, 18UTC.
- Screen level analysis increments are used for the soil moisture analysis (OI system, e.g. at Météo-France and ECMWF ERA-Interim),
- Screen level analysis fields are used as input of the SEKF soil moisture analysis (ECMWF)
- T2m and RH2m are diagnostic variables of the model, so their analysis only has an indirect effect on atmosphere through the soil and snow variables.
- Relevance of screen level analysis for evaluation purposes

OI Screen Level parameters analysis

Mahfouf, J. Appl. Meteo. 1991, & ECMWF News Lett. 2000

Same approach as snow analysis:

1. First guess departure ΔX_i estimated at each observation location i from the observation and the interpolated background field (6 h or 12 h forecast).
2. Analysis increments ΔX_j^a at each model grid point j are calculated from:

$$\Delta X_j^a = \sum_{i=1}^N w_i \times \Delta X_i$$

3. The optimum weights w_i are given by: $(\mathbf{B} + \mathbf{O}) \mathbf{w} = \mathbf{b}$

b : error covariance between observation i and model grid point j
(dimension of N observations)

B : error covariance matrix of the background field ($N \times N$ observations)
 $B(i_1, i_2) = \sigma_b^2 \times \mu(i_1, i_2)$ with the horizontal correlation coefficients $\mu(i_1, i_2)$
and $\sigma_b = 1.5 \text{ K} / 5 \% \text{ rH}$ the standard deviation of background errors.

$$\mu(i_1, i_2) = \exp\left(-\frac{1}{2} \left[\frac{r_{i_1 i_2}}{d} \right]^2\right)$$

O : covariance matrix of the observation error ($N \times N$ observations):

$\mathbf{O} = \sigma_o^2 \times \mathbf{I}$ with $\sigma_o = 2.0 \text{ K} / 10 \% \text{ rH}$ the standard deviation of obs. errors

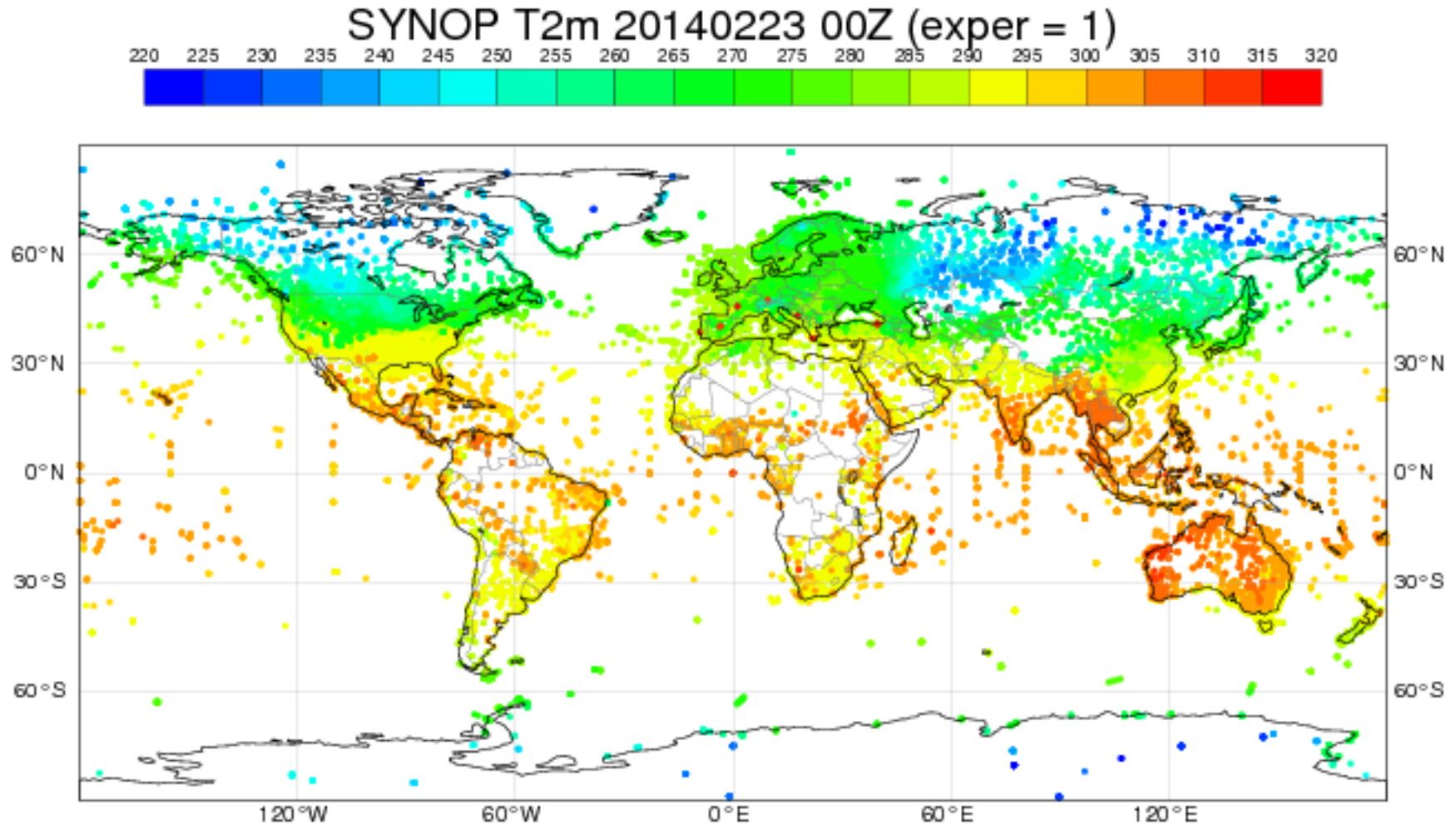
Screen Level parameters analysis

Quality control:

- Number of observations $N = 50$, $d = 300$ km, scanned radius 1000km.
- Gross quality checks as $rH \in [0,100]$ and $T > T_{\text{dewpoint}}$
- Observation points that differ more than 300 m from model orography are rejected.
- First-guess check:
Observation is rejected if : $|\Delta X_i| = \gamma \sqrt{\sigma_o^2 + \sigma_b^2}$ with $\gamma = 3$ (tolerance)
- Redundancy rejection
- Number of active observations > 16000 per 12 hour (less than 20% of the available observations).

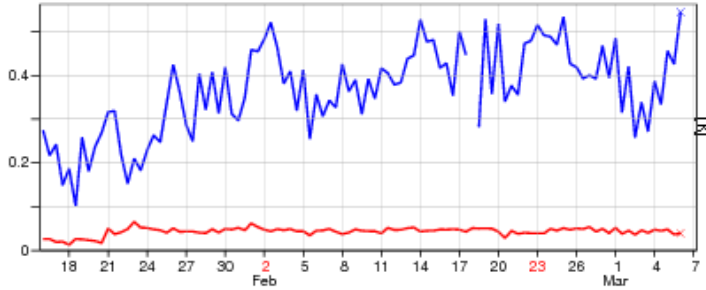
Screen level observations

All T2m observations available (>180000 per day)



Screen level observations monitoring

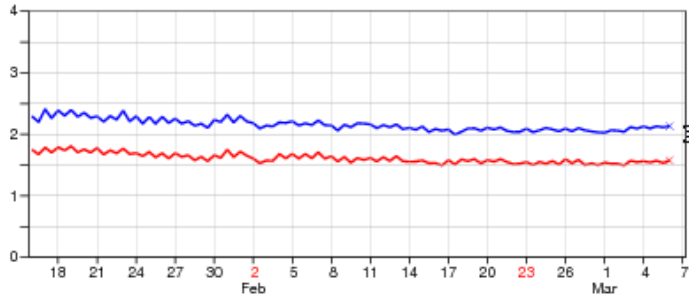
T2M FROM SYNOP
CHANNEL =1, USED DATA [TIME STEP = 12 HOURS]
Area: lon_w= 0.0, lon_e= 360.0, lat_s= -90.0, lat_n= 90.0 (over All_surfaces)
EXP = 0001
— (OBS-FG) corr — (OBS-AN) corr



Global first guess departure

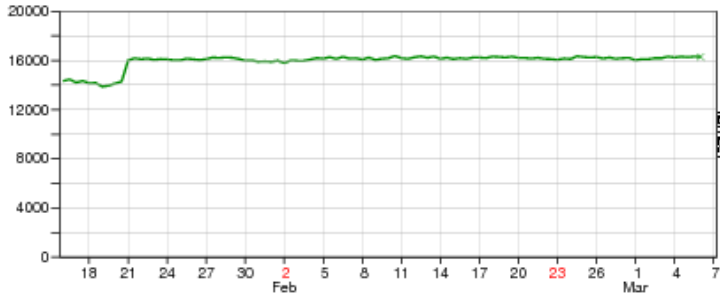
Global analysis departure

— stdv(OBS-FG) — stdv(OBS-AN)



Standard deviation of departure statistics

— n_displayed

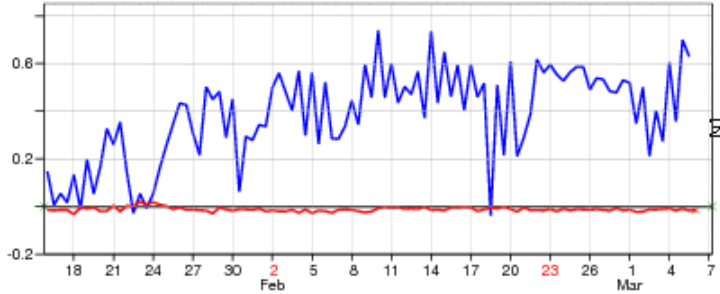


Number of observations used:
>16000 per 12 hours

2014

Screen level observations monitoring

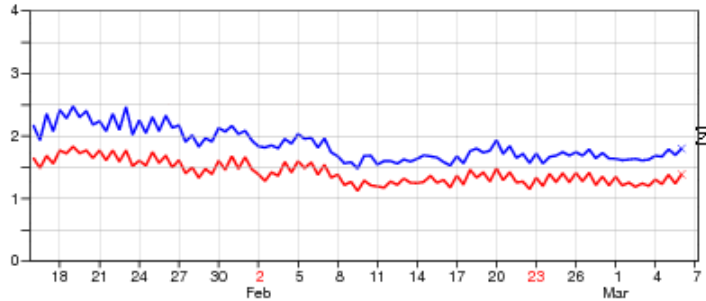
T2M FROM SYNOP
CHANNEL =1, USED DATA [TIME STEP = 12 HOURS]
Area: lon_w= 340.0, lon_e= 60.0, lat_s= 35.0, lat_n= 77.5 (over All_surfaces)
EXP = 0001
— (OBS-FG) corr — (OBS-AN) corr



Europe first guess departure

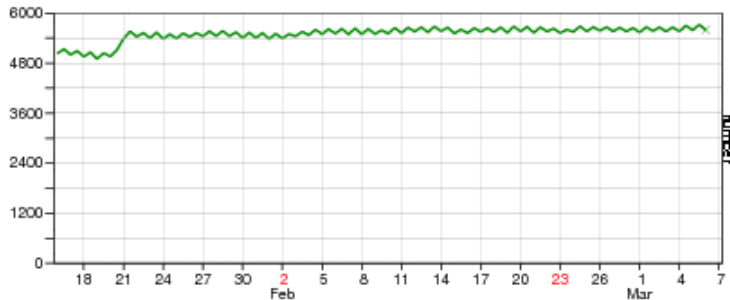
Europe analysis departure

— stdv(OBS-FG) — stdv(OBS-AN)



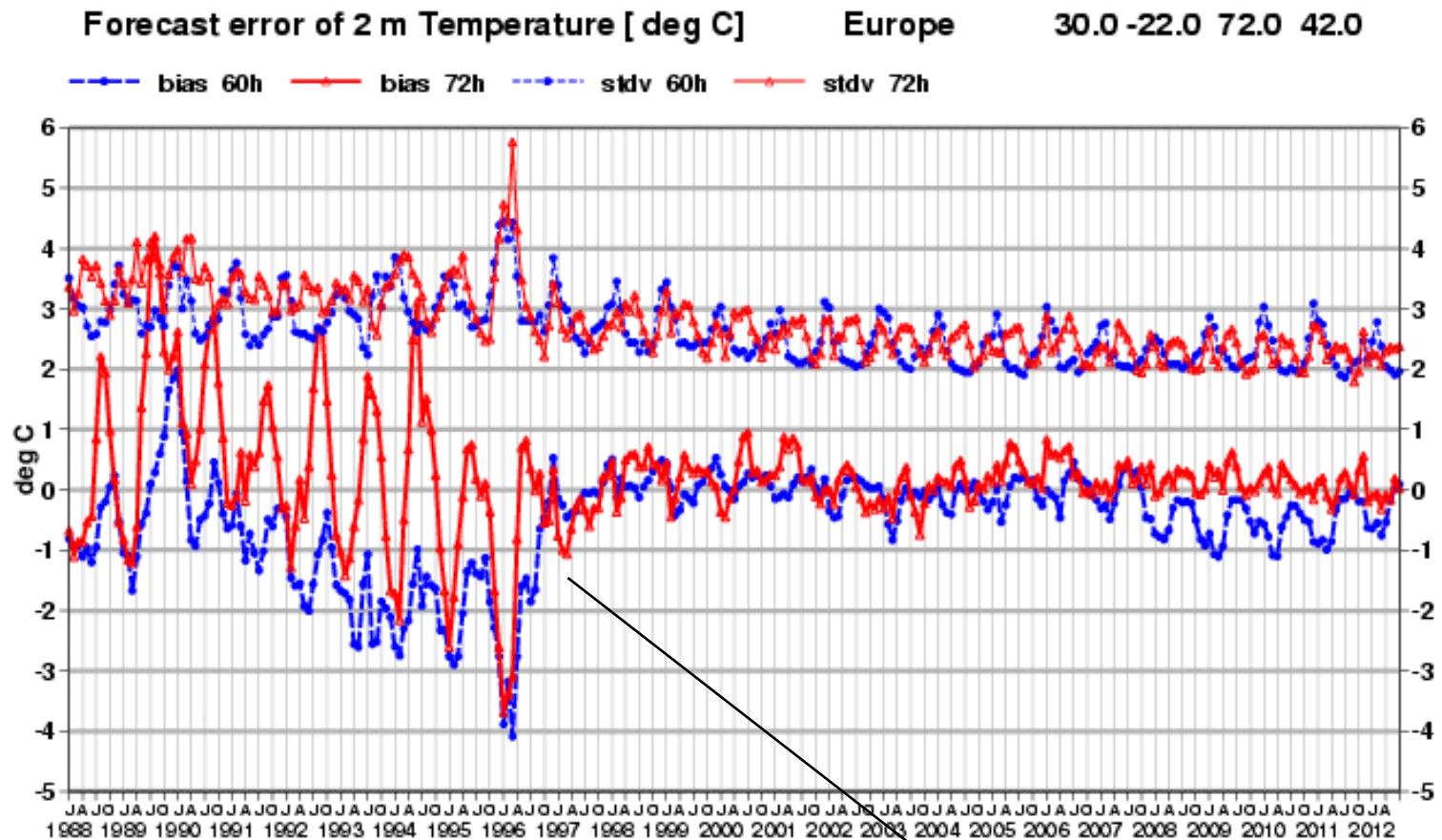
Standard deviation of departure statistics

— n_displayed



Number of observations used over Europe:
~5000 per 12 hours

Screen level analysis: 2m temperature forecast verification



Verification for 60h (night time) and 72h (day time)

Soil freezing parameterisation
Snow albedo parameterisation

From Richardson et al., 2012, ECMWF Tech. Memo 688

Outline

Part I (Monday)

- Introduction
- Snow analysis
- Screen level parameters analysis

Part II (Tuesday)

- Soil moisture analysis
- Summary and future plans