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Conclusions

he assimilation of SEVIRI radiances into the COSMO Model

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COSMO, Cracow 2008

18th January 2005 at 15:00 GMT

World Cloud Map



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Why bother with geostationaly satellite?





MSG2 CHANNEL: HRV

20080912 : 1000

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Why bother with geostationaly satellite?



Global influence (%) of satellite and in situ observations on backg analysis when assimilated by the ECMWF 4-D Var system. Synoo obs: Dritu affiting buoys: Paol: Southern Hemispheric bogus ob QuikSCAT; scatterometer sea-surface winds; Airep: com. aircraft Satob: satellite GeoMODIS winds; Temp: radiosondes, land/sea balloons, Amsua: AMSU temp:/rh. soundings; Hirs: hyper-spectr soundings; Ozone: radiative characteristics countegy: CarlaCard; com. EVENTY



MSG2 CHANNEL: HRV

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MSG2 CHANNEL: HRV

20080912:1000

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Outline

The recipe for SEVIRI assimilation

The cake ingredients: which channels ? The right dose: Bias Correction Remove lumps: Cloud Detection

1DVAR performaces

Single-Column experiments 3D model integration: two case studies

Longer time integration

Error specification

Methods Road-map for a new approach using COSMO-SREPS 8 days validation during the MAP-DPhase project Forecast scores

Conclusions



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Channel Selection

DFS: measures how much a channel in *isolation* is able to reduce the model error defined by **B** in the observational space $DFS = \frac{\mathbf{h}^T \mathbf{B} \mathbf{h}}{1 + \mathbf{h}^T \mathbf{B} \mathbf{h}}$ *DRM:* The DRM uses **A** to estimate which is the most useful channel in the analysis between *all* the ones used.



- The O_3 gas monitoring channels at $9.7\mu m$ blacklisted because large inaccuracies in the radiative transfer simulation are expected
- The 13.4 μ m blacklisted for persistent bias correction problems.



Analysis data period

1-20 September 2006

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Conclusions

Bias Correction

Air-Mass dependent bias:

Multi-linear regression coefficients based on 4 predictors:

- 900hPa-700hPa thickeness
- 200hPa-50hPa thickeness
- integrated total water mixing ratio
- T_{2m}

Coefficients dependent on weather regimes - updated Seasonally



The recipe for SEVIRI assimilation $\bigcirc \bullet \bigcirc$

1DVAR performaces

Bias Correction

Air-Mass dependent bias:

- Multi-linear regression coefficients based on 4 predictors:
 - 900hPa-700hPa thickeness
 - 200hPa-50hPa thickeness
 - integrated total water mixing ratio

T2m

Coefficients dependent on weather regimes - updated Seasonally Scan-Angle dependent bias: Negligible for geostationary satellite



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Cloud detection

Only cloud-free observations over sea points are used.



Cloud detection scheme based on a multi-spectral threshold technique SW from SAFNWC (Satellite Application Facility to support NoWCasting and very short range forecast)

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Conclusions

Cloud detection

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Cloud detection scheme based on a multi-spectral threshold technique SW from SAFNWC (Satellite Application Facility to support NoWCasting and very short range forecast)

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Extra quality checks: Pixels discarded:

Pixels whose background profiles possess saturated mixing ratio values

•
$$BT^{obs}_{10.8\mu m} - BT^{bg}_{10.8\mu m} < 3K$$



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Radiosound Comparison



 $F=(RMS_{RDS-B} - RMS_{RDS-A})/RMS_{RDS-B}$

Analysis data period

(1-20 September 2006)



1DVAR performaces

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Error specification

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Conclusions

Case Study

False allarm case: 8th of July 2004

False alarm occurred in North North-Eastern Italy, Trentino Alto Adige and Friuli-Venezia-Giulia. A risk scenario was diagnosed by LM outputs. In particular a large atmpspheric instability and convection events were forecasted. In reality the event was of minor intensity and drier winds with associated scattered thunderstorms were recorded only on the early morning of the 9th July.

Heavy precipitation case: 9th April 2005

Missed forecast of heavy precipitation in the liguria region. Typically produced by south-westerly up-stream flow due to orographic forcing



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Flag Processing

MSG1 field : flg_prc number of points : 13380 number of good points (i.e. where flg_prc has value 0): 466 date : 20040708 11:30 UTC MSG1 field : flg_prc number of points : 13380 number of good points (i.e. where flg_prc has value 0): 6 date : 20050409 20:45 UTC



ASSIMILATION CYCLE : +11 hrs and 30 mins



ASSIMILATION CYCLE : +8 hrs and 45 mins

1DVAR performaces

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Conclusions

Precipitation forecast















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1DVAR performaces

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Zonal Mean Temperature and Humidity

8 days zonal mean at +24 fc time



- Very smooth increments (mostly due to smooth error covariance matrix)
 - >Better model error specification required !!!
- effect only over sea (radious of influence set to around 50 km)
- increase in temperature and decrease in relative humidity should be only for non cloudy pixels. Dry increments can be spread over cloudy region. Nudging lacks a safety switch for this. ・ロット (中) ・ (中) ・



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Model error specificaion: some step forward

To improve the convergency rate of 1DVAR profiles and the realism of the retrieved profiles, **B** should:

- depend on weather regimes
- be scale-dependent

In regional models we could separate:

Large scale errors (at driving model resolution) due to uncertainties in initial and boundary conditions;

Small scale errors (at model resolution) due to regional model errors.



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B matrix calculations

Traditional approaches for estimating the background error covariance are built on background climatology (flow-independent)

- The Hollingsworth and Lönnberg (1986) method. Calculates the spatial covariance of differences between observations and the background.
 - (+) it is a direct diagnosis of background error covariance.
 - (-) it requires a uniform set of unbiased observations with spatially uncorrelated error.
 - (-) it only produces statistics for observable
- The NMC method (Parrish and Derber,1992). Calculates the spatial covariance of differences between 48h and 24h forecasts verifying at the same time.
 - (+) it is straightforward to calculate the required global statistics.
 - (-) it assumes that the statistical structure of 48h forecast error is similar to that of background error.
- The analysis-ensemble method.

Calculates statistics from pairs of backgrounds generated from perturbed analysis after few days from the initialisation.

(+) the effect of model error may be represented by introducing random perturbations to the physical parameterisations used in the assimilating model



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Climatological B calculation



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Conclusions

Road-map for a new approach using COSMO-SREPS

WHAT: Develop a simple technique to account both for the **time-dependent** and **space-dependent** components of the model error.



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Conclusions

Road-map for a new approach using COSMO-SREPS

WHAT: Develop a simple technique to account both for the **time-dependent** and **space-dependent** components of the model error.

- The time-dependency is accounted for by calculating **B** at the beginning of each assimilation cycle
- The **space-dependency** is accounted for by a localization procedure HOW:



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Road-map for a new approach using COSMO-SREPS

WHAT: Develop a simple technique to account both for the **time-dependent** and **space-dependent** components of the model error.

- The time-dependency is accounted for by calculating **B** at the beginning of each assimilation cycle
- The **space-dependency** is accounted for by a localization procedure

HOW:

- Use a multi-analysis model-perturbed ensemble system to identify areas of homogeneous spread ("island") where model errors are likely to be homogeneous.
- Calculate flow-dependent **B** matrices inside the selected islands.



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Conclusions

Road-map for a new approach using COSMO-SREPS

WHAT: Develop a simple technique to account both for the **time-dependent** and **space-dependent** components of the model error.

- The time-dependency is accounted for by calculating **B** at the beginning of each assimilation cycle
- The **space-dependency** is accounted for by a localization procedure HOW:
 - Use a multi-analysis model-perturbed ensemble system to identify areas of homogeneous spread ("island") where model errors are likely to be homogeneous.
 - Calculate flow-dependent B matrices inside the selected islands.

WHERE:

The new estimation of error covariance matrices are used to improve 1D variational retrieval of temperature and humidity profiles from MSG-SEVIRI radiances which are then assimilated into the nudging scheme of the regional model COSMO

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Conclusions

Sketch of the methodology



A two dimensional input spread field is computed: points with *spread* > *percmax* are identified as central points for possible islands.



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Conclusions

Sketch of the methodology



Then, the search for the island borders proceeds from this point in each of the four cardinal directions until the spread field reaches a value lower than *percmin*



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Conclusions

Sketch of the methodology



... the island is completed by filling a rectangular shape



The recipe for SEVIRI assimilation	1DVAR performaces	Longer time integration	Error specification ○○○○●○○○○	Conclusions		
COSMO-SRPES						

4 Global models drive 4 configuration of the regional model COSMO \rightarrow both errors in BC/IC and in the regional model are considered





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The ensemble spread index (ESI)

The ensemble spread index (ESI) is computed in terms of temperature and specific humidity on model levels (80 fields) by

- calculating the departure of each ensemble member from the ensemble mean
- ormalising the departures of the different variables
- convolving the departures with a vertical weighting function to obtain a unique integrated spread field.



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Validation framework

- Period: 9 days 8 16 of August 2007
- Two B calculations:
 - CFIX: NMC methods for the JJA period
 - EISL: island method using +12h FC time ensemble outputs
- only observations inside the island are retrieved to maximise the impact on the retrieved profiles



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B matrix columns



Ensemble error matrix correlation at around 850 hPa height for a sample of islands compared to the climatological calculation. Sharper features are clearly visible in the EISL experiment.

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Conclusions

Impact on forecast



Geopotential height and temperature scores at 850 hPa Forecast scores (forecast vs analysis) for the two experiments at different forecast time.

Slight improvements in the forecast scores



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Conclusion

• On the system set-up :

- Large inpact provided by the WV channels
- IR window channels can be used (all of them for robusteness of the system) if good knowledge of ground temperarure.
- Preliminary test have shown positive inpact in precipitation forecast
- Still needed some sensitivity tests to optimise the nudging coefficients
- *In general...* Expecially in regional model needs for assimilation over LAND and in CLOUDY conditions



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Forecast increments

EXP-CTRL in column integrated water vapour at +12 hr FORECAST

DIFF INT WV SIM_MSG1_20040708-SIM_NUDG_20040708 (kg/m²)



obove 5.0/10.0 4.0/5.0 3.0/4.0 2.5/3.0 2.0/2.5 1.5/2.0 1.0/1.5 0.5/1.0 0.2/0.5 0.1/0.2 -0.1/0.1 -0.2/-0.1 -0.5/-0.2 -1.0/-0.5 -1.5/-1.0 -2.0/-1.5 -2.5/-2.0 -3.0/-2.5 -4.0/-3.0 -5.0/-4.0 -10.0/-5.0







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Forecast increments

EXP-CTRL in column integrated water vapour at +36 hr FORECAST

DIFF INT WV SIM_MSG1_20040708-SIM_NUDG_20040708 (kg/m²)





DIFF INT WV SIM_MSG1_20050409-SIM_NUDG_20050409 (kg/m²)



