An Observing System Simulation Experiment to evaluate the future benefits of MTG-IRS data in a fine-scale weather forecast model

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Introduction

OSSE : Observing System Simulation Experiment

GOAL

software system to assess the value of an <u>atmospheric observing system</u> to operational mesoscale numerical forecasts

OSSE is useful for many applications, especially :

- \Rightarrow Evaluating current data assimilation and forecast systems;
- \Rightarrow Identifying observation need through sensitive experiments with synthetic observations.
- ⇒ Evaluating future observation system before deployment

Introduction



MTG-IRS : Meteosat Third Generation – IR Sounder

290 ---270 ---260 ---250 ---240 ---230 ---220 ---500

800

1000

1200

1400

1600

1800

2000

2200

Scheduled for launch in 2018/2020 ? GEO platform (36000 km) Full disk coverage / 30 min 1738 channels Spec. Res. = 0.625 cm⁻¹

Mission Band	Frequency range (cm ⁻¹)	Contribution
IRS-1	700-770	CO ₂
IRS-2	770-780	Surface, clouds
IRS-3	980-1070	O ₃
IRS-4	1070-1210	Surface, clouds
IRS-6	1600-2000	H ₂ O
IRS-7	2000-2175	СО



Introduction

All observations are simulated from free-run forecast produced by a NWP model

These forecasts represent truth and are denoted the \ll nature run (NR) \gg

Simulated obs. must mimic as close as possible, error characterics of obs. from the real system Simulated obs. are assimilated into an assimilation system that is independent of the NR model Performances in the forecast skill are evaluated



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- 1. The Nature Run
- 2. The 3D-Var Data assimilation system
- 3. OSSE calibration (tuning of obs. error)
- 4. Simulation of observations
- 5. Potential impact of IRS : Assimilation experiments

Conclusion, limitations and future work

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1. The Nature Run Characteristics

High Resolution ARPEGE/IFS Free-Run forecast

Spectral resolution : T1200 ~ 7 km over Europe Vertical levels: L105 Initial conditions: 20/06/2013 – 0h Model version : cy38op1 No data assimilation !



The ARPEGE domain

 \Rightarrow Simulated fields are available every hour for both pronostic and diagnostic model fields (3 periods of 3 months)

1. The Nature Run Comparison with ARPEGE OPER

Maps of averaged temperature fields produced by the Nature Run vs the ARPEGE OPER forecast model over 1 month (July 2013)



ARPEGE OPER forecast model :















1. The Nature Run Comparison with ARPEGE OPER

Averaged temperature and humidity profiles produced by the **Nature Run** vs the **ARPEGE OPER** forecast model over 1 month (July 2013)



1. The Nature Run Comments

 \Rightarrow This NR appears adequately realist with regard to the ARPEGE OPER model

Usage :

- Produce the initial first guess of the OSSE
- Provide coupling files for AROME OSSE boundary conditions
- Provide atmospheric states to simulate the full observing system (+ IRS)
- Verify the Forecast skills of the Limited Area model

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2. The 3D-Var data assimilation system Characteristics

3D-Var AROME/France forecast system

~ 2.5 km over France Vertical levels: L60 Initial conditions: 15/07/2013 – 0h Model version : cy38op1 3h-assimilation window Coupling (1h) : Nature Run Assimilation of the full atmospheric observing system (+ IRS) !



The AROME domain

2. The 3D-Var data assimilation system The operational observing system

Conventional:

- Radiosondes
- Aircraft .
- Ship / Buoy
- **Profilers**
- VAD winds
- Surface station .
- **Reflectivities**



-FRANCE couverture de donnees - TEMP - 2014/03/04 00H UTC Nombre total d'observations avant screening : 28



Satellite :

- ATMS
- AMSU-A ٠
- MHS / AMSU-B .
- **AMVs**
- **GPS-SOL** ٠
- IASI / CrIS / AIRS ٠
- **SEVIRI**
- HIRS

METEO-FRANCE couverture de donnees - CRIS - 2014/03/02 03H UTC Nombre total d'observations avant screening : 707

341



METEO-FRANCE converture de donnees - SATOB - 2014/03/04 00H UTC Nombre total d'observations avant screening : 1571

METEO-FRANCE couverture de donnees - SYNOP/SHIP - 2014/03/04 00H UTC Nombre total d'observations avant screening : 4823

FRANCE couverture de donnees - GPS - 2014/03/04 00H UTC lombre total d'observations avant screening : 11642



2. The 3D-Var data assimilation system Simulation of realistic observations



After the foreward model is applied to the grid point value of the NR (perfect obs.), a random contribution is added to the forward model output.

Error sources : Measurement, Forward model, Representativeness, Quality control errors ...

For instance, observation errors are assumed uncorrelated (as in OPER)

GOAL :

The response in the real and simulated system are to appear similar : CALIBRATION \Rightarrow Verifies the simulated data impact by comparing it to real data impact.

Can we use the same stdev error (sigma O) in DAS-OSSE as defined in DAS-OPER ?

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3. OSSE calibration Assignment of realistic errors for each obs. type

Configuration of assimilation experiments using simulated observations : sigma O scaling

	First Guess	Assim. OBS	Boundary condition	Stdev error
REF_OPER	AROME OPER	Real	ARPEGE OPER	AROME OPER * fact_oper
EXP_sig0.8	ARPEGE NR	Simulation	ARPEGE NR	AROME OPER * fact_0.8
EXP_sig0.5	ARPEGE NR	Simulation	ARPEGE NR	AROME OPER * fact_0.5
EXP_sig0.2	ARPEGE NR	Simulation	ARPEGE NR	AROME OPER * fact_0.2

METHOD :

- \Rightarrow Analysis increments (not shown)
- \Rightarrow comparison of obs-guess & obs-analyse statistics
- + specified stdev modifications if needed

<u>Note</u>: - conventional data fact_oper = 0.8 - satellite data fact_oper = 1.15

3. OSSE calibration Ex : Radiosonde data

13 days : July 2013 (104 assimilation) ~ 2000 obs/levels AROME domain



3. OSSE calibration Ex : Radiosonde data

13 days : July 2013 (104 assimilation) ~ 2000 obs/levels AROME domain



3. OSSE calibration Ex : Radiosonde data

13 days : July 2013 (104 assimilation) ~ 2000 obs/levels AROME domain



3. OSSE calibration Ex : Radiance data

AIRS



3. OSSE calibration Comment

No, we can NOT use the same stdev error (sigma O) in DAS-OSSE as defined in DAS-OPER since they produce too large perturbations.

Statistical properties of **EXP_sig0.5** for the OSSE vs the real world assimilation show the best match for each observing system.

 \Rightarrow All specified **sigma O are reduced by 50%.** Many of them were manually adapted to improve the fit.

This crucial evaluation was performed for the full observing system before to produce the final simulation ...

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4. Simulation of observations Full radiances observing system

Averaged Real Bt observations vs Simulated Bt observations (July 2013)



4. Simulation of observations Ex : AMSU-A

Maps of AMSU-A channel 5 Bt : observations vs simulations (20 July 2013)





4. Simulation of observations Preparations for IRS Bt assimilation

Out of the framework of the OSSE, we have conduced non-cycled assimilation experiments of real observations (IASI & SEVIRI) to model the **expected observation errors** for IRS radiances and the **optimal thinning distance**.

1) Hollingsworth-Lönnberg



2) Desroziers

$$\tilde{\mathbf{R}} = E \left[\mathbf{d}_a \ \mathbf{d}_b^T \right]$$

 $\mathbf{H}\tilde{\mathbf{B}}\mathbf{H}^T = E \left[\mathbf{d}_b \ \mathbf{d}_b^T \right] - E \left[\mathbf{d}_a \ \mathbf{d}_b^T \right]$

Assumption : Observation errors are spatially uncorrelated and background errors are spatially correlated.

- Spatially uncorrelated variance : Observation error
- Spatially correlated variance : Background error

Assumption : Because data assimilation follow linear estimation theory, the weight given to the observations in the analysis is in agreement with true observation and background errors + diagnostic of correlation length for the thinning distance

3) <u>Adapted background error method</u> Bormann and Bauer ,2010

4. Simulation of observations Preparations for IRS Bt assimilation : Stdev error estimate

Estimate of observation error amplitude using IASI real data as proxy for IRS



 \Rightarrow IRS stdev error estimate : ~0.4K for T channels and ~0.5/1K for Q channels

4. Simulation of observations Preparations for IRS Bt assimilation : Thinning distance

Error correlation length using real SEVIRI data as proxy for IRS



Result from the ensemble data assimilation : Lo = 50 km et Lb = 45 km

4. Simulation of observations Preparations for IRS Bt assimilation : Channel selection

Averaged IRS Bt simulated spectrum over the AROME domain



4. Simulation of observations Preparations for IRS Bt assimilation : Channel selection

Averaged IRS Bt simulated spectrum over the AROME domain



Normalized weighting function (a Transmittance / aln(P))

4. Simulation of observations **IRS** Bt simulations

Channel 1038





12°E

12°E

18°E

255

18°E

4. Simulation of observations IRS monitoring



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5. Potential impact of IRS : Assimilation experiments Configuration

- $\mathbf{REF} = \mathbf{Nature Run}$
- $\mathbf{CTL} = \mathbf{OSSE}$
 - ~ AROME-OPER with the full simulated observing system

+ Boundary conditions from NR

- IRS-80km = CTL + IRS (80 km, 25 Q channels)
- **IRS-15chan** = CTL + IRS (80 km, 15 Q channels)

Additional experiments :

- **IRS-40km** = CTL + IRS (**40 km**, 25 Q channels)
- **IRS-20km**= CTL + IRS (**20 km**, 25 Q channels)
- IRS-25chan= CTL + IRS (80 km, 25 Q channels)
- IRS-50chan= CTL + IRS (80 km, 50 Q channels)









IRS-80km IRS-40km IRS-20km CTL

1 day : 22 July 2013

(8 assimilation) AROME domain





1 day : 22 July 2013 (8 assimilation) AROME domain



IRS-15chan IRS-25chan IRS-50chan CTL



Summary and conclusion

- An Observing System Simulation Experiment was implemented at Météo-France to evaluate the **future benefits of MTG-IRS data in a fine-scale AROME forecast model**
- The Nature Run was shown to appear adequately realist with regard to the ARPEGE operational forecast model
- The full observing system (conv+radiances) was simulated from the NR
- Observation errors (perturbation) were tuned to mimic as close as possible, error characterics of obs. from the real system
- MTG-IRS data were simulated every 3h from the NR. The perturbation has been scaled using diagnosed obs. error from independent real obs assimilation exp. (IASI and SEVIRI)
- Several configurations (channel selection & thinning distances) were defined to evaluate the potential benefits of IRS on atmospheric analysis and forecast.
- ⇒ IRS showed strong and systematic positive impacts on the analysis of humidity on top of the whole satellite operational dataset, including IASI & SEVIRI
- \Rightarrow Negative impacts may occur if the density (channels, thinning) of IRS is inadequate ...

Limitations and future work

- The Nature Run which define the « true » atmospheric state needs further investigations about the realism of mid-latitude cyclone statistics, cloud amount ...
- \Rightarrow Comparison with the ECMWF free-run model run
- An **optimal channel selection** for MTG-IRS data is also highly recommanded to select the most informative channels and avoid redondancies and correlations (DFS ...). The potential of using PC scores instead of L1 radiance data may also be considered.
- Further development of our OSSE should focus on the **impact of clouds** on simulated Bt and assimilation. This problem was not considered here.
- The use of **different radiative transfer models** for simulation and assimilation may help to understand the error associated to radiances observations.
- These results do not guarantee positive impact MTG-IRS in forecast. Longer time experiments are required for forecast impact.
- Simulated observations require **more realistic observational error**. Further calibration is required to gain the confidence in results.

Future work : observation error correlation

- Neglecting **spatial-error correlations** in the assimilation can lead to sub-optimal analyse if the observation are used too densely (Liu and Rabier, 2003)
- In this work, the perturbation added to radiances simulations was assumed to be uncorrelated.
- Recently, the a posteriori desroziers diagnostic for **inter-channel error correlation** was run on IRS simulated WV data within the framework of this OSSE.

<u>Result</u> : Significant inter-channel error correlation were found even if the perturbation added to the observation was not correlated ...



Thank you

