Use of Satellite Observations in Numerical Weather Prediction. ECMWF Annual Seminar 2014

Convective-Scale Satellite Data Assimilation

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Convective Scale NWP: Why Bother? (© Barker)

Percentage benefit wrt UK Index

 $\frac{1}{2} - Monthly \rightarrow 12 \text{-month average}}{26 \text{-month average}} = 36 \text{-month average}}$

(forecast skill for surface weather)

Local and severe weather

- Moist processes
- Clouds, fog
- Visibility
- Precipitation
- Orographic effects

10% represents > 5-10 years lead over global model

Convective Scale DA: Spatial and Temporal Needs

Smaller spatial and temporal scales

- Rapid Update Cycling (hourly or sub-hourly)
- More timely use of satellite data (short cut-off)
- Quick turnaround (4DVar penalized)
- Uncertainties and predictability (probabilistic forecasts)

Thinning and/or super-obbing

- Ability to observe small structures
- Correlated errors (measure, processing, representativeness)
- "Big Data" paradigm

Cycling requirements

- Wait for valuable observations
- Wait for model spin-up to settle
- Hurry to get skillful forecast





Source: Dale Barker (Met Office)

Convective Scale Initialization: Two Approaches



Source: Sai Ravela (MIT)

MADCast: Multi-sensor Advection Diffusion nowCast



MADCast: Multi-sensor Advection Diffusion nowCast



MADCast: Simulated Observation Experiment





MADCast: Multi-sensor Advection Diffusion nowCast



MADCast: Real Observation Experiment

Multi-sensor synergistic analysis of cloud fraction (GOES, AIRS, IASI, CrIS, MODIS)



Vertical Cross Section





Rapid Refresh: Cloud Analysis Schematic



observation

Uses METAR, satellite, radar, lightning data
Updates RR 1h-fcst RR hydrometeor, water vapor fields
Generates latent heating from radar and lightning data

Source: Ming Hu (NOAA)

Convective Scale DA: Satellite Observations

- Real Observations
 - **IASI:** Guidard et al. (QJRMS 2011). AROME.
 - TRMM Microwave Imager (TMI): Aonashi and Eito (JMSJapan, 2011). Displaced MLEF. JMANHM.
 - MSG SEVIRI: Stengel et al. (QJRMS 2010). 4DVar HIRLAM. Schomburg and Schraff (QJRMS 2013). LETKF COSMO (retrieved cloud top). Barker (WSDA 2014). AMSU-B/MHS. 3DVar UKV

OSSEs

- GOES-R: Otkin (JGR 2010), Jones et al. (MWR, 2014). EnKF DART.
- **MTG IRS:** Guedj (EUMETSAT). Correlated obs error.
- GPM: Chambon et al. (QJRMS 2013). MLEF.
- **GLM:** Stefanescu et al. 1DVar+3DVar assimilation of total lightning. *WRFDA*.



Active obs in AROME for one rainy day

Source: T. Montmerle (Météo-France)

All-Sky Radiances: Challenges

Method

State augmentation to include model cloud microphysics variables in the analysis $(q_c, q_i, ...)$

Goals

- Fit observations at initial time
- Sustain cloud increments in forecast

All-sky radiances

- Assimilate cloud-and-precip-affected radiances
- Accuracy and efficiency of radiative transfer
- Non-linear observation operators
- Jacobian calculation: modified base state

Satellite radiances sensitive to land surface

- Forecast needed near populated areas
- Improved modeling of T_{skin} and emissivity over land, snow, sea-ice
- T_{skin} introduced as a sink variable





AIRS Channel #787



Satellite Field of View (FoV): Interpolation



Satellite Field of View (FoV): Interpolation



All-sky Radiances: Observation Error Covariances

Huber Norm: estimated via Iterated Reweighted Least Square (IRLS) = reweighting of observations according to OmF at each outer-loop



Figure from Fisher (2008)



All-sky Radiances: Observation Error Covariances

AIRS Diagnostic R Matrix



Correlated errors (esp. for moisture channels)

At least partly due to representativeness error (Waller et al. 2014)

Source: Weston (2011)



Representativeness Error

Simulated mismatch in resolution: (Daley 1993, Liu and Rabier 2002, Waller et al. 2014)

- Perfect observations (high resolution)
- Perfect Background (lower resolution)





Representativeness Error

Modified interpolation scheme:

Automatic detection of sharp gradients
New "proximity" for interpolation



Representativeness Error: Wavelet Scale Matching





(1) Linear regime: $M(\mathbf{x} + \Delta \mathbf{x}) \approx M(\mathbf{x}) + \Delta \mathbf{x} (\partial M / \partial \mathbf{x})$ $\rightarrow M(\mathbf{x} + k \Delta \mathbf{x}) - M(\mathbf{x}) \approx k [M(\mathbf{x} + \Delta \mathbf{x}) - M(\mathbf{x})]$

- ② Nonlin. regime: $M(\mathbf{x}+k\Delta\mathbf{x}) M(\mathbf{x}) \neq k [M(\mathbf{x}+\Delta\mathbf{x}) M(\mathbf{x})]$
- (3) Contradictory region: Adjusting x towards $x+k\Delta x$ at the initial time worsens the fit with the data from $M(x+\Delta x)$

Source: Fabry and Sun (2010)



Source: Wang et al. (2012)





Non-linear Perturbation: $H(x_t) - H(x_b)$

Tangent-linear Perturbation: $\hat{H}(x_t-x_b)$



AIRS Window Channel #787

Middle Loop: Fit to Observations















Update of q_{cloud} , q_{ice} in WRF





228 237 246 255 264 273 282 291 300 309 318

AIRS Window Channel #787

Cloud Analysis: Impact on Forecast



Background Error Covariances are required to update observed and unobserved model variables in a *balanced* way

BE Covariances: Raw Ensemble Auto-Correlations

Horizontal autocorrelations (mid-troposphere)



- Heterogeneous
- Anistropic
- Flow dependent
- Wide range of spatio-temporal scales

BE Covariances: Impact of Model Resolution



3 h forecasts of specific humidity $(g.kg^{-1})$ in low layer

Source: Benjamin Ménétrier

q at 945hPa

BE Covariances: Impact of Model Resolution



AEARO 90 - 🔾 : 25 km





BE Covariances: Variance Filtering



Ménétrier et al. (2014)

Background Error Covariances: Masked Statistics



Background Error Covariances: Masked Statistics



Montmerle and Berre (2010), Ménétrier and Montmerle (2011), Brousseau et al. (2012)

BE Covariances: Balance

Geostrophic balance

Hydrostatic balance

- Complex, non-linear, flow-dependent relationship b/w model variables
- Traditional balance not applicable at high-resolution



Also Betra-Carvalho et al. (QJRMS, 2012)

Mini-4DVar (10min)



Mini-4DVar (5min)



Mini-4DVar (1min)



Hybrid Ensemble/Variational Data Assimilation

$$J = \frac{1}{2} \mathbf{v}^T \mathbf{v} + \frac{1}{2} \left(\mathbf{y}^o - H \left(\mathbf{x}^b \right) + \mathbf{H} \mathbf{U} \mathbf{v} \right)^T \mathbf{R}^{-1} \left(\mathbf{y}^o - H \left(\mathbf{x}^b \right) + \mathbf{H} \mathbf{U} \mathbf{v} \right)$$
$$\mathbf{B} = \mathbf{U} \mathbf{U}^T \qquad \qquad \delta \mathbf{x} = \mathbf{U} \mathbf{v}$$

Ensemble Covariance included in 3D/ 4DVar via *state augmentation*

(Lorenc 2003, Buehner 2005, Wang et al. 2008, Fairbairn et al., 2012)



$$\delta \mathbf{x} = \beta \mathbf{U} \mathbf{v}^{c} + \beta_{e} \sum_{m=1}^{M} \delta \mathbf{x}_{m}^{f} \circ (\mathbf{U}^{e} \mathbf{v}_{m}^{e})$$

. .

Stationary multivariate covariance model including clouds

Localized [+ filtered] ensemble covariance

Background Error Covariances: Masked Statistics

Raw correlations

Localization

Localized correlations



Source: Benjamin Ménétrier

Background Errors: Non-Gaussianity

Anderson-Darling distance to a Gaussian PDF



Source: Raphael Lagrand (Météo-France)

Background Errors: Non-Gaussianity



Displacement Analysis: Grid Warping



Forecast

Observation

Displacement analysis in WRF (dWRF)



- Hurricane Katrina OSSE
- Synthetic observations (TPW)



Innovation



Displacement analysis in WRF (dWRF)



Assimilation system can operate in two modes:

- Standard (*i.e.* additive increments)
- Displacement

Nehrkorn et al. (MWR 2013, 2014)

P control



Initial time: 08-28-05 06:00:00z



Vortex displaced forward along track

P control



-60

-50

-40

-30

-20 -10

10

0

18 Hour forecast time: 08-29-05 00:00:00z



18 hours later vortex maintains forward position

dWRF DA: GOES All-Sky Radiances















RH-q_c



Model Level

Convective Scale: Model Spin-up



Initialization: Unbalanced analysis

- DFI, Incremental DFI, Diabatic DFI
- IAU, 4DIAU
- 4DVar, 4DEnVar
- Resolution gap b/w DA and model

Source: Craig Schwartz

Convective Scale: Model Error

GOES-13 10.7µm



Cintineo et al. (2014)

Convective Scale: Model Error

Joint PDFs: Observations, Parameters T=60 Minutes



Posselt and Vukicevic. (2010)

Convective Scale: Model Error

Joint PDFs: Observations, Parameters T=120 Minutes



Posselt and Vukicevic. (2010)

Advances

- Include cloud parameters in analysis state
- Multiple re-linearizations of observation operator
- Improved flow-dependent multivariate BE covariances
- Increasingly relying on information from (filtered) ensembles
- Displacement analysis for coherent features

Question

 At convective scale, will current DA methods be defeated by non-linearities (before we reach retirement age)???

Perspectives

Coupled Assimilation

- Land surface (temperature and soil moisture)
- Ocean (SST, mixing)
- Hydrology (run-off)
- Aerosols (visibility)
- Composition (air quality, photolysis)

Model error

• Accurately represent model error (weak constraint, stochastic model processes, ...)

Interaction with larger scales

 Multi-scale covariances → DA across scales (Jk Constraint, Lateral Boundary Conditions)



- http://mpas-dev.github.io

- New Global model
- Nonhydrostatic
- Voronoi meshes
- Variable resolution
- WRF, CAM, GFS physics
- Scalable code
- MPAS-A (NCAR)
- MPAS-O and MPAS-LI (LANL)

Source: Bill Skamarock



Source: Bill Skamarock

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