GEO-CAPE

Geostationary satellite mission for air quality and coastal ecosystems

Air quality

Ocean color from space

One of 15 missions recommended to NASA for the next decade by the U.S. National Research Council

Atmospheric GEO-CAPE Workshop at Columbia, MD 22 September 2009



Atmospheric Planning Team: J. Al-Saadi, K. Chance, R. Chatfield, S. Christopher, J. Crawford, B. Duncan, D. Edwards, A. Eldering, J. Fishman, S. Kondragunta, D, Jacob, L. Iraci, R. Kawa, X. Liu, D. Neil, M. Newchurch, K. Pickering, J. Rodriguez, R. Scheffe, J. Szykman, O. Torres, J. Wang

Atmospheric Variability sub-team: Fishman & Newchurch Co-chairs, Al-Saadi, Chatfield, Crawford, Christopher, Duncan, Kawa, Pickering, Scheffe.

PURPOSE: Defining the spatial and temporal scale requirements for GEO-CAPE to

1) characterize emission patterns and

2) observe the spatio-temporal evolution of pollution processes.

The critical precursors and pollutants are: O3, CO, NO2, SO2, HCHO, H2O2, PAN, HNO3, HNO4, acetylene, HCN, glyoxol, and formic acid.

The critical processes are biomass burning, lightning, biogenic VOC emission, dust events, and surface carbon flux.

The critical time frames range from hourly to multi-day episodic.

Data Sources and analyses methods for Atmospheric Variability:

- Space-borne observations of columns and profiles (O3-, NO2, CO, HCHO, CHOCHO, aerosols)
- EPA and USDA nets (CO, O3, NO2, SO2, aerosols)
- Lidar observations (O3 and aerosols)
- McDermid/LeBlanc: mid- & upper-trop DIAL O3 at TMF
- Ozonesonde observations
- Nested resolution WRF-chem (define regions and dates)
- Global GEOS-chem
- Univariate statistical climatologies
- Calculated variability: data and models (variograms)
- Pattern recognition
- Production/Destruction rate calculations
- Horizontal Variability of Trace Gases over the Eastern United States
- Spatial autocorrelation (vertical, horizontal, cross species)
- Vertical Autocorrelations of O3 from the Ozonesonde Record
- EOF and SVD analyses

Aerosol backscatter (right) and ozone (left) profiles from a/c DIAL



30, 35, 40, 45, 50, 55, 60, 65, 70, 15, 65, 85, 90, 95, 100, 105, 110, 115,







13 September 2008 Huntsville, AL



Time



4 October 2008 Huntsville, AL



Time [GMT]



10 August 2008 Huntsville Ozone DIAL







WRF calculation, Pickering et al.

Diurnal processes

Atlanta



WRF calculation, Pickering et al.



Variability Statistics Based on CMAQ 1.5-km Horizontal Resolution Simulation

Ken Pickering, NASA/GSFC Melanie Follette-Cook, UMBC/GEST Yasuko Yoshida, UMBC/GEST Chris Loughner, Univ. of MD

CMAQ 1.5 km run results

- 1.5 km domain covering the Baltimore/Washington area and upwind regions, nested within a 4.5 km domain, nested within a 13.5 km domain
- 30 levels in the vertical, up to 90 hPa
- Spatial analysis at 18 UTC to correspond with Aura overpass time. Temporal analysis from 13 -23 UTC.
- Data analyzed for a low pollution day July 7th, 2007, and a high pollution day July 9th, 2007
- All plots have two lines on them
 - Dashed line is r = 0.7 contour, indicating ~50% of variance explained
 - by neighboring grid cell
 - Solid line is r = 0.87 contour, indicating ~75% explained variance
 - Also some results for r = 0.95, indicating ~90% explained variance

CMAQ 1.5 km domain – surface ozone



CMAQ 1.5 km domain – surface NO₂



CMAQ 1.5 km domain – surface CO

Surface CO 20070707_018



Surface CO 20070709_018



CMAQ 1.5 km domain – surface SO₂



0.0 2.5 5.0 7.5 10.0 12.5 15.0 17.5 20.0 22.5 25.0

[ppbv]

Decay of horizontal autocorrelation of the tropospheric column





-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0



-1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0

Decay of temporal correlation as a function of pressure O_3 and CO



0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1.0

Decay of temporal correlation as a function of pressure NO_2 and SO_2



0.0 0.1 0.2 0.3 0.4 0.5 0.6 0.7 0.8 0.9 1.0

Decay of vertical correlation as a function of pressure – NO₂



Boundary layer NO2 not very well mixed. Correlations decay rapidly.

Decay of vertical correlation as a function of pressure – CO



CO much better mixed -1.0 -0.8 -0.6 -0.4 -0.2 0.0 0.2 0.4 0.6 0.8 1.0 in BL than NO2.

Correlation of surface ozone and tropospheric column ozone





OHIO POWER PLANTS EMISSIONS

OHIO POWER PLANTS EMISSIONS



Courtesy of Kostya Vinnikov



GOME-2 HCHO and CHOCHO



- Glyoxal and formaldehyde fields are very similar
- Good agreement with SCIAMACHY data
- Ratio CHOCHO / HCHO depends
 on sources



Vrekoussis et al., GOME-2 observations of oVOCs: What can we learn from the ratio CHOCHO to HCHO on a global scale, *paper in preparation*, 2009

TES Observations of Tropospheric Ammonia

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AER, Inc.

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AURA, Sept 14-18, 2009, Leiden, The Netherlands



Why Measure Ammonia from Space?

Lack of direct NH₃ obs. to help with large uncertainties in modeled emissions

- In situ (mostly surface) measurements are sparse
- Uncertainty in the seasonal and spatial variability
 - CMAQ (regional) : peak emissions during fertilization application in <u>spring</u> (April)
 - GEOS-Chem (global) : peak emissions with high temperatures in summer (July)

Satellite measurements have potential to constrain the NH₃ emissions

US EPA Monitoring Network

(Gary Lear)





Sensitivity Tests

How sensitive are TES measurements to changes in NH₃ emissions?

- Map shows the sensitivity of TES measurements (marked by X) to NH₃ emissions from any model grid box (from up to a week prior)
 - relative to the influence of the NH₃ directly underneath the TES track



Sensitivity of TES obs in the track to NH₃ emissions from the week prior

TES Obs are most sensitive to NH₃ emissions directly underneath the track (X)

TES is sensitive to NH₃ emissions away from obs.

 e.g. TES is ~40% as sensitive to emissions here compared with directly beneath (X)

NH₃ lifetime increased :

- NH₃ (gas) -> NH₄ (aerosol-phase)
- Bi-directional flux (biosphere and atmosphere)

NH₃ Spatial Gradients and Seasonal Variability: San Joaquin Valley -2008

July-August Transects

Monthly Mean values



TES captures spatial gradient

• High values (40 ppbv): much greater than GEOS-Chem

Seasonal variability

Peaks in April and September

- Typical of farming with fertilizer application?
- Will compare with in situ datasets where available

TES NH3 Validation Example : Transects over North Carolina USA

- Started in early February 2009
- Will run at least through Dec. 2009
- CAMNet NH₃ monitoring sites match-up with TES overpass
- Will allow detection of spatial variability and seasonal trends

05





Observing the troposphere with IASI: Emission, chemistry and transport



Products / Applications

Tropospheric sources

LATA

→ Ammonia Clarisse et al., Nature Geo, 2009 The link to agriculture





Pierre Coheur, Aura Science Team Meeting, September 15 2009



Global Modelling of NH₃ and the first comparison with satellite observations

Frank Dentener Lieven Clarisse



Aura, Leiden, September 2009,

MA. Sutton et al. / Environmental Pollution 156 (2008) 583-604





Clarisse et al, [2009]

Some new uses of OMI NO₂ observations

- Spatial Resolution
- Fires
- Farms

R.C. Cohen UC Berkeley

\$\$ NASA

11/21/2004: Unbinned 3km wide pixels



The Sensitivity of U.S. Surface Ozone Formation to NO_x and VOCs as Viewed from Space

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September 16, 2009 Leiden, Netherlands

OMI HCHO/NO₂ : August 2006



OMI captures gradient from downtown to suburbs to rural areas!

WRF-Chem: July 9th : 3 pm : Surface





Testing and improving OMI NO₂ using DANDELIONS and INTEX-B data



Evaluating TM4 a priori profile shapes





The Cabauw Intercomparison campaign of Nitrogen Dioxide measuring Instruments

















Ankie Piters, KNMI

and CINDI Organisation Team









Antehe



to wellowing

NO2 gradients <200m measured by lidar

• c) Measuring from a distance towards the tower to validate with the in-situ sensors at different latitude levels



NO2 surface gradients exceed satellite resolution



courtesy: T. wagner, MPI Mainz





Use of OMI Data in Monitoring Air Quality Changes Resulting from NO_x Emission Regulations over the United States

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Lightning NO_x Source Being Added to CMAQ

10 8 kilometers 6 4 CMAO 2 CMAQ + LNOx INTEX DC-8 0 10 20 50 100 200 500 NO_x concentration (ppt)

NO_x (ppt)

Lightning flash rates predicted for times and locations of convective precipitation in meteorological model.

Flash rates scaled on a monthly basis to the NLDN + IC estimate from Boccippio IC/CG climatology

Vertical distribution of LNOx production based on observed climatology and direct function of pressure. Production/flash = 500 moles NO

Comparison of CMAQ with INTEX-A aircraft data is good up to ~7 km. Aircraft emissions still needed in CMAQ.

Method

Airborne field data are statistically evaluated using a modified variogram technique to examine their spatial variability.

Classical Variogram Definition (Matheron, 1962)

$$2\gamma(h) = \frac{1}{N(h)} \sum_{N(h)} (Z(s_i) - Z(s_j))^2$$

Where N is the number of data pairs separated by distance h; Z(s) is the variable of interest at a given location s; and locations s_i and s_j denote location pairs separated by distance h

Variogram Definition used for this analysis (also called a semimadogram)

$$\gamma(h) \equiv \frac{1}{N(h)} \sum_{N(h)} |Z(s_i) - Z(s_j)|$$

Simply stated, it is the average difference for the variable of interest over a given distance. Future plans may include calculating other statistics (e.g., median and percentiles).

Distribution of Flight Data Collected in the Boundary Layer (below 2 km)



Longitude

Longitude

-60

-114

-112

-55

NO2 Variograms: Basic behavior is similar for all four campaigns, although magnitudes differ. Interpreting magnitude is difficult since it is influenced by both the magnitude of pollution encountered and the fraction of flight time in urban/polluted versus remote areas.



Normalized NO2 Variograms: Here, variograms have been calculated for the fractional difference in NO2 for values in excess of 1 ppbv. The similarity in these curves suggests that despite the differences in magnitude for the campaign-specific variograms, the variability in proximity to pollution plumes is consistent across campaigns.



The 2008-2009 cluster of North Pacific volcanic eruptions: A-Train observations and OMI validation

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June 17, 2009

Data SIO, NOAA, U.S. Navy, NGA, GEBCO Image © 2009 TerraMetrics Image IBCAO Data © 2009 MIRC/JHA





Okmok: SO₂ validation with WSU MF-DOAS

Backup

Conclusions

- Multiple species exhibit significant variation in time and space at increasingly finer scales.
- Autocorrelations and cross correlations are strong functions of species, altitude, horizontal distance, and time separation.

28 July 2008 Huntsville, AL



Time



26 June 2009 Huntsville, AL



26 Jun, 2009 Time [GMT]



Ozone and CO 4 days in Atlanta July



Basic variogram interpretation (taken from http://www.ems-i.com/gmshelp /Interpolation/Interpolation_Schemes/Kriging/Variogram_Editor.htm)



- The nugget represents a minimum variance. For this analysis, the nugget is likely dominated by the measurement uncertainty.
- The contribution (sometimes called the "sill") represents the average variance of points at such a distance away from the point in question that these is no correlation between the points.
- The range represents the distance at which there is no longer a correlation between the points.

For the airborne data analysis presented here, the distance (h) is considered to represent satellite resolution and the variogram (γ (h)= average difference) to be an indication of expected sub-grid variability for a given resolution.

Data filtering and assumptions:

Data assessed for all pairs below 2 km

Data pairs with distances of up to 100 km included

Data pairs must span less than 30 minutes which minimizes differences that may be attributed to chemistry (especially for NO2) and transport.

Assessed variables are measured at 1 hz (roughly 100 m resolution for NOAA P-3 and 150 m for NASA DC-8)

Data pairs are restricted to daylight conditions as defined by solar zenith angles of 70 degrees or less

Data are assumed to be isotropic (i.e., vector direction between data pairs is not important)

Data are assumed to represent a well-mixed boundary layer (i.e., vertical separation between data pairs is not used as a discriminator)

Normalized NO2 Distributions: CARB data shows the broadest distribution which helps corroborate the larger relative NO2 differences observed on the previous slide.

