Utilization of Satellite Measurements at Va rious Stages of Atmospheric Modeling to Improve Air Quality Simulations

Daewon W. Byun,

Chang-Keun Song, Hyun Cheol Kim, Soontae Kim, and Bonnie Cheng

¹Institute for Multi-dimensional Air Quality Studies (IMAQS), University of Houston, Houston, TX

Objectives

- Present examples of utilizing satellite data for improvi ng "inputs" for air quality modeling
 - Meteorological modeling: use satellite derived high-resoluti on land use/land cover data for improving surface exchange processes in MM5 (better dynamics input)
 - Emissions modeling: use of the satellite-derived high resolu tion vegetation data and compare biogenic emissions estima tes (more accurate emissions input)
 - Air quality modeling: data assimilation of total ozone colum n in regional modeling (improved lateral & top BCs)

Methods

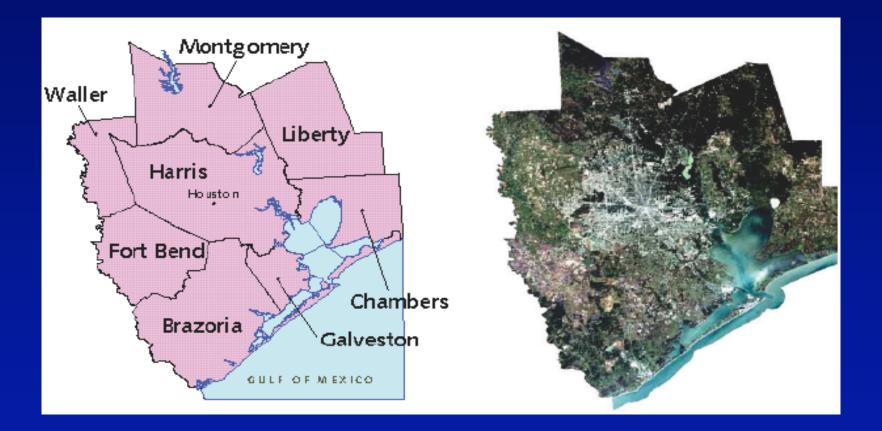
Conduct meteorological, emissions, and air quality sensitivity modeling by ...

- Processing the satellite-derived Land Use/Land Cover data for b ase and future years for meteorological and biogenic emissions.
- Simulating meteorological conditions using a modified MM5 wit h comprehensive NOAH land surface model.
- Estimating biogenic emissions with leaf biomass for new satellit e-derived LULC classes.
- Link satellite-derived ozone data from global chemistry model (RAQMS) for regional air quality modeling

TFS 2000 LULC data

- Texas Forest Service (TFS), with the support of Texas Commission on Environmental Quality (TCEQ), has compiled a new high-resolut ion land use and land cover (LULC) dataset for the eight counties in the HGA to characterize regional changes in vegetation and tree spe cies.
- The updated map of LULC was produced using LANDSAT satellite imagery and ancillary datasets for the base year 2000.
- A supervised classification process that uses an image processing so ftware was employed to define the 8 land cover (LC) classes and 15 land use (LU) classes (GEM, 2003)

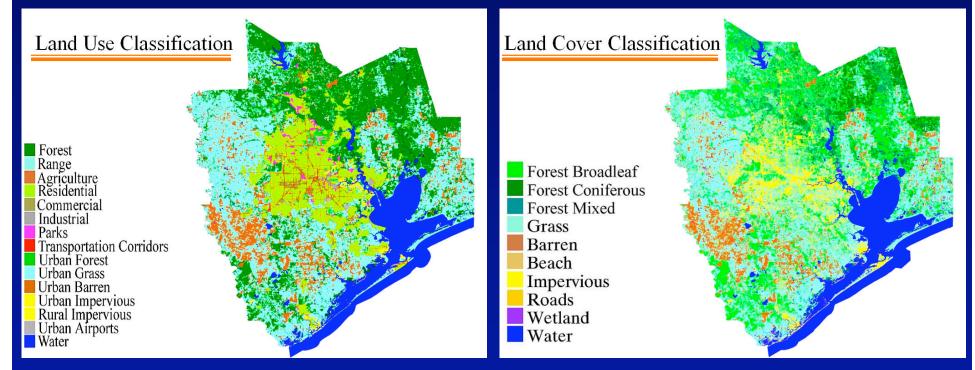
TFS 2000 LULC data



A high-resolution (~30 m) land cover and land use dataset from LANDSAT multi-sp ectral pictures taken in September 2000 for the purpose of managing the urban forest in HGA. (GEM, 2003)

TFS 2000 LULC data

Stetson (2003)



A supervised classification process using image processing software was employed to define the classes described below:

Land Use: forest, range, agriculture, urban/developed Land Cover: forest composition (coniferous, broadleaf, mixed), grass, wet land, water, barren (e.g., beach, bare soil), impervious (roads, parking lots, buildings)

Improve Meteorological Simulations

- To better represent surface-air exchange of momentum, moisture, and heat, w e need to have high quality LULC data
- High quality LULC data improves surface energy balance and thus better cha racterization of PBL mixing and dynamics.
- MM5 with the modified NOAH land surface model (Cheng et al., 2003) was used to improve meteorological input. (cf. Direct assimilation of satellite ob served ground temperature – McNider, UAH)

1 urban 2 dry/cropland 4_mixed crop/pasture 5 crop/grassland 6 crop/woodland 7_grassland 8_shrubland 10 savanna

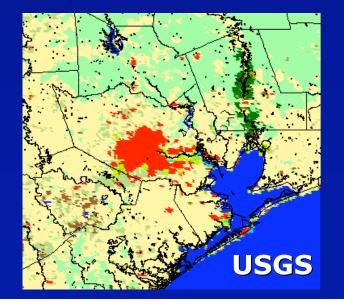
15_mixed forest 16 water

19 barren 22 residential

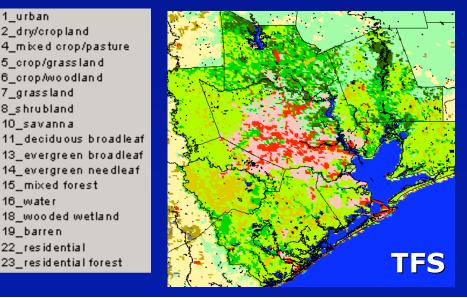
18 wooded wetland

23 residential forest

Original LULC in MM5 system



Improved 8-county LULC



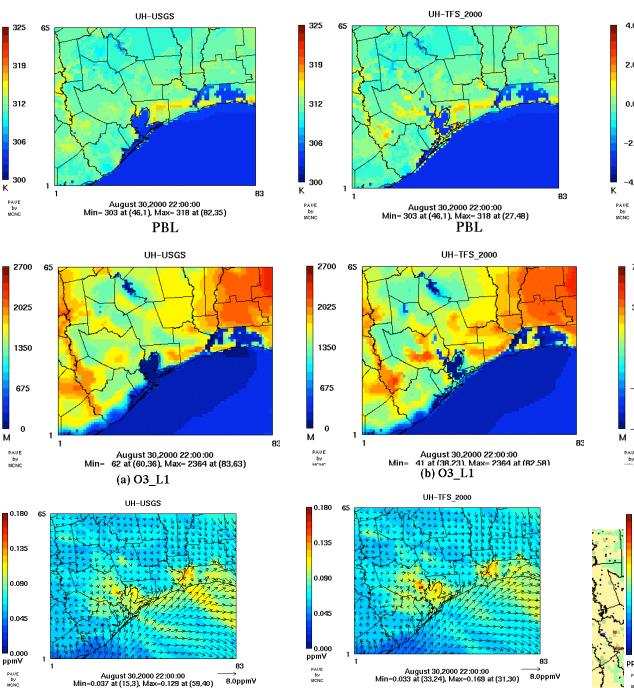
TEMPG

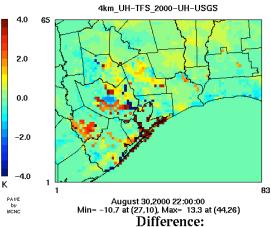
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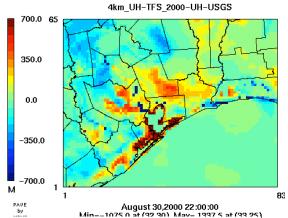
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TEMPG

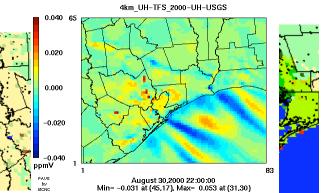
Difference:



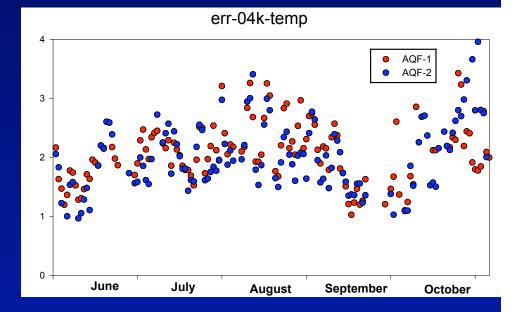




(c) Difference: (b) – (a)



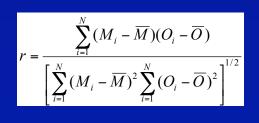
Temperature



Temp MM5 w/ Better LULC (AQF1) R : 0.6679, err : 2.0202 w/ original LUL (AQF2) R: 0.6562, err : 2.0415

Gross Mean Absolute Error (G_ERR)

 $E_{AG} = \frac{1}{N} \sum_{i=1}^{N} \left| M_i - O_i \right|$



Correlation coefficient (R),

Summary of Forecasting Simulatio ns June – October, 2005

Wind Speed

WS w/ better LULC R: 0.2272, err : 1.6862 w/ original LULC R : 0.1865, err : 1.7459

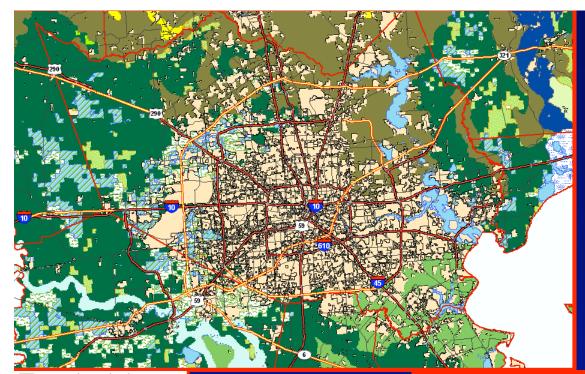
Ozone

Error point : Daily sum of absolute oz one conc. differences between obs. & model

Use of satellite-derived high-resolution LULC in MM 5 improves ozone forecasting!

Biogenic emissions

- The GloBEIS version 3.1 (<u>http://www.globeis.com/</u>) (Guenther et al., 1998; Yarwood, 1999 & 2002) was used.
- The internal database was revised to couple with the TFS LULC data and leaf mass density (LMD) data prepared for HGA.
- Meteorology inputs for GloBEIS3
 - Used the same TCEQ's photo synthetically active radiation (PASR) estimated from GOES satellite data
 - Used the canopy temperature from MM5 simulations with T FS LULC.

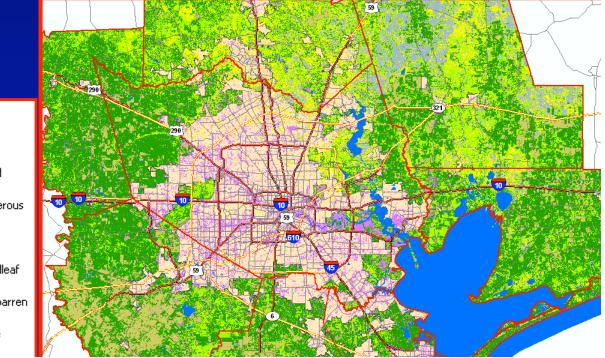


TCEQ Bio. LU/LC (Wiedinmyer et al., 2001)

TFS LANDSAT based Land Cover

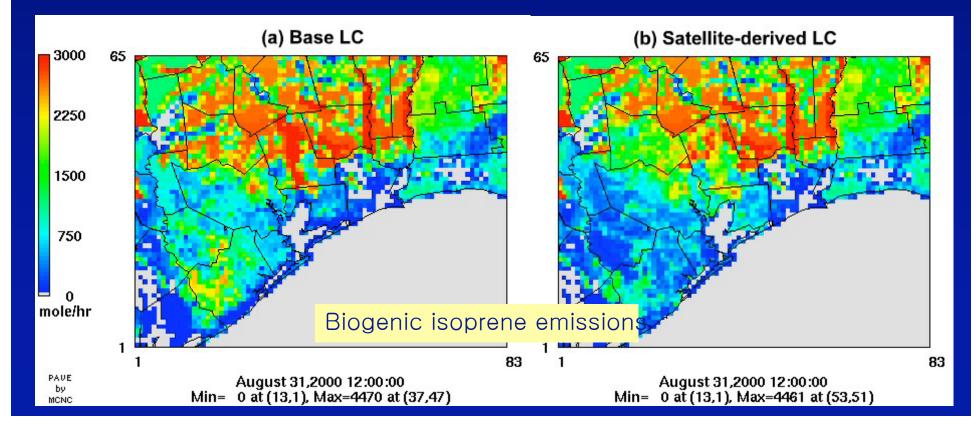
Crops w/ Grassland
 Crops w/ Wetlands
 Crops w/ Woodlands
 Bald Cypress Water Tupelo Swa
 Bluestem Grassland
 LCC Woods
 Evergreen Forest with Hardwood
 Grasslands and Young Forest
 Marsh Barrier
 Pecan Elm Forest
 Pine, Hardwood Forest
 Urban
 Water
 Willow Oak, Water Oak, Blackgu

Water Impervious Forest_Mixed Forest_coniferous Urban_Areas Urban_Areas Bare_fields_barren Grass_Range

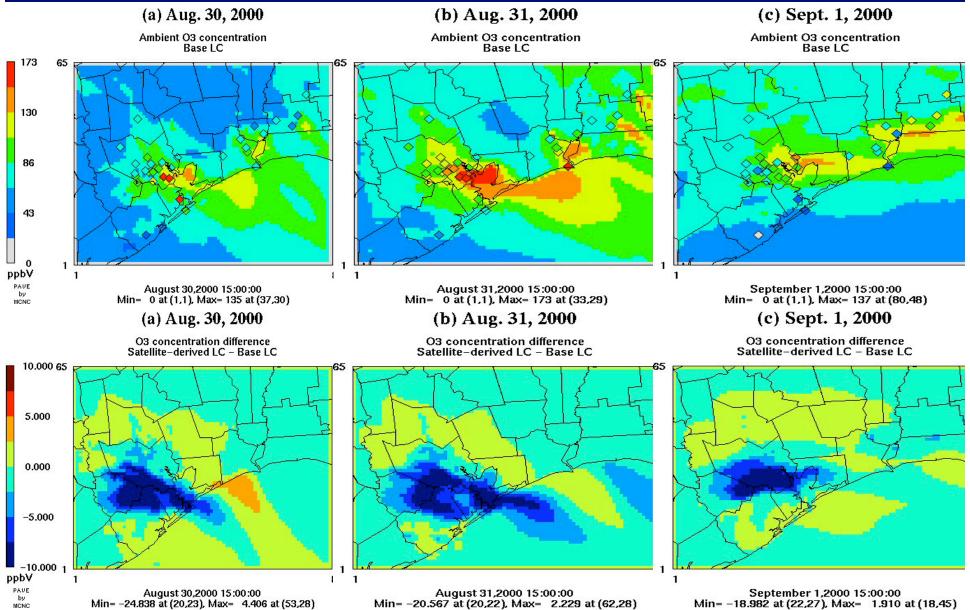


Effects of using different LU/LC data Difference – TCEQ-bio LU/LC vs. TFS LANDSAT LC

at least consistent with surface-air excha nge spatial distribution patterns in MM5



Base case O3 & Difference between LANSAT LC & TCEQ Biogenic emissions



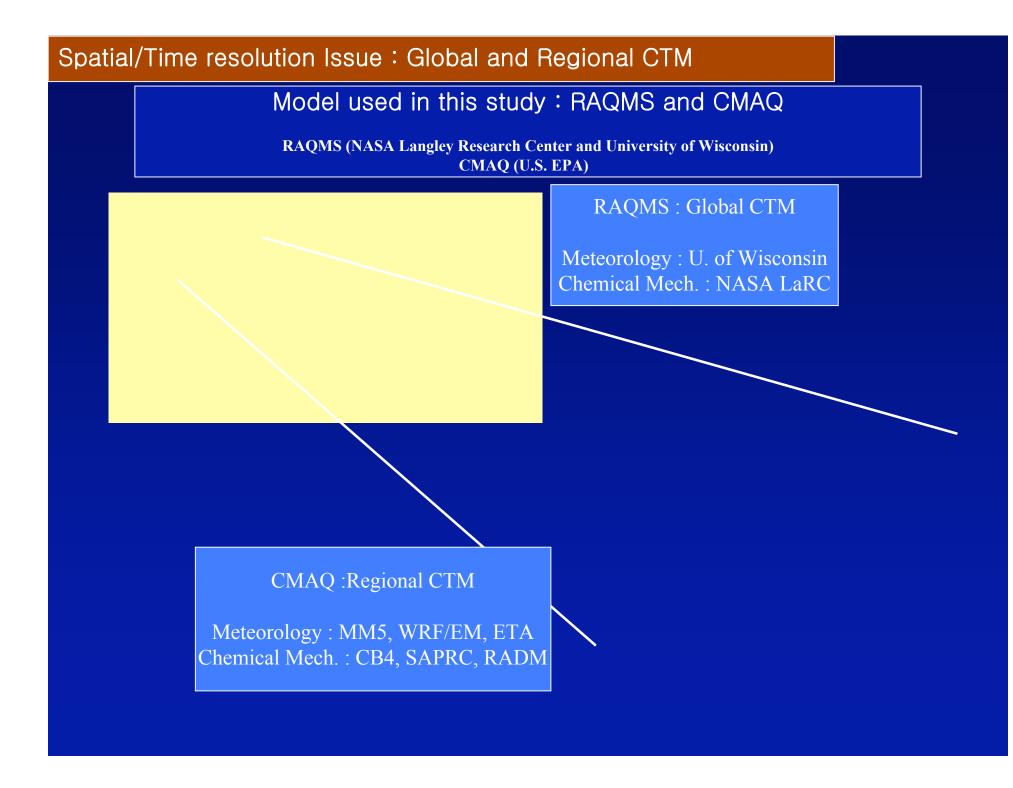
Min= -18.982 at (22,27), Max= 1.910 at (18,45)

Chemical data assimilation of satellite obs ervation for regional air quality modeling

 Use global scale tropospheric chemistry model with satellite data assimilated to improve lateral & top BCs

Link GEOS-CHEM or RAQMS → CMAQ

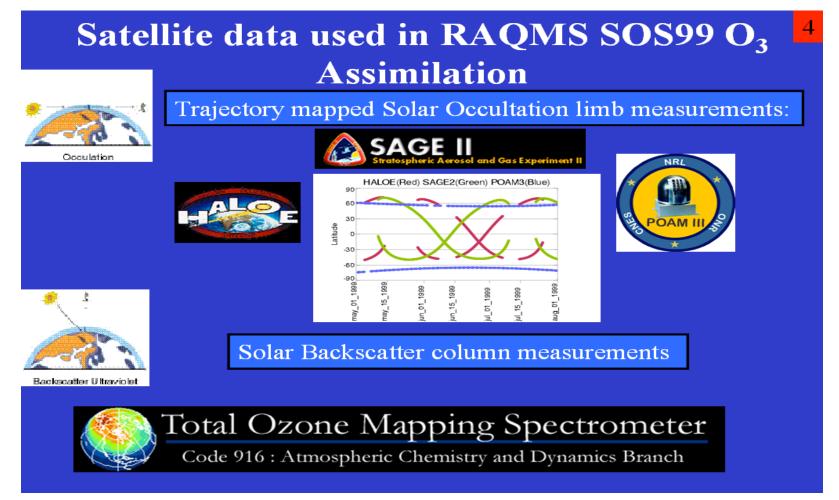
 Assimilate satellite observations directly in CMAQ



RAQMS Satellite Data Assimilation

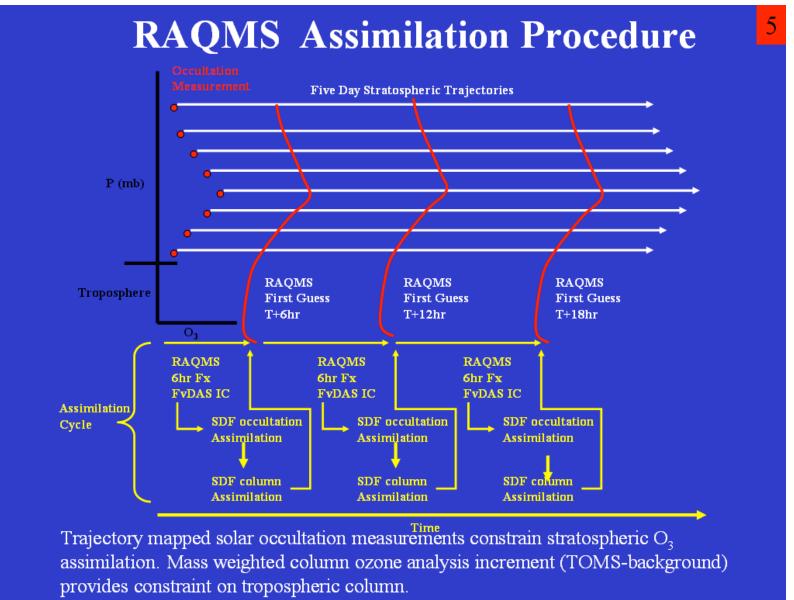
[Pierce et. al., 2004, NCEP Talk]

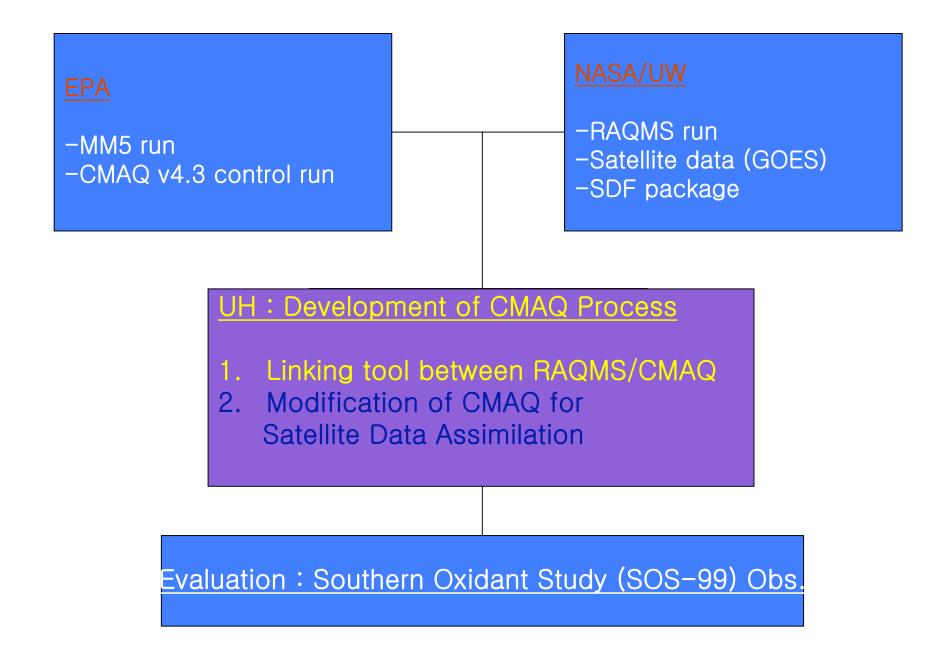
RAQMS uses the Statistical Digital Filter (SFD) [Stobie, 2000] (an Optimal Interpolation approach) and online chemical predictions, to evaluate the feasibility of assimilating trajectory mapped solar occultation and TOMS total column measurements.



RAMQMS Satellite Data Assimilation

[Pierce et. al., 2004, NCEP Talk]

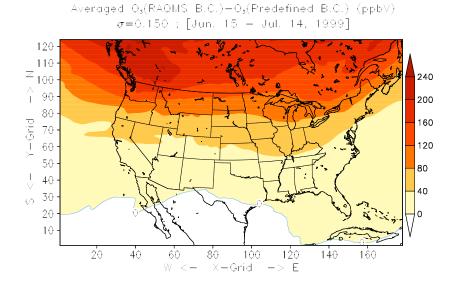




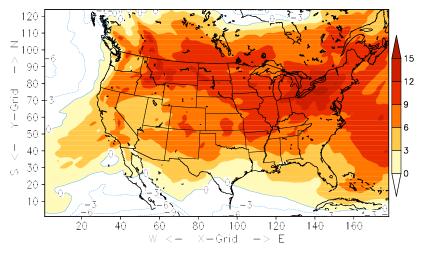
Evaluation

– CMAQ/Profile Results
– CMAQ/RAQMS Results

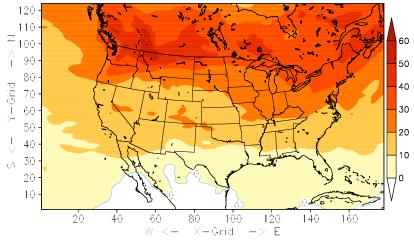
by using SFC and O3 radiosonde obs. (SOS -- June 15-July 14, 1999)



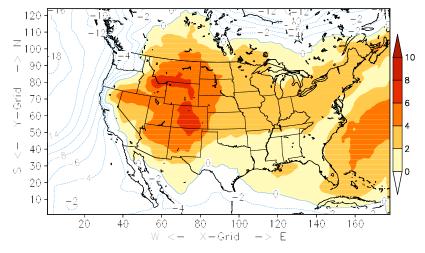
Averaged $O_3(RAQMS B.C.) - O_3(Predefined B.C.)$ (ppbV) σ =0.650 ; [Jun. 15 - Jul. 14, 1999]



Averaged O₃(RAQMS B.C.)-O₃(Predefined B.C.) (ppbV) σ =0.350 ; [Jun. 15 - Jul. 14, 1999]

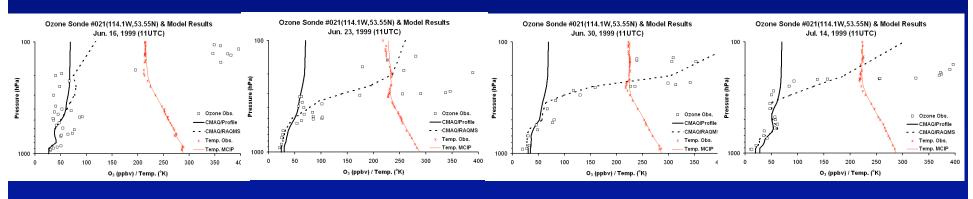


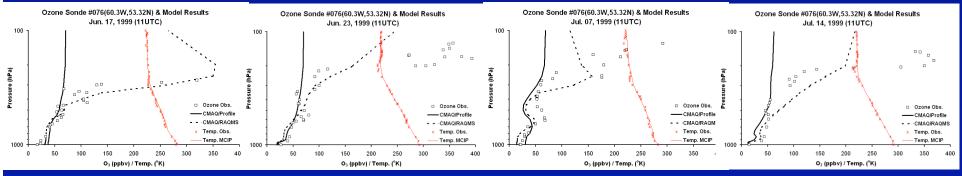
Averaged $O_3(RAQMS B.C.) - O_3(Predefined B.C.)$ (ppbV) σ =0.998 ; [Jun. 15 - Jul. 14, 1999]

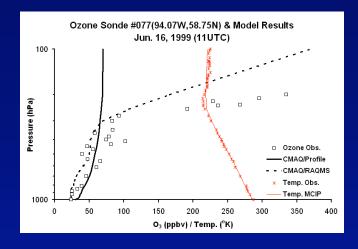


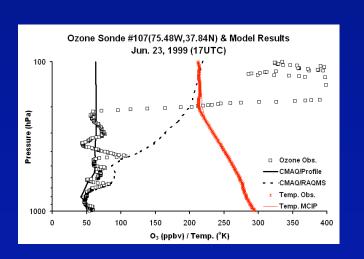
Results : Evaluation for Vertical Sounding

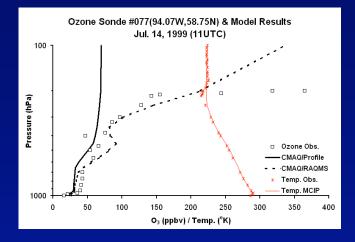
SITE #	Location	Nation	LAT.	LON.	ALT.
STN021	EDMONTON/STONY PLAIN	CAN	53.55	-114.10	766.00
STN076	GOOSE BAY	CAN	53.30	-60.36	40.00
STN077	CHURCHILL	CAN	58.75	-94.07	35.00
STN107	WALLOPS ISLAND	USA	37.90	-75.48	13.00

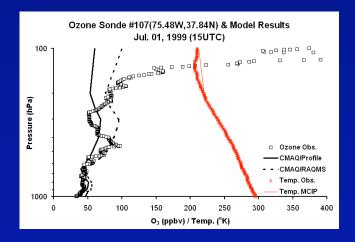






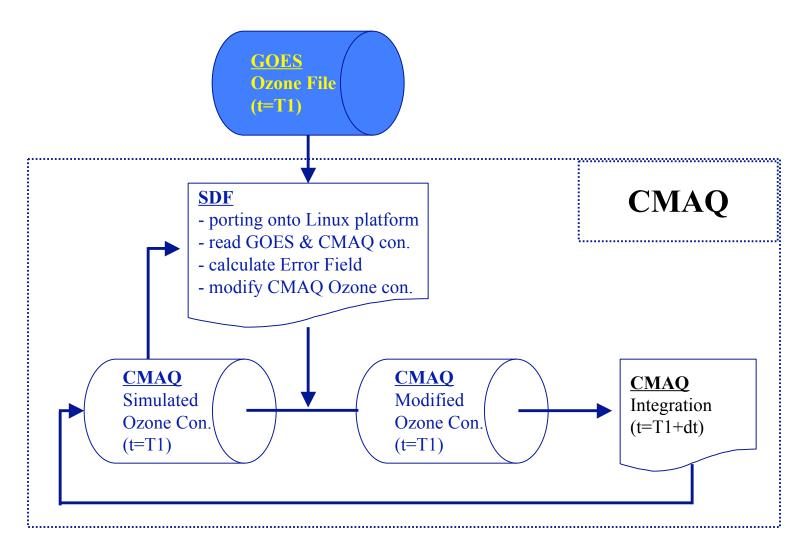




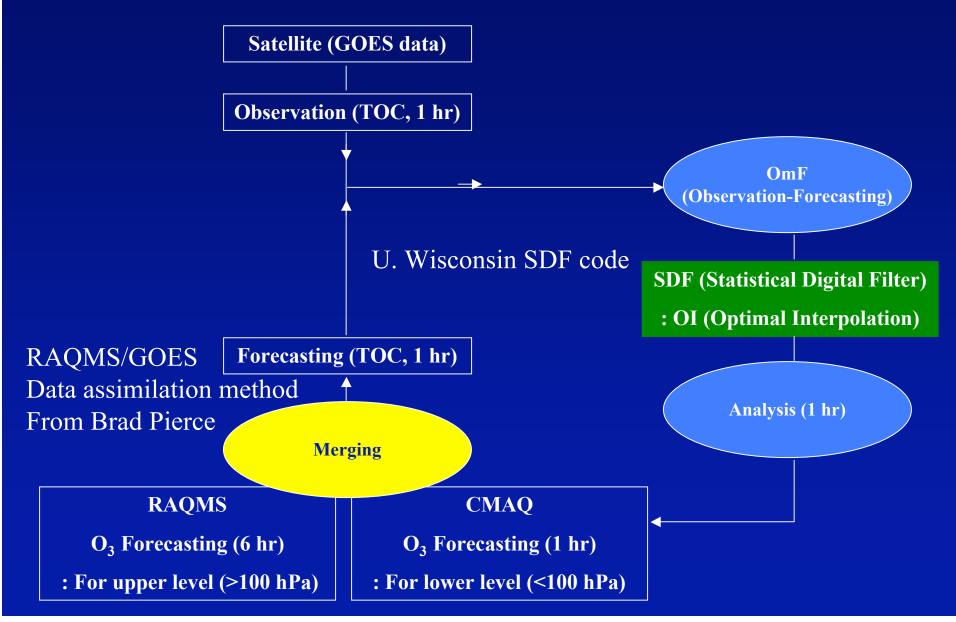


Direct assimilation of satellite data in regional modeling

CMAQ/SDF with GOES Ozone

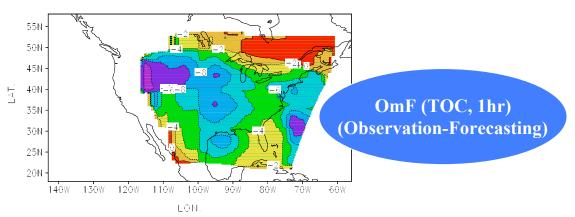


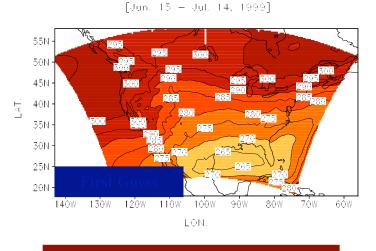
Assimilation of satellite column ozone data for regional air quality modeling: CMAQ/SDF



First guess (TOC, 1hr) **Observation (TOC, 1 hr)** Averaged Total O_3 Column from CMAQ and RAQMS (DU) Averaged Obervation (GOES data) of Total O₃ Column (DU) [Jun. 15 - Jul. 14, 1999] [Jun. 15 - Jul. 14, 1999] 55N 55N 295300 400 50N 50N · 204 380 45N 45N · 360 285 40N 140N · LAT. 340 320 35N · 35N · 300 30N · 30N · 280 25N· 25N · 260 20N 20N 130W 120W 110W 140W 130W 120W 110W -100W 90W 80W 60₩ 140W 100W 90W 80₩ 70W 60W 70W LON. LON.

Averaged OmF(Observation-Model) of Total O, Column (DU) [Jun. 15 - Jul. 14, 1999]

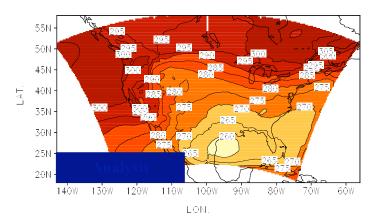




Averaged Total O3 Column from CMAQ and RAQMS (DU)

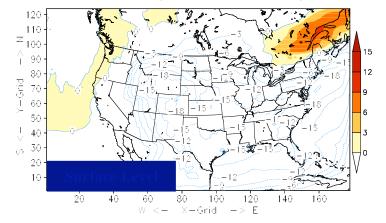
SDF (Statistical Digital Filter) : OI (Optimal Interpolation)

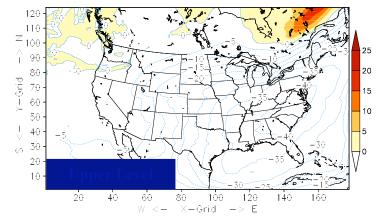
Averaged Total O₃ Column from CMAQ/SDF (DU) After SDF implementation [Jun. 15 - Jul. 14, 1999]



Result : [CMAQ/SDF – CMAQ/Default] O₃ concentration (ppb)

Averaged O₃(CMAQ/SDF)-O₃(CMAQ/Default) (ppbV) σ =0.998 ; [Jun. 15 - Jul. 14, 1999]





On-going work

(1)Utilization of satellite-derived sea-surface temperature in MM5 modeling to improve meteorological inputs

 (2)Inclusion of satellite-derived fire emissions to improve emissions inputs (NESDIS, ARL, NCAR - C. Wiedinmyer, Harvard - J. Logan & D. Jacob)

(3)CMAQ-4D VAR (in collaboration with Sandu (VT), Carmichael (U. Iowa), HARC, EPA, CalTech) Acknowledgement

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SAIC: Fred Vukovich,

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