

出國報告（出國類別：會議）

## 第18屆空氣品質模式發展研討會

服務機關：行政院環境保護署

姓名職稱：蔡宜君技士

派赴國家：美國

出國期間：民國108年10月20日至10月26日

報告日期：民國109年1月17日



## 摘要

本署全年度每日 3 次發布全臺七大空品區未來 3 日及離島隔日空氣品質預報，提供民眾外出活動及相關應變單位決策參考。而無論氣象觀測或是環境品質監測之技術，以及預報模式發展，皆隨科技發展日益進步，為提升空氣品質預報模擬技術及預報作業流程，並了解國際空氣品質預報及遙測環境品質發展情形，爰規劃至美國參加第 18 屆空氣品質模式研討會(The 18th Annual Community Modeling and Analysis System Conference，簡稱 CMAS)。

本次研討會從 2019 年 10 月 21 日至 23 日為期三天，計有 17 國環境部門代表、民間單位及學術研究人員與會，共有 270 多位專家、學者參與，總計有 82 篇學術演講以及 66 篇論文海報。本次研討會討論議題包含模式發展、環境品質監測、排放清單建立機制、空品模式如何耦合氣象模式、探討遙測環境品質技術、空氣品質對於健康之風險評估、經濟影響及物聯網感測器數據比較等等。藉由參加本次研討會了解國際專家、學者於環境品質相關研究及監測技術最新研究成果。本署與國立中央大學大氣科學學系鄭芳怡副教授亦共同以「Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan」(利用天氣型態校正臺灣 PM<sub>2.5</sub> 預報誤差之結果評估) 為主題發表，將本署空氣品質預報偏差校正成果展現於國際研討會，並吸取與會國際專家、學者意見及建議。

美國環保署作業用空氣品質模式(CMAQ)已發展至第 5.3 版，針對污染物模擬的化學機制更臻完善，並預計將預報長度從 48 小時延長至 72 小時，另美國國家太空總署(NASA)發展全球尺度模式結合衛星監測資料，針對臭氧、二氧化氮等氣體之三維分布(GEOS-CF)發布 5 天預報。國際間環保相關部門為提升預報能力，皆投入許多資源及研究人力於預報模式的精進。除邁向長天期預報，亦納入高解析度衛星資料輔助空氣品質預報，值得我國精進空品預報相關業務參考。

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## 壹、目的及背景說明

本署全年度每日 3 次發布全臺七大空品區未來 3 日及離島隔日空氣品質預報，提供民眾外出活動及相關應變單位決策參考。本署預報作業人員檢視氣象觀測資料（如降雨、衛星 AOD 產品等）及空氣監測資料，判斷空氣品質受氣象條件影響關係，並參考人為排放分布輔助了解污染分布狀況，再以氣象預報及空氣品質預報模式推估未來 1 至 3 日空氣品質變化情形，製作空氣品質預報後於本署空氣品質監測網發布。而無論氣象觀測或是環境品質監測之技術，以及預報模式發展，皆隨科技發展日益進步，為提升空氣品質預報模擬技術及預報作業流程，並了解國際空氣品質預報及遙測環境品質發展情形，爰規劃至美國參加第 18 屆空氣品質模式研討會(The 18th Annual Community Modeling and Analysis System Conference，簡稱 CMAS)。

CMAS 由美國環保署與北卡羅萊納大學主辦，自 2000 年舉辦第一屆以來，至 2019 年共辦理 18 屆研討會，舉辦地點皆在北卡羅來納大學教堂山分校，此研討會為全球空氣品質模式研發者及使用者提供交流討論的平台。來自全球各地的空氣品質模式的專家、學者藉此機會交流意見，討論議題多為模式發展、環境品質監測、排放清單建立機制、空品模式如何耦合氣象模式等，近年亦加入探討遙測環境品質技術、空氣品質對於健康之風險評估、經濟影響及物聯網感測器數據比較等等應用性議題。本次研討會從 2019 年 10 月 21 日至 23 日為期三天，共有來自 17 個國家、270 多位專家、學者參與，總計有 82 篇學術演講以及 66 篇論文海報。

此次參加研討會目的除前述所言，本署與國立中央大學大氣科學學系鄭芳怡副教授亦共同以「Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan」(利用天氣型態校正臺灣 PM<sub>2.5</sub> 預報誤差之結果評估) 為主題發表，將本署空氣品質預報偏差校正成果展現於國際研討會，並吸取與會國際專家、學者意見及建議。

## 貳、研習過程

日期	工作內容概要
108.10.20	啟程，出發至美國北卡羅萊納州教堂山。
108.10.21	第 18 屆空氣品質模式研討會於美國北卡羅萊納州教堂山星期五會議中心(Friday Center)舉行，由北卡羅萊納大學教堂山分校副校長 Terry Magnuson 教授進行開幕致詞拉開序幕。本次研討會共有 3 個主題演講，首先為哈佛-史密松天體物理中心 Kelly Chance 教授介紹利用地球同步衛星量測大氣中各種空氣污染物並解析污染物長程傳輸及污染物與氣候交互影響，接著由美國環保署 Wayne Cascio 博士說明空氣品質模式研發對於人體健康及經濟的重要性，最後為美國環保署 Sherri Hunt 博士介紹空氣品質預報模式研發所遭遇的問題及挑戰。本日議程主題分別為空氣品質預報模式 (Community Multiscale Air Quality, CMAQ) 最新研究發展趨勢及利用空品模式探討人體曝露對健康的風險等 20 篇演講，另於論文海報解說階段共 32 篇論文海報參加。
108.10.22	研討會第 2 天議程主題為排放量估計及多尺度模式應用及評估，共有 32 篇演講。美國太空總署全球模式與同化辦公室 Emma Knowland 博士介紹全球觀測系統成分預報模式(Global Earth Observing System Composition Forecast model, GEOS-CF)發展現況。下午時段為論文海報解說，共 34 篇論文海報參加，其中本署與國立中央大學鄭芳怡老師共同投稿「利用天氣分類偏差校正方法修正臺灣 PM <sub>2.5</sub> 預報結果評估」。
108.10.23	研討會第 3 天議程主題為空氣品質與氣候或能源應用之關係、模式個案診斷分析與敏感度分析、污染物遙測/感測技術研究等 3

項，共有 30 篇演講，其中美國太空總署 Laura Judd 博士報告於紐約長島地區進行機載遙測實驗，量測地面高空間解析度臭氧及 NO<sub>2</sub>，解析污染物時空分布。

108.10.24 返程，至美國舊金山轉機

108.10.25-26 返程，回到臺北。

### 參、心得與建議事項

(一) 本次會議旨在空氣品質預報模式研發成果及其在各領域的應用，藉由持續更新的空氣品質預報模式探討空氣品質預報的精進、污染物管制策略制定、增加衛星資料提高模式解析度以及應用於人體曝露之健康風險、經濟影響等等。參訪重點主要國外空氣品質預報模式及遙測/感測技術發展趨勢，成果評估如下：

本次投稿主題為「Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan」，由主辦單位安排以論文海報方式進行展示，為了改善空氣品質預報模式結果，利用風場及海平面氣壓做群集分析，計算出 6 種天氣型態，並歸納各種天氣型態分別於 PM<sub>2.5</sub> 預報誤差的特徵藉此進行偏差校正，對比原有利用統計模型的系統性偏差校正等模式後處理，準確度有明顯提升，於會場亦得到熱烈討論。

美國國家氣象局所發展的國家空氣品質格點預報系統(The National Air Quality Forecasting Capability, NAQFC)所使用的空氣品質預報模式(Community Modeling Air Quality, CMAQ)已發展至 5.3 版本，並且預計將預報長度從 48 小時延長至 72 小時。

美國近年積極發展衛星遙測技術並將其資料加入全球預報模式，例如全球觀測系統成分預報模式(Global Earth Observing System

Composition Forecast model, GEOS-CF), 此模式可合成三維的大氣成分結構如臭氧、氮氧化物、有機化合物等, 解析度為 25 公里, 預報長度可達 5 天, 除有利掌握大範圍的長程傳輸, 亦用於大範圍的能源開發評估。

本次會議安排數個有關感測器研究的演講, 顯示感測器數據研究亦為未來空氣品質預報模式研究趨勢。北卡羅萊納大學夏洛特分校 Brian Magi 助理教授進行感測器(PurpleAir PA-II)與 FEM 標準方法(BAM 1022)測值、濕度及溫度做多線性回歸校正, 結果發現在中、高濕度時有更明顯的改善, 顯示濕度越高越容易造成感測器數值偏差, 此與國內感測器數據表現類似, 但此研究目前僅計算低濃度表現(約 20 微克/立方公尺以下), 未來將持續研究較高濃度誤差之校正。同時亦顯示感測器數據研究為未來精進空氣品質預報模式研究趨勢。

## (二) 心得與建議：

考量 CMAQ 第 5 版以後針對污染物模擬的化學機制更臻完善, 建議將本署空氣品質預報系統更新至 CMAQv5.0 以上版本並介接氣象局作業化 WRF 之氣象預報資料。另本署空氣品質預報模式系統預報長度為 72 小時, 建議改善模式能力的同時可評估延長預報時段, 以利本署未來提供較長天期預報產品。

本署與交通部中央氣象局合作, 介接該局衛星資料及相關氣溶膠產品, 並於 2018 年新增日本向日葵衛星 8 號資料產品。建議持續發展衛星產品推估空氣污染物濃度技術, 除了透過衛星遙測技術具有大範圍的環境監測量能, 更可作為未來評估使用衛星資料改善模式排放量或精進空氣品質預報模式參考, 提升本署空氣品質預報能力。



附錄一、研討會相關照片



圖 1、研討會情形

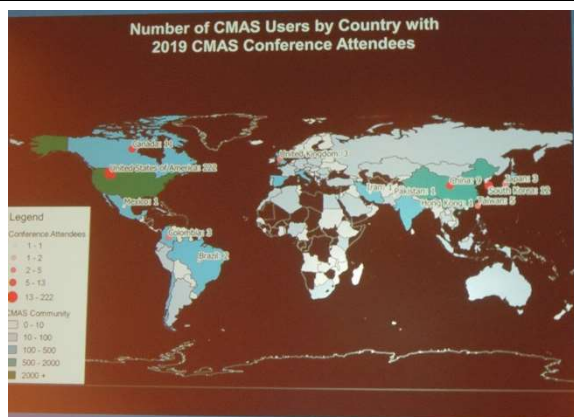


圖 2、本次研討會參與國家分布圖



圖 3、研討會中場休息時間學者討論情形(1)

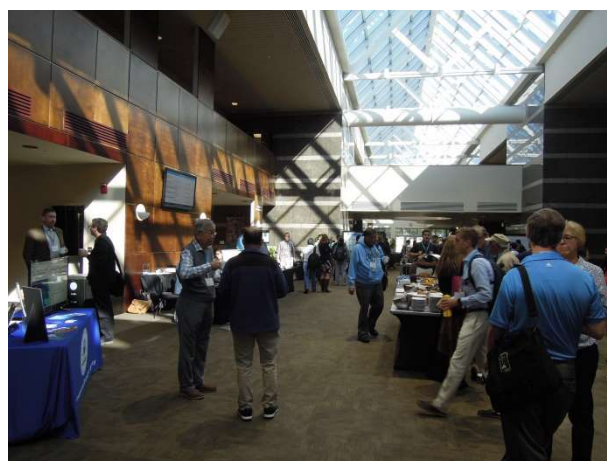


圖 4、研討會中場休息時間學者討論情形(2)

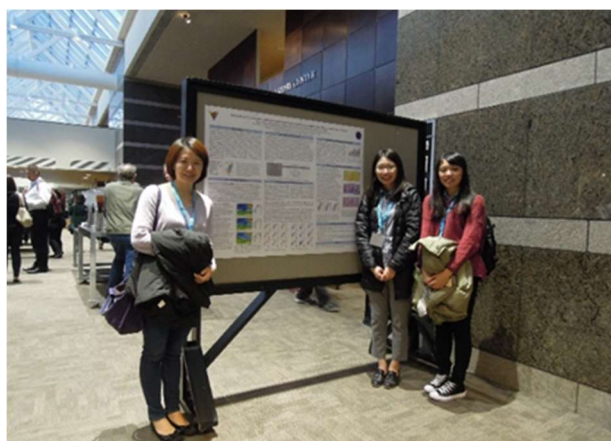


圖 5、與國立中央大學鄭芳怡老師（左 1）及李怡茹博士生（右 1）於投稿海報前留影

National Aeronautics and Space Administration

### Summary

- GEOS-CF produces daily global air quality forecasts at 25km (16 miles) horizontal resolution in near-real time !!
- Output available at [fluid.nccs.nasa.gov/cf](http://fluid.nccs.nasa.gov/cf)

**Under development:**

- Assimilation system for trace gases (O<sub>3</sub>, NO<sub>x</sub>, CO, & others)

EOS MLS

GMAO Global Modeling and Assimilation Office  
[http://gmao.gsfc.nasa.gov/val/val\\_gmao@GEOS-CF/](http://gmao.gsfc.nasa.gov/val/val_gmao@GEOS-CF/)  
 S.A. Kinnison@nasa.gov  
 ChrisFitzg@nasa.gov

圖 6、GEOS-CF 發展現況

附錄二、研討會議程

October 21, 2019		
	<b>Grumman Auditorium</b>	
7:30 AM	Registration and Continental Breakfast	
8:00 AM	A/V Upload	
	<b>Opening Remarks</b>	
8:30 AM	Dr. Terry Magnuson, Vice Chancellor for Research, UNC Chapel Hill "State of the CMAS Center - 18 years Serving the CMAS Community"	
8:40 AM	Sarav Arunachalam, Acting Director, CMAS Center "The TEMPO Green Paper: Applications in air quality and health, agriculture, forestry, and economics"	
8:50 AM	Dr. Kelly Chance, Harvard-Smithsonian Center for Astrophysics "New Insights into the Health Effects of Air Pollution and Persistent Knowledge Gaps: Why is CMAS more important now than ever?"	
9:20 AM	Wayne E. Cascio, M.D., Director, National Health and Environmental Effects Research Laboratory, U.S. EPA "Challenges in Atmospheric Science and Air Pollution: Directions and Questions for Modelers"	
9:50 AM	Dr. Sherri W. Hunt, National Center for Environmental Research, U.S. EPA "How to Enjoy the CMAS Conference"	
10:20 AM	BH Baek, UNC Institute for the Environment	
10:25 AM	<b>Break</b>	
	<b>Grumman Auditorium</b>	<b>Dogwood Room</b>
	<b>Model Development</b>	<b>Modeling to Support Exposure, Health Studies, and Community-scale Applications</b>
	Chaired by Ben Murphy (US EPA) and Talat Odman (Georgia Tech)	Chaired by Vlad Isakov (US EPA) and Amir Hakami (Carleton University)
10:40 AM	NOAAs National Air Quality Forecast Capability operational and experimental updates Jose Tirado	Air quality and health benefits of Chinas emission control policies on coal-fired power plants 2005-2020 Ruili Wu
11:00 AM	Implementation of new satellite-based source maps in the FENGSHA dust module and initial application with the CMAQ-based NAQFC system Daniel Tong	Comparing Projected and Modeled Health Benefits of Alternative Energy Futures Kristen E Brown
11:20 AM	Exploring the Vertical distribution of Wildland Fire Smoke in CMAQ Joseph L. Wilkins	Air Quality and Human Health Impacts of Prescribed Fire: Links to PM <sub>2.5</sub> and Asthma in Georgia, USA Talat Odman
11:40 AM	Impact of renoxification on air quality over Northern Hemisphere Golam Sarwar	Health and Economic Impacts of Air Pollution Induced by Weather Extremes over the Continental U.S. Yang Zhang
12:00 PM	<b>Lunch in Trillium</b>	
	<b>Model Development, cont.</b>	<b>Modeling to Support Exposure, Health Studies, and Community-scale Applications, cont.</b>
1:00 PM	Predicting the phase state and phase separation of atmospherically relevant aerosols and its impact on multiphase chemistry in a	Impacts of Medium-Range Transport of Biomass Burning Aerosols on Air Quality and Public Health in Colombia

	regionalscale atmospheric model (CMAQ) Quazi Ziaur Rasool	Karen Ballesteros-González
1:20 PM	CAMQ-AI: A computationally efficient deep learning model to improve CMAQ performance over the United States Ebrahim Eslami, Yunsoo Choi	Source Attribution of Global PM <sub>2.5</sub> Mortality Using the Adjoint of Hemispheric CMAQ Amir Hakami
1:40 PM	An efficient way in quantifying the nonlinear response of air pollution to emission changes using the indicator-based response surface model Song Liu	Traffic-Related PM <sub>2.5</sub> and NO <sub>2</sub> Health Risk Assessment in the United States: A Fine Scale Hybrid Modeling Approach Alejandro Valencia
2:00 PM	Estimating damages from climate change and air pollution for subnational incentives for green onroad freight Rebecca Kaarina Saari	Hybrid Air Quality Modeling in West Oakland, California, to Support the Development of an Emissions Reduction Program Stephen Reid
2:20 PM	Improving SCICHEM Pre- and Post-Processing Setup Speed Using Cloud Computing Amy McVey	Using mobile phone data to quantify the impact of spatiotemporal human mobility on air pollution exposure estimation Haofei Yu
2:40 PM	CMAQ 5.3b PARALLEL PERFORMANCE WITH MPI AND OPENMP George Delic	HyADS: A tool for estimating nationwide exposures to emissions from large numbers of sources Lucas Henneman
3:00 PM	<b>Break</b>	
3:30 PM	<b>Lightning Poster Talks: Day 1</b>	
4:00 PM	<b>Poster Session: Day 1</b>	
<b>October 22, 2019</b>		
	<b>Grumman Auditorium</b>	<b>Dogwood Room</b>
7:30 AM	Registration and Continental Breakfast	
8:00 AM	A/V Upload	
	<b>Multi-scale Model Applications and Evaluations</b> Chaired by Christian Hogrefe (US EPA) and Marina Astitha (University of Connecticut)	<b>Emissions Inventories, Models, and Processes</b> Chaired by Jeff Vukovich (US EPA) and Mike Moran (Environment and Climate Change Canada)
8:30 AM	Global Sources of North American Ozone Barron Henderson	Version 1.0 of the National Emissions Inventory Collaborative 2016 Emissions Modeling Platform Alison Eyth
8:50 AM	Local to Global Air Quality Simulations using the NASA GEOS Composition Forecast Model GEOS-CF Emma Knowland	Updates and Improvements to 2023 and 2028 Emission Inventory Projections Caroline Farkas
9:10 AM	CAMx Ozone Source Apportionment Application over the Northern Hemisphere Pradeepa Vennam	Development of a Year 2016 fire inventory for United States through a multi-agency inventory collaboration effort Jeff Vukovich
9:30 AM	Quantifying Economic Damages from Crop Productivity Loss Due to Ozone Precursor Emissions via Hemispheric Adjoint Analysis Yasar Burak Oztaner	Estimating Emissions from Wildland Fires for Air Quality Modeling: Status Update George Pouliot
9:50 AM	Incorporation of Volcanic SO <sub>2</sub> Emissions in H-CMAQ Modeling System and its Impacts on Sulfate Aerosol Concentration across the Northern Hemisphere	Evaluating Updated Tools for the Estimation of Wildland Fire Emissions James Beidler

	Syuichi Itahashi	
10:10 AM	<b>Break</b>	
10:40 AM	Evaluation of the Community Multiscale Air Quality (CMAQ) model version 5.3 K. Wyatt Appel	Development of a Fast Fire Emission Processor and Its application with HMS-Bluesky and GBBEPx Inventories Youhua Tang
11:00 AM	Evaluation of CMAQ Estimated NOx from 2002 to 2016 Kristen Foley	Impacts of a National Survey on the Development of the 2017 National Emissions Inventory Rich Mason
11:20 AM	Improved estimation of background ozone and emission impacts using chemical transport modeling and data fusion Nash Skipper	EPA's 2018 Emissions & Generation Resource Integrated Database (eGRID): Updates and Improvements Jonathan Dorn or David Cooley
11:40 AM	Evaluating Seasonality and Trends in Modeled PM <sub>2.5</sub> Concentrations Using Empirical Mode Decomposition Marina Astitha	Improvements to EPA's SPECIATE database: SPECIATE 5.0 Madeleine Strum
12:00 PM	<b>Lunch in Trillium</b>	
	<b>Multi-scale Model Applications and Evaluations, cont.</b>	<b>Emissions Inventories, Models, and Processes, cont.</b>
1:00 PM	A Novel Ensemble Design for PM <sub>2.5</sub> Probabilistic Predictions and Quantification of Their Uncertainty Rajesh Kumar	Expansion of a Size Distribution Profile Library for Particulate Matter (PM) Emissions Processing from Three to 30 Source Categories Junhua Zhang
1:20 PM	Lightning Assimilation in WRF4.0.2: Impact of Parameter Options and Introduction of New Lightning Data Daiwen Kang	The estimated impacts of volatile chemical products on particulate matter and ozone criteria pollutants in an urban atmosphere Havala Pye
1:40 PM	Comparing Extreme Weather Events Generated by 36-km and 12-km WRF Simulations Tanya Spero	Initial Development of a NOAA Emissions and eXchange Unified System (NEXUS) Patrick C Campbell
2:00 PM	Particulate matter sensitivity to local emissions and meteorology over a Latin American megacity for source apportionment and uncertainty analysis James East	Towards Refining Estimates of Ammonia Emissions: Modeling Framework Preparations Shannon Capps
2:20 PM	<b>Break</b>	
2:50 PM	Model-Measurement Comparison of Ozone and Precursors Along Land-Water Interfaces during the 2017 LMOS Field Campaign Liljegren, J	Inter-comparison of Mobile Source Emissions from the CARS and CAPSS in Seoul Metropolitan Area, South Korea Minwoo Park
3:10 PM	Application of ozone source apportionment using CMAQISAM during LISTOS Qian Shu	Impact of anthropogenic emissions on urban air quality over the East African big conurbations of Addis Ababa, Kampala and Nairobi. Andrea Mazzeo
3:30 PM	Ramboll Shair: Integrating real-time sensor measurements and regional/local-scale models in Richmond, California Justin Bandoro	Incorporating Isotope into Atmospheric Chemistry Models Huan Fang
3:50 PM	<b>Break</b>	

4:15 PM	<b>Lightning Poster Talks: Day 2</b>	
5:00 - 7:00 PM	<b>Poster Session: Day 2 and Reception with Research-Discussion Tables</b>	
<b>October 22, 2019</b>		
	<b>Grumman Auditorium</b>	<b>Dogwood Room</b>
7:30 AM	Registration and Continental Breakfast	
8:00 AM	A/V Upload	
8:30 AM	Analysis of Recent Ozone Exceedances in Atlanta, GA Byeong-Uk Kim	Air Quality Impacts of Oil and Gas Emissions in the United States Srinivas Reka
8:50 AM	Source Apportionment Modeling to Investigate Local and Non-Local Contributions to Ground-Level Ozone in Albuquerque, New Mexico Garnet Erdakos	Examining the air pollutant emission implications of electric vehicle market penetration scenarios Dan Loughlin
9:10 AM	Mutual comparison of source sensitivities and apportionments obtained by BFM, DDM, and ISAM on PM <sub>2.5</sub> and ozone concentrations over Japan Satoru Chatani	Impact of future electrification of passenger cars on air quality within the United States Huizhong Shen
9:30 AM	Assessing the Impacts of Emissions from Oil and Gas Extraction on Urban Ozone and Associated Health Risks Congmeng Lyu	Optimal use of grid-connected energy storage to reduce human health impacts Qian Luo
9:50 AM	Assessing Air Quality Impact on Non-attainment Regions in Ohio Resulting From Power Plant Closures and Shale Gas Activity Saikat Ghosh	Impact of using the updated detailed-level marine emissions on Canadian Air Quality Rabab Mashayekhi
10:10 AM	<b>Break</b>	
10:40 AM	Vehicle Emission Regulations Save Lives in California Shupeng Zhu	Relaxing energy policies on top of climate change will significantly undermine states efforts to attain U.S. ozone standards Huizhong Shen
11:00 AM	Exploring air emissions of per- and polyfluoroalkyl substances (PFAS): emission rates, chemical and depositional fate, and transport E.L. D'Ambro	Air pollution control strategies directly limiting national health damages in the U.S. Yang Ou
11:20 AM	Assessing PM <sub>2.5</sub> Model Performance for the Conterminous U.S. with Comparison to Model Performance Statistics from 2007-2015 James T. Kelly	The Fast Air Quality Responses from the Fine Particle Reduction Measures of South Korea Jinseok Kim
11:40 AM	Modeling of U.S. and International Contributions to Visibility Impairment at Class I Areas Brian Timin	Dynamical downscaling of a global chemistry-climate model to study the influence of climate change on mid-21st century PM <sub>2.5</sub> and Ozone distributions in the Continental US Surendra Kunwar
12:00 PM	Air Quality Modeling of a Typical Wintertime PM <sub>2.5</sub> Pollution Event in Cache Valley, Utah: Implications for Emission Control Strategies. Nancy Daher	Urban-Scale Source Attribution of Greenhouse Gases Using an Air Quality Model Mike Moran

12:20 PM	<b>Lunch in Trillium</b>	
	<b>Remote Sensing/Sensor Technology and Measurements Studies</b> Chaired by Daniel Tong (George Mason University) and Matthew Alvarado (AER)	<b>Air Quality, Climate, and Energy, cont.</b>
1:20 PM	Estimation of Surface NO <sub>2</sub> Using Airborne Remote Sensing Data and CMAQ Model Output from DISCOVER-AQ Campaigns Kenneth E. Pickering	
1:40 PM	A Retrieval Closure Study to Determine the Best Metrics for Evaluating the CMAQ Model with CrIS NH <sub>3</sub> Retrievals Matthew J. Alvarado (AER)	The impact of the direct effect of aerosols on meteorology and air quality using aerosol optical depth assimilation during the KORUS-AQ campaign Jia Jung
2:00 PM	High-spatial resolution airborne mapping of NO <sub>2</sub> column densities of New York City and Long Island Sound Laura Judd	Understanding the Impacts of Land Use and Land Cover Change on Regional Climate in Sub-Saharan Africa: A Cautionary Tale for Regional Climate Modeling Timothy Glotfelty
2:20 PM	Overview of measurements and preliminary findings from the Long Island Sound Tropospheric Ozone Study 2018 (LISTOS) Luke Valin	Intensification of Extreme Precipitation in Eastern North Carolina Projected in Dynamically Downscaled CESM and CM3 Models Under Future Scenarios (2025-2100) Using WRF. Anna Jalowska
2:40 PM	Improving Mexico Emissions Inventory Using Satellite Data Tejas Shah	Quantifying the air quality and human health benefits of GHG mitigation Pathways in California Shupeng Zhu
3:00 PM	Comparison of PM <sub>2.5</sub> measured in urban North Carolina settings using a low-cost optical particle counter and Federal Equivalent Methods Brian Magi	Estimation of nose-level NO <sub>2</sub> in the Houston-Galveston area using airborne remote sensing data from DISCOVER-AQ and output from a 4-km resolution simulation with CMAQ D. Allen (Dale Allen)
3:30 PM	<b>The CMAS Job and Career Fair!</b>	
<b>Poster Session 1: Monday at 4:00 PM</b>		
<b>Air Quality, Climate and Energy</b> <a href="#">Modeling the Contribution of Long Range Transport to Nitrogen Deposition in U.S. Hydrological Regions</a> Sharmin Akter <a href="#">Modelling the impact of aromatics on secondary pollutants over eastern China</a> Xiaoyang Chen <a href="#">Greater contribution from land-use related sources to reactive nitrogen deposition under future climate change</a> Yilin Chen <a href="#">Improving Air Quality Simulation for 2016 Summer Ozone Episode by Assimilating Geostationary Satellite Observations</a> Peiyang Cheng <a href="#">Projected Wildfire Impacts on Regional Air Quality in The Western United States</a> XINYI DONG <a href="#">Impact of dust emission on particulate matter in spring of 2018 in China</a> Song Liu <a href="#">Future ozone concentrations in the state of Delaware under climate change scenarios</a> Mojtaba Moghani <a href="#">A multi-scale model analysis of ozone formation in the Bangkok Metropolitan Region, Thailand</a> Pornpan Uttamang <a href="#">Air Quality Impacts of Energy Transitions: Analysis of Alternative Pathways Through 2050 Using the National Energy</a>		

<p><a href="#">Modeling System and Emissions Downscaling</a> Mr. Shen Wang, Ph.D. Student, Johns Hopkins University</p>
<p><b>Model Development</b>  <a href="#">Imputing air quality estimates for alternative meteorological years using relative response values from a single meteorological year.</a> G.E. Bowker  <a href="#">What's New with the I/O API</a> Carlie J. Coats, Jr.  <a href="#">Investigating aqueous production pathways of particulate sulfur in CMAQ with AQCHEM-KMT (version 2) and the sulfur tracking method</a> Kathleen Fahey  <a href="#">Assessment of Environmental Variables that Affect Dissolved Oxygen Concentrations in Lake Erie Using Multi-media Modeling and Machine Learning</a> Christina Feng Chang  <a href="#">Development of PM2.5 short-term prediction model using Artificial Intelligence.</a> JeongBeom Lee, Geon-Woo Yun  <a href="#">Modelling high-resolution NO2 concentrations over the United Kingdom using RapidAIR</a> Nicola Masey  <a href="#">The Effect of Source-Resolved Primary Size Distributions on Particle Concentrations and Size predicted CMAQ</a> Ben Murphy  <a href="#">Improvements to the Integrated Source Apportionment Method (ISAM) for CMAQv5.3 release.</a> S.L.Napelenok  <a href="#">Development of a Visibility Forecasting Product using Canada's Wildfire Smoke Prediction System (FireWork)</a> Rita So  <a href="#">Kalman Filter Bias Correction for Improving Wildfire Smoke 24-hr Average PM2.5 Forecasting for AIRPACT5</a> Joseph K Vaughan</p>
<p><b>Modeling to Support Exposure and Health Studies and Community-scale Applications</b>  <a href="#">Evaluating the air quality impacts of an operational prescribed burning program</a> Sadia Afrin  <a href="#">Development of Smartphone Application for Modeling Personal Exposures to Ambient PM2.5 and Ozone</a> Sarav Arunachalam  <a href="#">Fusing CMAQ with Surface Observations to Improve Estimates of Air Quality &amp; Health Impacts of Oct. 2017 California Wildfires</a> Stephanie Cleland  <a href="#">Mapping Global Surface Ozone Concentrations through the Statistical Fusion of Observations and Models</a> Marissa DeLang and Jacob Becker  <a href="#">CMAQ-Enhanced Estimates of Personal Exposure to Diesel Particulates</a> Cesunica Ivey  <a href="#">Fixed point versus time-activity based personal air pollution exposure assessment: Does the number of days of assessment matter?</a> Xiangyu Jiang  <a href="#">A comparison of land management modeling tools as used for estimating prescribed fire smoke impacts</a> Megan Johnson  <a href="#">Modeling Air Quality and Health Impacts of the 2017 California Wildfires</a> Bonyoung Koo</p>
<p><a href="#">The Impacts of Prescribed Burning on PM2.5 Air Quality and Human Health: Application to Georgia, USA</a> Talat Odman  <a href="#">A Comparative Study of Two-way and Offline Coupled WRF and CMAQ over the continental U.S.: Performance Evaluation and</a>  <a href="#">Impacts of Chemistry-Meteorology Feedbacks on Air Quality</a> Kai Wang  <a href="#">Effects of grid resolution on the global mortality burden of fine particulate matter and ozone</a> Tammy Ruozhang Xu</p>
<p><b>Poster Session 2: Tuesday at 5:00 PM</b></p>
<p><b>Emissions Inventories, Models, and Processes</b>  <a href="#">Using WRF-STILT to Determine the Relative Contributions of US and Mexican Emissions to High Ozone Events in El</a></p>

Paso, Texas  
 Matthew Alvarado  
 A Quantitative Assessment of Vehicle-class and Precursor-specific Contribution from Onroad Emissions to PM<sub>2.5</sub> and O<sub>3</sub> Formation in the Northeast and Mid-Atlantic U.S.  
 Calvin Arter  
 Implementation of a dynamical NH<sub>3</sub> emissions parameterization in CMAQ for improving PM<sub>2.5</sub> simulation in Taiwan  
 Fang-Yi Cheng  
 EPA's National Emissions Inventory for Nonpoint Sources: Process and Improvements for the 2017 NEI  
 Hannah Derrick or Jonathan Dorn  
 U.S. Residential Wood Combustion (RWC) Survey: Results and Use in Improving RWC Emissions in the 2017 National Emissions Inventory  
 David Cooley or Jonathan Dorn  
 Chemical Speciation of PM<sub>2.5</sub> in Bogot, Colombia: Experimental Determination and Simulation with the Chemical Transport Model WRF-Chem  
 Sebastian Orlando Espitia Cano  
 Emissions of PCDD/Fs from 1960 to 2014 in China  
 Ye Huang  
 Hamilton Airshed Modelling System  
 Janya Kelly  
 Practical Application of Python for Emissions Data Processing, QA/QC, and Documentation  
 Byeong-Uk Kim  
 Improving Spatial Resolution of EDGAR Emissions for Mexico  
 Amy McVey  
 Using <sup>15</sup>N in NO<sub>x</sub> emissions for source identification  
 Greg Michalski  
 Modeling and Evaluation of Quebecs Air Quality due to an Increase in the Electric Vehicle Fleet Combined with Off-Road Emission Reductions  
 Jessica Miville

**Multiscale Model Applications and Evaluations**

Modeling SOA Formation over Eastern China: Impacts of Updates in CMAQ v5.3b2 relative to CMAQ v5.1  
 Xiaoyang Chen  
 Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan  
 Fang-Yi Cheng  
 Performance evaluation of the updated and developed air quality forecasting system for South Korea.  
 Dae-Ryun Choi  
 Puget Sound Ports Air Quality Health Impacts Study  
 Mahshid Etesamifard  
 Fine-scale meteorological modeling of the land-sea breeze circulation of Long Island Sound and surrounding coastal areas  
 Robert Gilliam and Ana Torres-Vazquez  
 Evaluation and Intercomparison of Modeled Atmospheric Deposition over North America and Europe An Overview of Phase 4 of the Air Quality Model Evaluation International Initiative  
 Christian Hogrefe  
 Evaluation of GFS-driven CMAQ predictions of PM<sub>2.5</sub> and O<sub>3</sub> at NOAA  
 Jianping Huang  
 Organic mass predictions during PM<sub>2.5</sub> high pollution episodes in China and Korea: high resolution model experiments  
 Hyeon-Kook Kim  
 Changes in ozone production chemistry across the U.S. between 2007 and 2016: An integrated modeling assessment  
 Shannon Koplitz  
 Apportioning Emission Source-group Impacts among Individual Sources: Application to Prescribed Fires  
 Talat Odman  
 Long-range transport of air pollutants over Brazilian Southeastern: Interaction between metropolitan areas  
 Rizzieri Pedruzzi  
 Numerical evaluation of the impacts of emission and meteorological characteristics on air pollution in the Seto Inland Sea region, Japan



Hikari Shimadera

[EPA-State 2016 Model Evaluation Forum](#)

Heather Simon

[The Benefits of Using High-Resolution Sea Surface Temperatures for Simulating Historical and Future Climate Extremes](#)

Tanya Spero

**Regulatory Modeling and SIP Applications**

[Projecting Future Ground-Level Ozone Concentrations in Albuquerque, New Mexico](#)

Shih Ying Chang

[Particulate Matter Emissions from Building Destruction in Armed Conflict Regions](#)

Daniel Han

[Influence of the emission sources on O<sub>3</sub> and PM<sub>2.5</sub> concentrations in Taiwan using CMAQ-HDDM](#)

Yi-Ju Lee

[SCICHEM model performance evaluations for near-field regulatory applications](#)

Ryan Cleary and Chris Owen

**Remote Sensing/Sensor Technology and Measurements Studies**

[Atmospheric Monitoring of Methane Leaks from Shale Oil and Gas Production using Airborne Near-Infrared Spectrometer](#)

Saikat Ghosh

[Community Participatory Mapping Crowdsourced Platform for Monitoring Air Pollution](#)

Wansoo Im, Ph.D.

[Ambient measurements of biogenic VOCs near Cannabis facilities in Denver, CO](#)

Chi-tsan Wang

附錄三、"NOAA's National Air Quality Forecast Capability operational and experimental updates"

# NOAA's National Air Quality Forecast Capability operational and experimental updates

Dorothy Koch<sup>1</sup>, Ivanka Stajner<sup>2</sup>, Jeff McQueen<sup>2</sup>, Pius Lee<sup>3</sup>, Jianping Huang<sup>2,5</sup>, Ho-Chun Huang<sup>2,5</sup>, Li Pan<sup>2,5</sup>, Youhua Tang<sup>3,6</sup>, Daniel Tong<sup>3,6</sup>, Patrick Campbell<sup>3,6</sup>, Ariel Stein<sup>3</sup>, James Wilczak<sup>4</sup>, Irina Djalalova<sup>4,8</sup>, Dave Allured<sup>4,8</sup>, Phil Dickerson<sup>7</sup>, Jose Tirado<sup>1,9</sup>

- (1) NOAA NWS/STI
- (2) NOAA NWS/NCEP
- (3) NOAA ARL
- (4) NOAA ESRL
- (5) IMSG
- (6) Cooperative Institute for Satellite Earth System Studies (CISESS)
- (7) EPA
- (8) CIRES, University of Colorado
- (9) ERG

*with contributions from the entire NAQFC Implementation Team*

CMAS Conference, Chapel Hill, NC

October, 2019



## NAQFC related presentations



- Implementation of new satellite-based source maps in the FENGSHA dust module and initial application with the CMAQ-based NAQFC system, **Daniel Tong (next speaker)**.
- Development of a Fast Fire Emission Processor and Its application with HMS-Bluesky and GBBEPx Inventories, **Youhua Tang (Tue, Oct 22, 10:40 AM - 11:00 AM, Dogwood Room)**.
- Evaluation of GFS-driven CMAQ predictions of PM2.5 and O3 at NOAA, **Jianping Huang (poster session, Tue, Oct 22, 05:00 PM - 07:45 PM, Main Atrium)**.



## NWS context



The Office of Science and Technology Integration (OSTI) Modeling Division (about \$29M/yr)

Includes:

- NGGPS (Next Generation Global Prediction System)
  - HFIP (Hurricane Forecast Improvement Program)
  - "Weeks 3-4"
  - Air Quality
  - COASTAL Act
- 
- We are moving towards "unification", reducing the number of model component versions, for example using the FV3 dynamical core for all atmospheric applications, including AQ
  - We are collaborating closely with the Research office (OAR) and community, to bring "Research to Operations" and vice versa.
  - The Unified Forecast System (UFS) is becoming a central effort, started from our office, now also supported by OAR/OWAQ, and we are planning to extend the collaboration to other NOAA Offices and beyond: <http://ufsccommunity.org>
  - The UFS has 8 "applications", including air quality, and an atmospheric chemistry working group
  - Most of our air quality development work will now be supported under the UFS project, and we are in the project and proposal development stage now.



## National Air Quality Forecast Capability



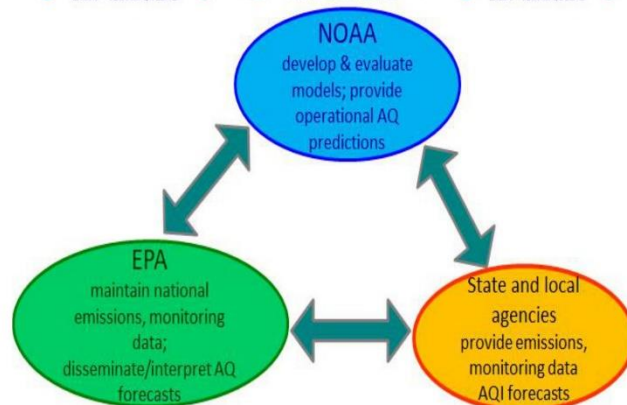
We improve the basis of air quality alerts and provide air quality information to people at risk to further NWS mission of protecting life and property and the enhancement of the national economy.

National Air Quality Forecast Capability (NAQFC) develops and implements operational air quality forecast guidance for the United States.

Current operational Prediction Capabilities are:

- Ozone nationwide
- Smoke nationwide
- Dust over CONUS
- Fine particulate matter (PM2.5) nationwide

These capabilities rely on a strategic partnership with the Environmental Protection Agency (EPA) and state and local air quality forecasters.



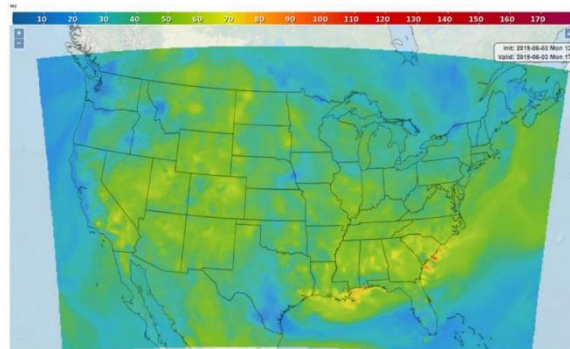
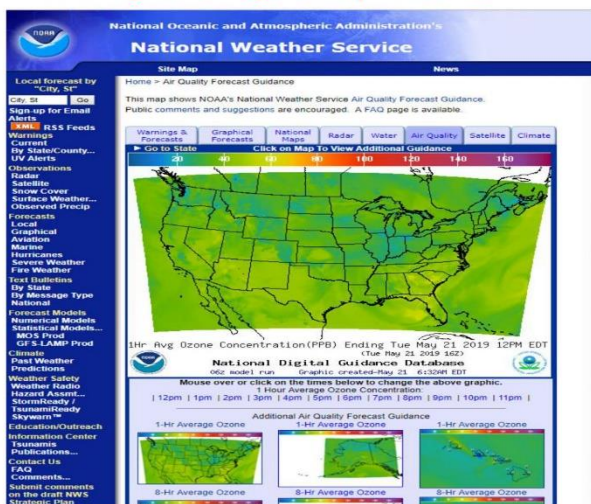


## Our products are available @



Realtime images @ [airquality.weather.gov](https://airquality.weather.gov) and as GRIB files from <ftp://tgftp.nws.noaa.gov/SL.us008001/ST.opnl/D.F.gr2/DC.ndgd/GT.aq/AR.conus/>

In GIS format @ [https://idpgis.ncep.noaa.gov/arcgis/rest/services/NWS\\_Forecasts\\_Guidance\\_Warnings](https://idpgis.ncep.noaa.gov/arcgis/rest/services/NWS_Forecasts_Guidance_Warnings)



Historical database available from the National Digital Guidance Database@ <https://www.ncdc.noaa.gov/data-access/model-data/model-datasets/national-digital-guidance-database-ndgd>

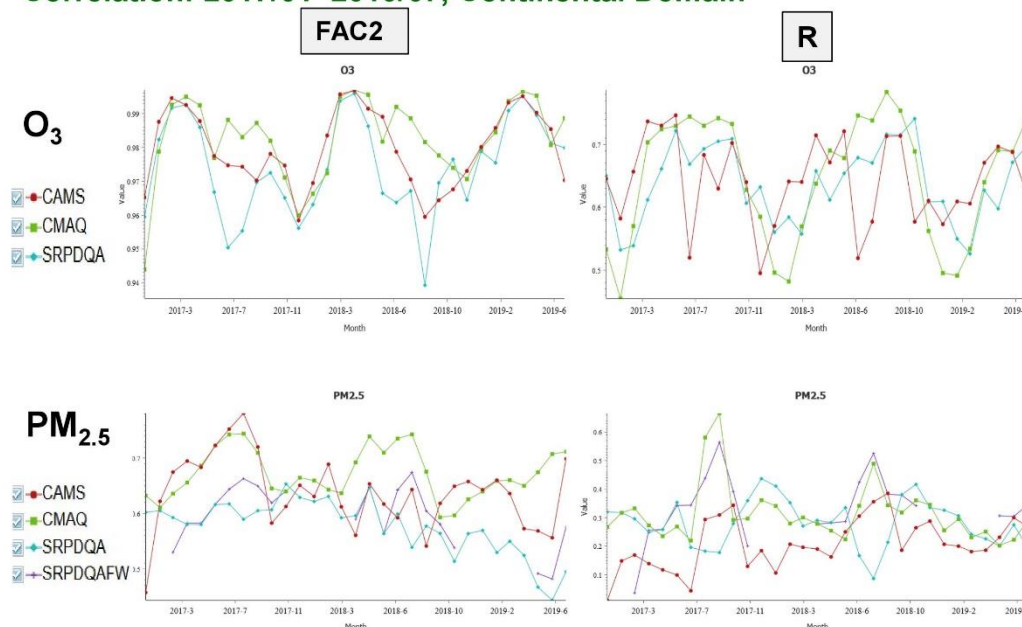


## Operational centers Forecast performance analysis



- NAQFC/CMAQ has the best performance based on several metrics against other operational centers models, ECCO (Regional AQ Deterministic Prediction system) Canada and ECMWF (CAMS-IFS)
- NAQFC/CMAQ has best overall Factor of 2 fraction (FAC2) scores for O3 and PM2.5.
- For ozone in the summertime, NAQFC/CMAQ has the best correlations.
- NAQFC/CMAQ has best (closest to 0) Mean Fractional Bias for daily max O3 and PM2.5
- NAQFC/CMAQ has best overall performance (AQPI) for summertime daily max O3
- NAQFC/CMAQ has best overall performance (AQPI) for daily max PM2.5
- All comparison statistics based on CMAQ raw output

## Time Series of O<sub>3</sub> and PM<sub>2.5</sub> Mean Monthly Values for Factor-of-2 and Correlation: 2017/01–2019/07, Continental Domain



Moran, et al

Statistics are calculated using **daily MAX** observed and forecasted concentrations

## Operational updates implemented last December

### Updated fine particulate matter (PM<sub>2.5</sub>) bias correction system to use:

- Consistent model predictions for training of the unified KFAN correction system
- Increased number of observation sites for model bias correction to over 900 monitors
- Improvements to forecast extreme events by adding the difference between the current raw model forecast and historical analogs' mean to the KFAN bias-corrected predictions

### New ozone bias correction with the same unified codes and configuration

- Uses ozone, wind direction, wind speed, temperature, solar radiation, NO<sub>x</sub>, NO<sub>y</sub> and PBL height as parameters to identify analogs
- airquality.weather.gov display bias corrected ozone instead of raw ozone

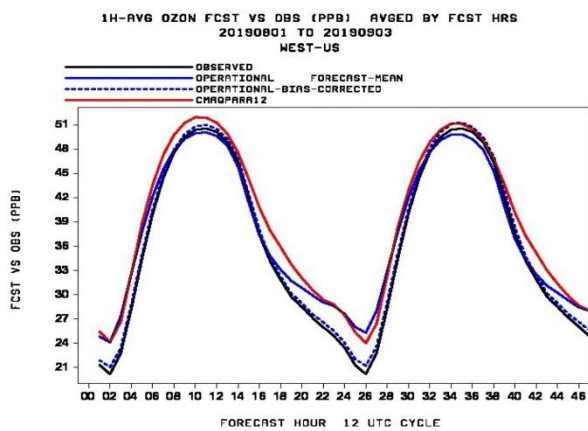
### Updated anthropogenic emissions (NEI2014v2)

### Update Alaska and Hawaii domain CMAQ code to the same version used for CONUS:

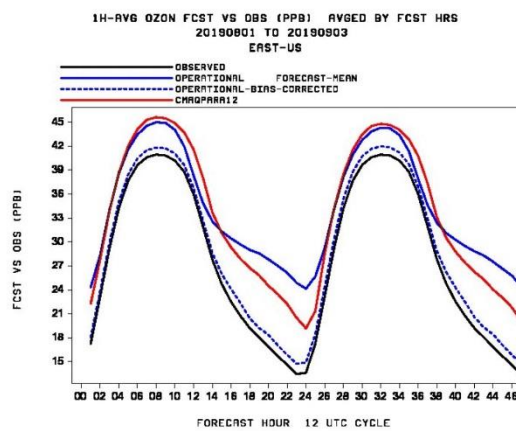
- CB05 gas-phase and aero6 aerosol chemistry (155 species)
- Improved heterogeneous, aqueous, winter-time reactions
- Improved SOA and coarse mode PM



### Performance of Ozone predictions: Observed Vs predicted 1 hr averaged Diurnal variability, August 2019



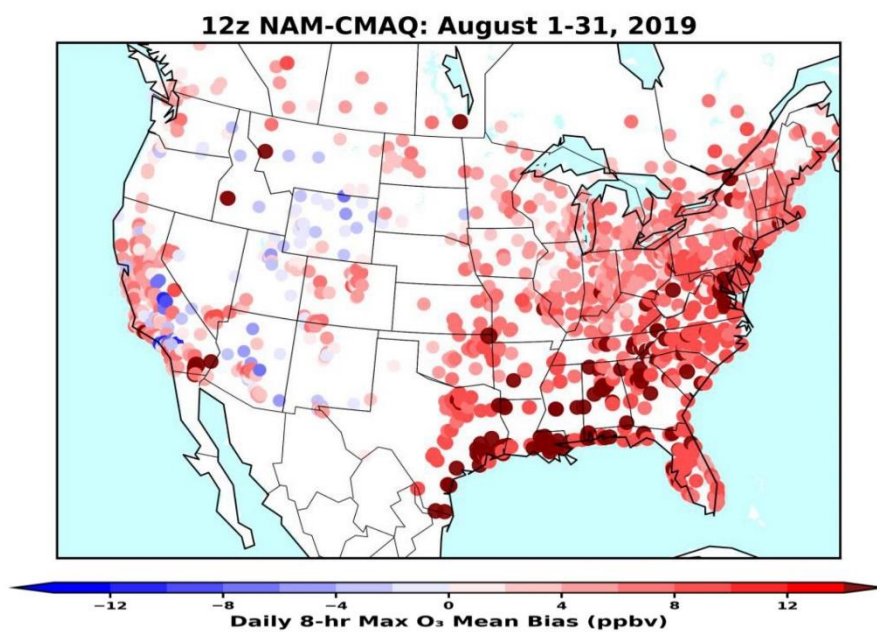
Western U.S.



Eastern U.S.

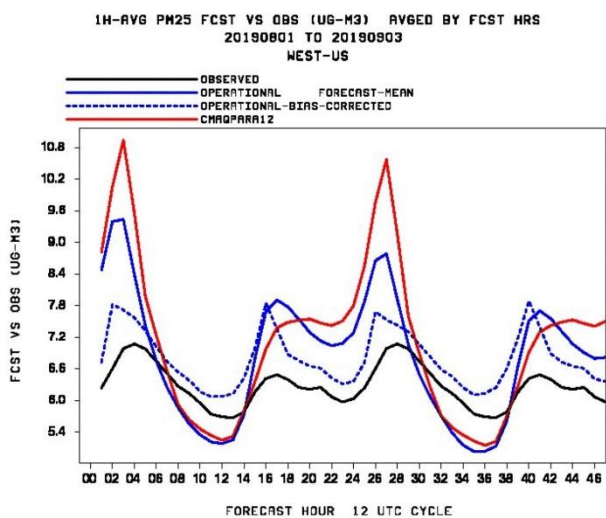


### Performance of Ozone predictions: Daily 8 hr max mean bias, August 2019

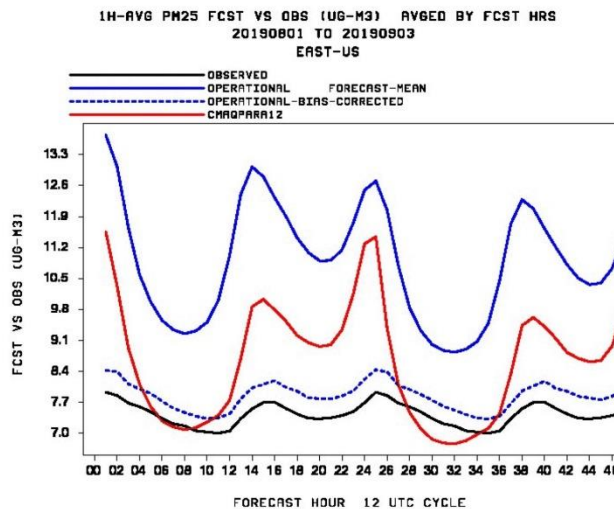




### Performance of PM predictions: Observed Vs predicted 1 hr averaged Diurnal variability, August 2019



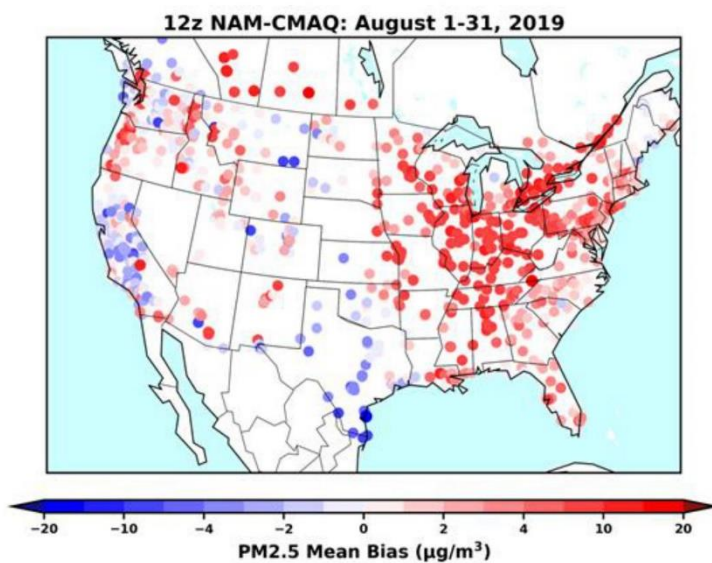
Western U.S.



Eastern U.S.



### Performance of PM predictions: 1 hr daily max PM2.5 August 2019 Monthly Average Bias





## NAQFC Future updates and work in progress



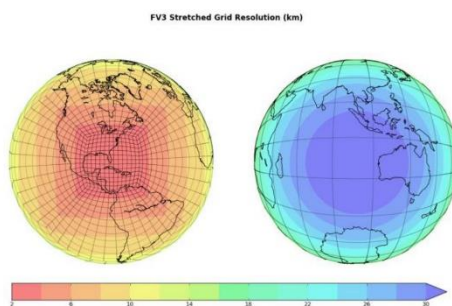
- Couple and drive CMAQ with FV3GFS, NOAA's Next Generation Global Prediction System.
- Extend the range of CMAQ predictions from 48 hours to 72 hours
  - Including the KFAN/Bias corrected products
- Updates to fire emission scheme (Global Biomass Burning Emissions Product eXtended (GBBEPx))
- Use of GEFS-Aerosols for lateral boundary conditions
- Develop new probabilistic forecast product for ozone and PM2.5
- Working on initiating display of operational PM 2.5 products on NWS websites and ozone bias correction on GIS web services



## Transitioning to FV3GFS-CMAQ



- The new dynamic core, Finite-Volume on a Cubed-Sphere (FV3)
- NOAA next generation global prediction system
- Allows for higher resolution and extension of weather forecast through 14 days
- Implemented last June

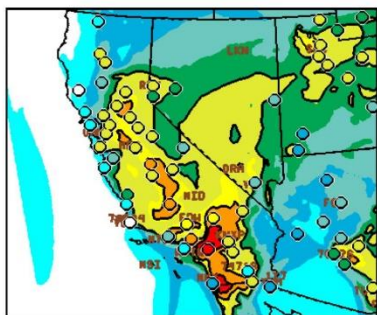
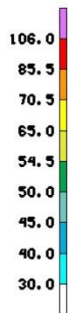


noaa.gov

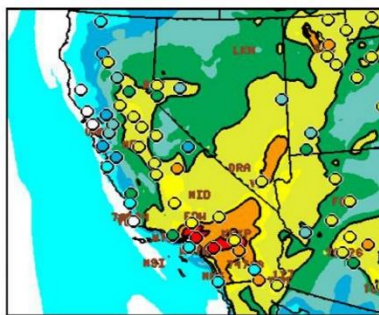




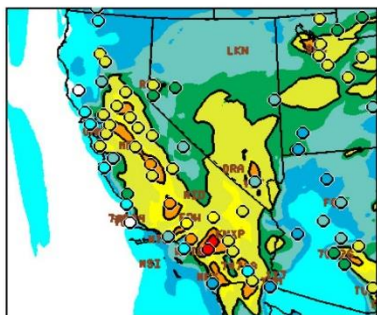
## CMAQ Ozone driven by FV3GFS 13 km



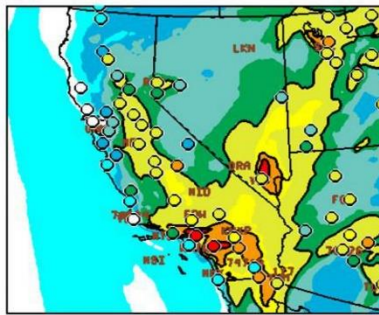
FV3GFS V502 PARR12 DAY1 OZHX08 (PPB) 20190813 12Z CYC



FV3GFS V502 PARR12 DAY2 OZHX08 (PPB) 20190820 12Z CYC



NAM V502 PROD DAY1 OZHX08 (PPB) 20190813 12Z CYC



NAM V502 PROD DAY2 OZHX08 (PPB) 20190820 12Z CYC

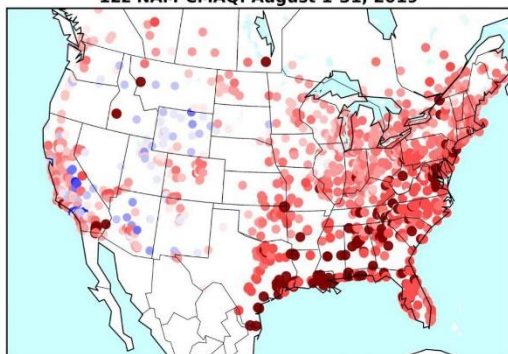
- FV3 and operational predictions are similar



## 8 hr daily max Ozone (day 1) August 2019 Monthly Average Bias



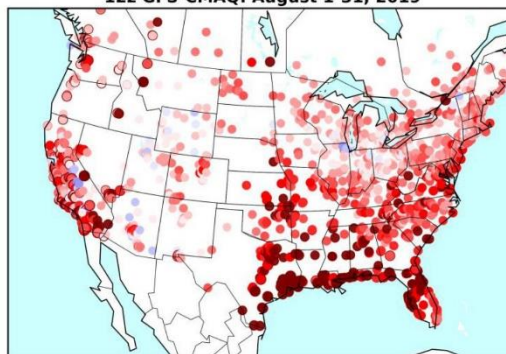
12z NAM-CMAQ: August 1-31, 2019



Daily 8-hr Max O<sub>3</sub> Mean Bias (ppbv)

NAM driven

12z GFS-CMAQ: August 1-31, 2019



Daily 8-hr Max O<sub>3</sub> Mean Bias (ppbv)

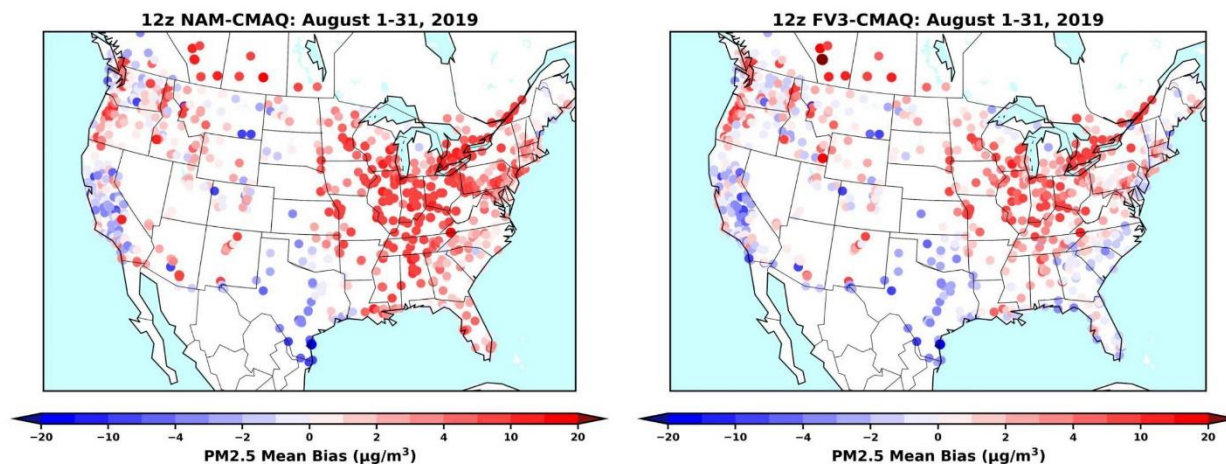
FV3GFS driven

Both models over predict Ozone, for FV3-CMAQ in SE and Mid-Atlantic slightly larger over prediction

Jianping Huang, EMC  
Benjamin Yang, PSU



## 1 hr daily max PM2.5 August 2019 Monthly Average Bias



Both models over predict PM over north  
FV3-CMAQ: under prediction over south

Jianping Huang, EMC  
Benjamin Yang, PSU



## Global Biomass Burning Emissions Product eXtended (GBBEPx)

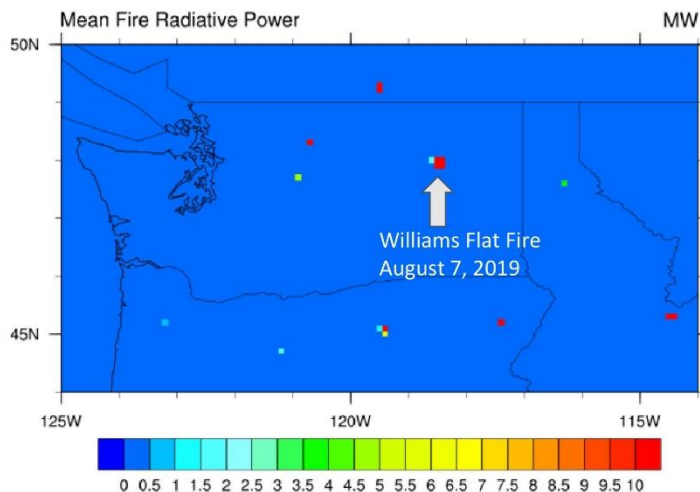


- Currently also testing a new fire emissions scheme with FV3GFS-CMAQ
- GBBEPx calculates biomass burning emissions from wildfires using the Fire Radiative Power (FRP) derived from satellites.
- Uses observations from MODIS, VIIRS and Geostationary satellites like GOES
- Developed by NOAA, NESDIS and scientist from NASA and South Dakota State University
- Currently testing 2 configurations in near real time and a retrospective run from last year's Camp fire event in California
- We also participated in this year multiagency Fire Influence on Regional to Global Environments and Air Quality (FIREX-AQ) field campaign and will be using the data to evaluate our system

## Case study GBBEPx Mean Fire Radiative Power

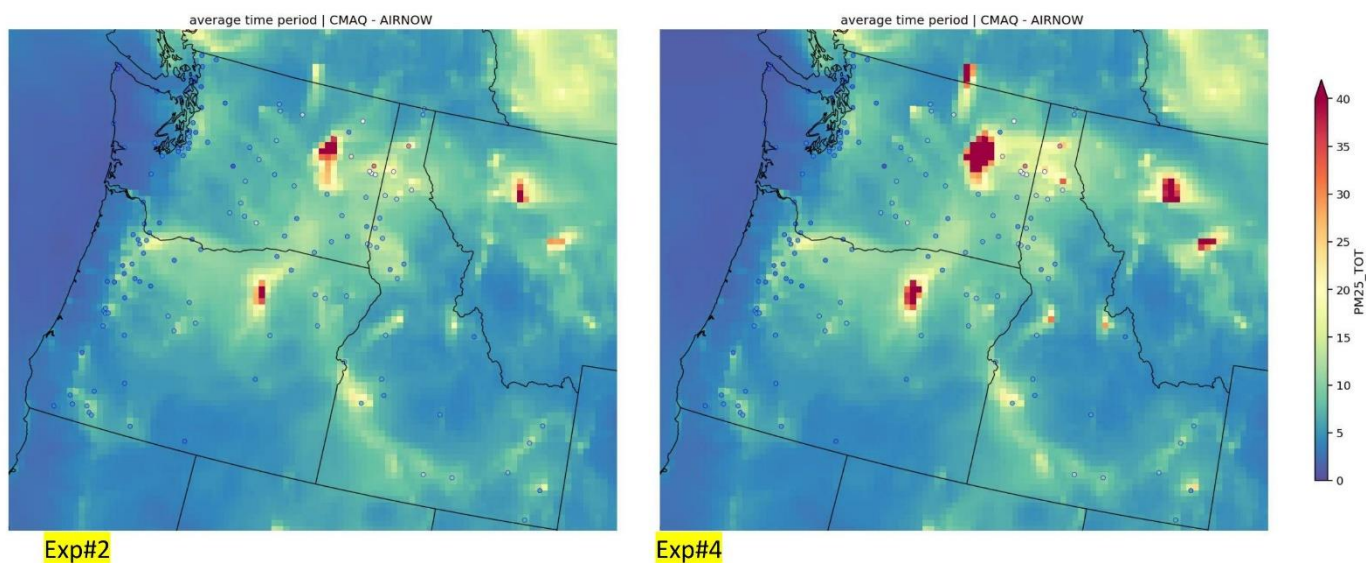


Image credit: NASA



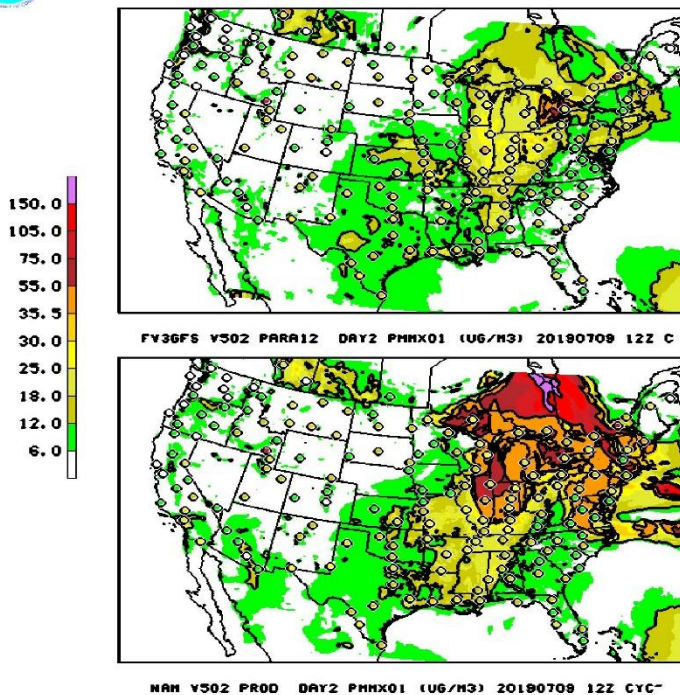
Suomi NPP satellite natural-color image using the VIIRS (Visible Infrared Imaging Radiometer Suite)

## CMAQ-AIRNOW PM<sub>2.5</sub> Evaluation - Region 10





## NAM vs FV3 fire cases



- Para12 GBx run performs best especially during fire events
- FV3 runs under predict over SE away from fires



## Probabilistic forecast

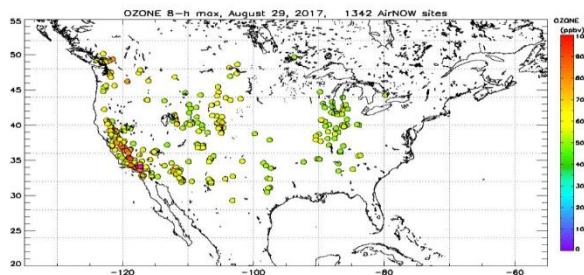
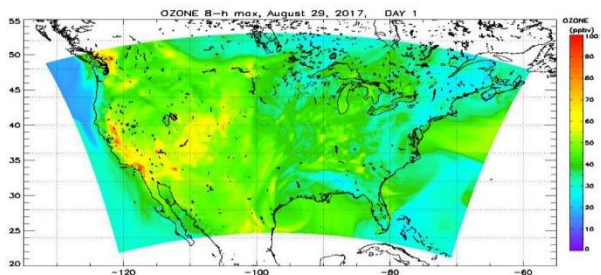
- In collaboration with partners from NOAA Earth System Research Laboratory probabilistic forecast for Ozone and PM2.5 is being developed
- The probabilities are calculated from a climatology first guess of previous CMAQ predictions and computed analogs from AirNOW observation sites around the nation.
- The following are examples for Ozone > 50 ppbv and Ozone > 70 ppbv respectively



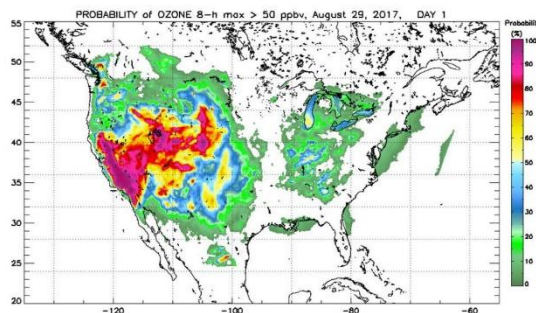
CMAQ guidance

Ozone > 50 ppbv

AirNow stations



Probability



James Wilczak & Irina Djalalova, ESRL

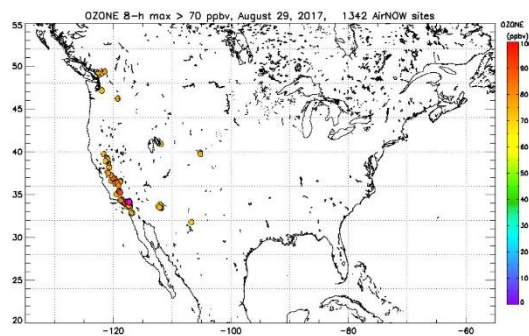
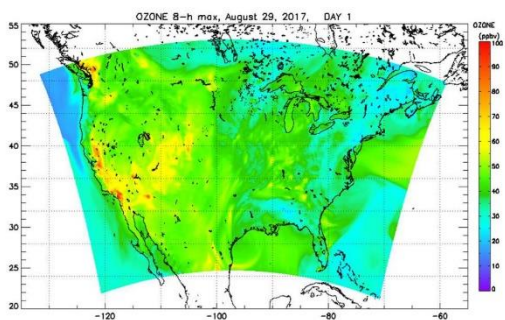
Experimental product



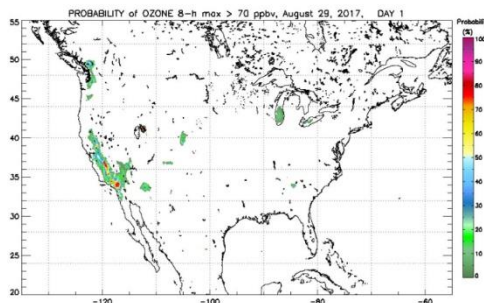
CMAQ guidance

Ozone > 70 ppbv

AirNow stations

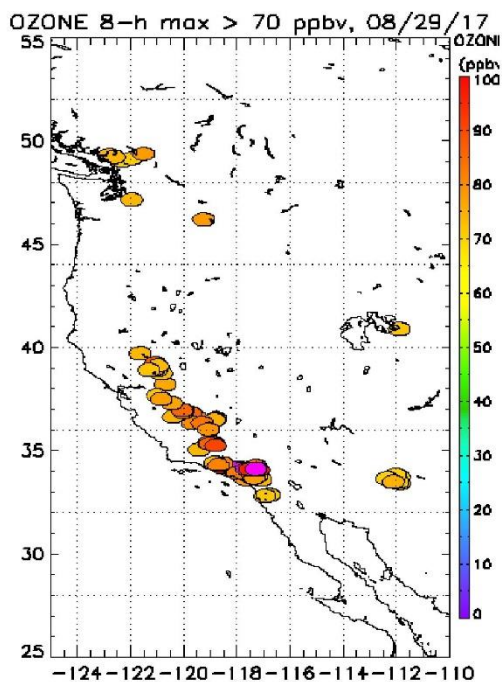


probability

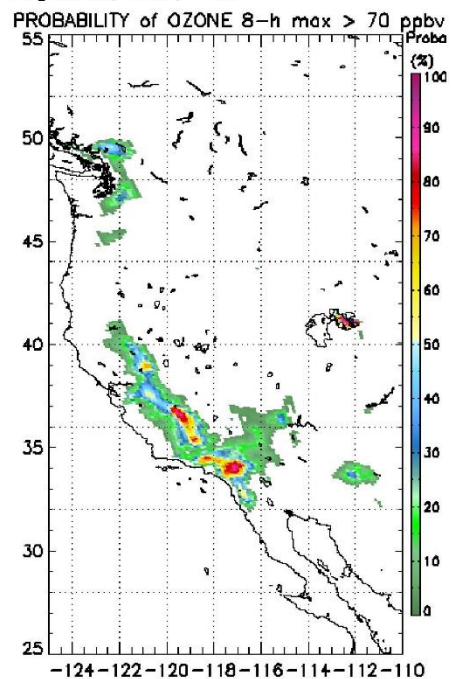


James Wilczak & Irina Djalalova, ESRL

Experimental product



August 29, 2017, DAY 1



## Summary



- We provide real time hourly predictions of ozone, particulate matter, smoke and dust to the nation
- Our work relies on a strategic partnership with the Environmental Protection Agency (EPA) and state and local air quality forecasters.
- Our products are freely available online (airquality.weather.gov, others)

### December 18 implementation:

- New bias-corrected ozone predictions
- Updated fine particulate matter (PM<sub>2.5</sub>) bias correction
- Updated anthropogenic emissions from NEI2014v2
- Updated Alaska and Hawaii domain CMAQ code to the same version used for CONUS

### Work in progress:

- Future improvements will include coupling of CMAQ with FV3GFS, extension to 72 hour predictions and new fire emissions processing



Thank you for your attention

Questions?

Contact me @ [jose.tirado-delgado@noaa.gov](mailto:jose.tirado-delgado@noaa.gov)



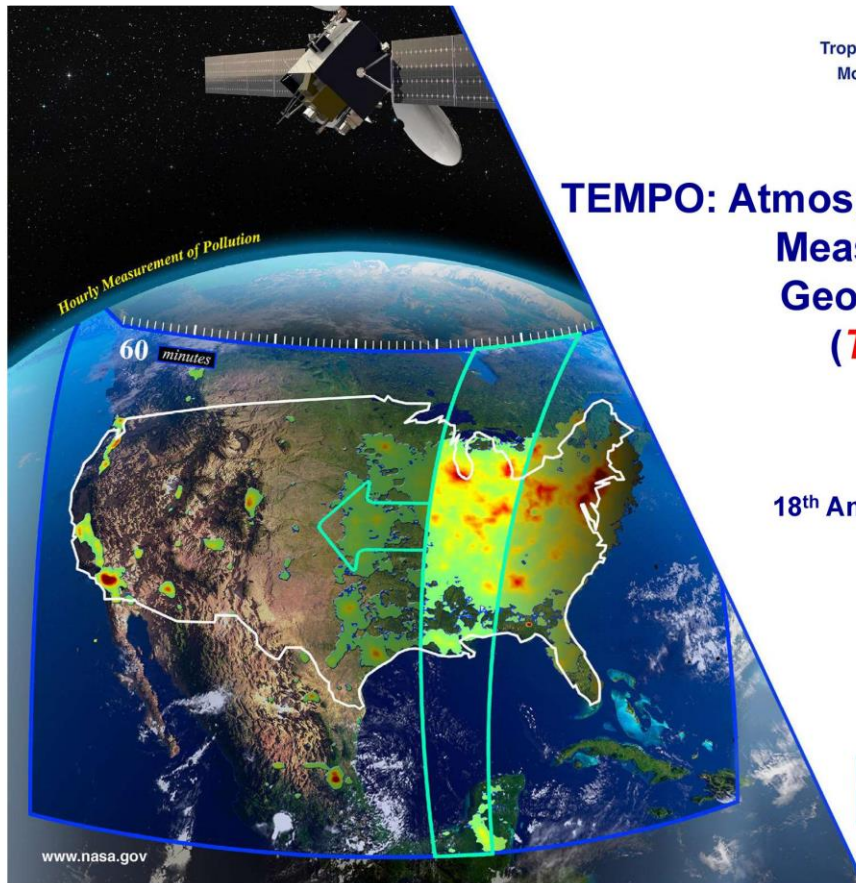
## Acknowledgments: *AQF implementation team members*

*Special thanks to new and previous NOAA and EPA team members who contributed to the system development*

<b><u>NOAA/NWS/STI</u></b>	<i>Dorothy Koch</i>	<i>STI modeling program lead</i>
<b><u>NWS/OD</u></b>	<i>Cynthia Jones</i>	<i>Data Communications</i>
<b><u>NWS/OSTI/MDL</u></b>	<i>David Miller, Dave Ruth</i>	<i>Dev. Verification, NDGD Product Development</i>
<b><u>NWS/STI</u></b>	<i>Jose Tirado-Delgado</i>	<i>Program Support</i>
<b><u>NESDIS/NCDC</u></b>	<i>Alan Hall</i>	<i>Product Archiving</i>
<b><u>NWS/NCEP</u></b>	<i>Ivanka Stajner, Jeff McQueen, Jianping Huang, Ho-Chun Huang</i>	<i>AQF model interface development, testing, &amp; integration</i>
	<i>Jun Wang, Li Pan, *Sarah Lu</i>	<i>Global dust aerosol and feedback testing</i>
	<i>*Brad Ferrier, *Eric Rogers,</i>	<i>NAM coordination</i>
	<i>*Hui-Ya Chuang, Perry Shafran, Boi Young</i>	
	<i>Geoff Manikin</i>	<i>Smoke and dust product testing and integration</i>
	<i>Rebecca Cosgrove, Steven Earle, Chris Magee</i>	<i>NCO transition and systems testing</i>
	<i>Mike Bodner, Andrew Orrison</i>	<i>HPC coordination and AQF webdrawer</i>
<b><u>ESRL/PSD</u></b>	<i>Jim Wilczak, Irina, Djalalova, Dave Allerud,</i>	<i>bias correction development</i>
<b><u>NOAA/OAR/ARL</u></b>	<i>Pius Lee, Daniel Tong, Youhua Tang</i>	<i>CMAQ development, adaptation of AQ simulations for AQF</i>
	<i>Barry Baker, Patrick Campbell</i>	
	<i>Ariel Stein</i>	<i>HYSPLIT adaptations</i>
<b><u>NESDIS/STAR</u></b>	<i>Shobha Kondragunta</i>	<i>Smoke and dust verification product development</i>
<b><u>NESDIS/QSDPD</u></b>	<i>Liqun Ma</i>	<i>Production of smoke and dust verification products</i>
<b><u>EPA/OAQPS partners:</u></b>		
	<i>Chet Wayland, Phil Dickerson, Brad Johns, John White</i>	<i>AIRNow development, coordination with NAQFC</i>

\* Guest Contributors

附錄四、"New Insights into the Health Effects of Air Pollution and Persistent Knowledge Gaps: Why is CMAS more important now than ever?"



Tropospheric Emissions: Monitoring of Pollution

**TEMPO: Atmospheric Pollution Measurements from Geostationary Orbit**  
**(TEMPO.SI.EDU!)**

Kelly Chance

18<sup>th</sup> Annual CMAS Conference  
UNC Chapel Hill  
October 21, 2019

Ball  
Smithsonian  
NASA

www.nasa.gov



**Hourly atmospheric pollution from geostationary Earth orbit**

Smithsonian  
NASA

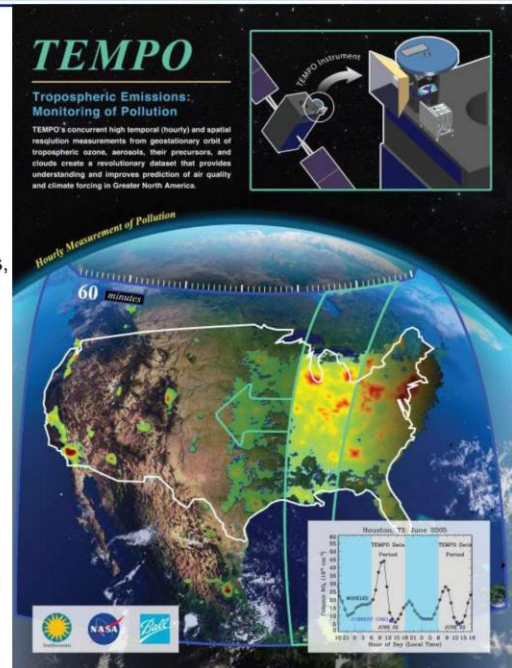
PI: Kelly Chance, Smithsonian Astrophysical Observatory  
 Deputy PI: Xiong Liu, Smithsonian Astrophysical Observatory  
 Instrument Development: Ball Aerospace  
 Project Management: NASA LaRC  
 Other Institutions: NASA GSFC, NOAA, EPA, NCAR, Harvard, UC Berkeley, St. Louis U, U Alabama Huntsville, U Nebraska, RT Solutions, Carr Astronautics  
 International collaboration: Mexico, Canada, Cuba, Korea, U.K., ESA, Spain

**Selected Nov. 2012 as NASA's first Earth Venture Instrument**

- Instrument delivery 2018
- NASA has arranged hosting on a commercial geostationary communications satellite with launch expected 2/2022

**Provides hourly daylight observations to capture rapidly varying emissions & chemistry important for air quality**

- Distinguishes boundary layer from free tropospheric & stratospheric ozone



**TEMPO**  
Tropospheric Emissions: Monitoring of Pollution

TEMPO's concurrent high temporal (hourly) and spatial resolution measurements from geostationary orbit of tropospheric ozone, aerosols, their precursors, and clouds create a revolutionary dataset that provides understanding and improves prediction of air quality and climate forcing in Greater North America.

Hourly Measurement of Pollution

60 minutes

Hourly O<sub>3</sub> (DU) - Zone 0005

TEMPO Data - Period  
TEMPO Data - Period

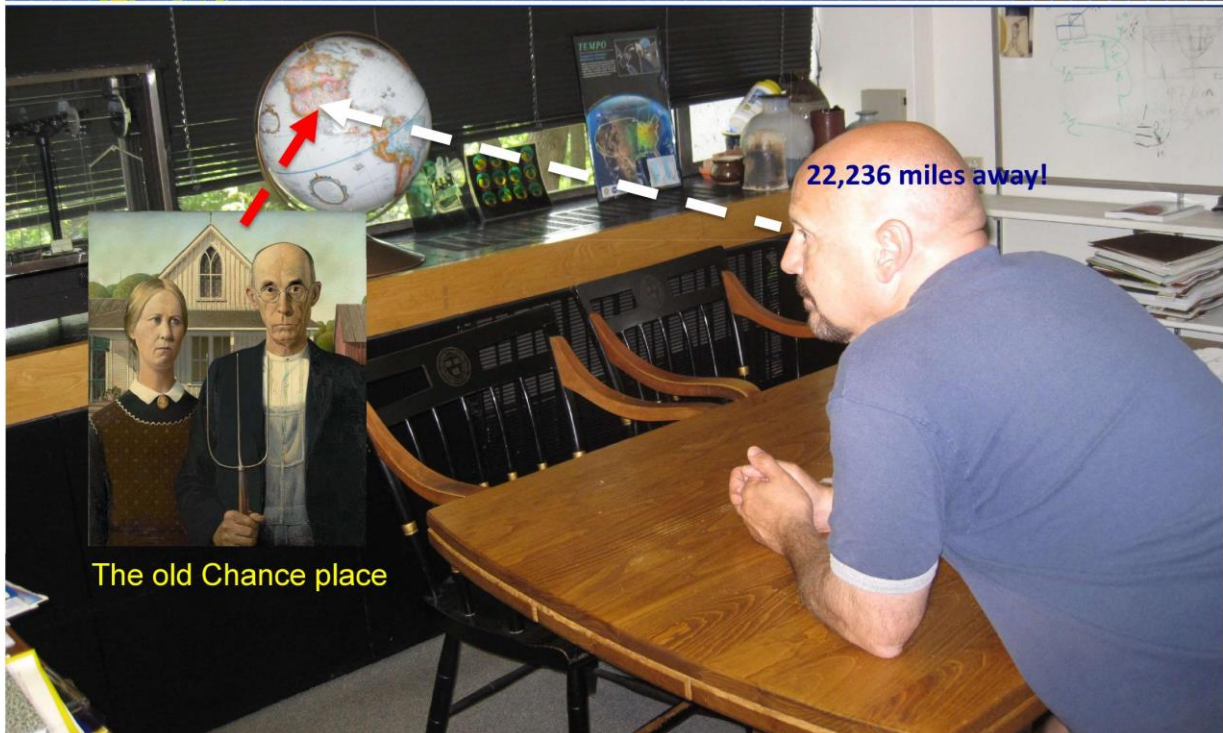
Smithsonian  
NASA  
Ball

North American component of an international constellation for air quality observations



**TEMPO** Smithsonian NASA

# The view from GEO



10/21/19

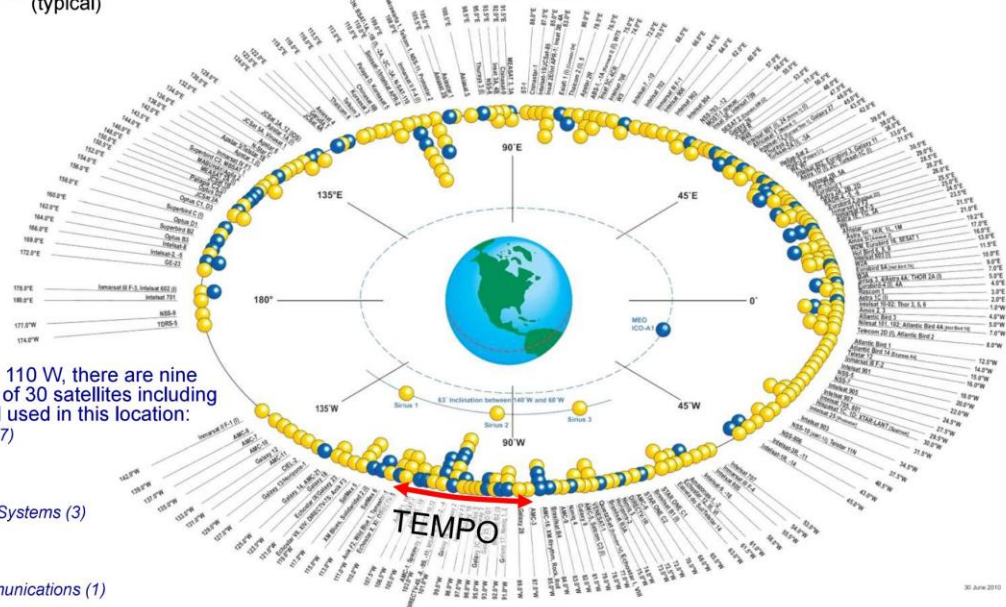
3

**TEMPO** Smithsonian NASA

# Geostationary orbit opportunities of interest

**BOEING**  
(typical)

**Commercial Communications Satellites  
Geosynchronous Orbit**



30 June 2010

10/21/19

4

**TEMPO will be located at 91° West**



- Instrument completed, accepted, delivered, now in storage
- Commercial geostationary satellite host selected for launch in February 2022 to 91°W

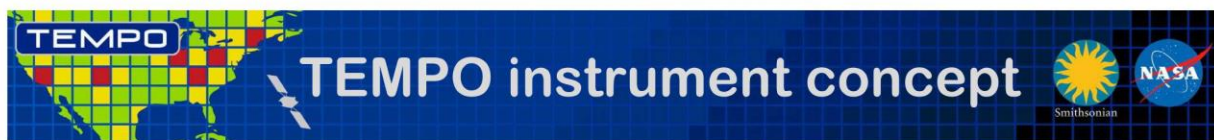
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6



- **Measurement technique**
  - Imaging grating spectrometer measuring solar backscattered Earth radiance
  - Spectral band & resolution: 290-490 + 540-740 nm @ 0.6 nm FWHM, 0.2 nm sampling
  - 2 2-D, 2k × 1k, detectors image the full spectral range for each geospatial scene
- **Field of Regard (FOR) and duty cycle**
  - Mexico City/Yucatan, Cuba to the Canadian oil sands, Atlantic to Pacific
  - Instrument slit aligned N/S and swept across the FOR in the E/W direction, producing a radiance map of Greater North America in one hour
- **Spatial resolution**
  - 2.1 km N/S × 4.7 km E/W native pixel resolution (9.8 km<sup>2</sup>)
  - Co-add/cloud clear as needed for specific data products
- **Standard data products and sampling rates**
  - Most sampled hourly, including eXcel O<sub>3</sub> (troposphere, PBL)
  - NO<sub>2</sub>, H<sub>2</sub>CO, C<sub>2</sub>H<sub>2</sub>O<sub>2</sub>, SO<sub>2</sub> sampled hourly (average results for ≥ 3/day if needed)
  - Measurement requirements met up to 50° for SO<sub>2</sub>, 70° SZA for other products

10/21/19

7




1. What are the temporal and spatial variations of **emissions of gases and aerosols important for air quality** and climate?
2. What are the physical, chemical, and dynamical **processes that transform tropospheric composition and air quality** over scales ranging from urban to continental, diurnally to seasonally?
3. How does air pollution drive **climate forcing** and how does climate change affect **air quality** on a continental scale?
4. How can observations from space **improve air quality forecasts and assessments** for societal benefit?
5. How does **intercontinental transport** affect air quality?
6. How do **episodic events**, such as wild fires, dust outbreaks, and volcanic eruptions, affect atmospheric composition and air quality?

10/21/19

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附錄四、"New Insights into the Health Effects of Air Pollution and Persistent Knowledge Gaps: Why is CMAS more important now than ever?"



Team Member	Institution	Role	Responsibility
K. Chance	SAO	PI	Overall science development; <b>Level 1b, H<sub>2</sub>CO, C<sub>2</sub>H<sub>2</sub>O<sub>2</sub></b>
X. Liu	SAO	Deputy PI	Science development, data processing; <b>O<sub>3</sub> profile, tropospheric O<sub>3</sub></b>
J. Al-Saadi	LaRC	Deputy PS	Project science development
J. Carr	Carr Astronautics	Co-I	<b>INR Modeling and algorithm</b>
M. Chin	GSFC	Co-I	Aerosol science
R. Cohen	U.C. Berkeley	Co-I	NO <sub>2</sub> validation, atmospheric chemistry modeling, process studies
D. Edwards	NCAR	Co-I	VOC science, synergy with carbon monoxide measurements
J. Fishman	St. Louis U.	Co-I	AQ impact on agriculture and the biosphere
D. Flittner	LaRC	Project Scientist	Overall project development; STM; instrument cal./char.
J. Herman	UMBC	Co-I	Validation (PANDORA measurements)
D. Jacob	Harvard	Co-I	Science requirements, atmospheric modeling, process studies
S. Janz	GSFC	Co-I	Instrument calibration and characterization
J. Joiner	GSFC	Co-I	<b>Cloud, total O<sub>3</sub>, TOA shortwave flux research product</b>
N. Krotkov	GSFC	Co-I	<b>NO<sub>2</sub>, SO<sub>2</sub>, UVB</b>
M. Newchurch	U. Alabama Huntsville	Co-I	Validation (O <sub>3</sub> sondes, O <sub>3</sub> lidar)
R.B. Pierce	NOAA/NESDIS	Co-I	AQ modeling, data assimilation
R. Spurr	RT Solutions, Inc.	Co-I	<b>Radiative transfer modeling for algorithm development</b>
R. Suleiman	SAO	Co-I, Data Mgr.	Managing science data processing, <b>BrO, H<sub>2</sub>O, and L3 products</b>
J. Szykman	EPA	Co-I	AIRNow AQI development, validation (PANDORA measurements)
O. Torres	GSFC	Co-I	<b>UV aerosol product, AI</b>
J. Wang	U. Iowa	Co-I	Synergy w/GOES-R ABI, <b>aerosol research products</b>
J. Leitch	Ball Aerospace	Collaborator	Aircraft validation, instrument calibration and characterization
D. 10/21/19	LaRC	Collaborator	GEO-CAPE mission design team member



Team Member	Institution	Role	Responsibility
Randall Martin	Dalhousie U.	Collaborator	Atmospheric modeling, air mass factors, AQI development
Chris McLinden	Environment Canada	Collaborator	Canadian air quality coordination
Michel Grutter de la Mora	UNAM, Mexico	Collaborator	Mexican air quality coordination
Gabriel Vazquez	UNAM, Mexico	Collaborator	Mexican air quality, algorithm physics
Amparo Martinez	INECC, Mexico	Collaborator	Mexican environmental pollution and health
J. Victor Hugo Paramo Figueo	INECC, Mexico	Collaborator	Mexican environmental pollution and health
Brian Kerridge	Rutherford Appleton Laboratory, UK	Collaborator	Ozone profiling studies, algorithm development
Paul Palmer	Edinburgh U., UK	Collaborator	Atmospheric modeling, process studies
Alfonso Saiz-Lopez	CSIC, Spain	Collaborator	Atmospheric modeling, process studies
Juan Carlos Antuña Marrero	GOAC, Cuba	Collaborator	Cuban Science team lead, Cuban air quality
Osvaldo Cuesta	GOAC, Cuba	Collaborator	TEMPO validation, Cuban air quality
René Estevan Arredondo	GOAC, Cuba	Collaborator	TEMPO validation, Cuban air quality
J. Kim	Yonsei U.	Collaborators, Science Advisory Panel	Korean GEMS, CEOS constellation of GEO pollution monitoring
C.T. McElroy	York U. Canada		CSA PHEOS, CEOS constellation of GEO pollution monitoring
B. Veihelmann	ESA		ESA Sentinel-4, CEOS constellation of GEO pollution monitoring
J.P. Veefkind	KNMI		ESA Sentinel-5P (TROPOMI)



# Air quality requirements from the GEO-CAPE Science Traceability Matrix

11-28-2011 DRAFT GEO-CAPE aerosol-atmospheres Science Traceability Matrix BASELINE and THRESHOLD

Science Questions	Measurement Objectives (color flag maps to Science Questions)	Measurement Requirements (mapped to Measurement Objectives)	Measurement Rationale																																																																					
1. What are the temporal and spatial variations of emissions of gases and aerosols important for air quality and climate?	<b>Baseline measurements:</b> O3, NO2, CO, SO2, HCHO, CH4, NH3, CHOCHO, different temporal sampling frequencies, 4 km x 4 km product horizontal spatial resolution at the center of the domain; and AOD, AAOD, AI, aerosol optical depth (AODCH), hourly for SZA<70 and 8 km x 8 km product horizontal spatial resolution at the center of the domain. <b>Threshold measurements:</b> CO hourly day and night; O3, NO2 hourly with SZA<70; AOD hourly (SZA<45); AI 8 km x 8 km product horizontal spatial resolution at the center of the domain.	<b>Geostationary Observing Location:</b> 100 W +/- 10 <b>Collocate measurements:</b> <b>A to K</b> All the baseline and threshold species <b>Cloud Camera:</b> 1 km x 1 km horizontal spatial resolution, two spectral bands, baseline only <b>Vertical information:</b> <b>A to K</b> Two pieces of information in the troposphere in daylight with sensitivity to the lowest 2 km Almade (+/- 1km) AODCH (total line only)	Provides optimal view of North America Continue the current state of practice in vertical, add temporal resolution Improve retrieval accuracy, provide diagnostics for gases and aerosol Separate the lower-most troposphere from the free troposphere for O3, CO, AODCH Detect aerosol plume height. Improve retrieval accuracy.																																																																					
2. How do physical, chemical, and dynamical processes determine tropospheric composition and air quality over scales ranging from urban to continental, diurnally to seasonally?	4. Measure the threshold or baseline species or properties with the temporal and spatial resolution specified (see next column) to quantify the underlying emissions, understand emission processes, and track transport and chemical evolution of air pollutants. <b>A, B, C, D, E, F, G, H, I, J, K</b> 5. Measure AOD, AAOD, and NH3 to quantify aerosol and nitrogen deposition to land and coastal regions. <b>A, B, C, D, E, F, G, H, I, J, K</b> 6. Measure AOD, AAOD, and AODCH to relate surface PM concentration, UV-B level and visibility to aerosol column loading. <b>A, B, C, D, E, F, G, H, I, J, K</b>	<b>Product horizontal spatial resolution at the center of the domain, (nominal 1020 x 35 N):</b> <b>A to K</b> 4 km x 4 km (baseline), 8 km x 8 km (threshold), 8 km x 8 km (baseline, threshold), 16 km x 16 km (baseline only) <b>Spectral region:</b> <b>A to K</b> UV-Vis or UV-TIR SWIR, MWIR CD SO2 TIR NH3 Vis UV-deep blue UV-deep blue Vis-NIR	Separate spatial/temporal variability, obtain better yields of products Capture spatial variability, sufficient to link to EC observations Inherently larger spatial scales, sufficient to link to EC observations Typical use Provide multi-spectral retrieval information in daylight Retrieve gas species from their atmospheric spectral signatures (typical)																																																																					
3. How does air pollution drive climate forcing and how does climate change affect air quality on a continental scale?	7. Determine the instantaneous radiative forcings associated with aerosols and trace gases on the continental scale and relate them quantitatively to natural and anthropogenic emissions. <b>A, B, C, D, E, F, G, H, I, J, K</b> 8. Observe plumes of CH4 emission from biogenic and anthropogenic releases, CO anthropogenic and wildfire emissions, AOD, AAOD, and AI from fires, AOD, AAOD, and AI from dust storms, SO2 and AOD from volcanic eruptions. <b>A, B, C, D, E, F, G, H, I, J, K</b>	UV-Vis or UV-TIR SWIR, MWIR CD SO2 TIR NH3 Vis UV-deep blue UV-deep blue Vis-NIR	Obtain spectral dependence of AOD for particle size and type information Obtain spectral dependence of AAOD for aerosol type information Retrieve absorbing aerosol information Detect aerosol height																																																																					
4. How can observations from space improve air quality forecasts and assessments for societal benefit?	9. Quantify the inflows and outflows of O3, CO, SO2, and aerosols across continental boundaries to determine their impacts on surface air quality and climate. <b>A, B, C, D, E, F, G, H, I, J, K</b> 10. Characterize aerosol particle size and type from spectral dependence measurements of AOD and AAOD. <b>A, B, C, D, E, F, G, H, I, J, K</b> 11. Acquire measurements to improve representation of processes in air quality models and improve data assimilation in forecast and reanalysis modes. <b>A, B, C, D, E, F, G, H, I, J, K</b> 12. Synthesize the GEO-CAPE measurements with information from in-situ and ground-based remote sensing networks to construct an enhanced observing system. <b>A, B, C, D, E, F, G, H, I, J, K</b> 13. Leverage GEO-CAPE observations into an integrated observing system including geostationary satellites over Europe and Asia together with LEO satellites and suborbital platforms for accessing the hemispheric transport. <b>A, B, C, D, E, F, G, H, I, J, K</b> 14. Integrate observations from GEO-CAPE and other platforms into models to improve representation of processes in the models and to link the observed composition, deposition, and radiative forcing to the emissions from anthropogenic and natural sources. <b>A, B, C, D, E, F, G, H, I, J, K</b>	<b>Atmospheric measurements over Land/Coastal areas, baseline and threshold:</b> <b>A to K</b> <table border="1"> <thead> <tr> <th>Species</th> <th>Time resolution</th> <th>Typical value<sup>2</sup></th> <th>Precision<sup>2</sup></th> <th>Description</th> </tr> </thead> <tbody> <tr> <td>O3</td> <td>Hourly, SZA&lt;70</td> <td>9 x 10<sup>18</sup></td> <td>0-2 km: 10 ppbv 2km-tropopause: 15 ppbv Stratosphere: 5%</td> <td>Observe O3 with two pieces of information in the troposphere with sensitivity to the lowest 2 km for surface AQ; also transport, climate forcing</td> </tr> <tr> <td>CO</td> <td>Hourly, day and night</td> <td>2 x 10<sup>18</sup></td> <td>0-2 km: 20ppbv 2km-tropopause: 20 ppbv</td> <td>Track anthropogenic and biomass burning plumes; observe CO with two pieces of information in the vertical with sensitivity to the lowest 2 km in daylight</td> </tr> <tr> <td>AOD</td> <td>Hourly, SZA&lt;70</td> <td>0.1 - 1</td> <td>0.05</td> <td>Observe total aerosol; aerosol sources and transport; climate forcing</td> </tr> <tr> <td>NO2</td> <td>Hourly, SZA&lt;70</td> <td>6 x 10<sup>15</sup></td> <td>1 x 10<sup>15</sup></td> <td>Distinguish background from enhanced/polluted scenes; atmospheric chemistry</td> </tr> </tbody> </table> <b>Additional atmospheric measurements over Land/Coastal areas, baseline only:</b> <b>A to K</b> <table border="1"> <thead> <tr> <th>Species</th> <th>Time resolution</th> <th>Typical value<sup>2</sup></th> <th>Precision<sup>2</sup></th> <th>Description</th> </tr> </thead> <tbody> <tr> <td>HCHO*</td> <td>3/day, SZA&lt;50</td> <td>1.0x10<sup>16</sup></td> <td>1 x 10<sup>16</sup></td> <td>Observe biogenic VOC emissions, expected to peak at midday; chemistry</td> </tr> <tr> <td>SO2*</td> <td>3/day, SZA&lt;50</td> <td>1 x 10<sup>16</sup></td> <td>1 x 10<sup>16</sup></td> <td>Identify major pollution and volcanic emissions; atmospheric chemistry</td> </tr> <tr> <td>CH4</td> <td>2/day</td> <td>4 x 10<sup>19</sup></td> <td>20 ppbv</td> <td>Observe anthropogenic and natural emissions sources</td> </tr> <tr> <td>NH3</td> <td>2/day</td> <td>2x10<sup>16</sup></td> <td>0-2 km: 2ppbv</td> <td>Observe agricultural emissions</td> </tr> <tr> <td>CHOCHO*</td> <td>2/day</td> <td>2x10<sup>14</sup></td> <td>4 x 10<sup>14</sup></td> <td>Detect VOC emissions, aerosol formation, atmospheric chemistry</td> </tr> <tr> <td>AAOD</td> <td>Hourly, SZA&lt;70</td> <td>0 - 0.05</td> <td>0.02</td> <td>Distinguish smoke and dust from non-UV absorbing aerosols; climate forcing</td> </tr> <tr> <td>AI</td> <td>Hourly, SZA&lt;70</td> <td>-1 - +5</td> <td>0.1</td> <td>Detect aerosols near/above clouds and over snow/ice; aerosol events</td> </tr> <tr> <td>AOCH</td> <td>Hourly, SZA&lt;70</td> <td>Variable</td> <td>1 km</td> <td>Determine plume height; large scale transport, conversions from AOD to PM</td> </tr> </tbody> </table>	Species	Time resolution	Typical value <sup>2</sup>	Precision <sup>2</sup>	Description	O3	Hourly, SZA<70	9 x 10 <sup>18</sup>	0-2 km: 10 ppbv 2km-tropopause: 15 ppbv Stratosphere: 5%	Observe O3 with two pieces of information in the troposphere with sensitivity to the lowest 2 km for surface AQ; also transport, climate forcing	CO	Hourly, day and night	2 x 10 <sup>18</sup>	0-2 km: 20ppbv 2km-tropopause: 20 ppbv	Track anthropogenic and biomass burning plumes; observe CO with two pieces of information in the vertical with sensitivity to the lowest 2 km in daylight	AOD	Hourly, SZA<70	0.1 - 1	0.05	Observe total aerosol; aerosol sources and transport; climate forcing	NO2	Hourly, SZA<70	6 x 10 <sup>15</sup>	1 x 10 <sup>15</sup>	Distinguish background from enhanced/polluted scenes; atmospheric chemistry	Species	Time resolution	Typical value <sup>2</sup>	Precision <sup>2</sup>	Description	HCHO*	3/day, SZA<50	1.0x10 <sup>16</sup>	1 x 10 <sup>16</sup>	Observe biogenic VOC emissions, expected to peak at midday; chemistry	SO2*	3/day, SZA<50	1 x 10 <sup>16</sup>	1 x 10 <sup>16</sup>	Identify major pollution and volcanic emissions; atmospheric chemistry	CH4	2/day	4 x 10 <sup>19</sup>	20 ppbv	Observe anthropogenic and natural emissions sources	NH3	2/day	2x10 <sup>16</sup>	0-2 km: 2ppbv	Observe agricultural emissions	CHOCHO*	2/day	2x10 <sup>14</sup>	4 x 10 <sup>14</sup>	Detect VOC emissions, aerosol formation, atmospheric chemistry	AAOD	Hourly, SZA<70	0 - 0.05	0.02	Distinguish smoke and dust from non-UV absorbing aerosols; climate forcing	AI	Hourly, SZA<70	-1 - +5	0.1	Detect aerosols near/above clouds and over snow/ice; aerosol events	AOCH	Hourly, SZA<70	Variable	1 km	Determine plume height; large scale transport, conversions from AOD to PM
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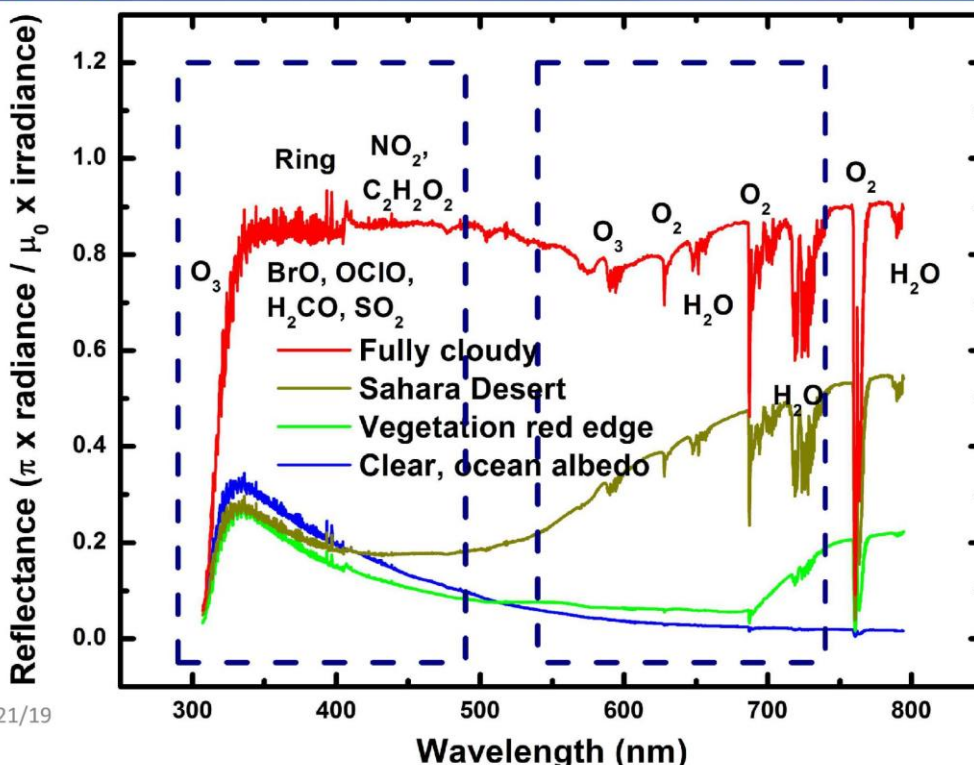
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Ultraviolet/visible species (GOME, SCIA, OMI, OMPS, TEMPO, etc.)

Atmospheric measurements over Land/Coastal areas, baseline and threshold: <b>A to K</b>				
Species	Time resolution	Typical value <sup>2</sup>	Precision <sup>2</sup>	Description
O3	Hourly, SZA<70	9 x 10 <sup>18</sup>	0-2 km: 10 ppbv 2km-tropopause: 15 ppbv Stratosphere: 5%	Observe O3 with two pieces of information in the troposphere with sensitivity to the lowest 2 km for surface AQ; also transport, climate forcing
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**TEMPO** Typical TEMPO-range spectra (from ESA GOME-1)



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**TEMPO** Low Earth orbit: Sun-synchronous nadir heritage

Instrument	Detectors	Spectral Coverage [nm]	Spectral Res. [nm]	Ground Pixel Size [km <sup>2</sup> ]	Global Coverage
GOME-1 (1995-2011)	Linear Arrays	240-790	0.2-0.4	40 × 30 (10 × 30 zoom)	3 days
SCIAMACHY (2002-2012)	Linear Arrays	240-2380	0.2-1.5	30 × 30/60/90 30 × 20/240	6 days
OMI (2004)	2-D CCD	270-810	0.42-0.63	18 × 24 - 42 × 162	daily
GOME-2a,b (2006, 2012)	Linear Arrays	240-790	0.24-0.53	40 × 80 (40 × 10 zoom)	near-daily
OMPS-1 (2011)	2-D CCDs	250-380	0.42-1.0	50 × 50	daily

**More than 10 billion measurements of spectral!**  
**Previous experience (since 1985 at SAO and MPI)**  
 Scientific and operational measurements of pollutants O<sub>3</sub>, NO<sub>2</sub>, SO<sub>2</sub>, H<sub>2</sub>CO, C<sub>2</sub>H<sub>2</sub>O<sub>2</sub> (& CO, CH<sub>4</sub>, BrO, OCIO, ClO, IO, H<sub>2</sub>O, O<sub>2</sub>-O<sub>2</sub>, Raman, aerosol, ....)

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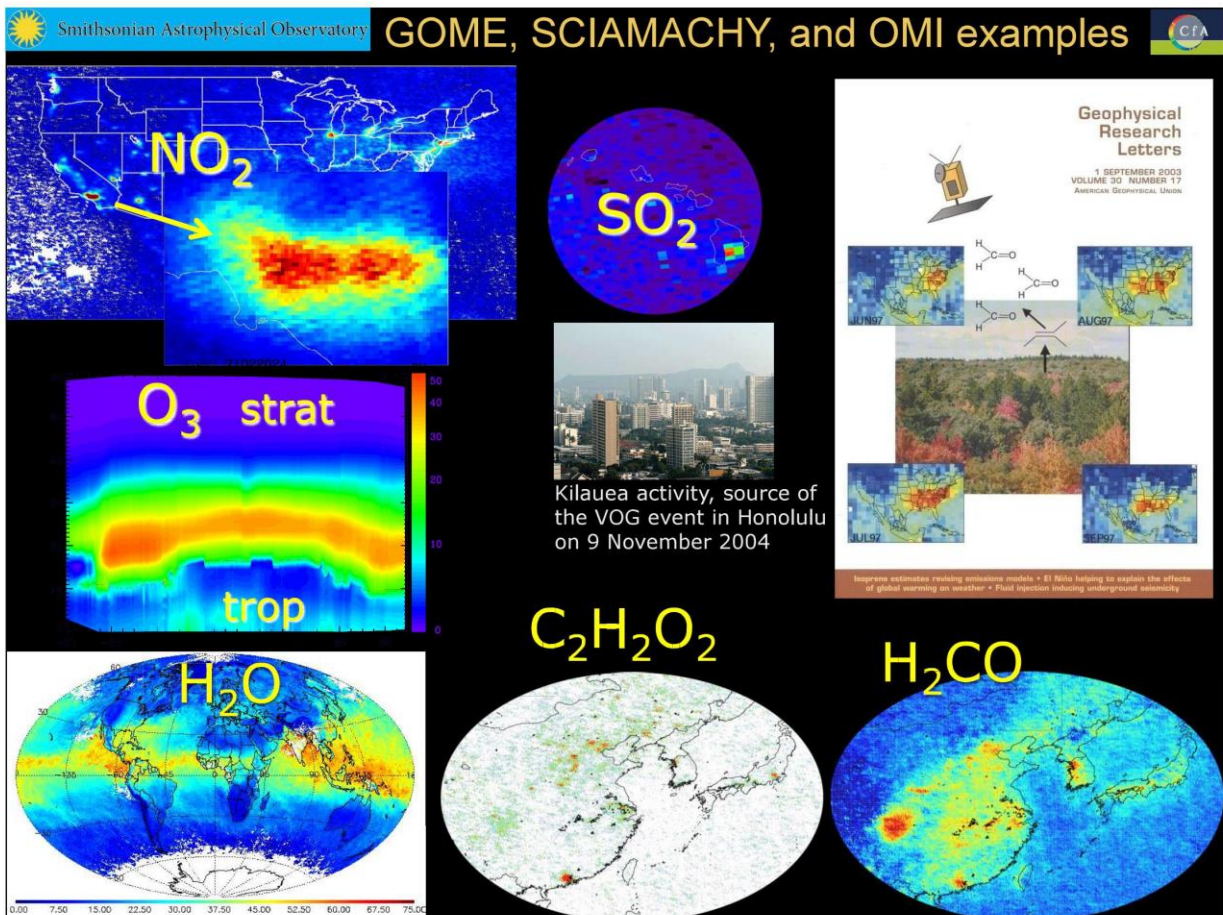
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A full, minimally-redundant, set of polluting gases, plus aerosols and clouds is now measured to very high precision from satellites. Ultraviolet and visible spectroscopy of backscattered radiation provides  $O_3$  (including profiles and tropospheric  $O_3$ ),  $NO_2$  (for  $NO_x$ ),  $H_2CO$  and  $C_2H_2O_2$  (for VOCs),  $SO_2$ ,  $H_2O$ ,  $O_2$ ,  $O_2-O_2$ ,  $N_2$  and  $O_2$  Raman scattering, and halogen oxides (BrO, ClO, IO, OCIO). Satellite spectrometers we planned since 1985 began making these measurements in 1995.

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# Baseline and threshold data products



Species/Products	Required Precision	Temporal Revisit
0-2 km O <sub>3</sub> (Selected Scenes) <b>Baseline only</b>	10 ppbv	2 hour
Tropospheric O <sub>3</sub>	10 ppbv	1 hour
Total O <sub>3</sub>	3%	1 hour
Tropospheric NO <sub>2</sub>	$1.0 \times 10^{15}$ molecules cm <sup>-2</sup>	1 hour
Tropospheric H <sub>2</sub> CO	$1.0 \times 10^{16}$ molecules cm <sup>-2</sup>	3 hour
Tropospheric SO <sub>2</sub>	$1.0 \times 10^{16}$ molecules cm <sup>-2</sup>	3 hour
Tropospheric C <sub>2</sub> H <sub>2</sub> O <sub>2</sub>	$4.0 \times 10^{14}$ molecules cm <sup>-2</sup>	3 hour
Aerosol Optical Depth	0.10	1 hour

- **Minimal set of products sufficient for constraining air quality**
- **Across Greater North America (GNA): 18°N to 58°N near 100°W, 67°W to 125°W near 42°N**
- **Data products at urban-regional spatial scales**
  - Baseline ≤ 60 km<sup>2</sup> at center of Field Of Regard (FOR)
  - Threshold ≤ 300 km<sup>2</sup> at center of FOR
- **Temporal scales to resolve diurnal changes in pollutant distributions**
- **Geolocation uncertainty of less than 4 km**
- **Mission duration, subject to instrument availability**
  - Baseline 20 months
  - Threshold 12 months

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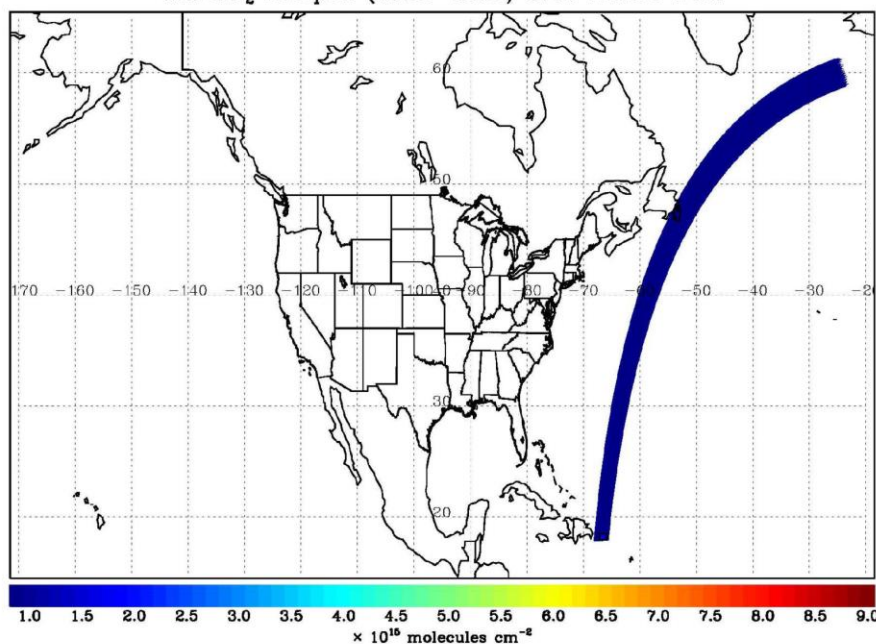
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# TEMPO hourly NO<sub>2</sub> sweep (GEO @92.85W)



OMI NO<sub>2</sub> in April (2005–2008) over TEMPO FOR



**Boresight: 34N, 91W  
~ 2034 good N/S pixels**

- ~ 1282 scans/hr
- ~ 2.6 M pixels/hr
- **Data rate: ~31.2 Mbs (~20 times of OMI data, comparable to TROPOMI)**
- **Scanning partial FOV at ≤ 10 min allowed up to 25% of time**

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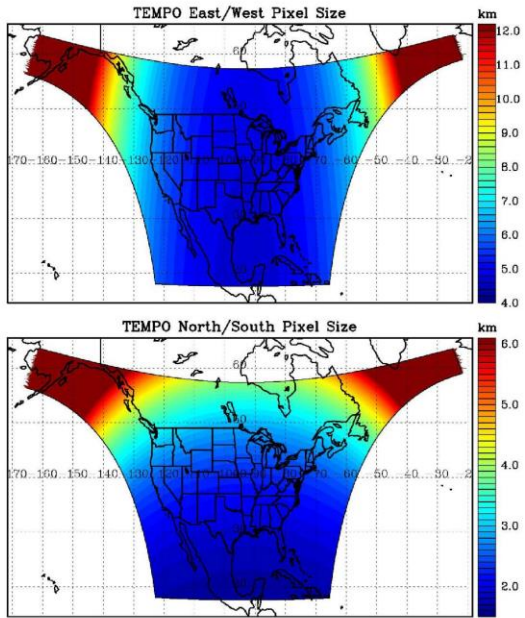
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**TEMPO footprint (GEO @91° W)**



• **Boresight at 33.76°N, 92.85°W**



Location	N/S (km)	E/W (km)	GSA (km <sup>2</sup> )	VZA (°)
Boresight	2.0	4.8	9.5	39.3
36.5°N, 100°W	2.1	4.8	10.1	42.4
Washington, DC	2.3	5.1	11.3	48.0
Seattle	3.2	6.2	16.8	61.7
Los Angeles	2.1	5.6	11.3	48.0
Boston	2.5	5.5	13.0	53.7
Miami	1.8	4.9	8.6	33.2
San Juan	1.7	5.6	9.2	37.4
Mexico City	1.6	4.7	7.7	23.9
Can. tar sands	4.1	5.6	20.8	67.0
Juneau	6.1	9.1	33.3	75.3

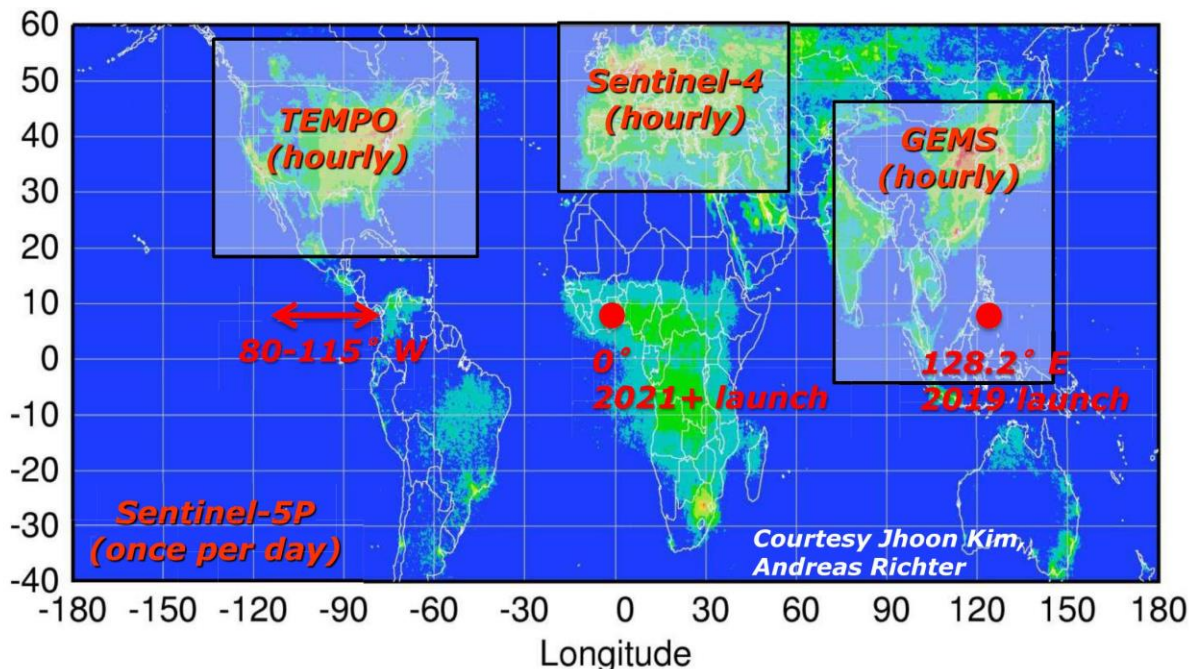
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# Global pollution monitoring constellation



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# The TEMPO Green Paper

Chemistry, physics, and meteorology experiments with the Tropospheric Emissions: Monitoring of Pollution instrument

Now at: <https://www.cfa.harvard.edu/atmosphere/publications.html>

K. Chance<sup>a</sup>, X. Liu<sup>a</sup>, C. Chan Miller<sup>a</sup>, G. González Abad<sup>a</sup>, G. Huang<sup>b</sup>, C. Nowlan<sup>a</sup>, A. Souri<sup>a</sup>, R. Suleiman<sup>a</sup>, K. Sun<sup>a</sup>, H. Wang<sup>a</sup>, L. Zhu<sup>a</sup>, P. Zoogman<sup>a</sup>, J. Al-Saadi<sup>d</sup>, J.-C. Antuña-Marrero<sup>a</sup>, J. Carr<sup>f</sup>, R. Chatfield<sup>g</sup>, M. Chin<sup>h</sup>, R. Cohen<sup>i</sup>, D. Edwards<sup>j</sup>, J. Fishman<sup>k</sup>, D. Flittner<sup>l</sup>, J. Geddes<sup>m</sup>, M. Grutter<sup>m</sup>, J.R. Herman<sup>n</sup>, D.J. Jacob<sup>o</sup>, S. Janz<sup>h</sup>, J. Joiner<sup>h</sup>, J. Kim<sup>p</sup>, N.A. Krotkov<sup>q</sup>, B. Lefer<sup>r</sup>, R.V. Martin<sup>a,s</sup>, O.L. Mayol-Bracero<sup>t</sup>, A. Naeger<sup>u</sup>, M. Newchurch<sup>u</sup>, G.G. Pfister<sup>j</sup>, K. Pickering<sup>v</sup>, R.B. Pierce<sup>w</sup>, C. Rivera Cárdenas<sup>a</sup>, A. Saiz-Lopez<sup>x</sup>, W. Simpson<sup>y</sup>, E. Spinei<sup>z</sup>, R.J.D. Spurr<sup>aa</sup>, J.J. Szykman<sup>bb</sup>, O. Torres<sup>h</sup>, J. Wang<sup>cc</sup>

<p><b>NORMAL TIME RESOLUTION STUDIES</b></p> <ul style="list-style-type: none"> <li>Air quality and health</li> <li>Ultraviolet exposure</li> <li>Biomass burning</li> <li>Synergistic GOES-16/17 Products</li> <li>Advanced aerosol products</li> <li>Soil NO<sub>x</sub> after fertilizer application and after rainfall</li> <li>Solar-induced fluorescence from chlorophyll</li> <li>Foliage studies</li> <li>Mapping NO<sub>2</sub> and SO<sub>2</sub> dry deposition at high resolution</li> <li>Crop and forest damage from ground-level ozone</li> <li>Halogen oxide studies in coastal and lake regions</li> <li>Air pollution from oil and gas fields</li> <li>Night light measurements resolving lighting type</li> <li>Ship tracks, drilling platform plumes, and other concentrated sources.</li> <li>Water vapor studies</li> </ul>	<ul style="list-style-type: none"> <li>Volcanoes</li> <li>Socio-economic studies</li> <li>National pollution inventories</li> <li>Regional and local transport of pollutants</li> <li>Sea breeze studies for Florida and Cuba</li> <li>Transboundary pollution gradients</li> <li>Transatlantic dust transport</li> </ul> <p><b>HIGH TIME RESOLUTION EXPERIMENTS</b></p> <ul style="list-style-type: none"> <li>Lightning NO<sub>x</sub></li> <li>Morning and evening higher-frequency scans</li> <li>Dwell-time studies and temporal selection to improve detection limits</li> <li>Exploring the value of TEMPO in assessing pollution transport during upslope flows</li> <li>Tidal effects on estuarine circulation and outflow plumes</li> <li>Air quality responses to sudden changes in emissions</li> <li>Cloud field correlation with pollution</li> <li>Agricultural soil NO<sub>x</sub> emissions and air quality</li> </ul>
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**Chemistry, physics, and meteorology experiments with the Tropospheric Emissions: Monitoring of Pollution instrument**

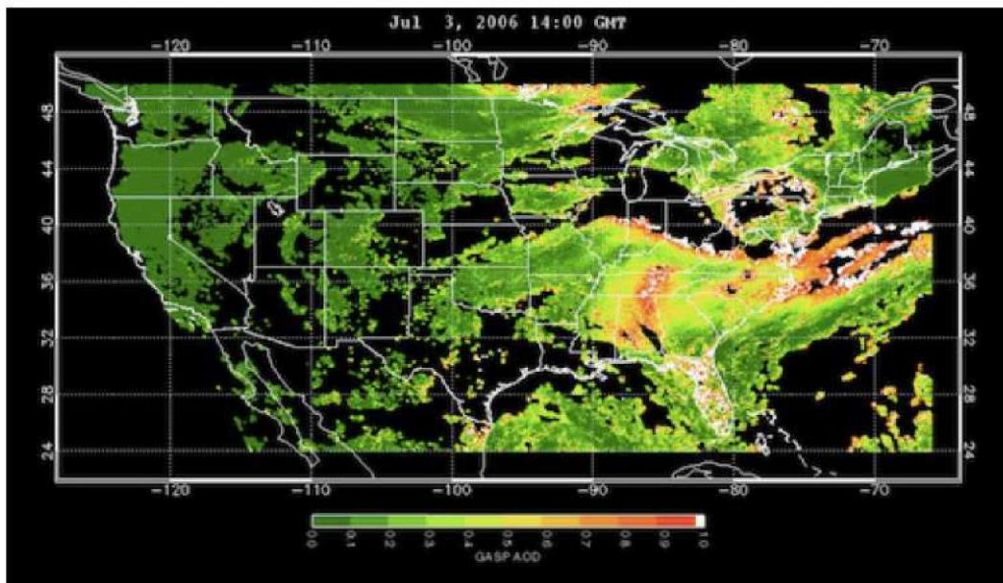
Now at: <https://www.cfa.harvard.edu/atmosphere/publications.html>

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<b>NORMAL TIME RESOLUTION STUDIES</b>	<b>Volcanoes</b>
Air quality and health	Socio-economic studies
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Ship tracks, drilling platform plumes, and other concentrated sources.	Cloud field correlation with pollution
Water vapor studies	Agricultural soil NO <sub>x</sub> emissions and air quality



**TEMPO will use the EPA's Remote Sensing Information Gateway (RSIG) for subsetting, visualization, and product distribution – to make TEMPO YOUR instrument**

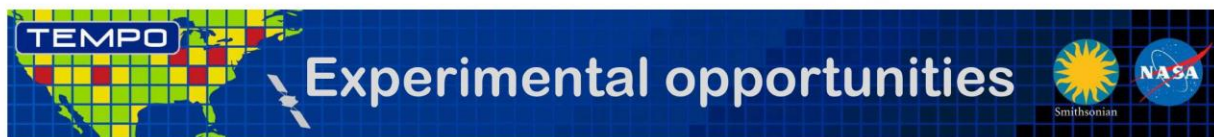




TEMPO's hourly measurements allow better understanding of the complex chemistry and dynamics that drive **air quality on short timescales**. The density of TEMPO data is ideally suited for data assimilation into chemical models for both air quality forecasting and for better constraints on emissions that lead to air quality exceedances. Planning is underway to combine TEMPO with regional air quality models to **improve EPA air quality indices and to directly supply the public with near real time pollution reports and forecasts through website and mobile applications**. As a case study, an OSSE for the Intermountain West was performed to explore the potential of geostationary ozone measurements from TEMPO to improve monitoring of ozone exceedances and the role of background ozone in causing these exceedances (Zoogman *et al.* 2014).

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The TEMPO Green Paper living document is at <http://tempo.si.edu/publications>. Please feel free to contribute

1. Up to 25% of observing time can be devoted to non-standard operations: Time resolution higher, E/W spatial coverage less
2. Two types of studies under regular or non-standard operations
  1. Events (e.g., eruptions, fires, dust storms, etc.)
  2. Experiments (e.g., agriculture, forestry, NO<sub>x</sub>, ....)
3. TEMPO team will work with experimenters concerning Image Navigation and Registration (*i.e.*, pointing resolution and accuracy)
4. Experiments could occur during commissioning phase
5. Hope to include SO<sub>2</sub>, aerosol, H<sub>2</sub>O, and C<sub>2</sub>H<sub>2</sub>O<sub>2</sub> as operational products
6. Can initiate a non-standard, pre-loaded scan pattern within several hours
7. Send your ideas into a TEMPO team member

10 October, 2019



# Traffic, biomass burning

**Morning and evening higher-frequency scans** The optimized data collection scan pattern during mornings and evenings provides multiple advantages for addressing TEMPO science questions. The increased frequency of scans coincides with peaks in vehicle miles traveled on each coast.

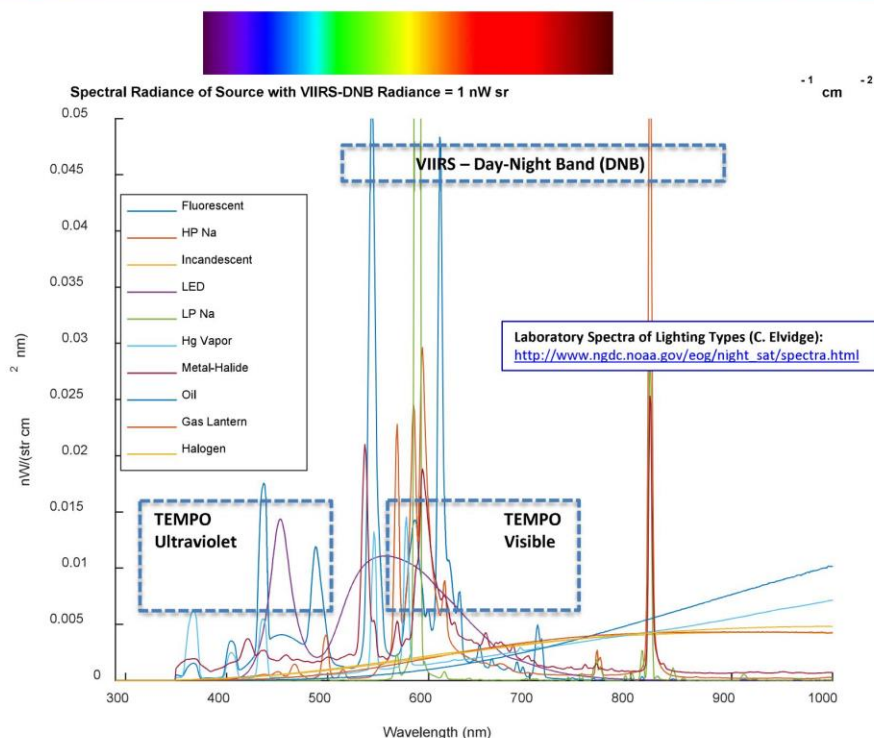
**Biomass burning** The unexplained variability in ozone production from fires is of particular interest. The suite of  $\text{NO}_2$ ,  $\text{H}_2\text{CO}$ ,  $\text{C}_2\text{H}_2\text{O}_2$ ,  $\text{H}_2\text{O}$ ,  $\text{O}_3$ , and aerosol measurements from TEMPO is well suited to investigating how the chemical processing of primary fire emissions effects the secondary formation of VOCs and ozone. For particularly important fires it is possible to command special TEMPO observations at even shorter than hourly revisit time, as short as 10 minutes.

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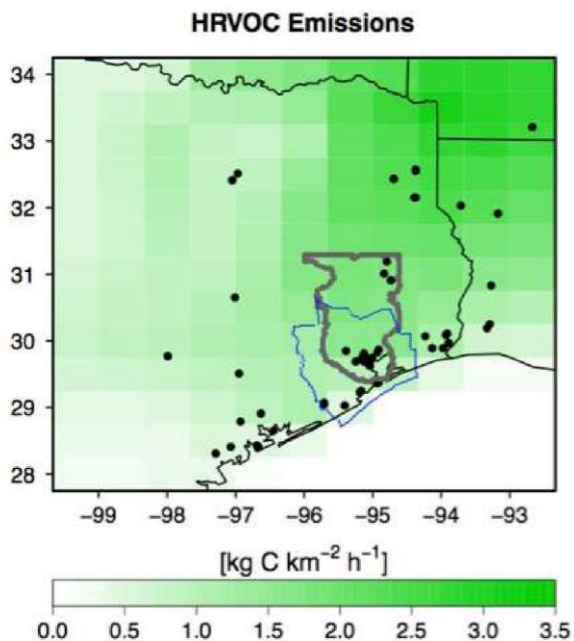
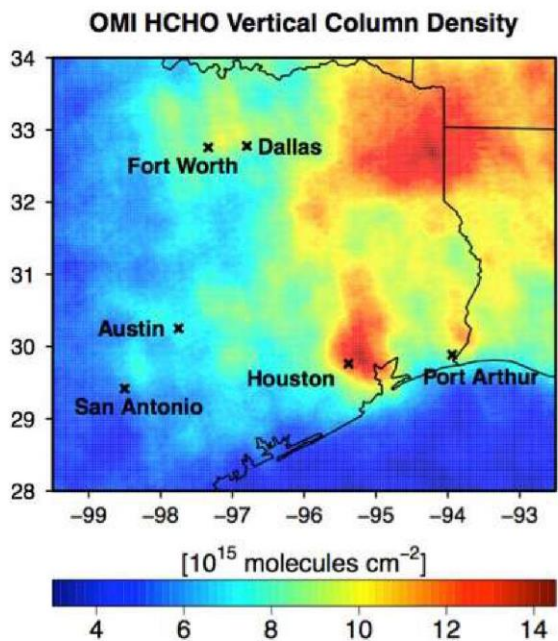
# City lights spectroscopic signatures



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**TEMPO** **Oversampling**  
**Lei Zhu et al., 2014**

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**TEMPO** **The end!**  
 Thanks to NASA, ESA, Maxar, Ball  
 Aerospace & Technologies Corp., ESA




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**Lightning NO<sub>x</sub>** Interpretation of satellite measurements of tropospheric NO<sub>2</sub> and O<sub>3</sub>, and upper tropospheric HNO<sub>3</sub> lead to an overall estimate of  $6 \pm 2$  Tg N y<sup>-1</sup> from lightning [Martin et al., 2007]. TEMPO measurements, including tropospheric NO<sub>2</sub> and O<sub>3</sub>, can be made for time periods and longitudinal bands selected to coincide with large thunderstorm activity, including outflow regions, with fairly short notice.

**Soil NO<sub>x</sub>** Jaeglé et al. [2005] estimate  $2.5 - 4.5$  TgN y<sup>-1</sup> are emitted globally from nitrogen-fertilized soils, still highly uncertain. The US a posteriori estimate for 2000 is  $0.86 \pm 1.7$  TgN y<sup>-1</sup>. For Central America it is  $1.5 \pm 1.6$  TgN y<sup>-1</sup>. They note an underestimate of NO release by nitrogen-fertilized croplands as well as an underestimate of rain-induced emissions from semiarid soils.

TEMPO is able to follow the temporal evolution of emissions from croplands after **fertilizer application** and from rain-induced emissions from semi-arid soils. Higher than hourly time resolution over selected regions may be accomplished by special observations. Improved constraints on soil NO<sub>x</sub> emissions may also improve estimated of lightning NO<sub>x</sub> emissions [Martin et al. 2000].



**Fluorescence and other spectral indicators** Solar-induced fluorescence (SIF) from chlorophyll over both land and ocean will be measured. In terrestrial vegetation, chlorophyll fluorescence is emitted at red to far-red wavelengths (~650-800 nm) with two broad peaks near 685 and 740 nm, known as the red and far-red emission features. Oceanic SIF is emitted exclusively in the red feature. SIF measurements have been used for studies of tropical dynamics, primary productivity, the length of carbon uptake period, and drought responses, while ocean measurements have been used to detect red tides and to conduct studies on the physiology, phenology, and productivity of phytoplankton. TEMPO can retrieve both red and far-red SIF by utilizing the property that SIF fills in solar Fraunhofer and atmospheric absorption lines in backscattered spectra normalized by a reference (e.g., the solar spectrum) that does not contain SIF.

TEMPO will also be capable of measuring **spectral indices developed for estimating foliage pigment contents and concentrations**. Spectral approaches for estimating pigment contents apply generally to leaves and not the full canopy. A single spectrally invariant parameter, the Directional Area Scattering Factor (DASF), relates canopy-measured spectral indices to pigment concentrations at the leaf scale.

**UVB** TEMPO measurements of daily UV exposures build upon heritage from OMI and TROPOMI measurements. Hourly cloud measurements from TEMPO allow taking into account diurnal cloud variability, which has not been previously possible. The OMI UV algorithm is based on the TOMS UV algorithm. The specific product is the downward spectral irradiance at the ground (in  $W m^{-2} nm^{-1}$ ) and the erythemally weighted irradiance (in  $W m^{-2}$ ).

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**Aerosols** TEMPO's launch algorithm for retrieving aerosols will be based upon the OMI aerosol algorithm that uses the sensitivity of near-UV observations to particle absorption to retrieve **absorbing aerosol index (AAI)**, **aerosol optical depth (AOD)** and **single scattering albedo (SSA)**. TEMPO will derive its pointing from one of the **GOES-16** or **GOES-17** satellites and is thus automatically co-registered. TEMPO may be used together with the advanced baseline imager (ABI) instrument, particularly the  $1.37\mu m$  bands, for aerosol retrievals, reducing AOD and fine mode AOD uncertainties from 30% to 10% and from 40% to 20%.

**Clouds** The launch cloud algorithm is based on the rotational Raman scattering (RRS) cloud algorithm that was developed for OMI by NASA GSFC. Retrieved cloud pressures from OMCLDRR are not at the geometrical center of the cloud, but rather at the optical centroid pressure (OCP) of the cloud.

**Additional** cloud products are possible using the  $O_2-O_2$  collision complex and/or the  $O_2 B$  band.

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**BrO** will be produced at launch, assuming stratospheric AMFs. Scientific studies will correct retrievals for tropospheric content. **IO** was first measured from space by SAO using SCIAMACHY spectra [Saiz-Lopez *et al.*, 2007]. It will be produced as a scientific product, particularly for coastal studies, assuming AMFs appropriate to lower tropospheric loading.

**The atmospheric chemistry of halogen oxides over the ocean, and in particular in coastal regions**, can play important roles in ozone destruction, oxidizing capacity, and dimethylsulfide oxidation to form cloud-condensation nuclei [Saiz-Lopez and von Glasow, 2012]. The budgets and distribution of reactive halogens along the coastal areas of North America are poorly known. Therefore, providing a measure of the budgets and diurnal evolution of coastal halogen oxides is necessary to understand their role in atmospheric photochemistry of coastal regions. Previous ground-based observations have shown enhanced levels (at a few pptv) of halogen oxides over coastal locations with respect to their background concentrations over the remote marine boundary layer [Simpson *et al.*, 2015]. Previous global satellite instruments lacked the sensitivity and spatial resolution to detect the presence of active halogen chemistry over mid-latitude coastal areas. TEMPO observations together with atmospheric models will allow examination of the processes linking ocean halogen emissions and their potential impact on the oxidizing capacity of coastal environments of North America.

TEMPO also performs **hourly measurements one of the world's largest salt lakes: the Great Salt Lake in Utah**. Measurements over Salt Lake City show the highest concentrations of BrO over the globe. Hourly measurement at a high spatial resolution can improve understanding of BrO production in salt lakes.

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- **Geostationary orbit, operating on a commercial telecom satellite**
  - NASA will arrange launch and hosting services (per Earth Venture Instrument scope)
    - 80-115° W acceptable latitude
    - Specifying satellite environment, accommodation
  - Hourly measurement and telemetry duty cycle for at least  $\leq 70^\circ$  SZA
- **TEMPO is low risk with significant space heritage**
  - We proposed SCIAMACHY in 1985, as suggested by the late Dr. Dieter Perner
  - All proposed TEMPO measurements except eXceL O<sub>3</sub> have been made from low Earth orbit satellite instruments to the required precisions by SAO and Science Team members
  - All TEMPO launch algorithms are implementations of currently operational algorithms
    - NASA TOMS-type O<sub>3</sub>
    - SO<sub>2</sub>, NO<sub>2</sub>, H<sub>2</sub>CO, C<sub>2</sub>H<sub>2</sub>O<sub>2</sub> from fitting with AMF-weighted cross sections
    - Absorbing Aerosol Index, UV aerosol, Rotational Raman scattering cloud
    - SAO eXceL profile/tropospheric/PBL O<sub>3</sub> for selected geographic targets
- **Example higher-level products: Near-real-time pollution/AQ indices, UV index**
- **TEMPO research products will greatly extend science and applications**
  - **Example research products:** BrO and IO from AMF-normalized cross sections; height-resolved SO<sub>2</sub>; additional cloud/aerosol products; vegetation products; additional gases; city lights

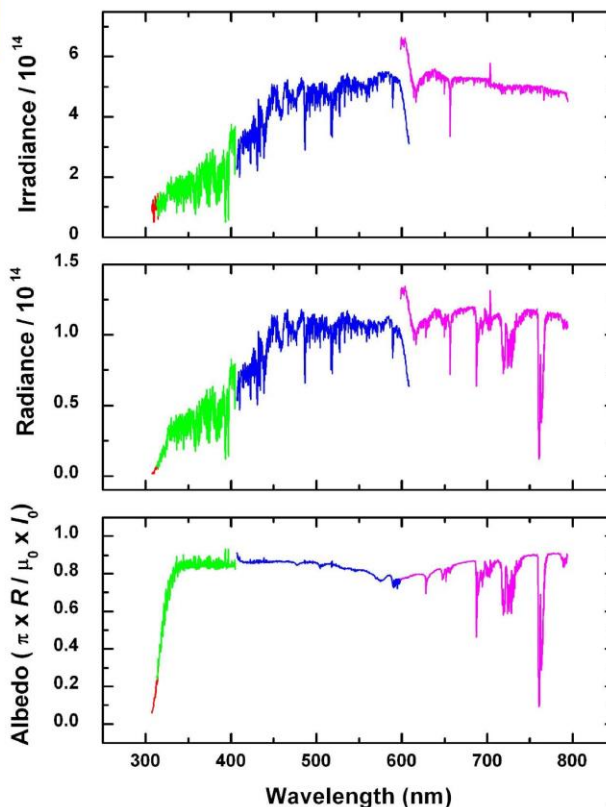
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## What do we measure?

**GOME irradiance, radiance, and reflectance spectrum for high-albedo (fully cloudy) ground pixel**



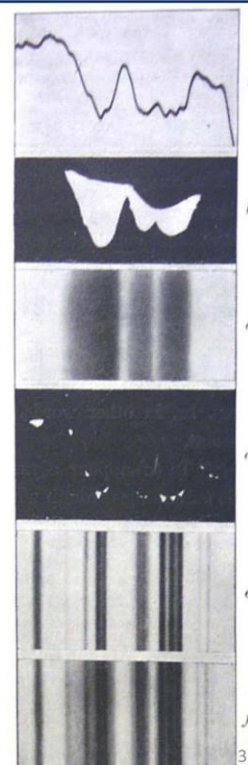
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Langley, S.P., and C.G. Abbot, *Annals of the Astrophysical Observatory of the Smithsonian Institution, Volume 1* (1900).

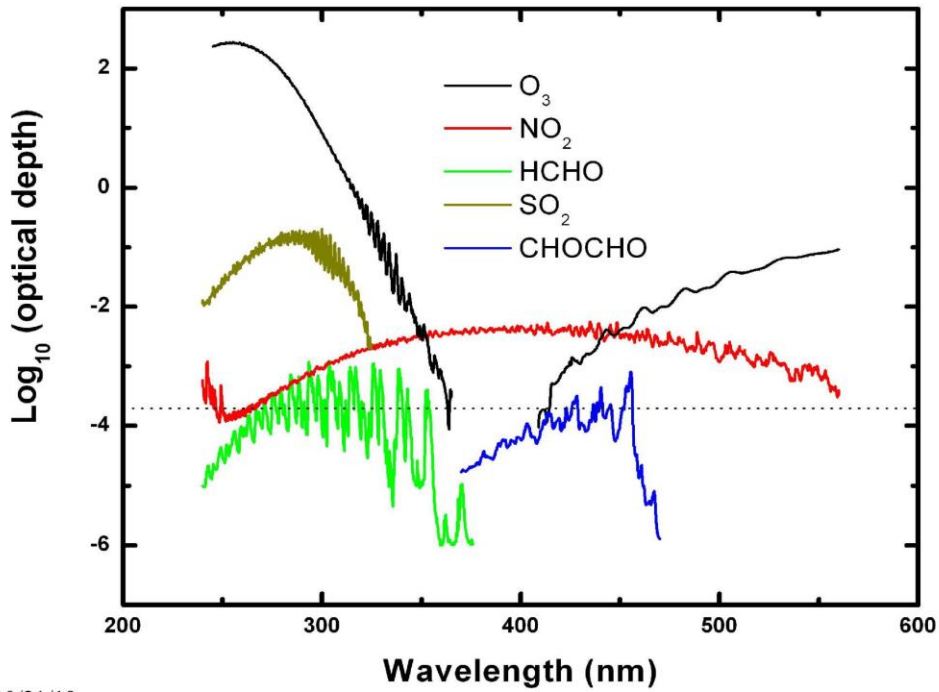
Langley's recently invented bolometer was used to make measurements from the infrared through the near ultraviolet in order to determine the mean value of the solar constant and its variation. Langley and Abbot also developed substantial new experimental techniques (such as an early chart recorder) and various analysis techniques (e.g., the "Langley plot"), including photographic techniques for high and low pass filtering to produce line spectra from "bolographs" (spectra), illustrated, foreshadowing the high pass filtering used today by researchers employing the DOAS technique for analyzing atmospheric spectra.



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### Optical Depths for Typical GEO Measurement Geometry



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The slide features a satellite map of Puerto Rico with a grid overlay. Major cities labeled include Mayagüez, Ponce, Patillas, Caguas, Carolina, Vega Baja, San Juan, and Fajardo. A large yellow text overlay reads "¡Cada hora!". Logos for TEMPO, Smithsonian, and NASA are in the top right. A scale bar for 16.31 miles is in the bottom left. Data sources listed at the bottom include SIO, NOAA, U.S. Navy, NGA, GEBCO, Landsat, Copernicus, Google, and LDEO-Columbia, NSF, NOAA.

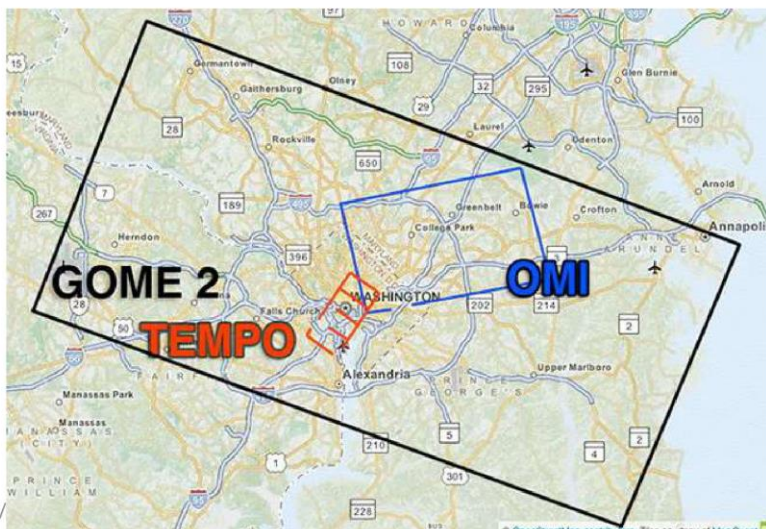
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# TEMPO Coverage comparisons




- **Spatial resolution: allows tracking pollution at sub-urban scale**
  - GEO at 100°W: 2.1 km N/S × 4.7 km E/W = 9.8 km<sup>2</sup> (native) at center of FOR (36.5°N, 100°W)
  - Full resolution for NO<sub>2</sub>, HCHO, total O<sub>3</sub> products
  - Co-add 4 N/S pixels for O<sub>3</sub> profile product: 8.4 km N/S × 4.7 km E/W





~ 1/300 of GOME-2  
~ 1/30 of OMI

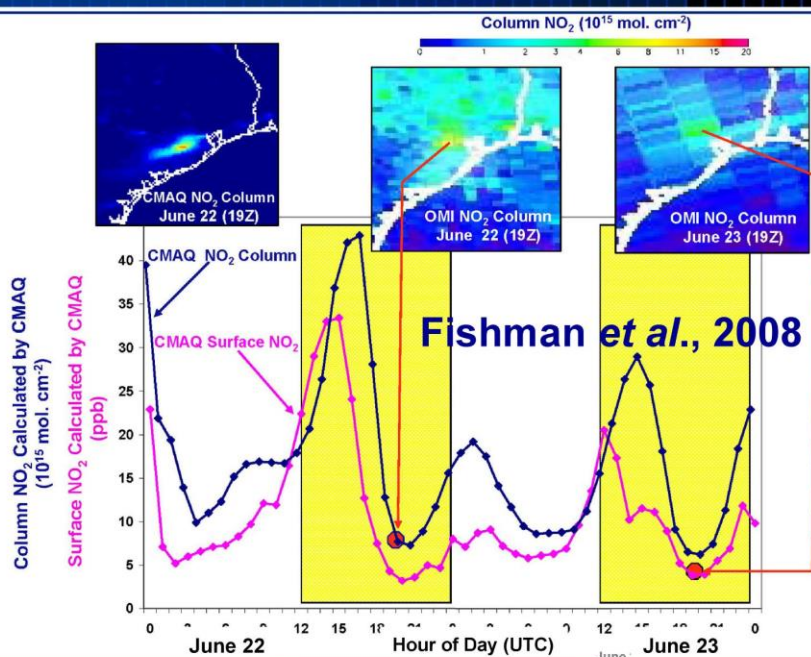
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# Why geostationary? High temporal and spatial resolution

Hourly NO<sub>2</sub> surface concentration and integrated column calculated by CMAQ air quality model: Houston, TX, June 22-23, 2005



LEO observations provide limited information on rapidly varying emissions, chemistry, & transport

GEO will provide observations at temporal and spatial scales highly relevant to air quality processes

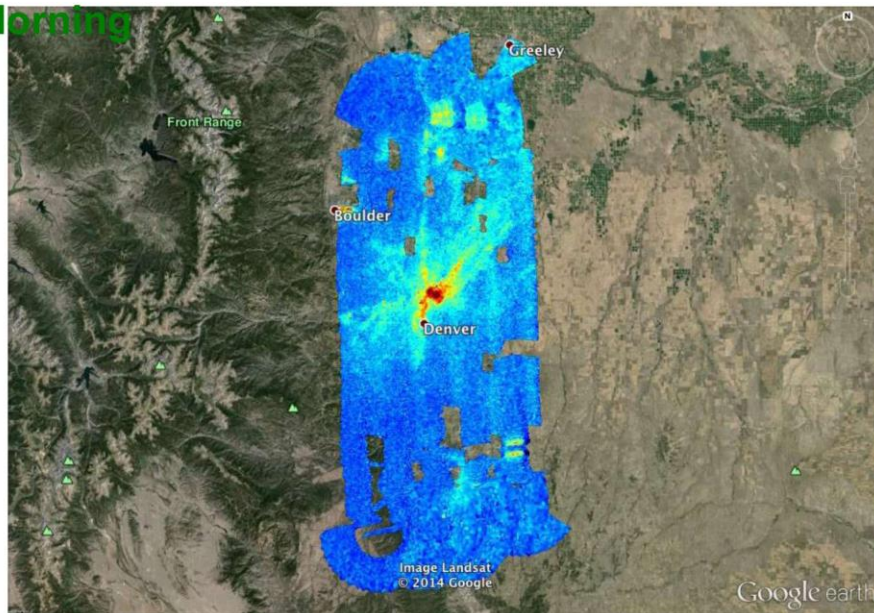
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### GeoTASO NO<sub>2</sub> Slant Column, 02 August 2014

Morning



Co-added to approx.  
500m x 450m  
10/21/19

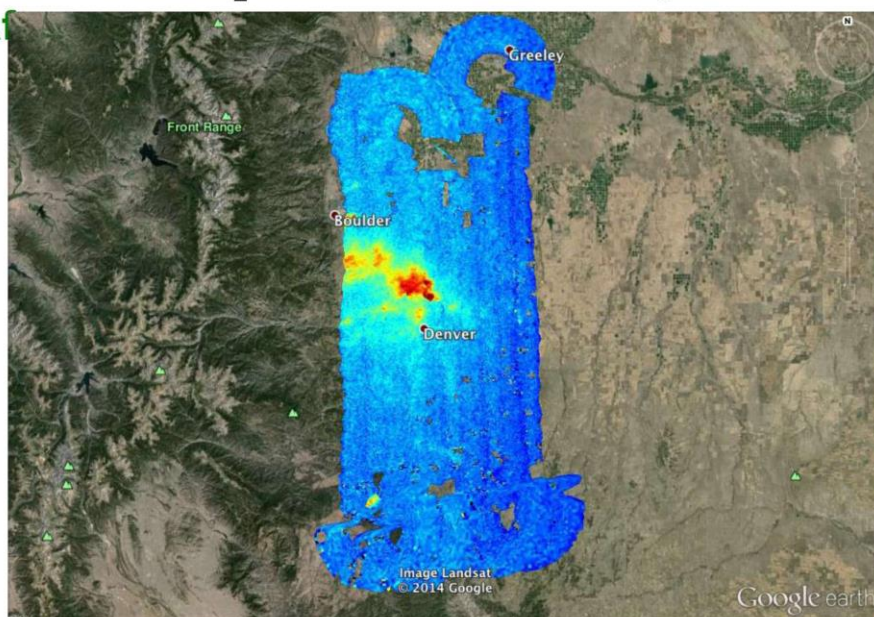
Morning vs. Afternoon

Preliminary data,  
C. Nowlan, SAO<sup>43</sup>



### GeoTASO NO<sub>2</sub> Slant Column, 02 August 2014

Afternoon




Co-added to approx.  
500m x 450m  
10/21/19

Morning vs. Afternoon

Preliminary data,  
C. Nowlan, SAO<sup>44</sup>



附錄五、"Local to Global Air Quality Simulations using the NASA GEOS Composition Forecast Model GEOS-CF"

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# Local to Global Air Quality Simulations using the NASA GEOS Composition Forecast Model, GEOS-CF


**K. Emma Knowland**  
USRA/GESTAR  
NASA Global Modeling and Assimilation Office (GMAO)

**In collaboration with:**  
GMAO: Christoph Keller, Lesley Ott, Steven Pawson, Emily Saunders, Pamela Wales  
Atmospheric Chemistry and Dynamics Lab: Bryan Duncan, Melanie Follette-Cook, Junhua Liu, Julie Nicely, Daniel Anderson

22 October 2019  

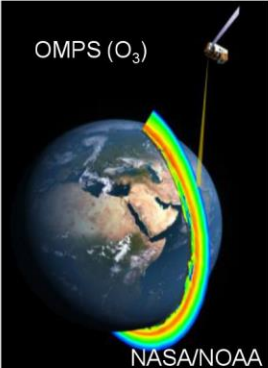
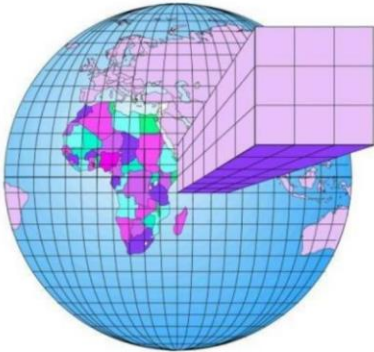
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## NASA GMAO global meteorology and chemistry products

**GEOS**



OMPS (O<sub>3</sub>) 

NASA/NOAA  
www.nasa.gov

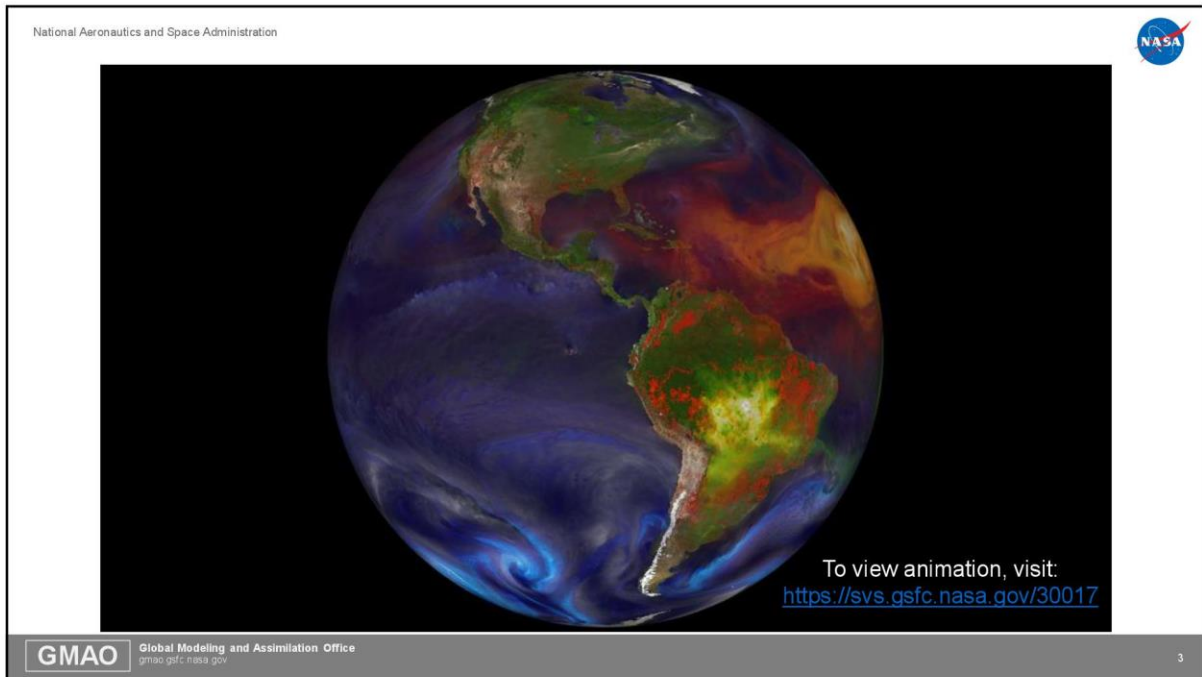
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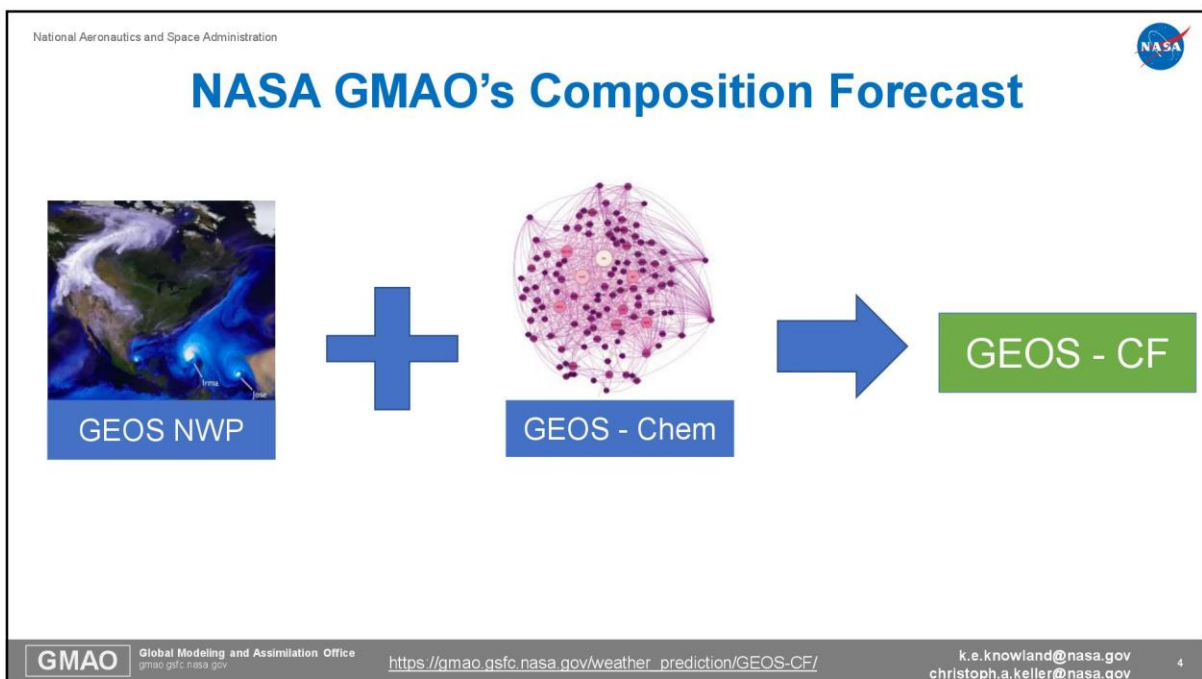
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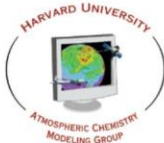

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## GEOS-Chem is a state-of-the science chemistry transport model

Tropospheric and Stratospheric full chemistry

- 250 reactive species, 725 reactions
- 100+ user/developer groups worldwide
- Updated version is released about every year
- Version 12.0.1



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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

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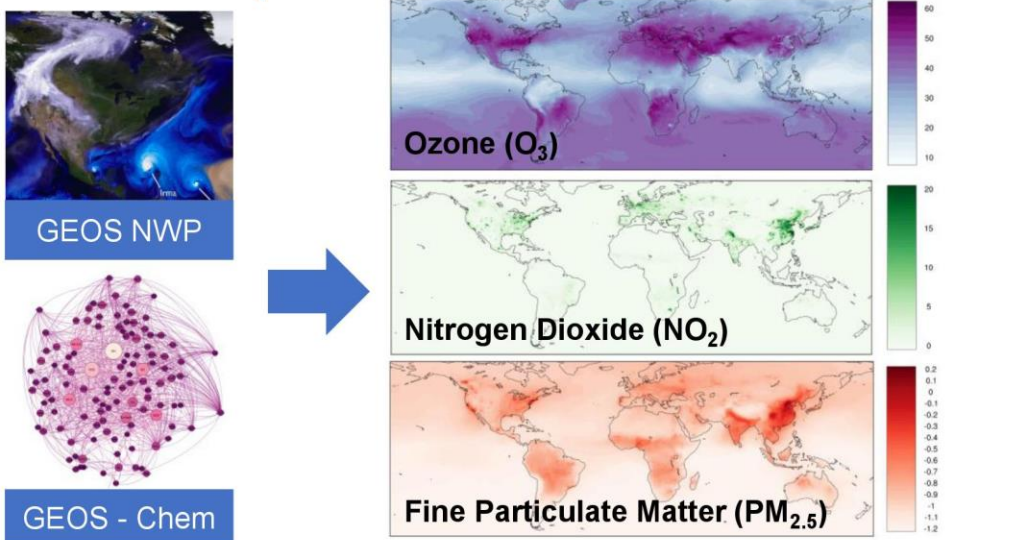
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## NASA's composition forecast (GEOS-CF)



**GEOS NWP**

**GEOS - Chem**

**Ozone ( $O_3$ )**

**Nitrogen Dioxide ( $NO_2$ )**

**Fine Particulate Matter ( $PM_{2.5}$ )**

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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

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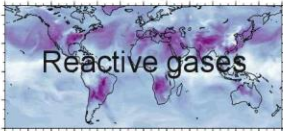
## Aerosol and Gas Phase Chemistry



**Aerosols**

- Particulate matter:
  - Carbon
  - Sea salt
  - Dust
  - Sulfate
  - (Nitrate)
  - (Secondary Organics)

GOCART



**Reactive gases**

- Ozone (O<sub>3</sub>)
- Nitrogen dioxide (NO<sub>2</sub>)
- Carbon monoxide (CO)
- Volatile organic compounds (VOCs):
  - Formaldehyde
  - Benzene / Toluene
  - And many more!

GEOS-Chem

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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

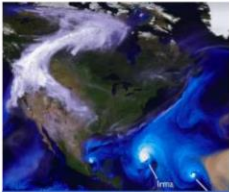
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christoph.a.keller@nasa.gov

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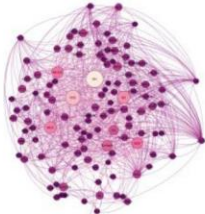
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## Daily composition forecast



GEOS NWP



GEOS - Chem



GEOS - CF

One **5-day forecast** per day

GMAO

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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

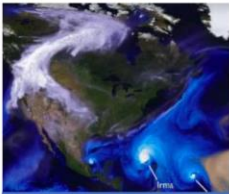

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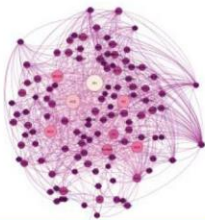
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## Daily composition forecast



GEOS NWP



GEOS - Chem

➔

**GEOS - CF**

One **5-day forecast** per day

- 1-day meteorological replay "analysis"
- 5-day forecast

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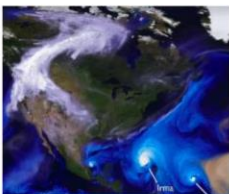

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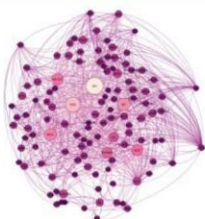
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## Daily composition forecast



GEOS NWP



GEOS - Chem

➔

**GEOS - CF**

One **5-day forecast** per day

- 1-day replay
- 5-day forecast
- c360 (0.25°, ~**25x25 km<sup>2</sup>**) resolution, 72 model layers

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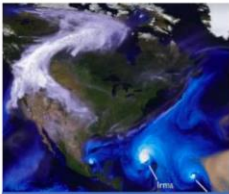

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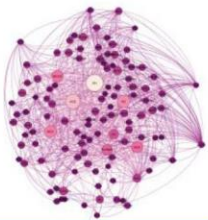

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## Daily composition forecast



GEOS NWP



GEOS - Chem

**GEOS - CF**

One **5-day forecast** per day

- 1-day replay
- 5-day forecast
- c360 (0.25°, ~**25x25 km<sup>2</sup>**) resolution, 72 model layers
- O<sub>3</sub>, NO<sub>x</sub>, VOCs, PM ...
- T, U, V, RH ....

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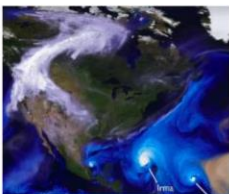

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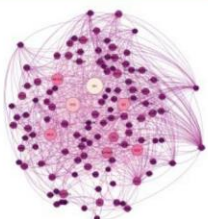

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## Daily composition forecast



GEOS NWP



GEOS - Chem

**GEOS - CF**

One **5-day forecast** per day

- 1-day replay
- 5-day forecast
- c360 (0.25°, ~**25x25 km<sup>2</sup>**)
- **15 minute "surface"**
- **1-hour** average and instantaneous 2D & 3D

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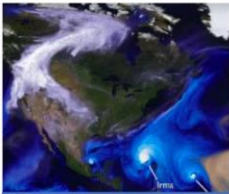

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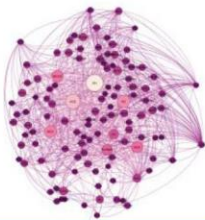
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## Daily composition forecast



GEOS NWP



GEOS - Chem

➔

**GEOS - CF**

One **5-day forecast** per day

- 1-day replay
- 5-day forecast
- c360 (0.25°, ~25x25 km<sup>2</sup>)

➤ **Available since 1 January 2018**

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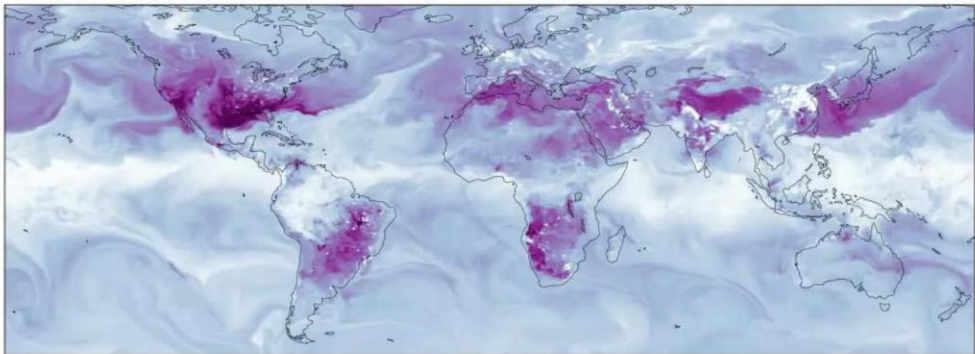
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## High-Resolution Global Simulation

2017-10-01 00:30 UTC



10 20 30 40 50 60

Surface ozone [ppbv]

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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

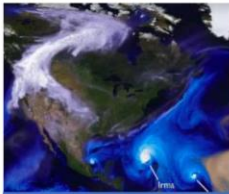
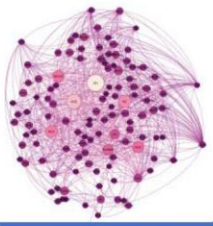
k.e.knowland@nasa.gov  
christoph.a.keller@nasa.gov

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## Daily composition forecast

GEOS - CF

➔


- Currently **no direct data assimilation of constituents** in GEOS-CF
- Stratospheric O<sub>3</sub> (above ~75 hPa) nudged to GEOS assimilated O<sub>3</sub>

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k.e.knowland@nasa.gov  
christoph.a.keller@nasa.gov

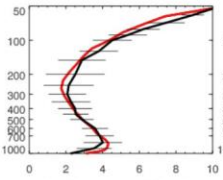
15

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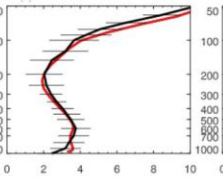
## Comparison of GEOS-CF 2017 O<sub>3</sub> against ozone sondes



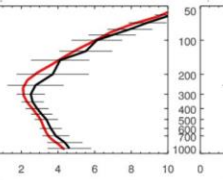
Trinidad Head, CA



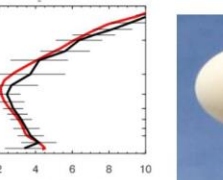
Sapporo, Japan



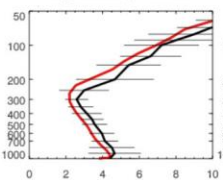
Payerne, Switzerland



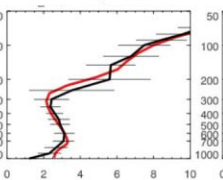
Hohenpeissenberg



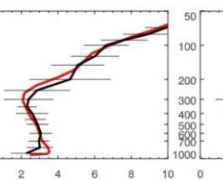
Legionowo, Poland



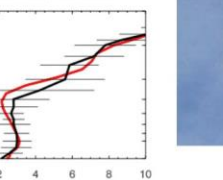
Goose Bay, Canada



Edmonton, Canada



Churchill, Canada



Sondes
  GEOS-CF

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## Daily composition forecast

### GEOS - CF

- GOCART aerosols constrained by satellite measurements of AOD
- Biomass burning emissions from QFED

**GEOS NWP**

**GEOS - Chem**

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## Case study: agricultural fires in India

**Delhi, India, 2017-10-26 00:00 UTC**

**MODIS fires Nov 01, 2017**

Observation      0 +1 +2 +3 +4 +5      GEOS-CF

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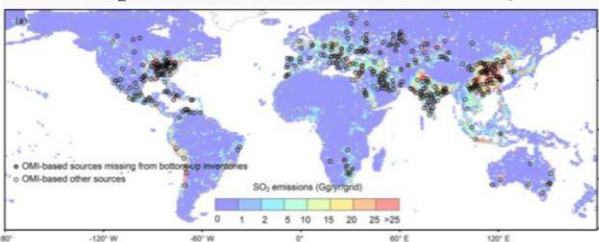
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## Daily composition forecast



### GEOS - CF

SO<sub>2</sub> emissions in the OMI-HTAP inventory




● OMI-based sources missing from bottom-up inventories  
○ OMI-based other sources

SO<sub>2</sub> emissions (Gg/yr/lat) scale: 0, 1, 2, 5, 10, 15, 20, 25, >25

➤ Annual gridded scale factors based on satellite data are applied to the emissions of CO (Oda et al., 2017) and SO<sub>2</sub> (Liu et al., 2018).

**A new emission inventory, OMI-HTAP, combines OMI-based SO<sub>2</sub> emissions for large sources and the bottom-up inventory, HTAP, for smaller sources.** Liu, F., et al., *Atmos. Chem. Phys.*, 18, 2018



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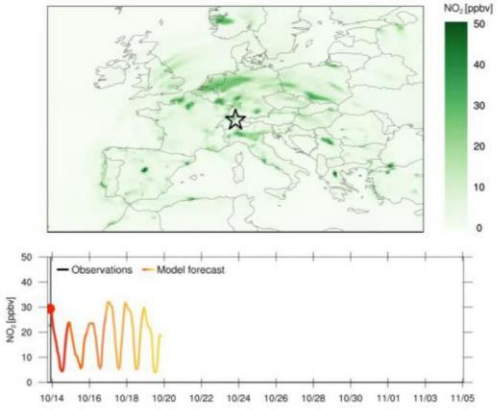
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## Daily composition forecast



### GEOS - CF

Zurich, Switzerland, 2017-10-14 00:00 UTC




NO<sub>2</sub> [ppbv] scale: 0, 10, 20, 30, 40, 50

— Observations — Model forecast

Legend: Observation (black line), GEOS-CF (0, +1, +2, +3, +4, +5 scale)

➤ Scale factors applied to emissions for diurnal and weekly variations




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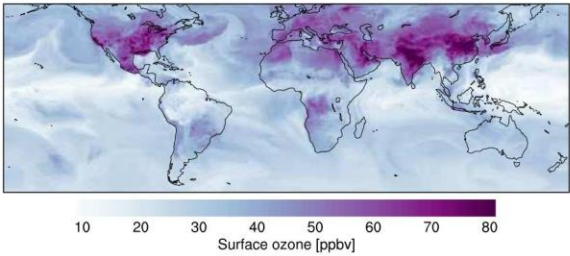
20

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## Air quality and health applications

➤ **How good is the model?**

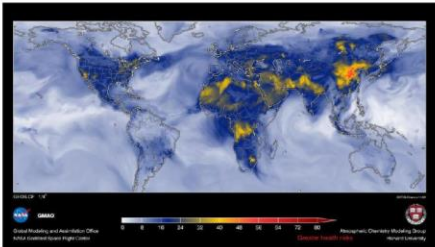
2019-06-09



10 20 30 40 50 60 70 80  
Surface ozone [ppbv]

**Optimize model predictions**


➤ **How bad is the air pollution?**



**Global exposure assessment**

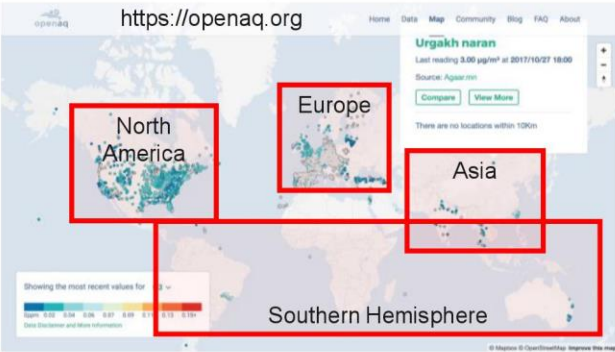
**GMAO** Global Modeling and Assimilation Office [https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/) [k.e.knowland@nasa.gov](mailto:k.e.knowland@nasa.gov)  
[christoph.a.keller@nasa.gov](mailto:christoph.a.keller@nasa.gov)

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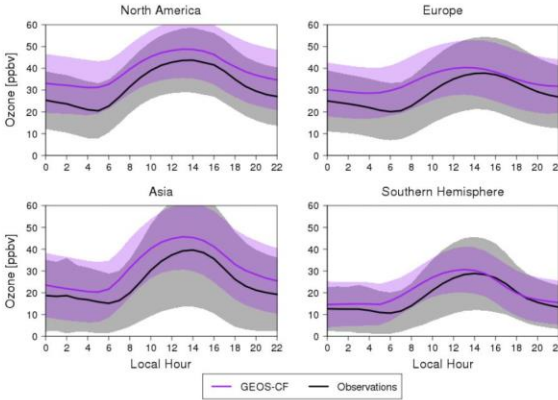
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## Surface evaluation

<https://openaq.org>



OpenAQ is a non-profit compiling publicly available air quality data in near-real time into an open-source data base

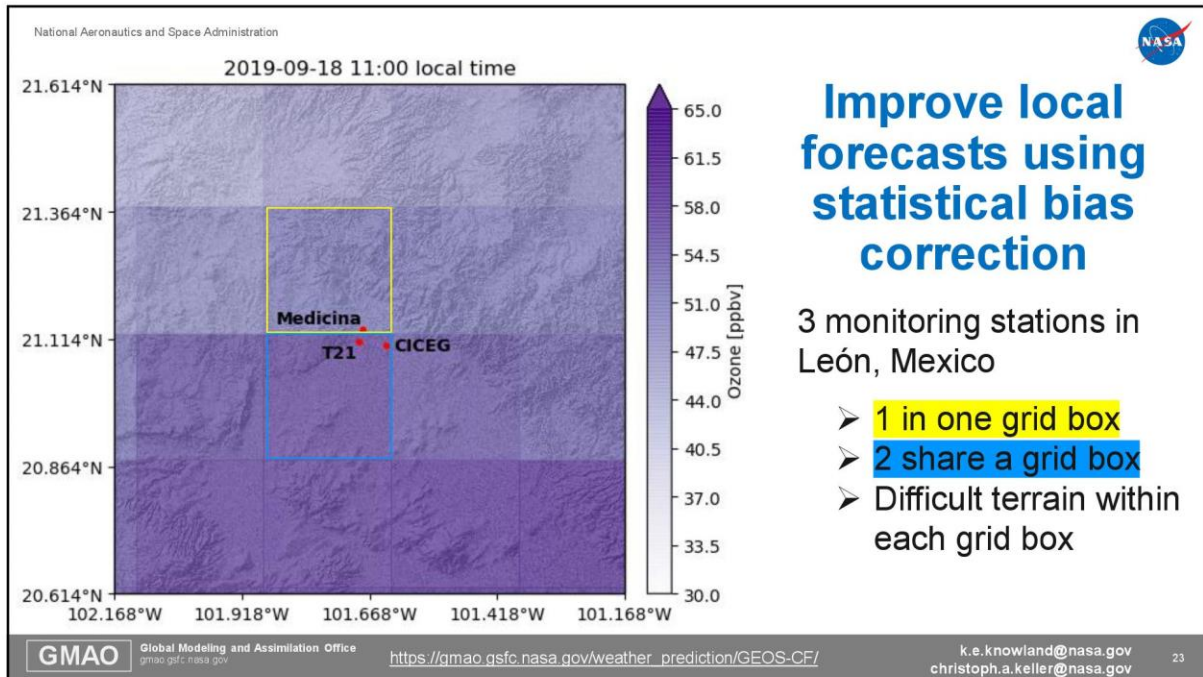


— GEOS-CF — Observations

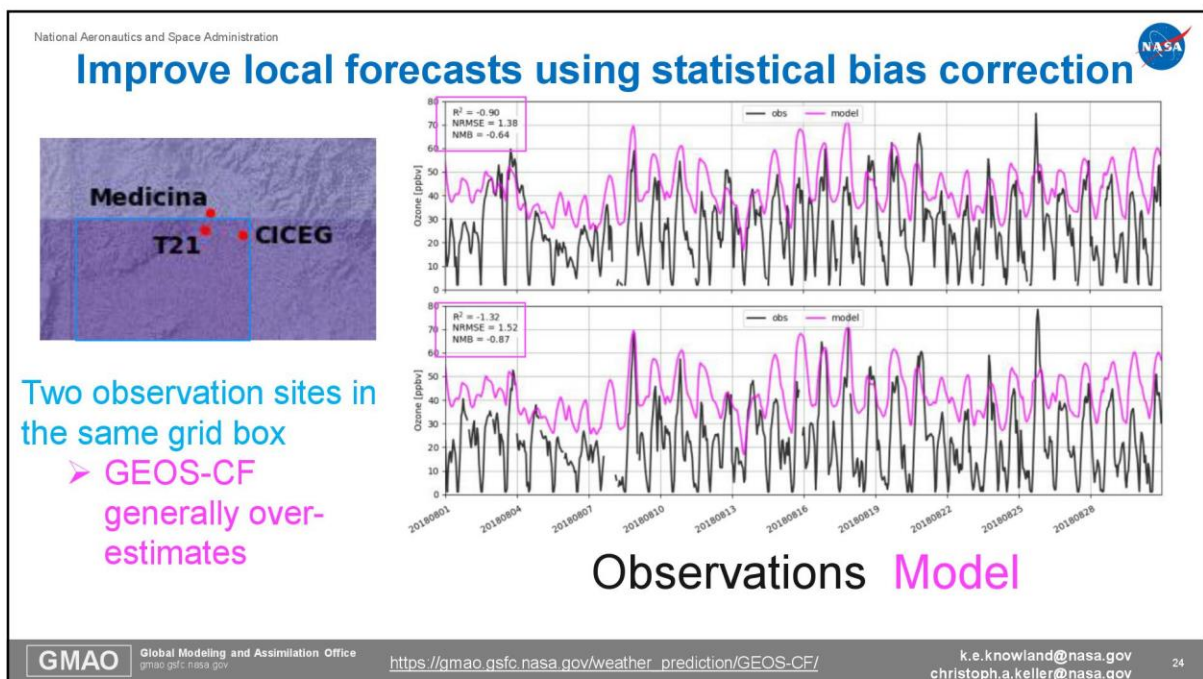
**GMAO** Global Modeling and Assimilation Office [https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/) [k.e.knowland@nasa.gov](mailto:k.e.knowland@nasa.gov)  
[christoph.a.keller@nasa.gov](mailto:christoph.a.keller@nasa.gov) 22

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## Use machine learning to correct for small scale variability and/or model biases

Inputs

Meteorology (10 vars)

Chemistry (42 vars)

Emissions (not yet used)

Geography (not yet used)


→

Output

Local concentration adjustment

ML

- Algorithm: gradient boosted decision trees (XGBoost)
- Train separate algorithm for each site




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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

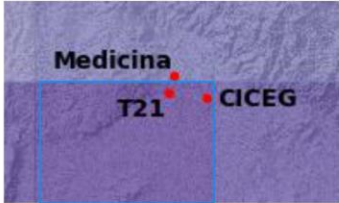
k.e.knowland@nasa.gov  
christoph.a.keller@nasa.gov

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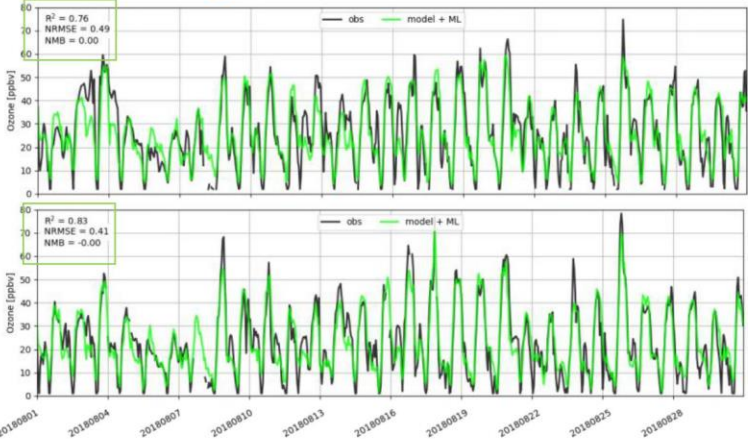
25

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## Improve local forecasts using statistical bias correction




**Medicina**  
T21 CICEG



Two observation sites in the same grid box

- GEOS-CF+AI captures diurnal variability at sub-grid scale

Observations Model + AI



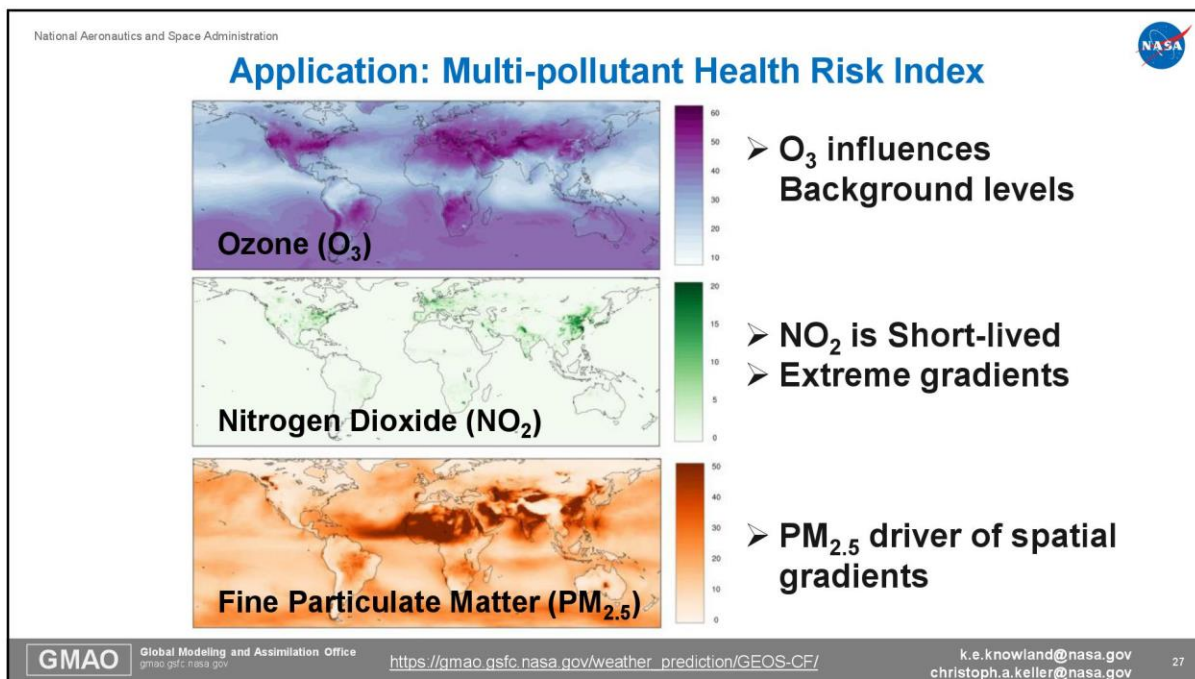
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[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)

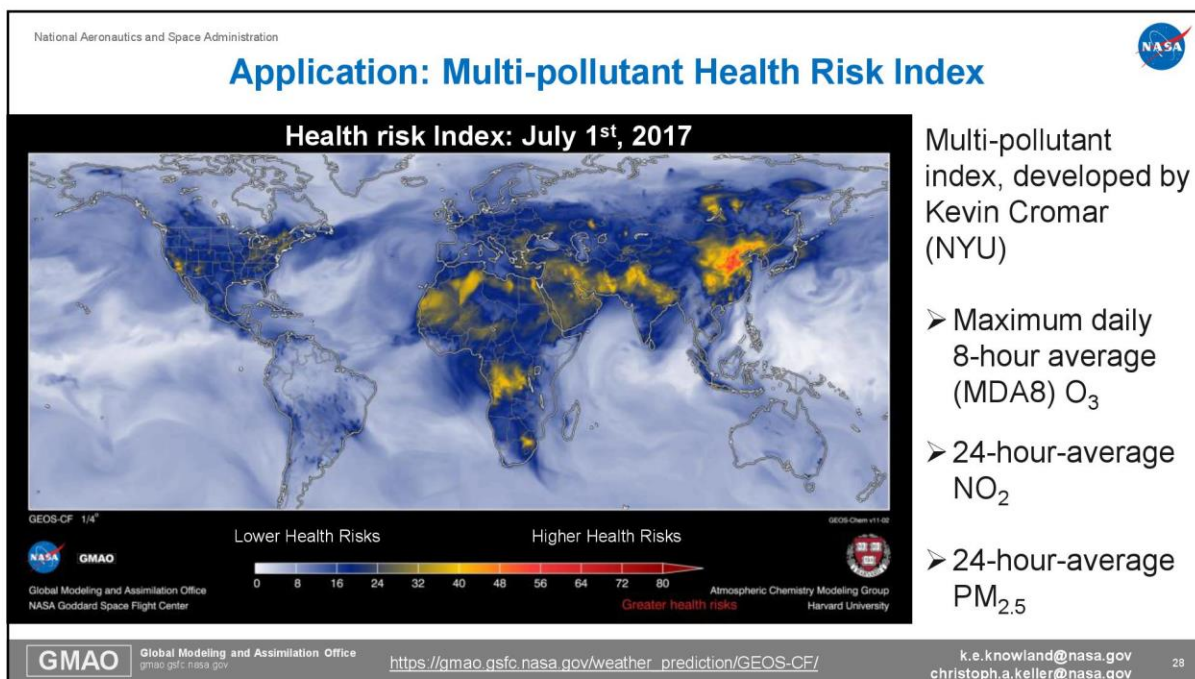
k.e.knowland@nasa.gov  
christoph.a.keller@nasa.gov

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## GMAO Mission Support for Field Campaigns



Global Modeling and Assimilation Office **GMAO**

Weather Mission Support CF Reanalysis Forecast

**Navigation**

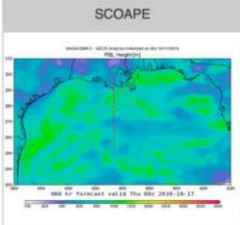
- FLUID Overview
- Contact

**Active Missions**

- SCOAPE
- FIREX-AQ
- CAMP2EX
- AEOLUS-CALVAL
- SOCRATES
- ORACLES
- EPOCH
- ACE-ENA
- ABOVE
- ATOM
- MOSAIC

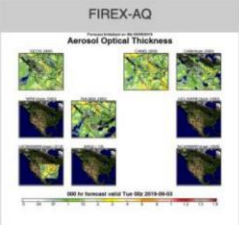
### GMAO Active Mission Support

SCOAPE

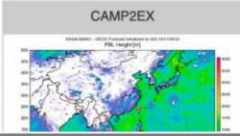


FIREX-AQ

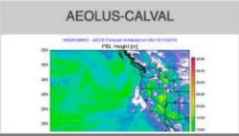
Aerosol Optical Thickness




CAMP2EX



AEOLUS-CALVAL





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## Satellite Coastal and Oceanic Atmospheric Pollution Experiment (SCOAPE)

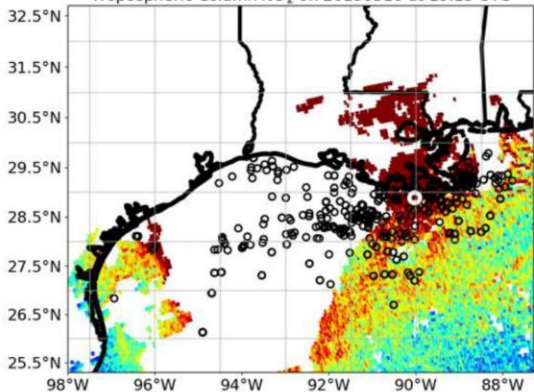


"Can we use NASA resources to monitor off shore sources in the Gulf of Mexico which are impacting air quality onshore?"




- **The Gulf of Mexico is a complex region for both meteorology and composition**
- **Frequent cloud cover**
- **Onshore and off shore flow**
  - **Recirculation of onshore pollution**

Tropospheric Column NO<sub>2</sub> on 20190510 at 19:13 UTC



Tropospheric Column NO<sub>2</sub> (DU)



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## Satellite Coastal and Oceanic Atmospheric Pollution Experiment (SCOAPE)

**BOEM** *"Can we use NASA resources to monitor off shore sources in the Gulf of Mexico which are impacting air quality onshore?"*

NASA/GMAO - GEOS CF Forecast Initialized on 12z 05/15/2019  
Surface O<sub>3</sub>

- May 2019 measurement campaign
- GSFC support
  - Mission planning
  - Post mission analysis

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## SCOAPE "Tools"

- Post mission: Ongoing analysis by BOEM depends on the "tools" available and knowledge to use them

1. Model
2. Satellite
3. Trajectories

**BOEM**  
BUREAU OF OCEAN ENERGY MANAGEMENT

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 christoph.a.keller@nasa.gov

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## Where to find GEOS-CF

Maps available at [fluid.nccs.nasa.gov/cf](https://fluid.nccs.nasa.gov/cf)

Global Modeling and Assimilation Office **GMAO**

Weather | Mission Support | **CF** | Reanalysis | Seasonal

NASA/GMAO - GEOS CF Forecast initialized on 12z 08/13/2019

Surface NO<sub>2</sub>

018 hr forecast valid Wed 06z 2019-08-14

0 0.1 0.2 0.5 1 2 5 10 15 20 50


**FLUID is a mobile-friendly website**

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Global Modeling and Assimilation Office  
[https://gmao.gsfc.nasa.gov/weather\\_prediction/GEOS-CF/](https://gmao.gsfc.nasa.gov/weather_prediction/GEOS-CF/)  
 k.e.knowland@nasa.gov  
 christoph.a.keller@nasa.gov

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## Where to find GEOS-CF


GEOS-CF 3D and 2D data since 1 January 2018:

- 1) Download from the NASA Data Portal  
<https://portal.nccs.nasa.gov/datashare/gmao/geos-cf/>
- 2) Remote access through OPeNDAP  
<https://opendap.nccs.nasa.gov/dods/gmao/geos-cf>

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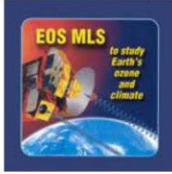

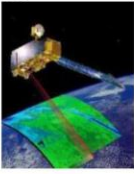
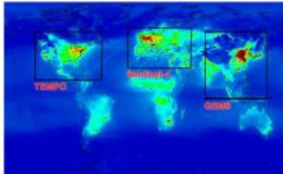
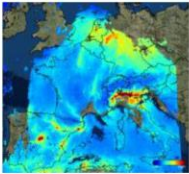


## Summary

- GEOS-CF produces daily global air quality forecasts at 25km (16 miles) horizontal resolution in near-real time !!
- Output available at [fluid.nccs.nasa.gov/cf](http://fluid.nccs.nasa.gov/cf)

**Under development:**

- Assimilation system for trace gases (O<sub>3</sub>, NO<sub>x</sub>, CO, & others)



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**Thank you!**

<https://svs.gsfc.nasa.gov/>

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# 附錄六、"Ramboll Shair: Integrating real-time sensor measurements and regional/local-scale models in Richmond, California"



## Ramboll Shair: Integrating realtime sensor measurements and regional/local-scale models in Richmond, California


Justin Bandoro, Julia Luongo, Mike Dvorak, Drew Hill,  
Tasko Olevski, Kurt Richman, Chris Emery, Greg Yarwood,  
Shari Libicki

18<sup>th</sup> CMAS Conference  
October 22<sup>nd</sup>, 2019

Ramboll Shair  
Make sense of your air quality data



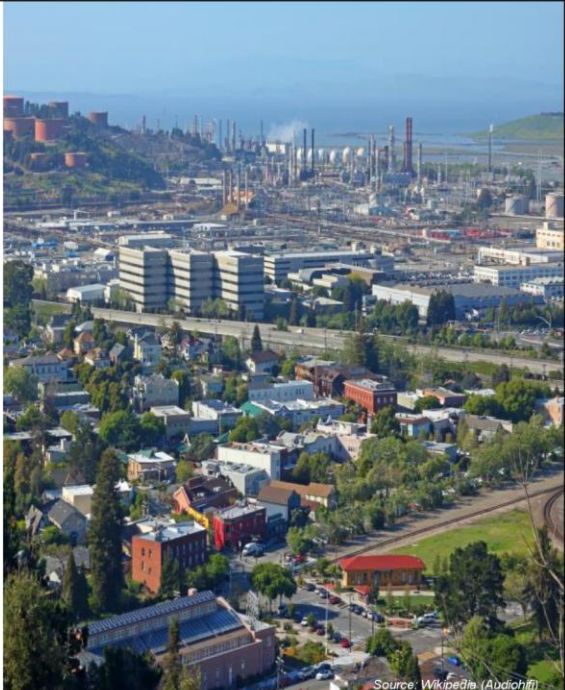
1



## Outline

- **Background**
  - AB617 overview
  - Air Rangers Community Air Protection
- **Ramboll Shair Model**
  - Emission estimates with near-realtime traffic emissions
  - CAMx regional and Shairstreet roadway models
  - Sensor integration
- **Feedback and Next Steps**
  - Communicating results to the Richmond community


Ramboll Shair  
Make sense of your air quality data



Source: Wikipedia (Abdichiff)

2

1



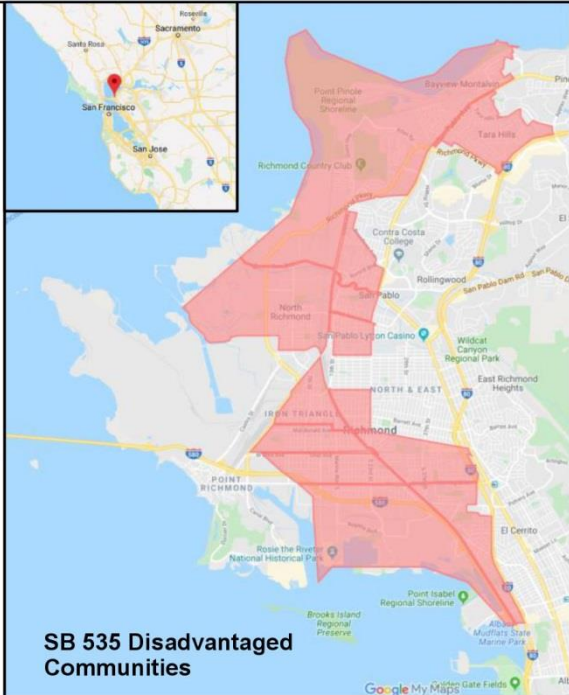
## Background

**AB617**

- California Assembly bill
- Funds community level air measurement, education, and emissions reductions
- Targets disadvantaged communities

**AB617 Funded Project in Richmond, CA ("Air Rangers")**

- Fill PM and NO<sub>2</sub> monitoring gaps
- Attribute hotspots to sources
- Communicate actionable data to the public in **near-realtime**
- Facilitate healthy outdoor recreation
- Local work force development for disadvantaged youth



**SB 535 Disadvantaged Communities**

3







## Air Rangers Community Air Protection Grant

**Action Plan**

Converge strategic partners

Site **50** PM & NO<sub>2</sub> sensors;  
take **70** toxic metals samples

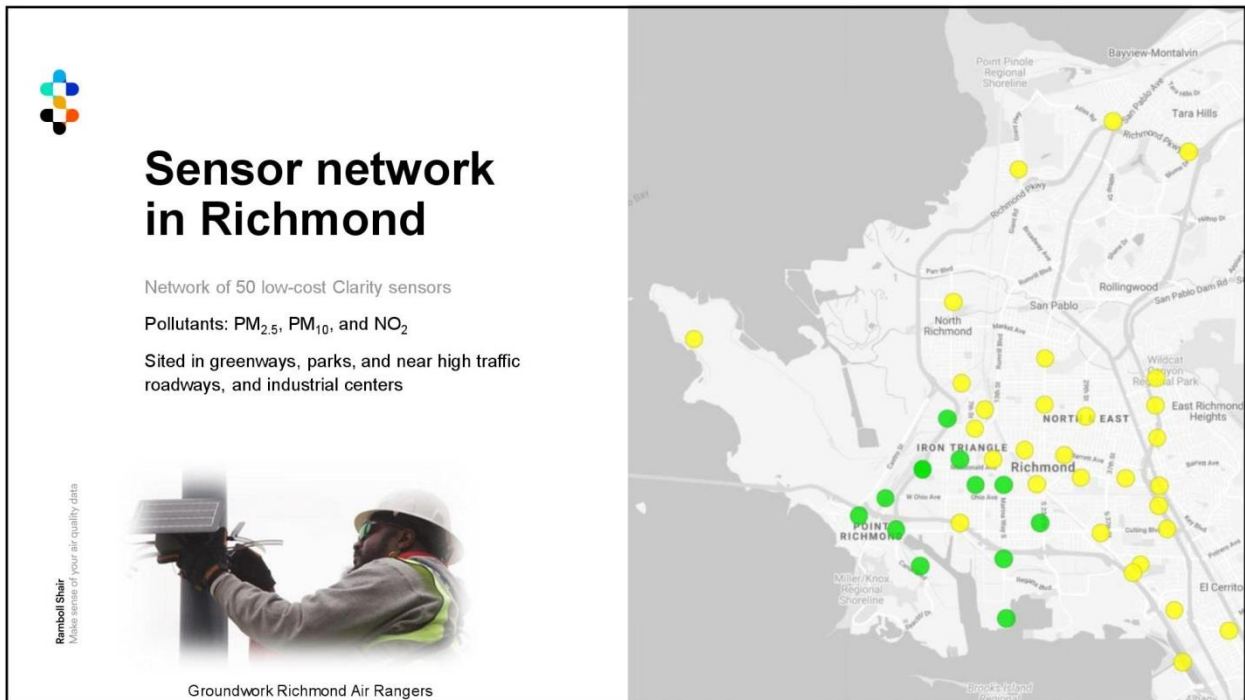
**Communicate** hyper-local air quality to the public, **hourly**

**Hourly hotspot** identification and **source attribution**





4



**Sensor network in Richmond**

Network of 50 low-cost Clarity sensors

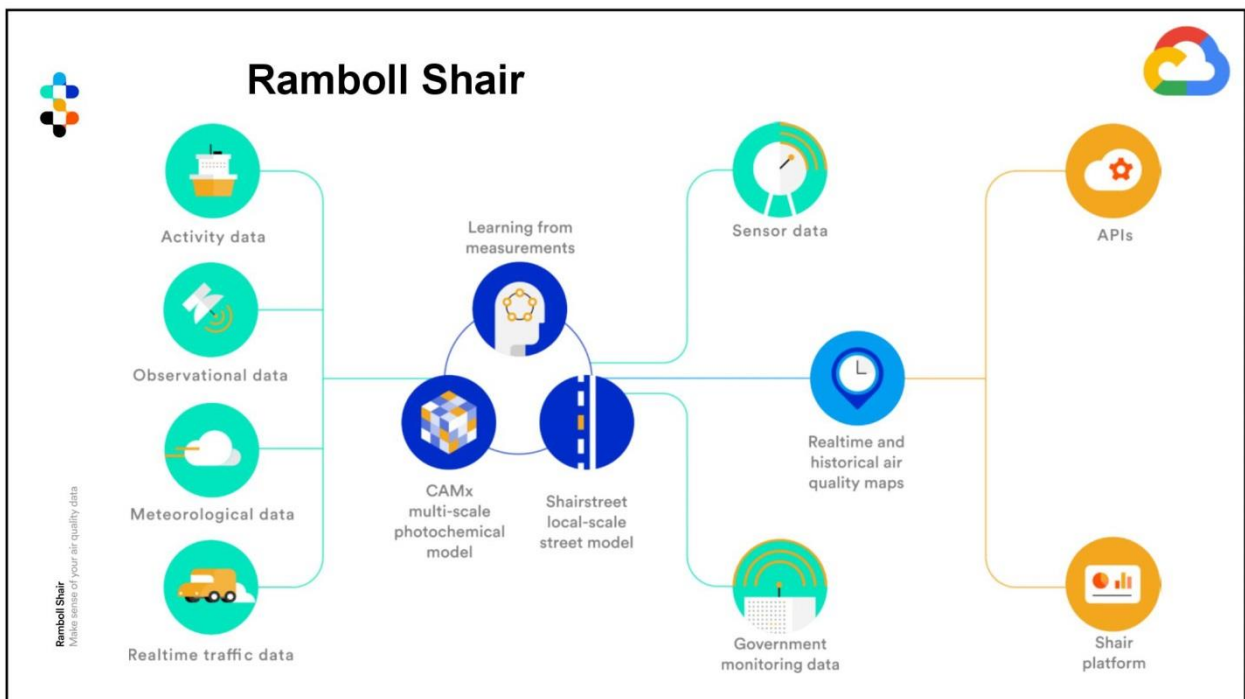
Pollutants: PM<sub>2.5</sub>, PM<sub>10</sub>, and NO<sub>2</sub>

Sited in greenways, parks, and near high traffic roadways, and industrial centers

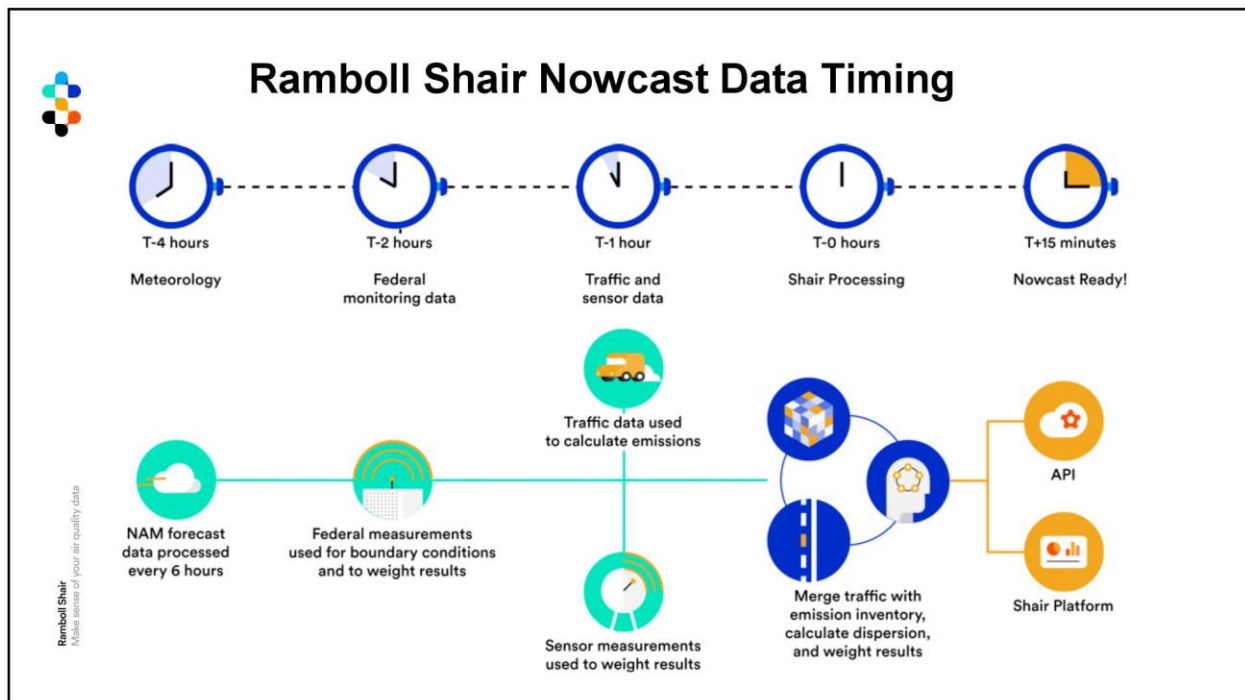
Groundwork Richmond Air Rangers

Map labels: Point Pinole Regional Shoreline, Bayview-Montalvin, Tara Hills, Richmond Park, North Richmond, San Pablo, Rollingwood, East Richmond Heights, Iron Triangle, Richmond, Point Richmond, Miller/Knox Regional Shoreline, El Cerrito.

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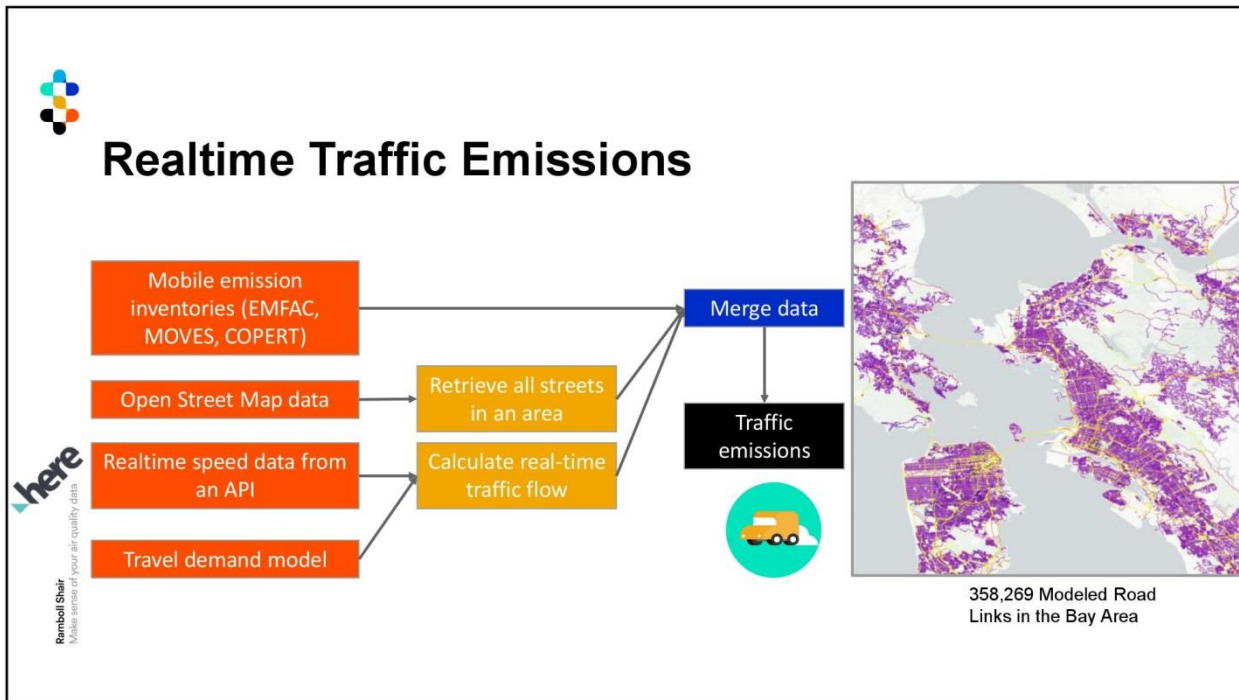
7

## CAMx Modeling

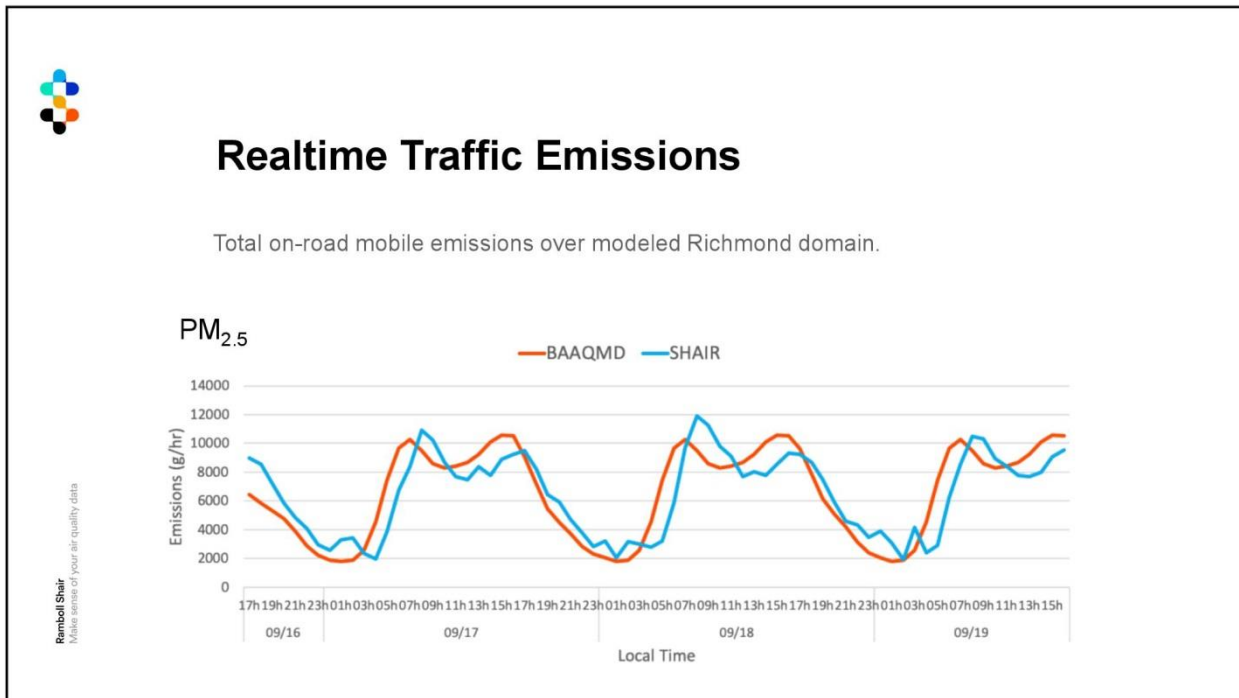
- **Meteorology**
  - WRF NAM forecast refreshed every 6 hours
  - Nested grids: 9km, 3km, and 1km resolution
- **CAMx Grids**
  - 1km and 200m resolution
  - Captures region's most influential sources
- **Chemistry and Deposition**
  - Fast chemical scheme for O3-NO-NO2
  - Wet and dry deposition with ZHANG03 model
- **Source Attribution**
  - Tag emissions by source category for source apportionment
- **Modeled every hour**
  - Latest emissions combined with previous hour dispersion results

Ramboll Shair  
Make sense of your air quality data

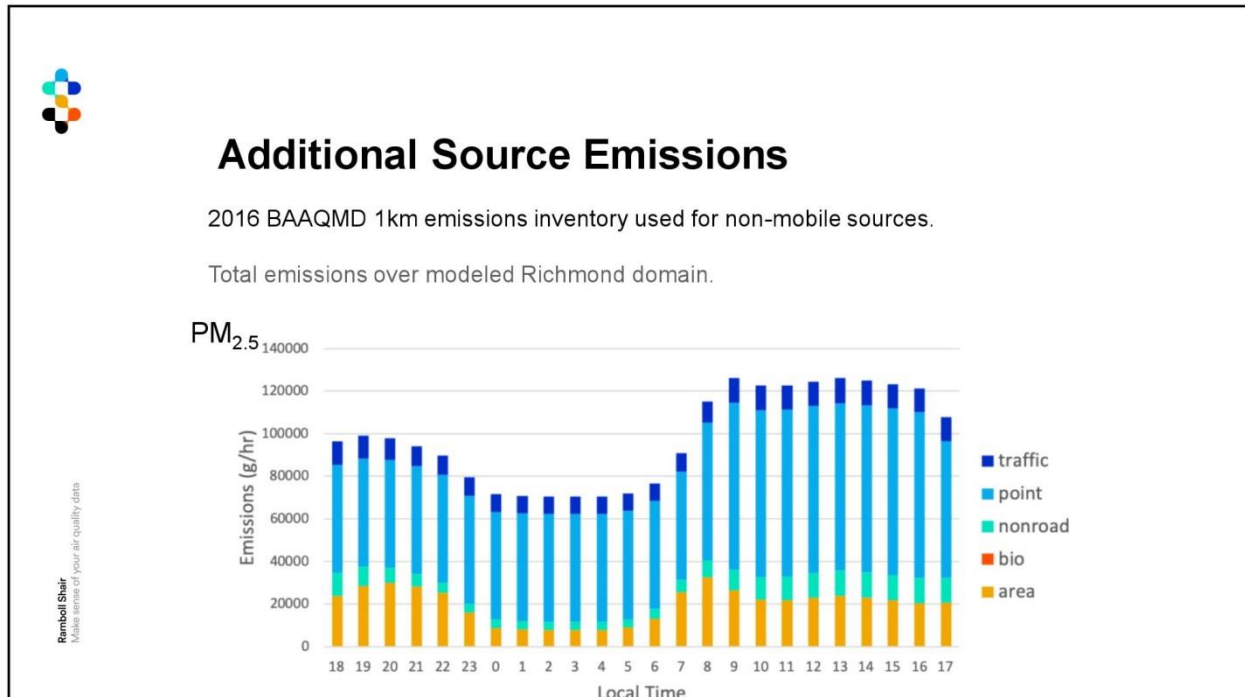
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**Shairstreet Modeling**

Fine scale resolution of road emissions

**1. Street canyon model**


Calculates additional recirculation contribution for road links that have street canyon-like geometry, using wind speed and direction at top of the canyons. Takes into account vehicle generated mechanical turbulence.

Modeled buildings in Richmond (view from Chanslor Ave and 9<sup>th</sup> St)

Source: Berkowicz 1997.

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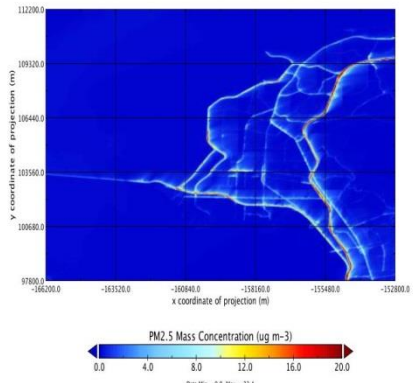
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## Shairstreet Modeling

Fine scale resolution of road emissions

- 1. Street canyon model**  
Calculates additional recirculation contribution for road links that have street canyon-like geometry, using wind speed and direction at top of the canyons. Takes into account vehicle generated mechanical turbulence.
- 2. Line source dispersion model**  
Road links are modeled as line sources based on Venkatram and Horst 2006 approximate Gaussian dispersion solution at **10x10m resolution**.
- 3. NOx chemistry model**  
Simplified NO-NO<sub>2</sub>-O<sub>3</sub> model based on During et al. 2011, with parameterized mixing times based on street-canyon geometry.




$$\text{NO} + \text{O}_3 \rightarrow \text{NO}_2 + \text{O}_2 \quad (k)$$

$$\text{NO}_2 \rightarrow \text{NO} + \text{O}_3 \quad (J)$$

Reaction and photolysis rates,  $k$  and  $J$ , taken from CAMx 200m grid along with background NO<sub>x</sub> and O<sub>3</sub>

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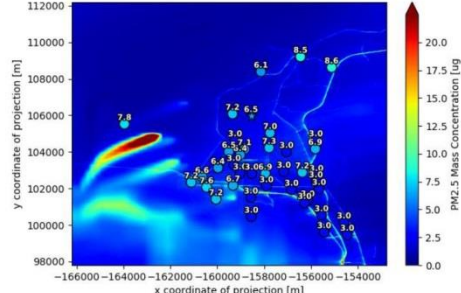
## Sensor Integration

- Latest Clarity sensor and BAAQMD monitor data pulled
- Measurement data undergoes internal quality control checks and adjustment
- Objective analysis via Barnes interpolation is used to assimilate sensor measurements

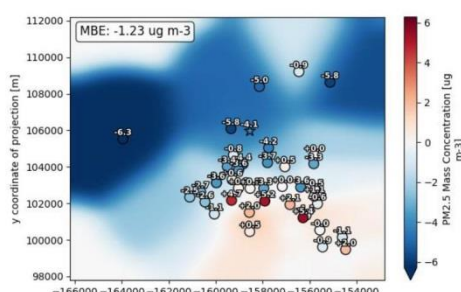
$$\varphi(\vec{x}) = \sum_{i=1}^{N_d} w_i d_i, \quad w_i = \frac{r_i^{-2}}{\sum_i r_i^{-2}}$$

October 12<sup>th</sup> 2019, 7:00pm PDT


Raw model + sensors



Interpolated model error



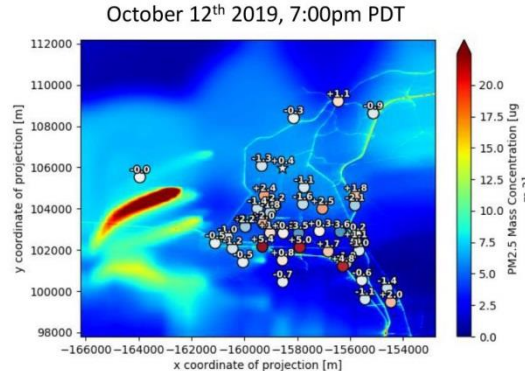
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## Sensor Integration


Assimilated map



- Latest Clarity sensor and BAAQMD monitor data pulled
- Measurement data undergoes internal quality control checks and adjustment
- Error between model and observations are calculated
- Objective analysis via Barnes interpolation is used to assimilate sensor measurements

$$\varphi(\vec{x}) = \sum_{i=1}^{N_d} w_i d_i \quad , w_i = e^{-\frac{r_i^2}{2R^2}} / \sum_i e^{-\frac{r_i^2}{2R^2}}$$


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## Communicating Results


Web/phone app for Richmond is being developed





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## Next Steps

**In the coming months, Shair output will be put to use by stakeholders**

**Groundwork Richmond**

- Targeted tree planting as an air pollution intervention

**City of Richmond**


- Integrate hourly Shair output into the City's web data dashboard
- Investigate air quality data-driven land use planning activities
- Communicate information to the public

**Richmond citizens**

- Engage with their environment
- Plan healthier outdoor activities
- Take precautionary actions during localized pollution events


**Richmond-San Pablo Community Air Monitoring Steering Committee**

- Identify major emissions sources
- Inform their official Community Emissions Reduction Plan (December 2020)



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## “


We're excited to partner with Ramboll Shair to better understand air quality on a community scale. We need more information to better direct our resources and achieve our climate action goals.

*- Adam Lenz, Environmental Services Manager, City of Richmond*



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# Thank you

Justin Bandoro  
[jbandoro@ramboll.com](mailto:jbandoro@ramboll.com)

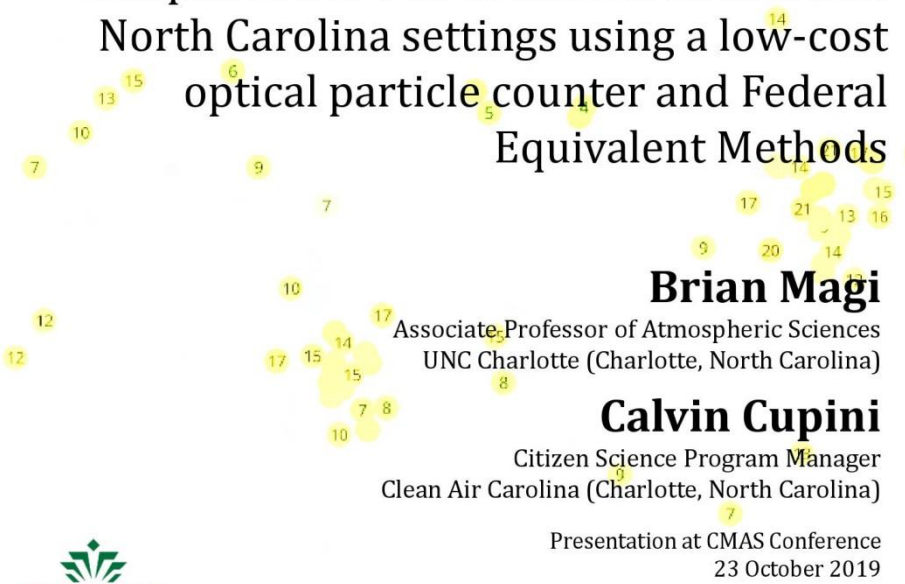
Get a copy of our whitepaper:  
<https://ramboll-shair.com>



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附錄七、"Comparison of PM<sub>2.5</sub> measured in urban North Carolina settings using a low-cost optical particle counter and Federal Equivalent Methods"


**Comparison of PM<sub>2.5</sub> measured in urban North Carolina settings using a low-cost optical particle counter and Federal Equivalent Methods**



**Brian Magi**  
Associate Professor of Atmospheric Sciences  
UNC Charlotte (Charlotte, North Carolina)

**Calvin Cupini**  
Citizen Science Program Manager  
Clean Air Carolina (Charlotte, North Carolina)


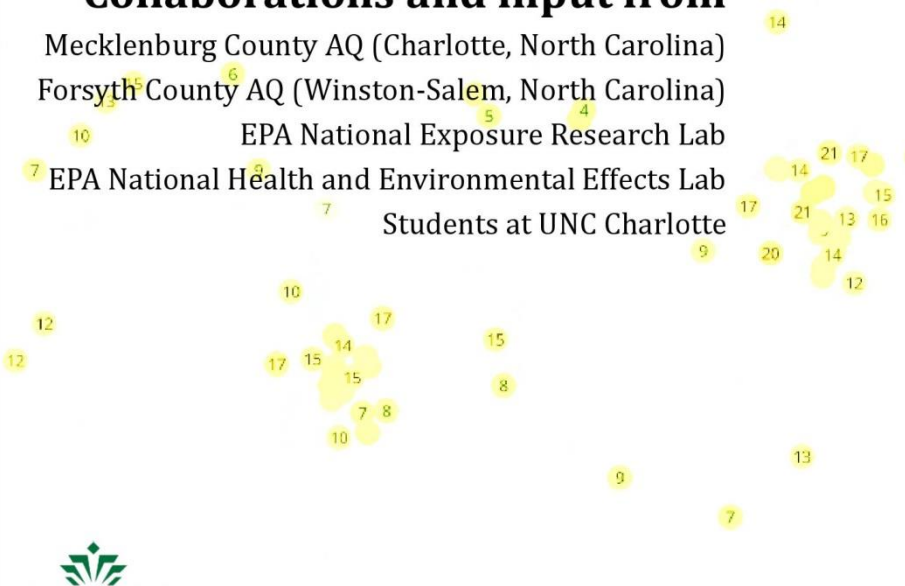
Presentation at CMAS Conference  
23 October 2019



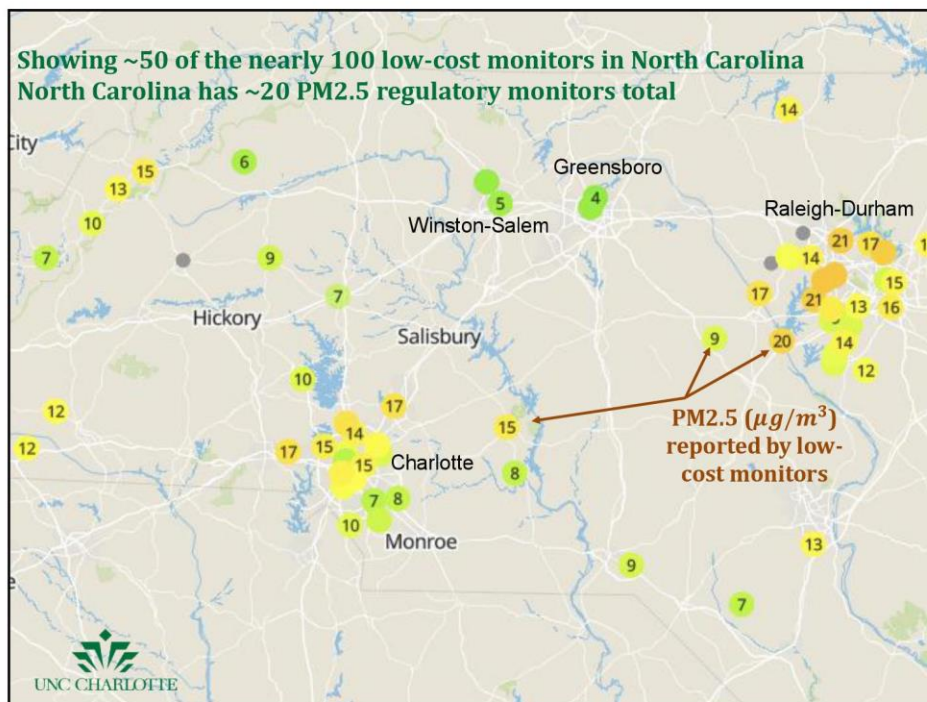
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**Collaborations and input from**

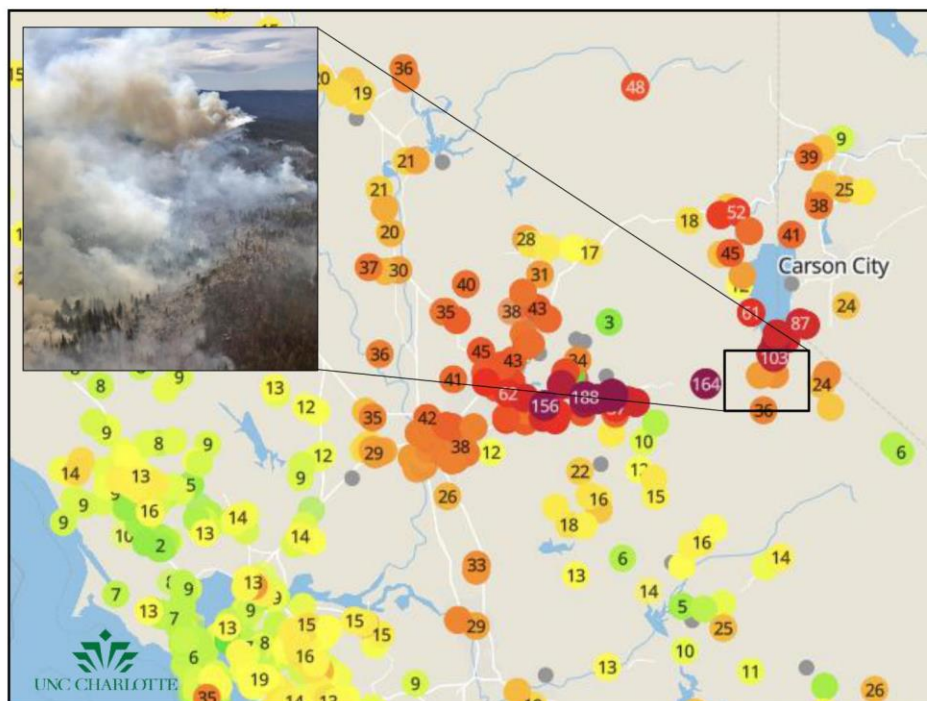
- Mecklenburg County AQ (Charlotte, North Carolina)
- Forsyth County AQ (Winston-Salem, North Carolina)
- EPA National Exposure Research Lab
- EPA National Health and Environmental Effects Lab
- Students at UNC Charlotte



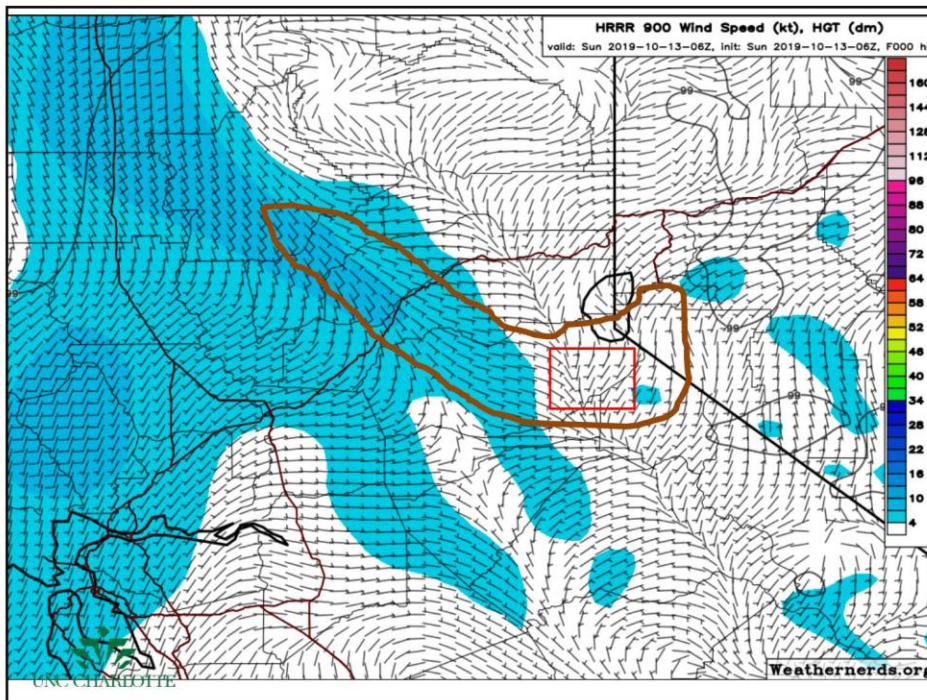
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Purple Air **PA-II** sensor package

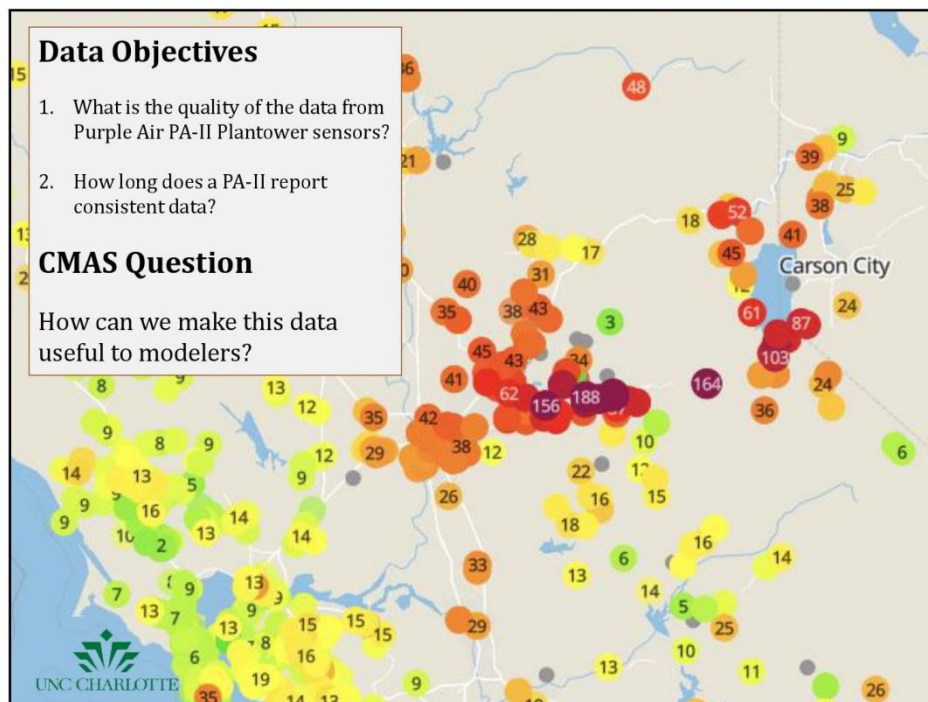
Measures number concentration using 2 Plantower PMS5003 sensors. Reports number and mass concentrations.

Also measures temperature and relative humidity

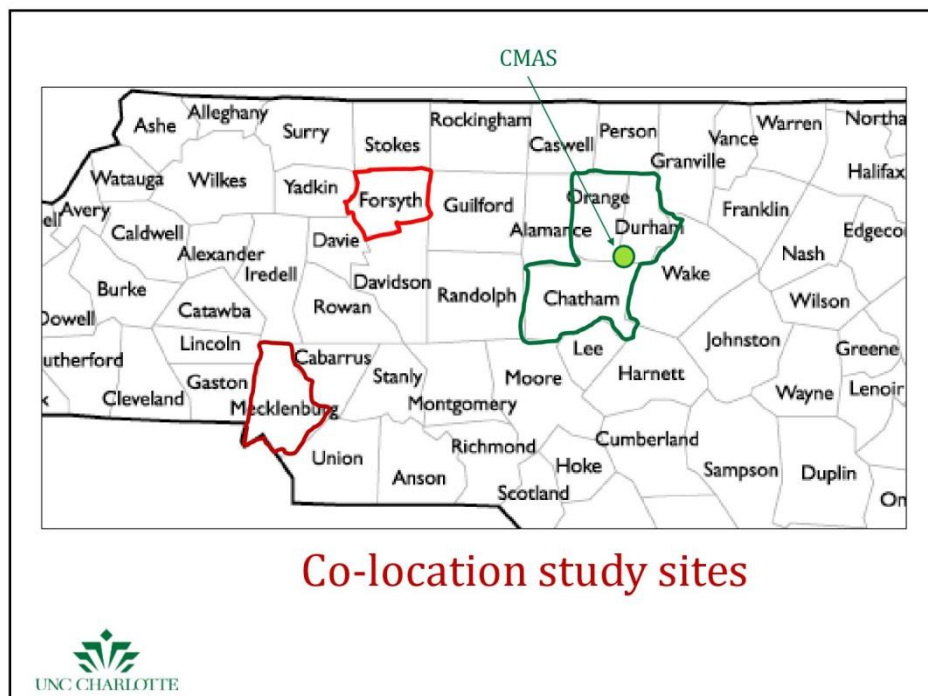
UNC CHARLOTTE

Good description at <https://www.aqmd.gov/aq-spec/product/purpleair-pa-ii> but also at Zamora et al 2019 *Environmental Science & Technology* <https://doi.org/10.1021/acs.est.8b05174>

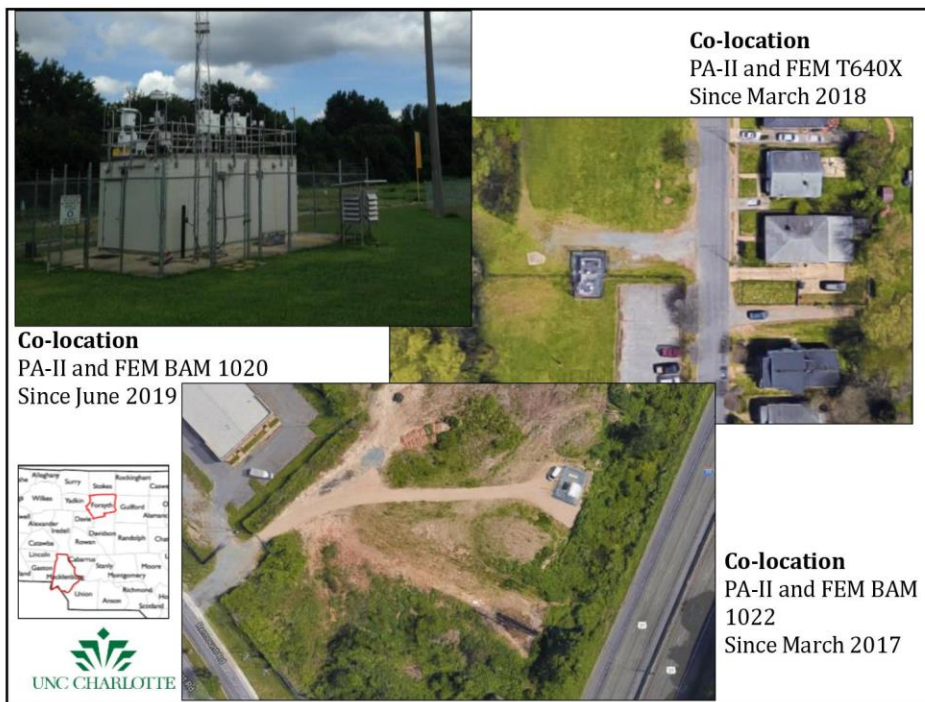
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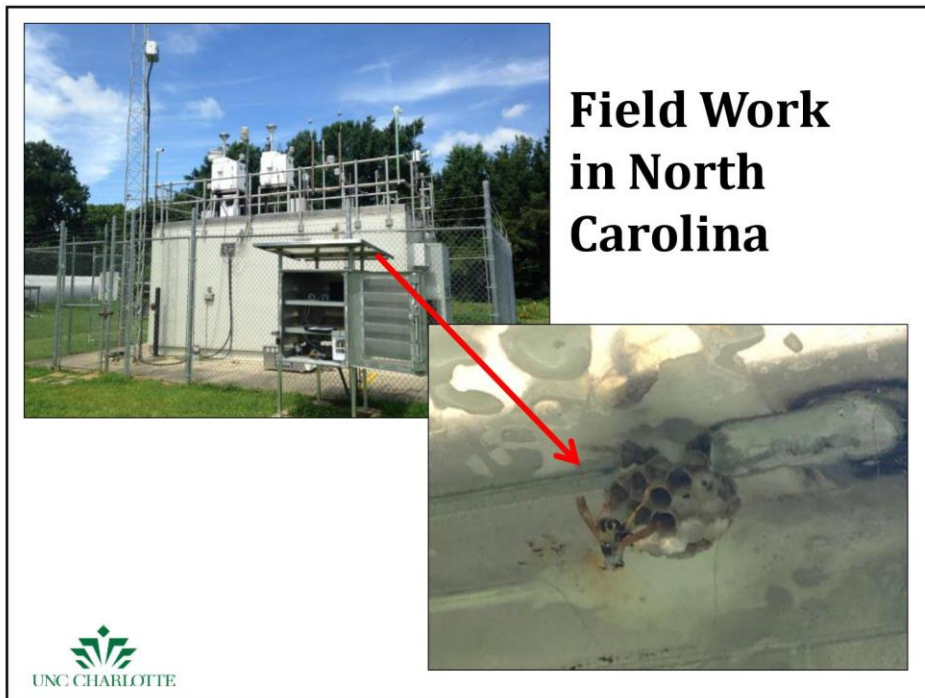
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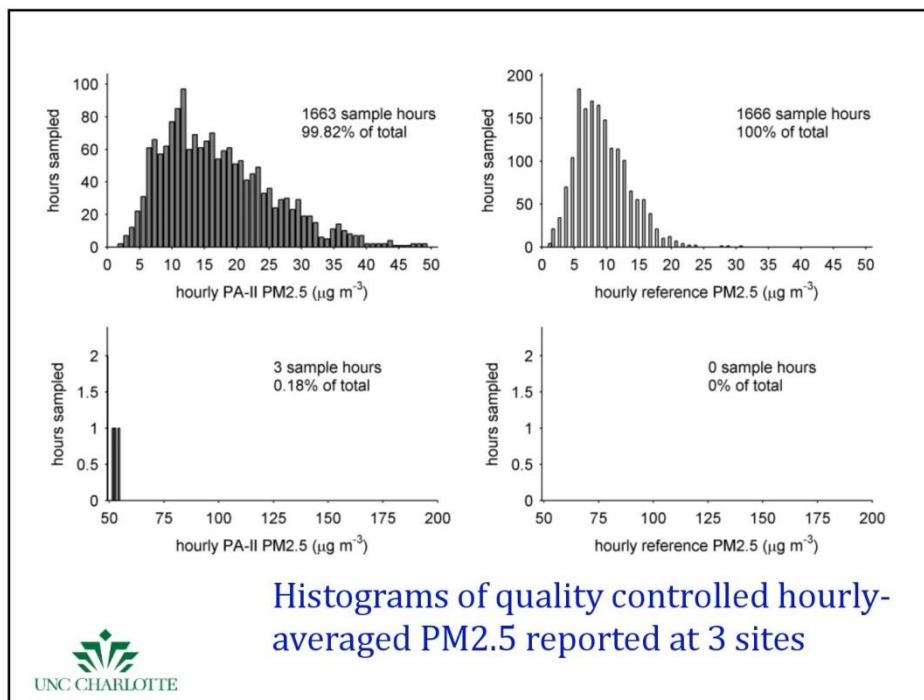
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## Quality control and data processing

1. Collect raw sub-hourly data
2. QC Step 1 (non-physical reporting) ~5% of raw data discarded
3. QC Step 2 (repeating artifact)
4. Hourly average the sub-hourly
5. Discard values less than limit of detection ~20% of remaining data discarded
6. Discard values greater than limit of sampling
7. Use remaining "uncorrected" PA-II data as input to a linear regression

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## Correction of PA-II PM2.5 data using a linear regression

$$PM2.5_{PAII} = a_0 + a_1 PM2.5_{ref} + a_2 RH + a_3 T$$

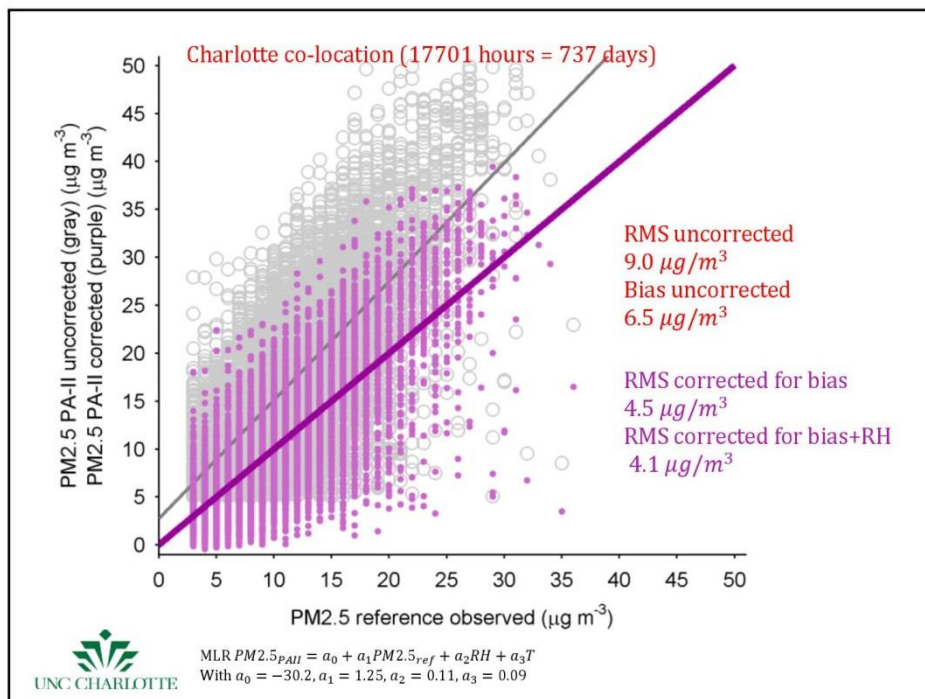
bias correction
RH correction
T correction

$$PM2.5_{PAII} = b_0 + b_1 PM2.5_{ref}$$

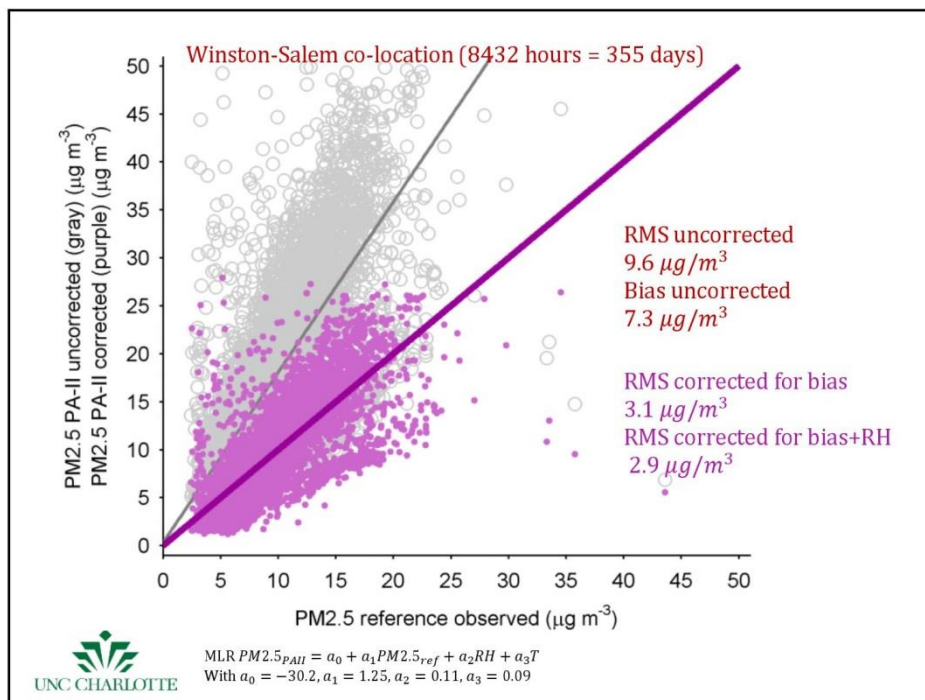
bias correction



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Comparison of all FEM co-location hourly-averaged results

Site Sensor A	N (hours)	Bias before ( $\mu\text{g}/\text{m}^3$ )	RMS before ( $\mu\text{g}/\text{m}^3$ )	RMS after ( $\mu\text{g}/\text{m}^3$ )
Remount1	13964	5.3	7.8	4.2
Remount2	2892	9.8	12	4.5
Forsyth	8432	7.3	9.6	2.9
Garinger1	1593	8.1	10	3.1
Garinger2	1435	7.1	9.5	3.2

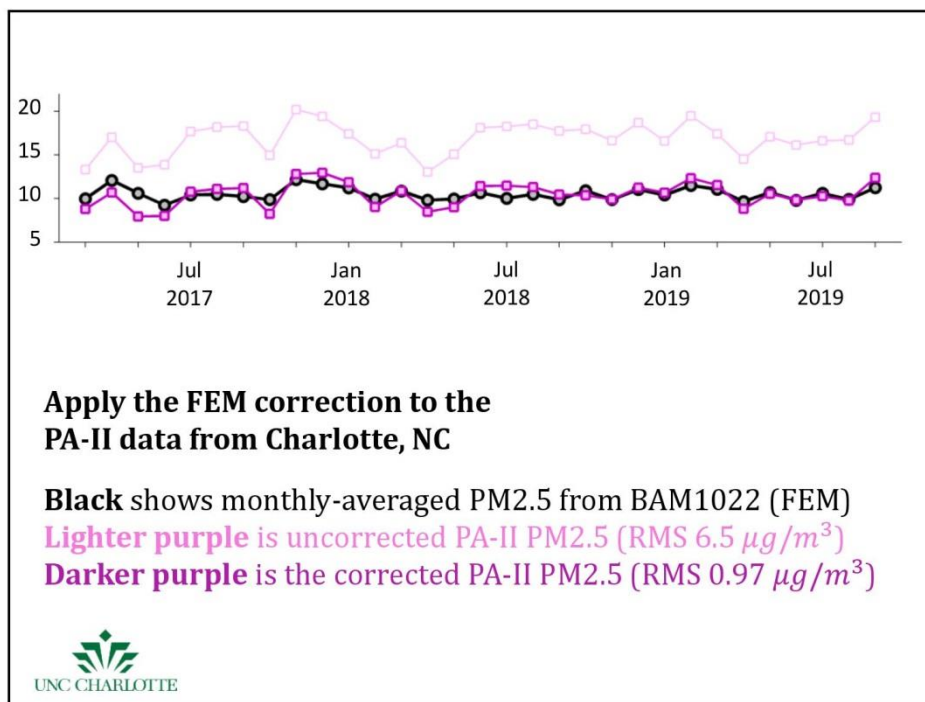
  

Site Sensor B	N (hours)	Bias before ( $\mu\text{g}/\text{m}^3$ )	RMS before ( $\mu\text{g}/\text{m}^3$ )	RMS after ( $\mu\text{g}/\text{m}^3$ )
Remount1	17701	6.5	9	4.1
Remount2	2882	8.2	11	3.9
Forsyth	8312	6.8	8.9	2.6
Garinger1	1611	7.8	9.9	3.1
Garinger2	1446	7.8	9.9	3.2

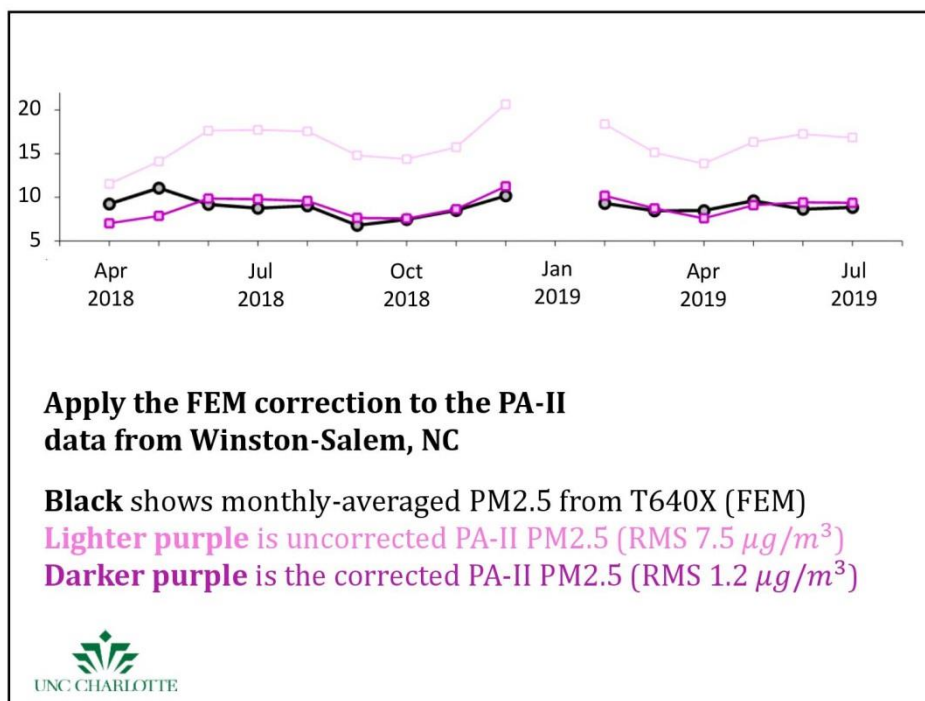
Average bias  $7.5 \pm 1.2 \mu\text{g}/\text{m}^3$   
 Average RMS before  $9.8 \pm 1.2 \mu\text{g}/\text{m}^3$   
**Average RMS after  $3.5 \pm 0.6 \mu\text{g}/\text{m}^3$**

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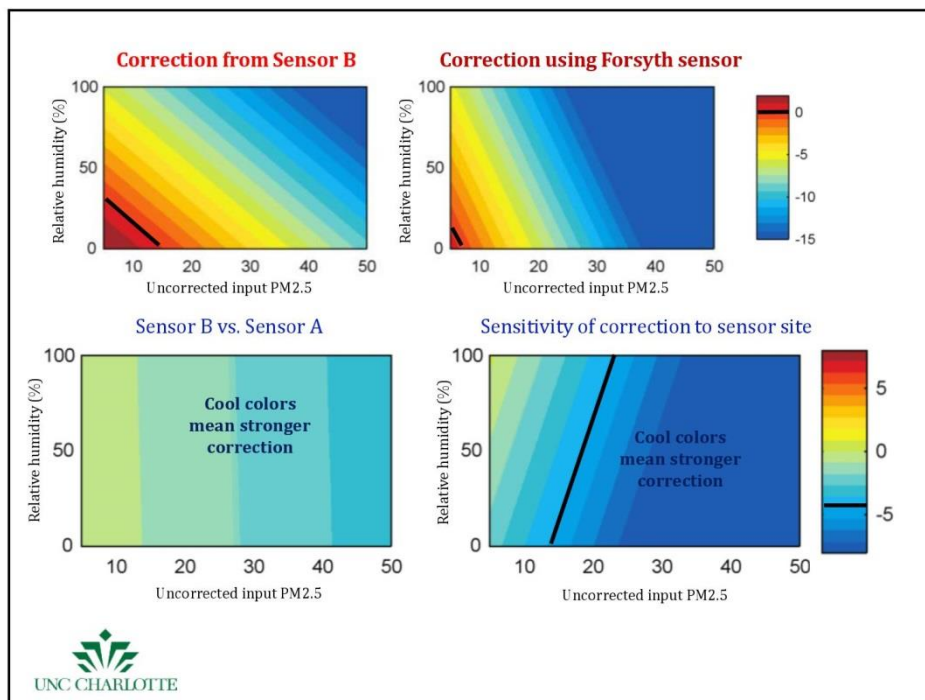
2020/1/12



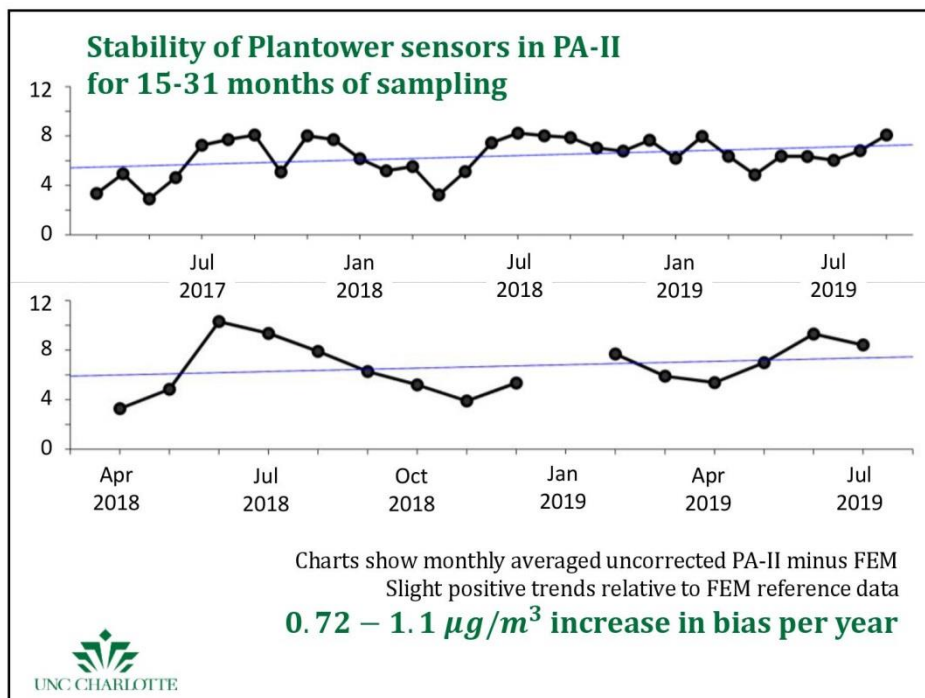
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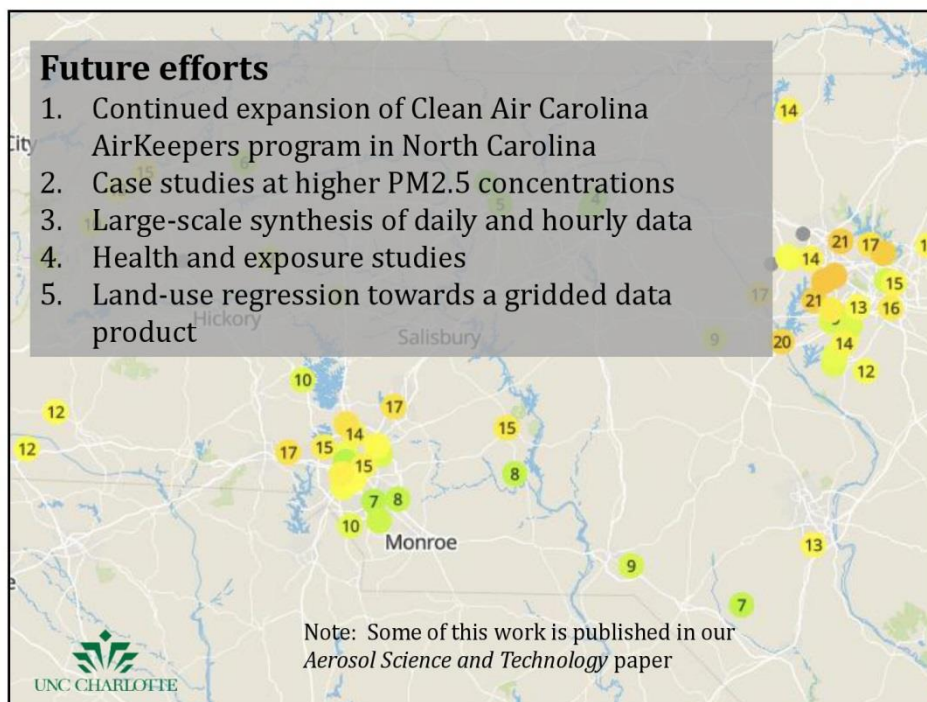
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# 附錄八、" Assessment of a weather pattern-dependent bias correction method to the PM2.5 prediction in Taiwan "

## Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan

Fang-Yi Cheng<sup>1</sup>, Chih-Yung Feng<sup>2</sup>, Zhih-Min Yang<sup>2</sup>, Chia-Hua Hsu<sup>1,3</sup>, Shuenn-Chin Chang<sup>4</sup>, Yi-Chun Tsal<sup>4</sup>  
<sup>1</sup>Department of Atmospheric Sciences, National Central University, Taiwan; <sup>2</sup>Manysplendid Infotech, Taipei, Taiwan  
<sup>3</sup>Department of Mechanical Engineering, University of Colorado, Boulder, CO, USA  
<sup>4</sup>Environmental Protection Administration, Taipei, Taiwan

### Introduction

A real-time air quality forecasting system (namely (AQF) using Weather Research and Forecasting (WRF) meteorological model and Community Multiscale Air Quality (CMAQ) model version 5.2 was developed in Taiwan. A bias correction method is typically applied to remove the systematic biases due to the unresolved meteorological dynamics and emission patterns. We developed a bias-correction method, which combines a cluster-analysis-based synoptic weather classifications, and considers the systematic PM<sub>2.5</sub> biases associated with each weather type. This study investigates the performance of the AQF system and the capability of the bias correction technique on improving the PM<sub>2.5</sub> predictions.

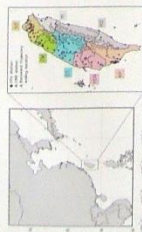


Fig. 1. Black points identify locations of EPA surface stations. Blue triangles identify the locations of CMAQ surface stations. The seven AQZs (NT, CM, CT, YCN, KP, YI and HD) are identified by name. The contour line indicates the mountain range (CMR).

### Bias Correction Method

**Model Output Statistics (MOS)** MOS is a type of statistical post-processing techniques which considering the bias between station observation (predictand) and variables from numerical model forecasts (predictors). The predictors currently are selected from CMAQ and WRF forecasts of four nearest grids surrounding the observation. The statistical method used for correcting CMAQ forecast is multiple linear regression with 30 days archived model forecast output through a forward selection procedure.

### Adjusted bias correction method

Further classification of bias characteristics is developed based on synoptic weather patterns (Hsu and Cheng, 2019).  
 Fig. 2 presented composite map of mean sea level pressure and surface wind fields for three classified synoptic weather patterns. C1, Taiwan is affected by the northeasterly monsoonal flow; C2, Taiwan is affected by the continental high-pressure peripheral circulation; C3, which corresponds to a weak synoptic weather pattern and is affected by a weak easterly flow.

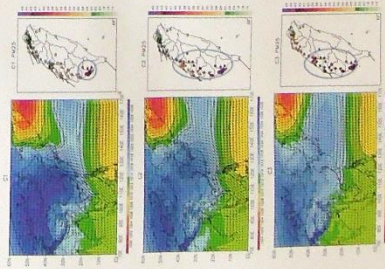


Fig. 2. Composite plot of SLP (hPa) (and surface wind vectors (m s<sup>-1</sup>)) for C1-C3 clusters (left) and associated observed PM<sub>2.5</sub> distributions (right).

### Air Quality Forecast System

The WRF version 3.8.1 and Community Multiscale Air Quality (CMAQ) model, version 5.2, was applied to provide the PM<sub>2.5</sub> forecasts. The initial conditions for all domains are from the CMAQ forecasting output of the previous day, and boundary conditions of D1 were established based on the default background profile datasets. The boundary conditions of D2 were provided from D1. The anthropogenic emissions in D1 were obtained from the 2010 Model Inter-Comparison Study for Asia (MICS-Asia) emissions inventory and the anthropogenic emissions in D2 were adopted from TEDS-9.0 with reference year 2013. Biogenic emissions were provided by the model of emissions of gases and aerosols.

WRF version	WRF3.8.1
DOI (Fig. 1)	D02 (Fig. 1)
resolution	15-km
CMAQ version	CMAQV5.2
Anthropogenic emissions	MICS-Asia + TEDS9.0
Chemical mechanism	CB05S1-AE6
Biogenic emissions	MEGAN2.04

Table 1. Model configuration of WRF-CMAQ air quality forecasting system

### Distributions of Bias Characteristics

The biases of the wind speed (Fig. 3) and PM<sub>2.5</sub> concentrations (Fig. 4) from WRF-CMAQ forecast system associated with C1, C2 and C3 clusters are developed for the 7-months forecasting period (2018 October to 2019 April) when the air pollution problem is more serious compared to that in the summer season. There is common variability within the same weather pattern and distinct variability between different weather patterns.

There is consistent wind speed overestimation (Fig. 3). The overestimation can be attributed to too strong synoptic prevailing wind and the local land-sea breeze.

There is underestimation of PM<sub>2.5</sub> in southwestern Taiwan in C1, C2 and C3 clusters (Fig. 4). The underestimation is the most apparent during the daytime hours. The overestimation of PM<sub>2.5</sub> is seen in the CT AQZ due to the application of the emission control strategy applied in the recent years. A significant underestimation of the PM<sub>2.5</sub> is seen in YCN AQZ in C3 cluster, which is associated with the worst PM<sub>2.5</sub> problem.

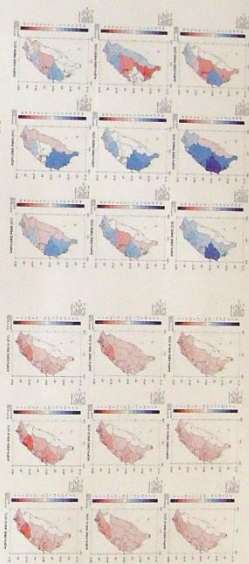


Fig. 3. Forecast wind speed biases.

### Evaluation of forecast PM<sub>2.5</sub> components

Fig. 5 depicts the comparison of PM<sub>2.5</sub> components between forecast and observations in C3 cluster. Douliu and Chiayi are located in YCN AQZ. The serious PM<sub>2.5</sub> underestimation problem is due to the deficiencies in forecasting organic carbon and sulfate species. Underestimation of the OC is due to the deficiency of simulating the SOA formation. Underestimation of the sulfate can be attributed to emission problem.

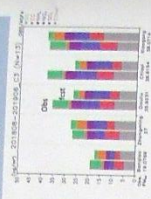


Fig. 5. Comparison of PM<sub>2.5</sub> components between observation (left bar) and forecast (right bar) in C3 cluster.

### Assessment of Bias-Correction Method

The new bias-correction method is applied to correct forecast PM<sub>2.5</sub> bias. There is a significant reduction of the RMSE of the forecast PM<sub>2.5</sub> (Fig. 5) after the application of the new bias-correction method, which also outperforms the DMOS technique.

The improvement is the most effective in southwestern Taiwan where the high PM<sub>2.5</sub> concentrations are frequently occurred. The assessment of the 2019 March CMAQ forecast indicates that the new bias-correction method is able to reduce the bias of the raw model forecast PM<sub>2.5</sub> concentrations (Fig. 6).

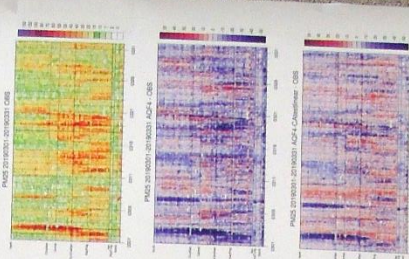


Fig. 6. Observed PM<sub>2.5</sub> concentrations in Taiwan in March 2019 (top); bias of the forecast PM<sub>2.5</sub> from raw model (middle); and bias of forecast PM<sub>2.5</sub> with new bias-correction method (bottom).

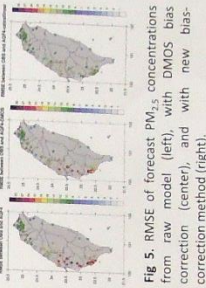


Fig. 5. RMSE of forecast PM<sub>2.5</sub> concentrations from raw model (left), with DMOS bias correction (center), and with new bias-correction method (right).

### Summary and Conclusion

- A WRF-CMAQ model system is developed to forecast the PM<sub>2.5</sub> concentration in Taiwan.
- A new bias-correction method is developed to correct the forecast PM<sub>2.5</sub> bias.
- The new bias-correction method successfully reduces the bias and RMSE of the forecast PM<sub>2.5</sub> concentrations and outperforms the original DMOS technique.

### Reference

Hsu, C.-H. and F.-Y. Cheng, 2019. Synoptic weather patterns and associated air pollution in Taiwan. *Aerosol and Air Quality Research*, 19, 1139-1151.

# Assessment of a weather pattern-dependent bias correction method to the PM<sub>2.5</sub> prediction in Taiwan

Fang-Yi Cheng<sup>1</sup>, Chih-Yung Feng<sup>2</sup>, Zhih-Min Yang<sup>2</sup>, Chia-Hua Hsu<sup>1,3</sup>, Shuenn-Chin Chang<sup>4</sup>, Yi-Chun Tsai<sup>4</sup>

<sup>1</sup>Department of Atmospheric Sciences, National Central University, Taiwan

<sup>2</sup>Manysplendid Infotech, Taipei, Taiwan

<sup>3</sup>Department of Mechanical Engineering, University of Colorado, Boulder, CO, USA

<sup>4</sup>Environmental Protection Administration, Taipei, Taiwan

## Abstract

A real-time air quality forecasting (AQF) system using Weather Research and Forecasting (WRF) meteorological model and Community Multiscale Air Quality (CMAQ) model version 5.2 was developed in Taiwan. There is a systematic over-prediction of the wind speed in western Taiwan. The too strong wind can be attributed to both synoptic prevailing wind and the local sea breeze. The sulfate and organic carbon are under-predicted in western Taiwan while the nitrate is over-predicted in northern Taiwan and under-predicted in southern Taiwan.

To improve the PM<sub>2.5</sub> forecast, a bias correction method based on the dynamic model output statistics (DMOS) is developed to reduce the predicted PM<sub>2.5</sub> bias. It was found that under the influence of the strong northeasterly monsoonal flow, AQF tends to underestimate the PM<sub>2.5</sub> concentration in southwestern Taiwan. On the other hand, when the synoptic weather condition is weak, there is systematic PM<sub>2.5</sub> over-prediction in northern Taiwan and under-prediction in southern Taiwan. As the PM<sub>2.5</sub> biases vary with the synoptic weather patterns, an alternative bias correction method to account for the influence of the synoptic weather types was developed (namely DMOS\_W). A certain number of synoptic weather patterns were identified using the cluster analysis method. The bias correction method considers the systematic PM<sub>2.5</sub> biases associated with each weather type.

The performance of the PM<sub>2.5</sub> prediction is further improved with the approach using the synoptic weather pattern based method (DMOS\_W) compared to that from the original DMOS method. More details will be introduced during the workshop.

**Keywords:** PM<sub>2.5</sub> forecast, bias correction, dynamic model output statistics, synoptic weather patterns.