

Multi-tracer Measurements and Inversions Quantifying California's GHG Emissions Budget

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CH₄ measurement network and emission inventory development
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WRF-STILT development
- JPL: C. Frankenberg
SCIAMACHY CH₄ retrievals
- Caltech-UHouston: S. Newman, B. Leffer
Pasadena Lidar PBL depths

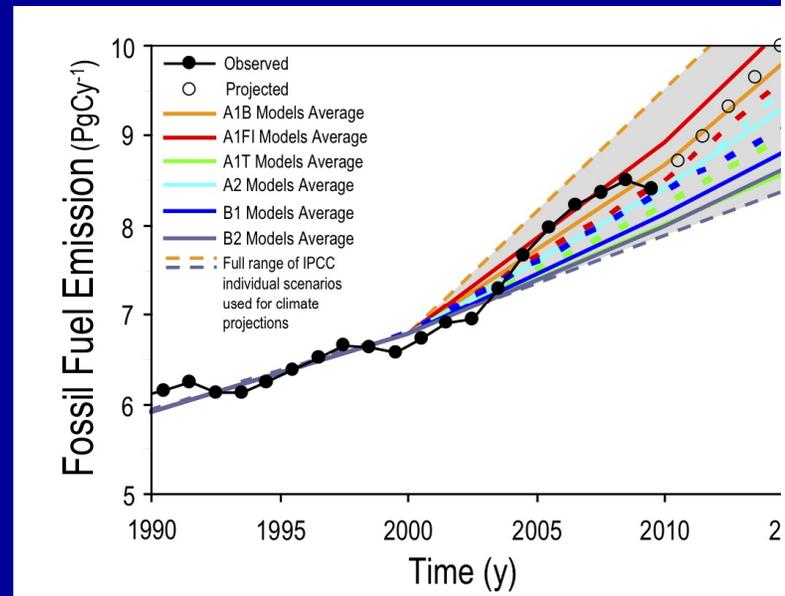
Outline

- Need for GHG emissions control validation
- Initial work for CEC and US-DOE in Central California
- Progress on CARB Inverse Modeling Project
- Discussion of New Research

Emissions Reduction Validation is Crucial for Future Climate Control

- Global GHG emissions near IPCC A1F1 “high growth” scenario
- UNFCCC reporting of GHG emissions currently optional
- Progress at Copenhagen on emissions reductions agreements limited, in part, by lack of verification capability. President Obama: “[verification] must, however, ensure that an accord is credible, and that we are living up to our mutual obligations.”
- National Academy study: “Verifying Greenhouse Gas Emissions: Methods to Support International Climate Agreements”

Net Global Carbon Dioxide Emissions (pgC y



Global Carbon Project
Adapted from Raupach et al. 2007, PN

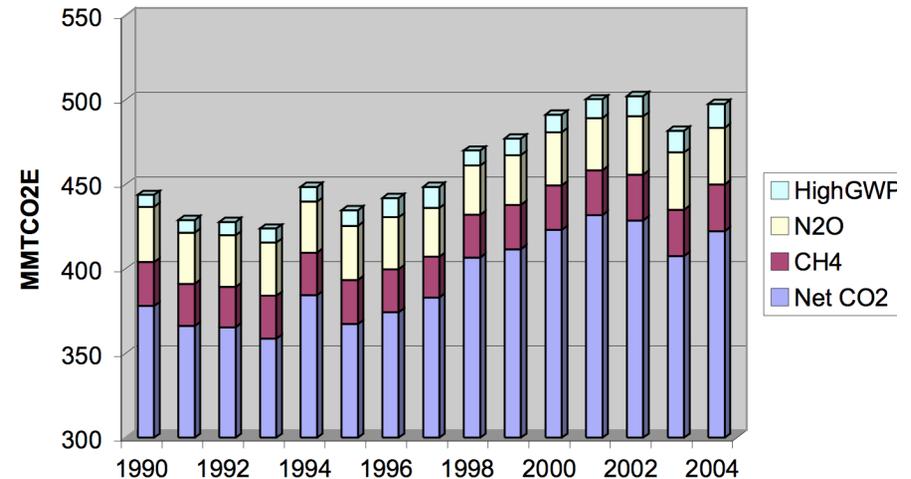
California emissions reductions policy (AB-32) drives the need for validation capability today

California GHG Emissions

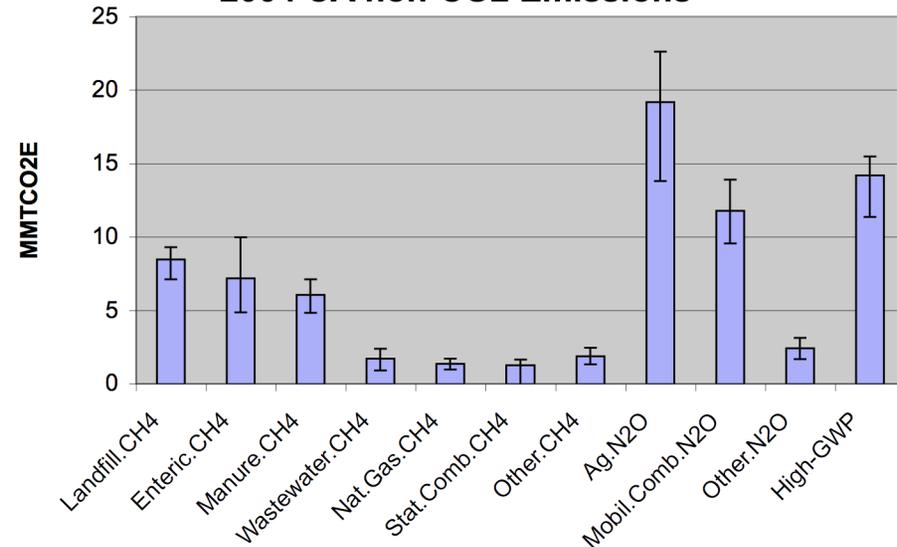
- California is the first state in US to legislate GHG controls
 - AB-32: Return to 1990 levels by 2020
- Sum of non-CO₂ GHG emissions comparable to fossil fuel CO₂ but
 - Industrial and biological sources are not readily metered
 - Uncertainties in inventories are large (even using US average fractional error estimates)
- Atmospheric inverse method provides independent check

CEC, 2006 ; USEPA, 2007

CA GHG Emission Trends

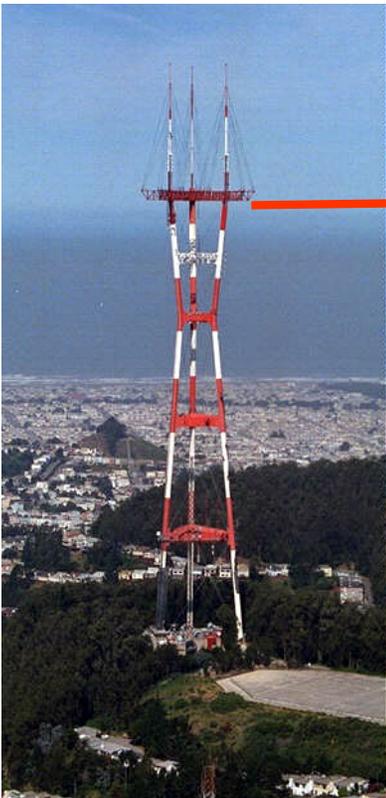


2004 CA non-CO₂ Emissions



Initial Work for CEC Began in 2003 California Greenhouse Gas Emissions Measurement Project (calgem.lbl.gov)

**Sutro Tower
(232 m agl)
Oceanic +
urban**



115 km
lat 37.876289° lon -121.525128°



**Walnut Grove
(483 m agl)
Central Valley +
Bay Area**



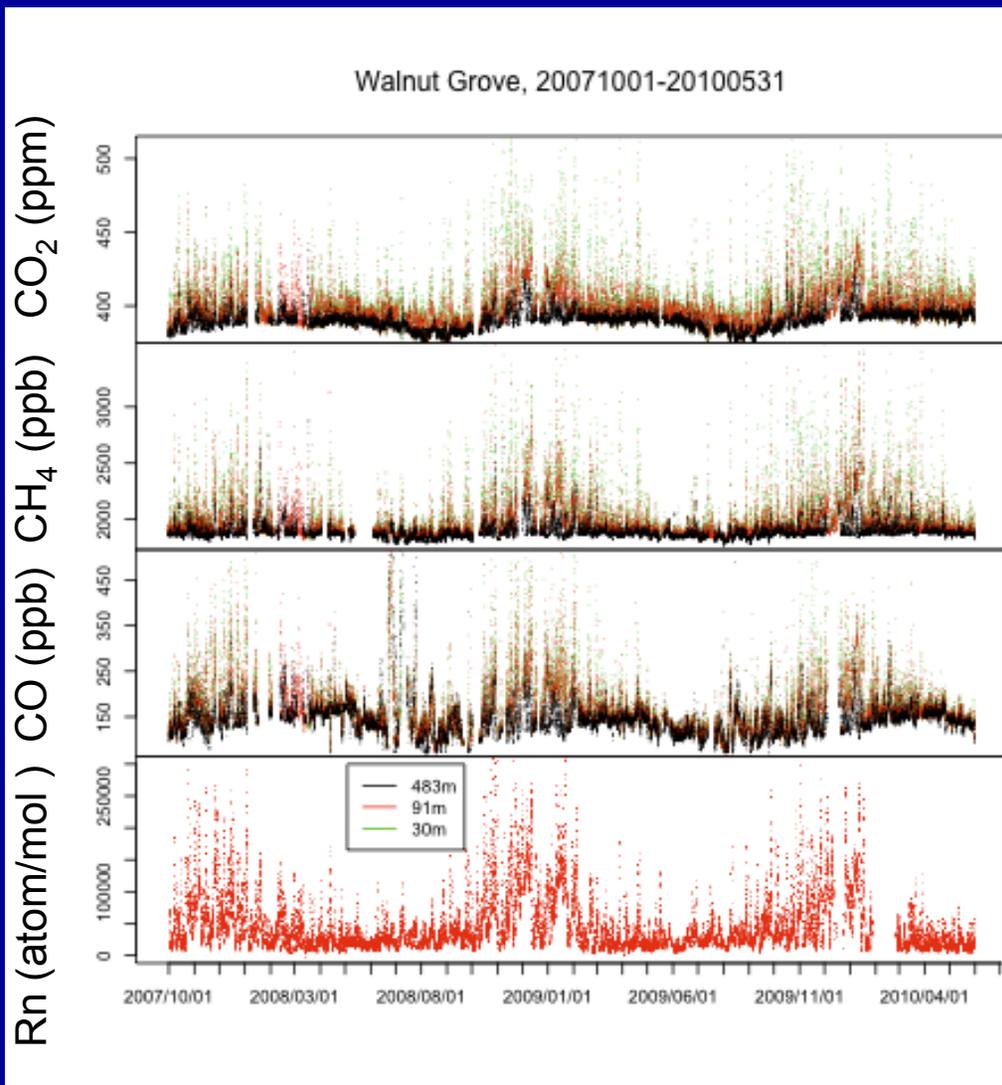
**Daily flasks sample:
CO₂, CO, CH₄, N₂O,
SF₆, halocarb, VOC,
¹³CO₂, ¹³CH₄
Radiocarbon ¹⁴CO₂**

**In-situ instruments
measure:
CH₄, CO₂, CO, ²²²Rn**

Image NASA
2008 DigitalGlobe
Aurora Technologies
US Tele Atlas

In-situ Measurements at Walnut Grove

- Elevated mixing ratios above background (black) indicate strong regional-local emissions
- Synoptic variations offer opportunity to extract emissions information
- Multi-year time series will probe inter-annual variations in emissions

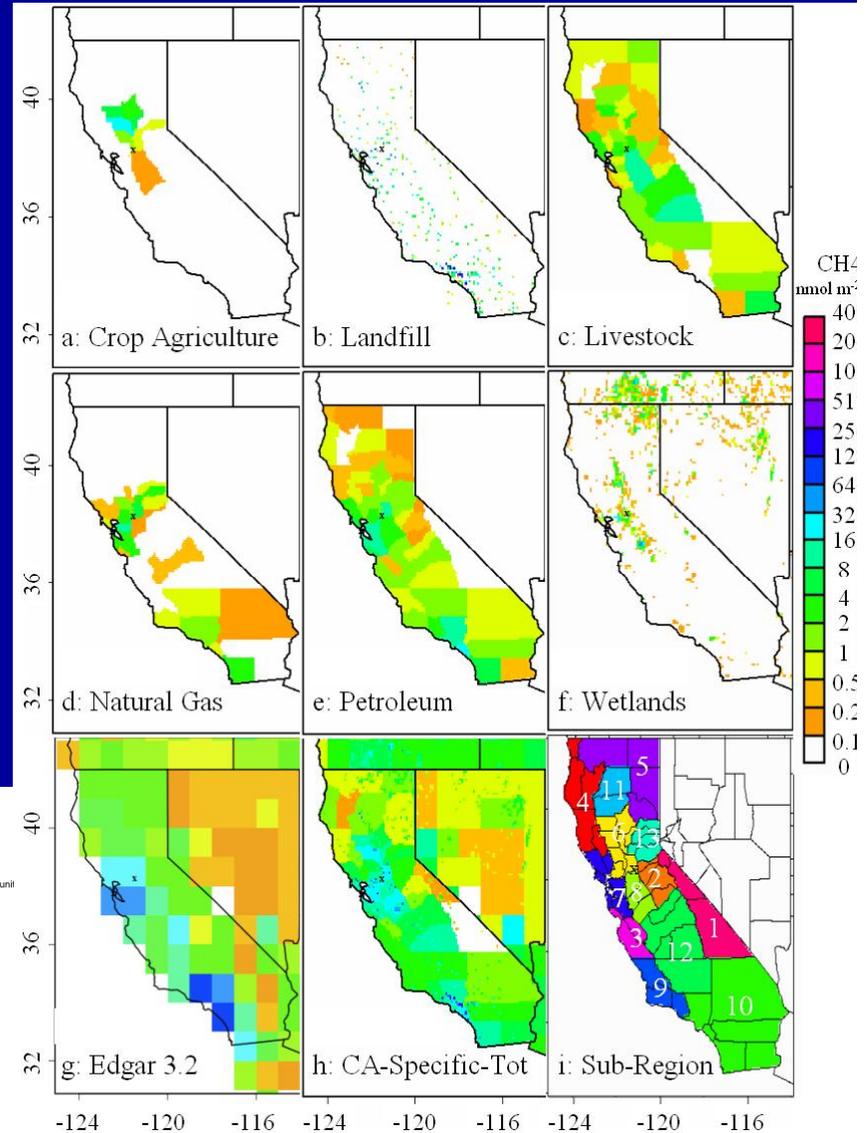
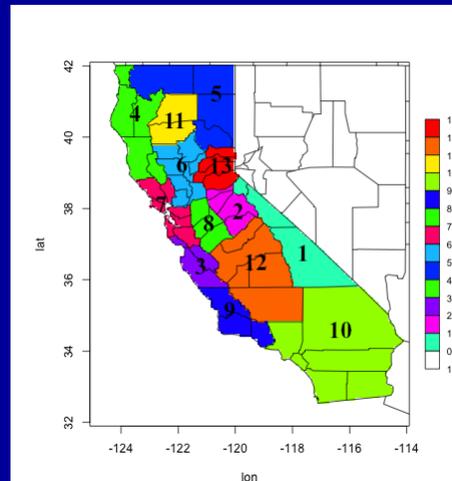


Inverse Model Estimates of Central California CH₄ and N₂O Emissions

- Emissions Inventories
 - Incorporating data-driven updates
- Meteorological Model
 - Advancing accuracy and resolution
- Estimates of CH₄ and N₂O Emissions
 - Single tower CH₄ over seasons
 - Initial work with N₂O

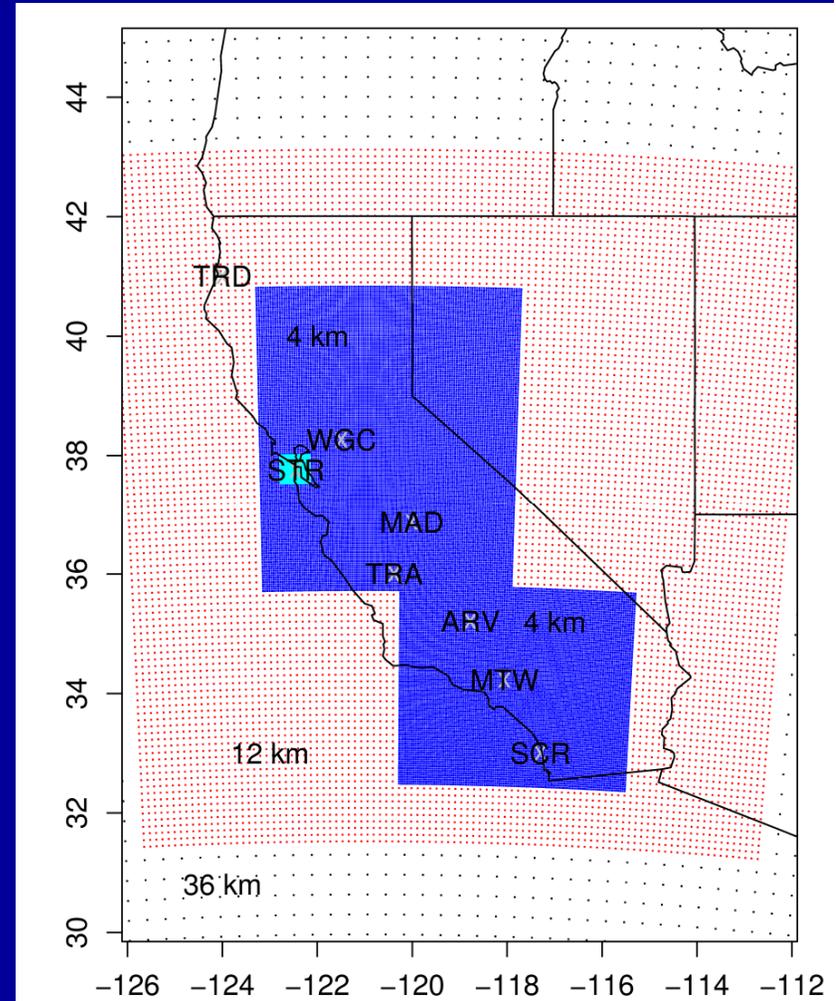
Early *a priori* CH₄ Emission Maps

- Landfill point sources (CARB)
- Crop Agriculture (Salas et al., 2006)
- 10 km wetland fluxes (Potter et al.)
- County level livestock (USDA)
- County level natural gas dist./use
- County level petroleum refining and use (CARB VOC data)
- EDGAR3.2 (1x1degree)
 - Landfills and petroleum extraction and refining ~ 2 x CA estimates
- Regions for spatial analysis (Fischer et al., 2009)



Meteorological Model for CA Domain

- Weather Research Forecast Model (WRF)
 - Domains (extension of Zhao et al., 2009):
 - 36 km (W. US), 12 km (CA)
 - 4 km (Central Valley)
 - 1.3 km (Sutro, WGC)
 - NARR boundary forcing and internal nudging
 - Daily runs, hourly output



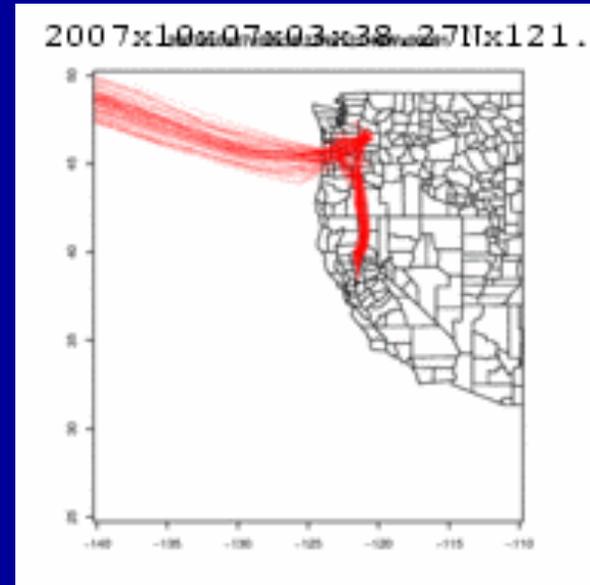
WRF-STILT Footprints for WGC Tower

- Footprint from ensemble of particle trajectories, p

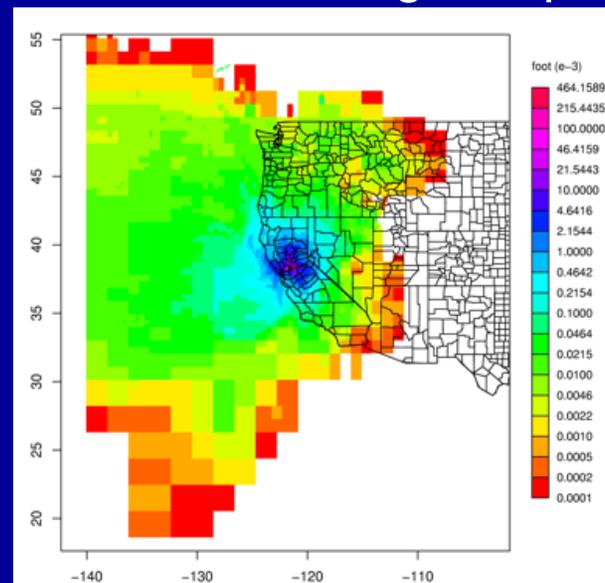
$$f(\underline{X}_r, t_r | x_i, y_j, t_m) \propto \sum_{p=1..N} \left(\frac{\Delta t}{Z_i} \right) |i, j, m, p$$

- Largest surface influences (purple) for Bay Area and Central Valley
- Predict local CH₄ signals

$$C_l(\underline{X}_r, t_r) = \sum_{i, j, m} f(\underline{X}_r, t_r | x_i, y_j, t_m) \cdot F(x_i, y_j)$$



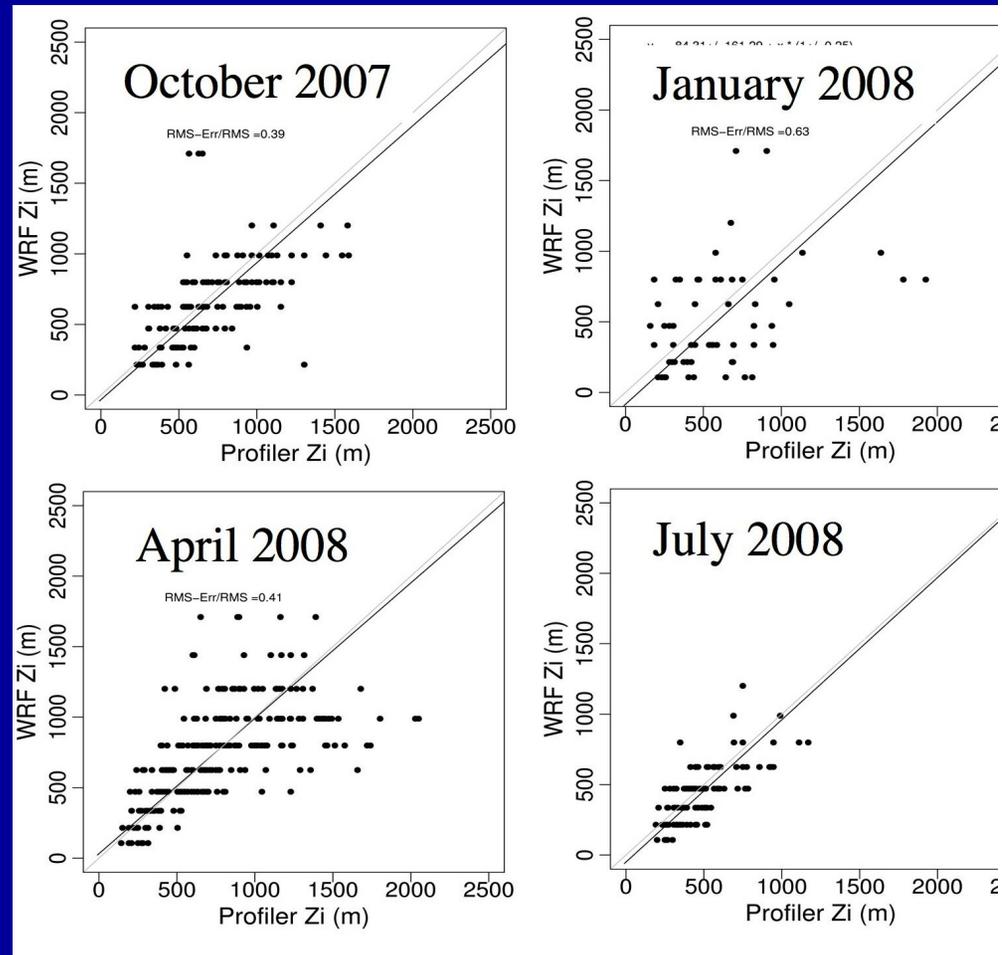
Oct-Dec, 2007 avg. footprint



Uncertainty Estimation

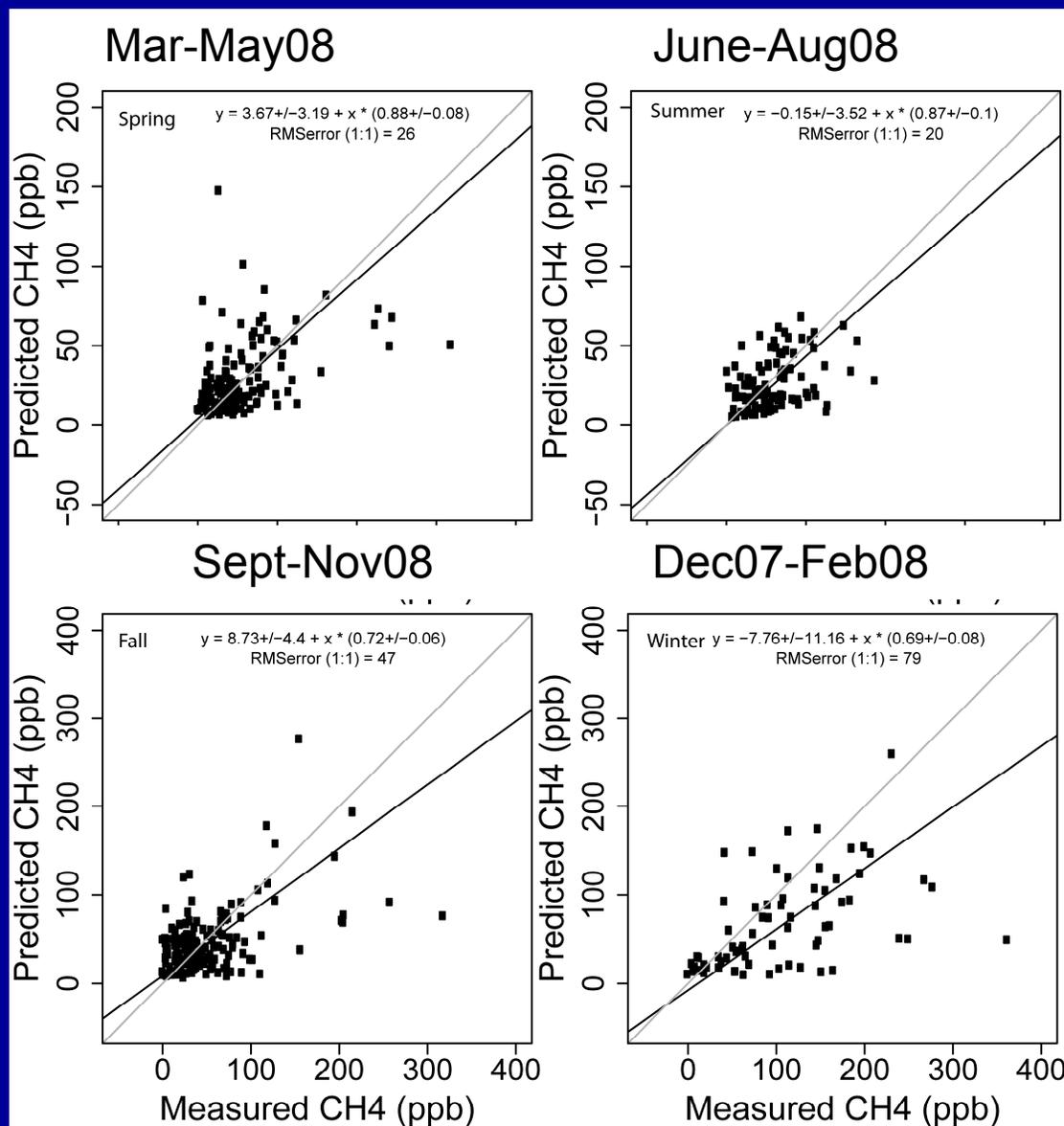
- Quantify error sources
- Propagate errors through modeling system to provide quantitative uncertainties
 - Boundary layer ~ 25 %
 - Wind Velocity ~ 10%
 - GHG background ~ 15 %
 - Inventory resolution ~ 8 %
 - Other ~ 8%
- Quadrature sum ~ 32%
of signal for individual time points

WRF-STILT versus Profiler PBL Depth



Comparison of Measured and Predicted CH₄ at Walnut Grove by Season

- Scatter approximately consistent with estimated uncertainties
- CH₄ emissions appear underestimated in CA inventory for most periods



Bayesian Linear Inverse Model

- Local signal $C_{\text{local}} = C_m - C_{\text{bg}} \sim \sum \lambda_i F_i f + \varepsilon$
 - Write cost function for $C_{\text{local}} = y$
 - Balance predicted-measured mismatch with error S_ε , and posterior-prior mismatch with error S_{prior}

$$\underline{J} = (\underline{y} - \underline{K}\underline{\lambda})^T \underline{S}_\varepsilon^{-1} (\underline{y} - \underline{K}\underline{\lambda}) + (\underline{\lambda} - \underline{\lambda}_{\text{prior}})^T \underline{S}_{\text{prior}}^{-1} (\underline{\lambda} - \underline{\lambda}_{\text{prior}})$$

where $\underline{K} \underline{\lambda} = \sum_{i,j,m} f(\underline{X}_r, t_r | x_i, y_j, t_m) \cdot F(x_i, y_j, t_m)$

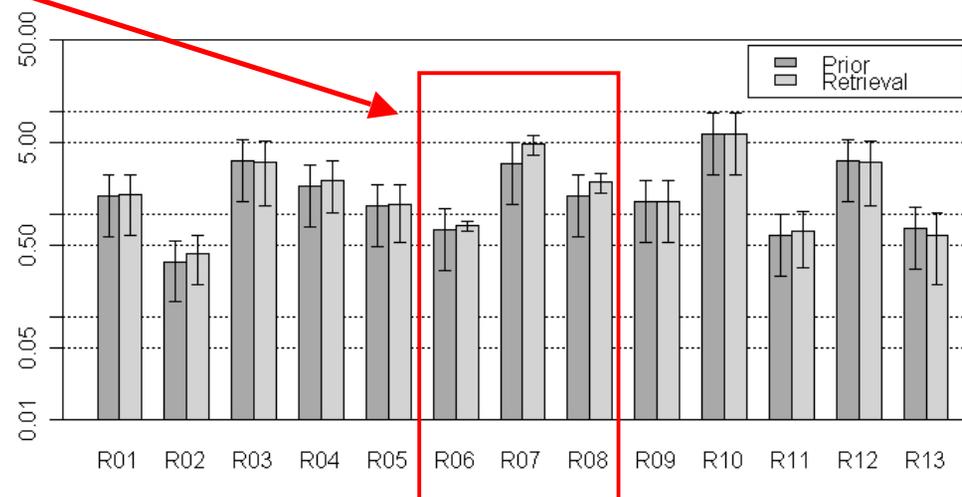
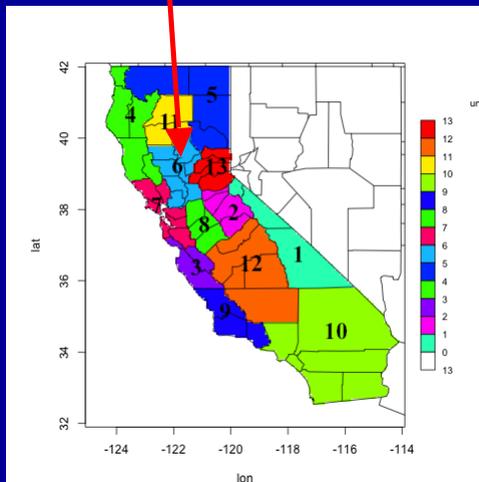
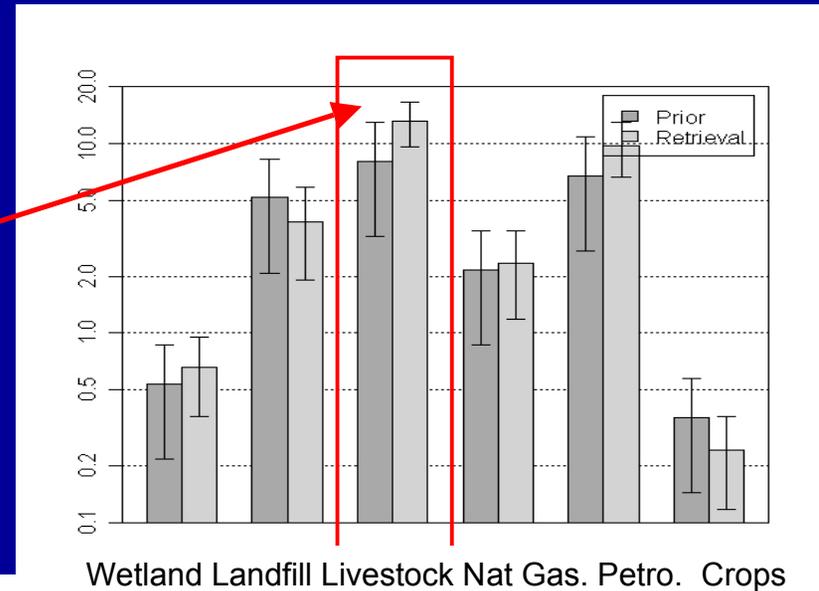
- If errors random ($\varepsilon \sim N(0, \sigma)$), solution algebraic

$$\hat{\underline{\lambda}} = (\underline{K}^T \underline{S}_\varepsilon^{-1} \underline{K})^{-1} (\underline{K}^T \underline{S}_\varepsilon^{-1} \underline{y} + \underline{S}_{\text{prior}}^{-1} \underline{\lambda}_{\text{prior}})$$

$$\hat{\underline{S}}_\lambda = (\underline{K}^T \underline{S}_\varepsilon^{-1} \underline{K} + \underline{S}_{\text{prior}}^{-1})^{-1}$$

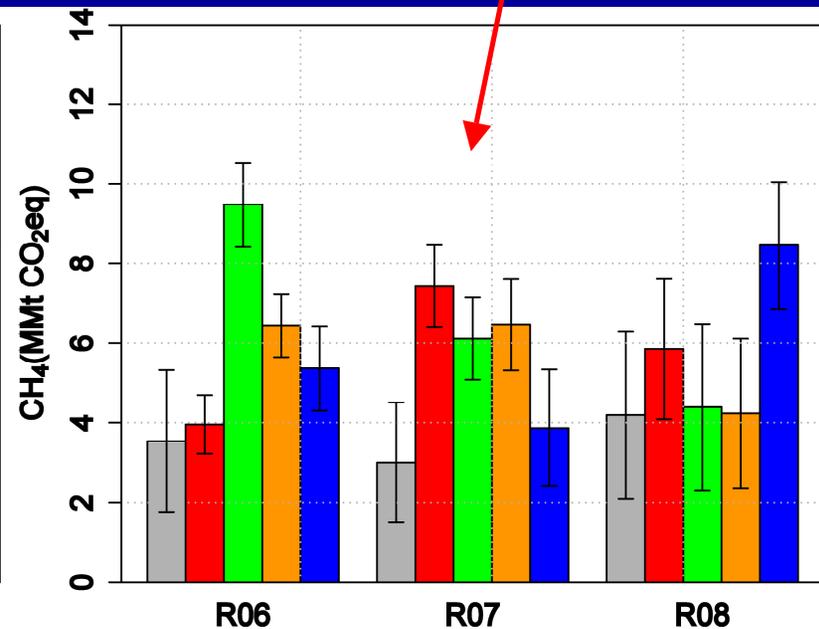
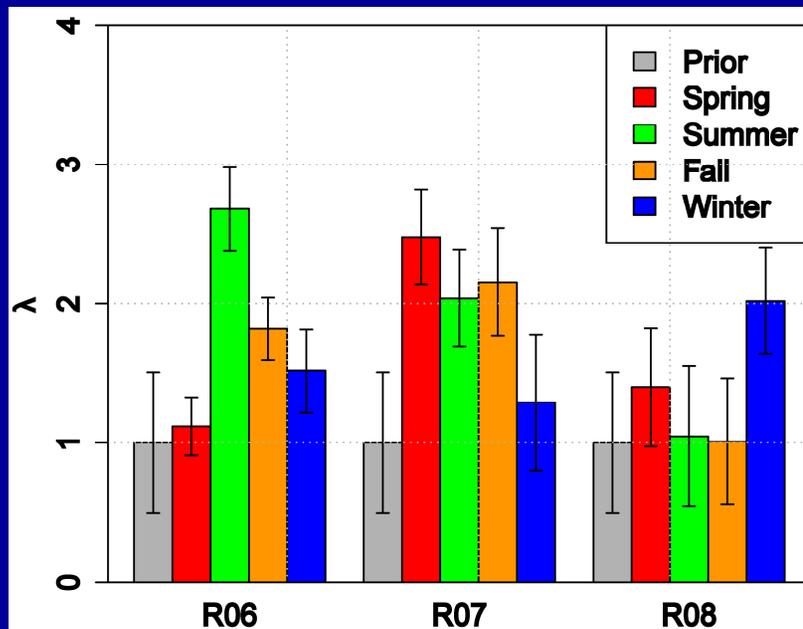
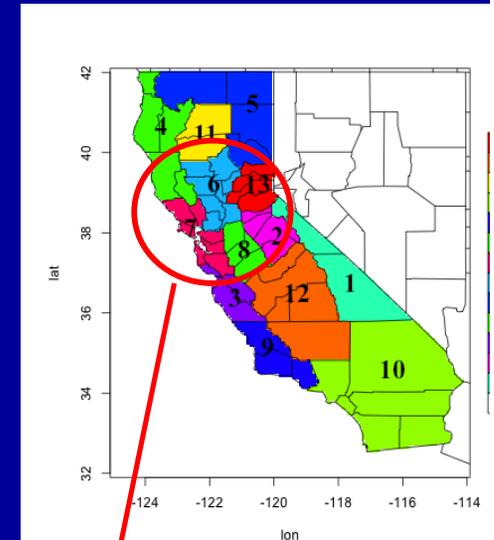
Oct-Dec07: Estimated CH₄ Emissions (MMT CO₂equiv yr⁻¹)

- Bayesian estimate of scaling factor for each emission source or region (Zhao et al., 2009)
- Source analysis: only livestock significantly different from prior ($\times 1.6 \pm 0.15$)
- Region analysis:
 - errors reduced only for regions 6, 7, 8 near tower



Capturing Seasonality of CH₄ Emissions

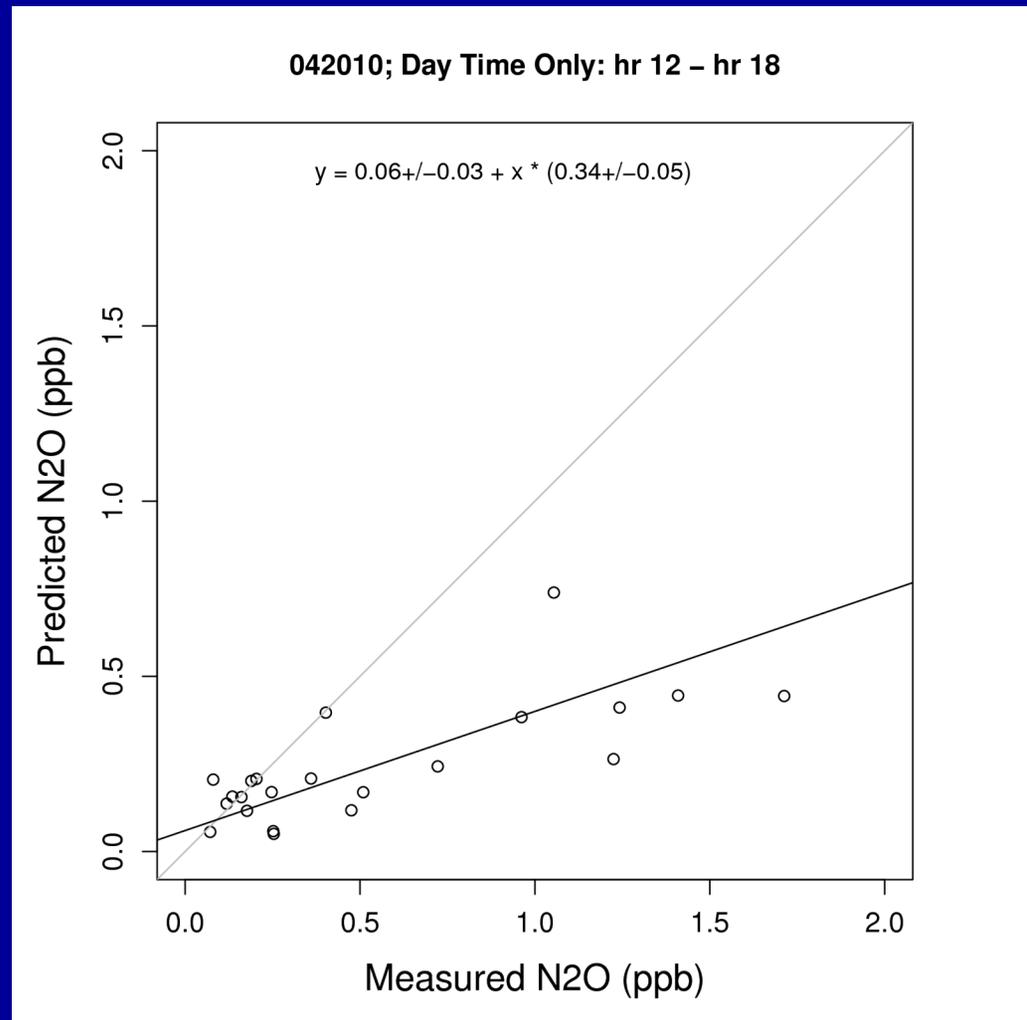
- Emissions from regions 6,7,8 show statistically significantly higher emissions in warm months
 - Seasonality of rice and wetlands in regions 6,7
 - Exception is region 8, dominated by livestock



Measured and Predicted N₂O

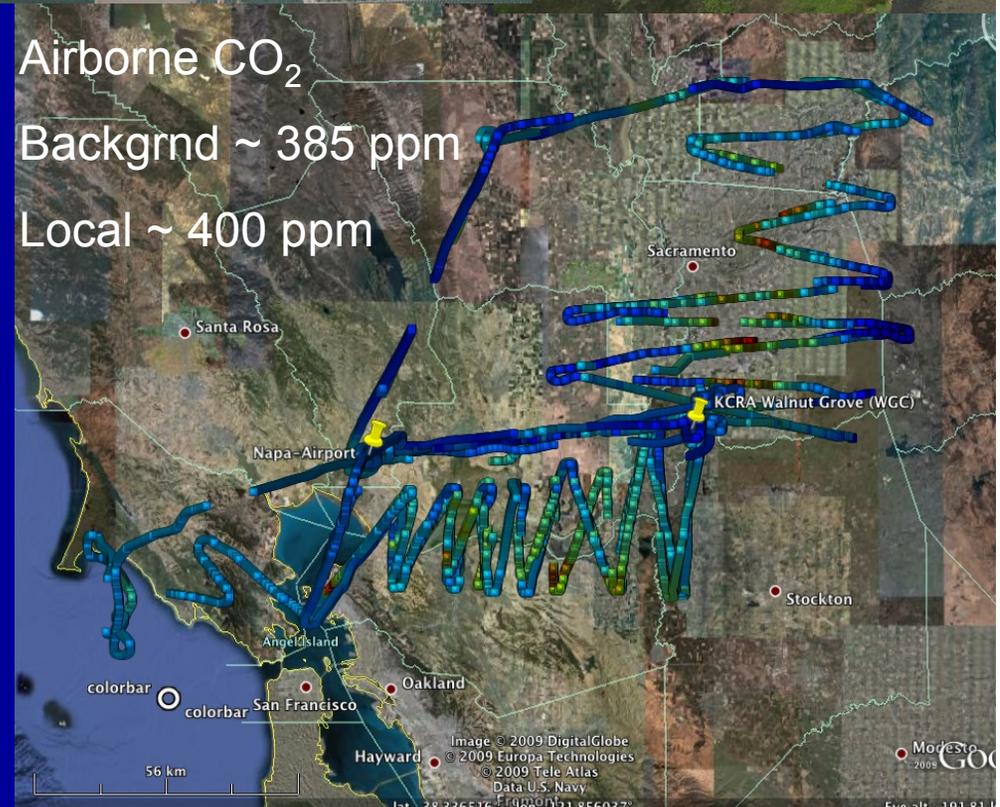
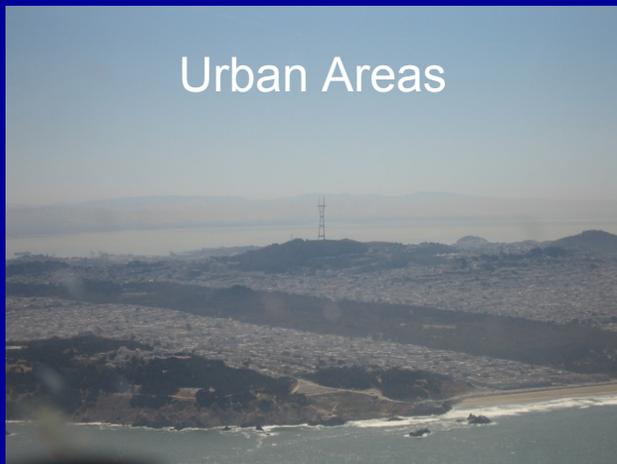
Initial continuous N₂O measurements April, 2010

- Comparison with existing prior model (Edgar 3.2)
- Slope suggests actual emissions ~ 3 x higher than inventory
- Need to extend analysis 2007- present

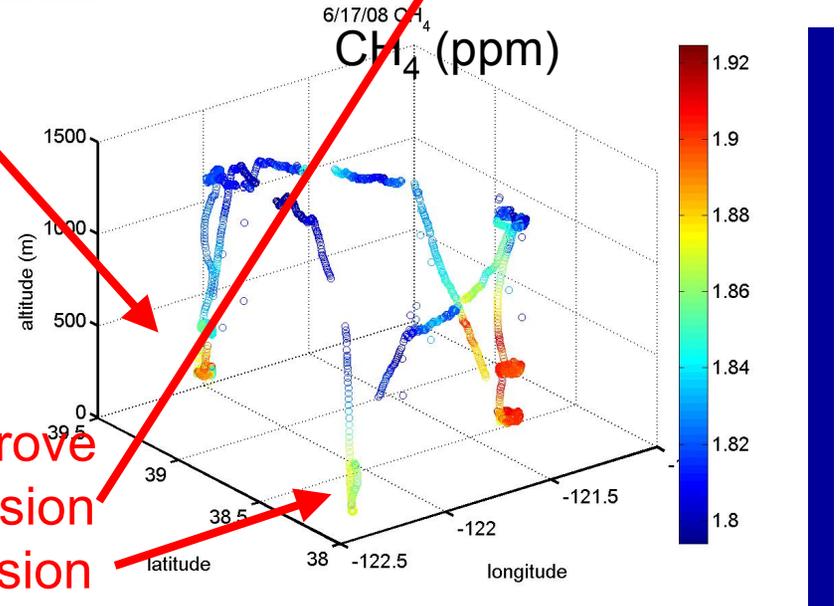
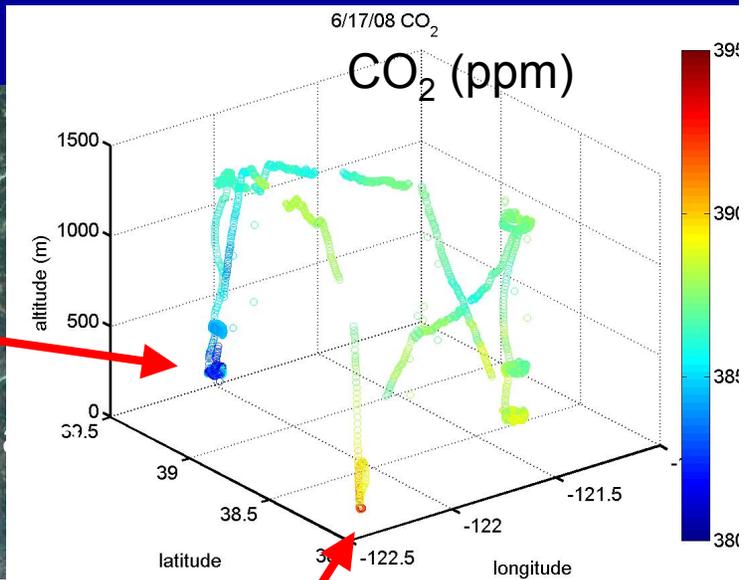
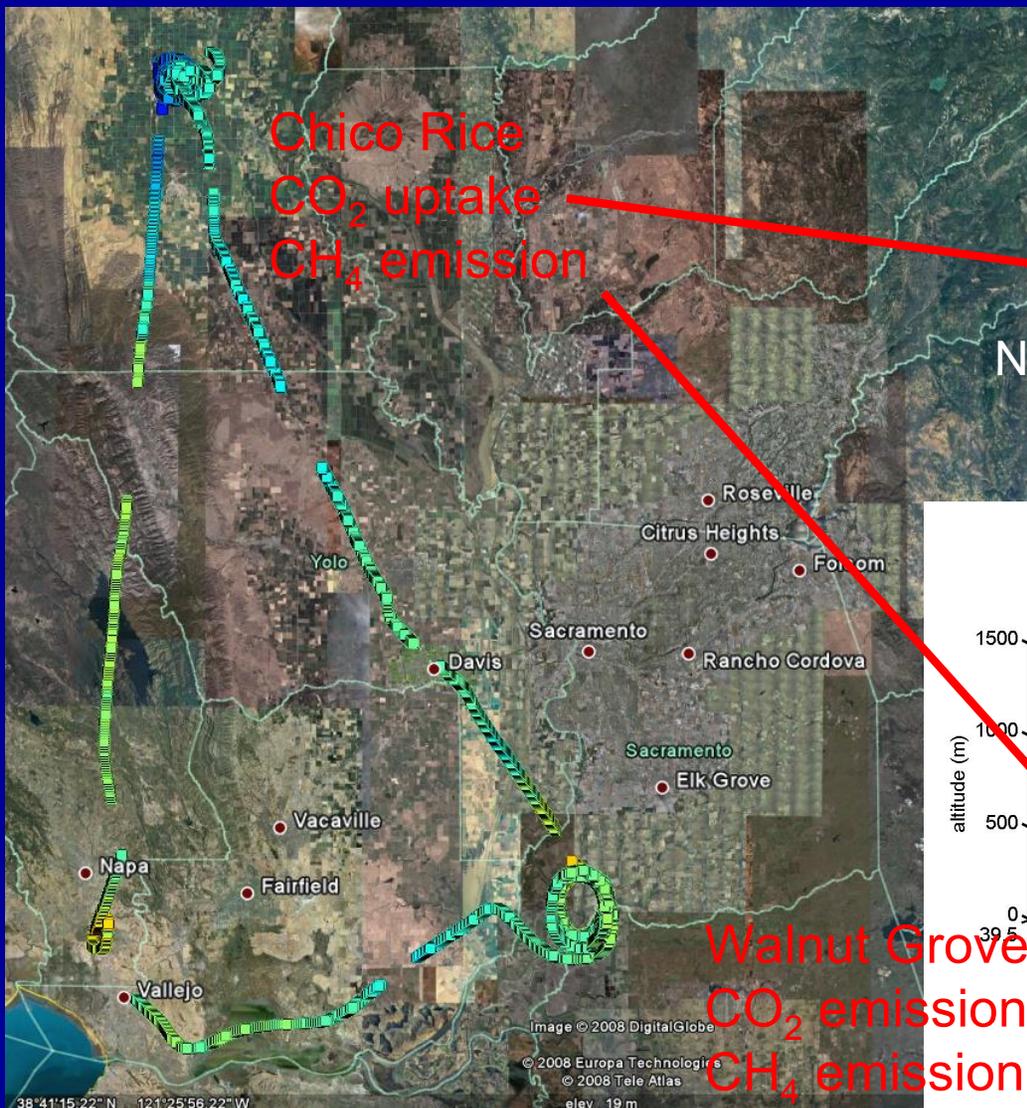


Identifying GHG Sources with Airborne Greenhouse Gas Survey (AGES)

- June, 2008 and March, 2009 campaigns from Napa CA
- Observations of Sac. river delta, Sacramento urban area, and Central Valley agriculture
- Follow-on work being planned with UCB (PI Goldstein)
- Multi-species measurements
 - Insitu CO_2 , CH_4 , CO
 - Flask CO_2 , CH_4 , CO , N_2O , $^{14}\text{CO}_2$

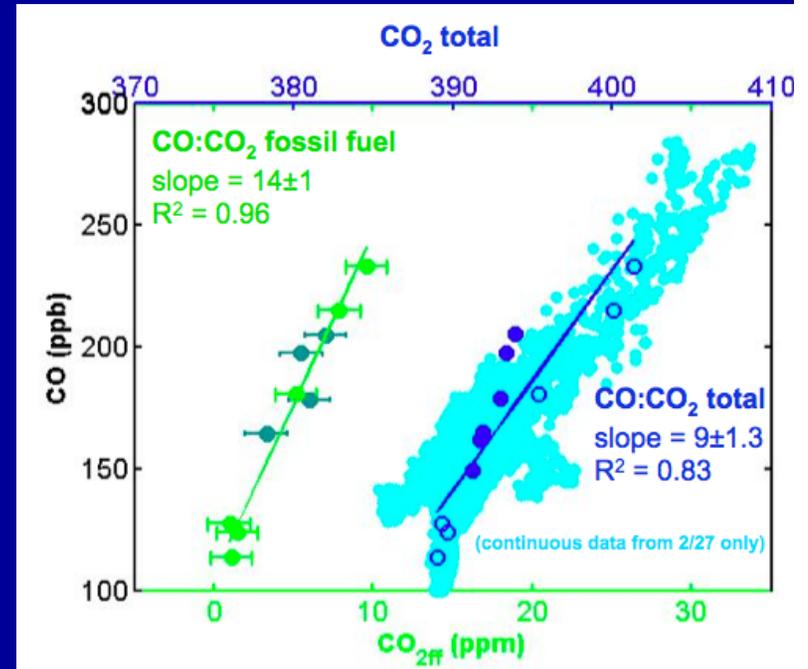


Contrasting CO₂ and CH₄ Signals over Rice and WGC



Airborne Estimate of Sacramento Fossil Fuel CO₂

- AGES flights also measured Sacramento fossil fuel CO₂ using ¹⁴CO₂, HGWPs
- Continuous fossil fuel CO₂ (ffCO₂) from tight correlation of flask CO:ffCO₂
- Slab-model estimate of Sacramento CO_{2ff} emissions consistent with CARB and Vulcan ffCO₂ inventories



2009 Sacramento CO_{2ff} Emissions (MtC yr)

AGES	Vulcan	CARB
3 (2-6)	3.0	2.6

$$Q = \int_{\text{acrossplume}} \int_{z=0}^{Z_{PBL}} u(y,z)n(z)CO_{2ff}(y,z)dydz$$

(Turnbull et al., 2011)₂₁

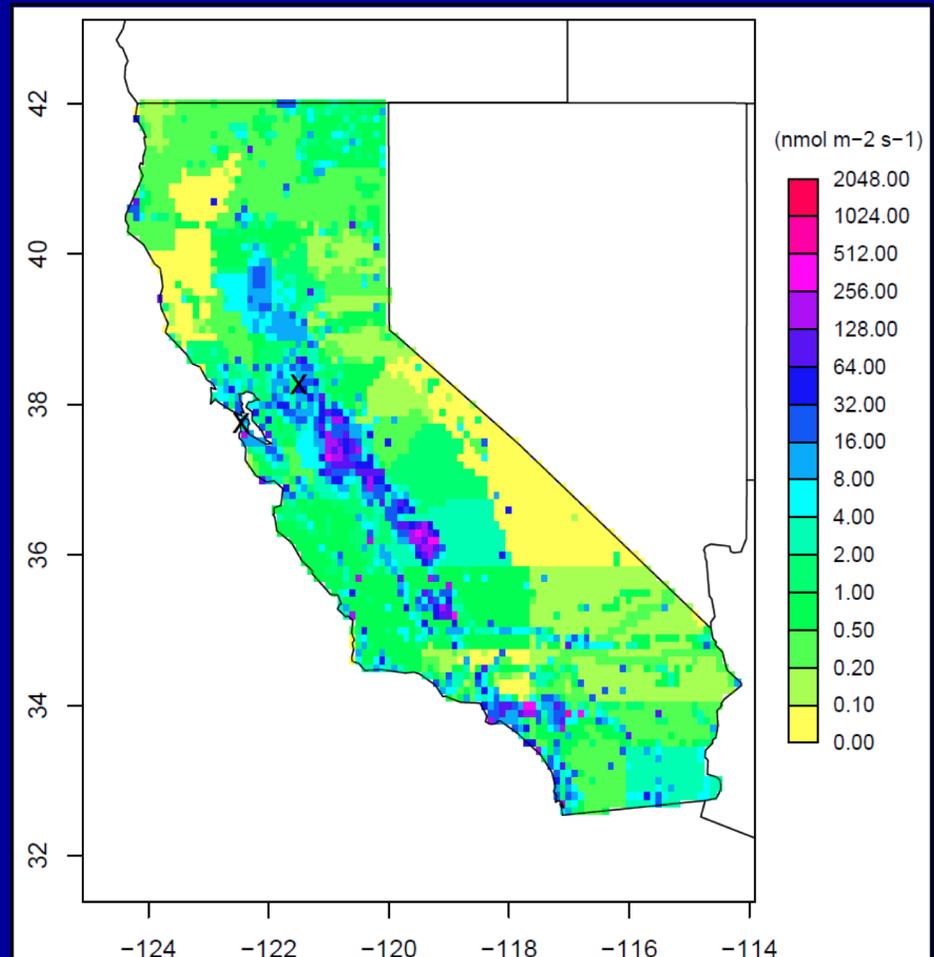
New CH₄ Inverse Modeling with CARB

- Enhanced CH₄ emission maps
- Improved meteorological model
- Initial inverse analysis of multi-site Central Valley CH₄ data

Enhanced *a priori* CH₄ Emission Maps

- Facility specific livestock counts and updated emission factors (DWR, CEC, CARB)
- GIS driven natural gas pipelines, storage, and use (CEC, CARB)
- Facility specific petroleum refining and distribution (CARB)
- Facility specific waste water treatment (CARB)
- Mobil source CH₄ model (following Vulcan fossil fuel CO₂, Gurney et al., 2009)

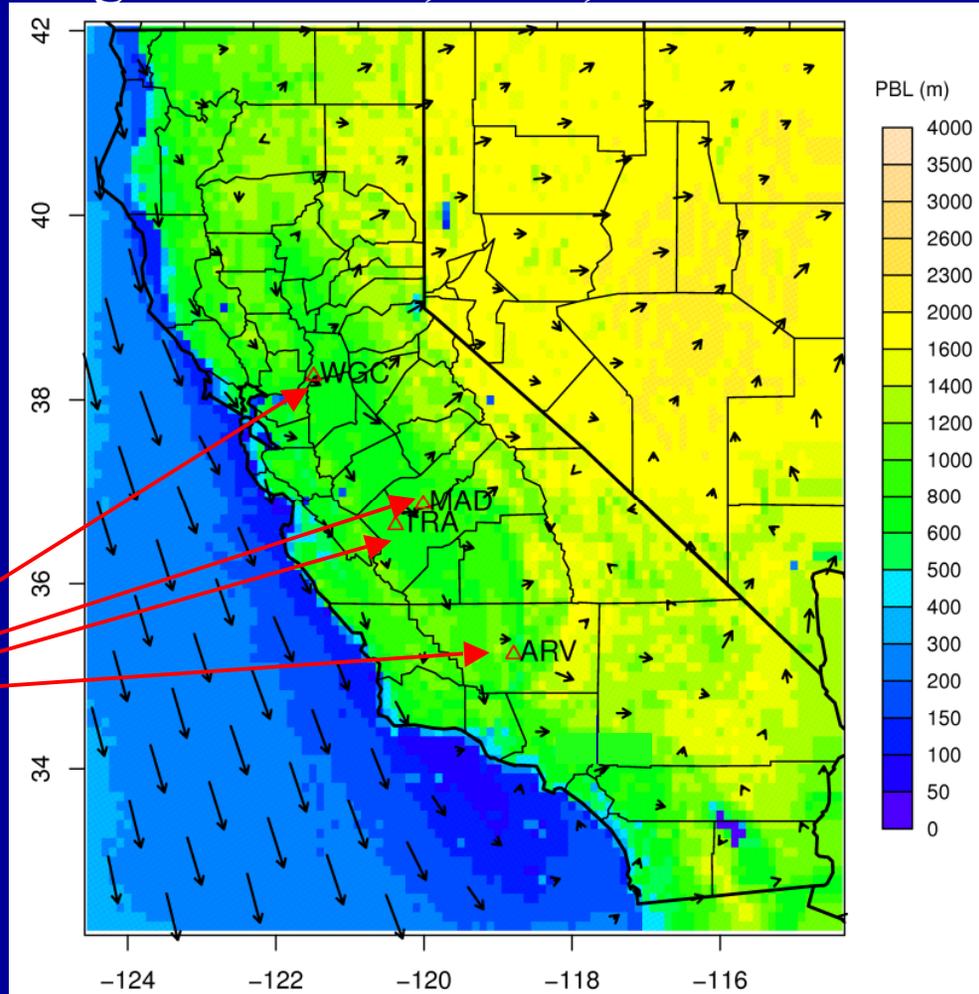
Map of California specific CH₄ emissions at 0.1 degree resolution



Improved Meteorological Model

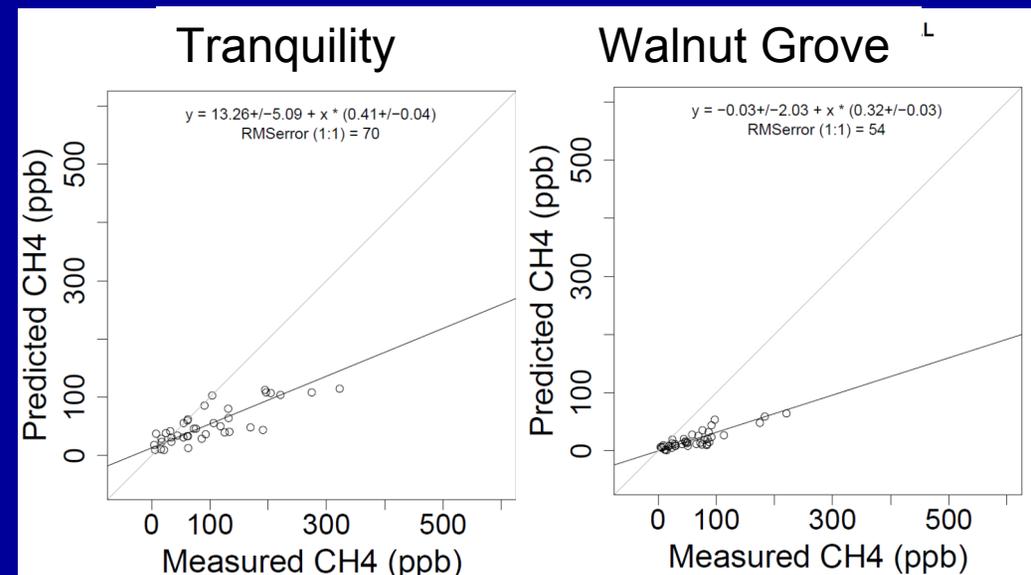
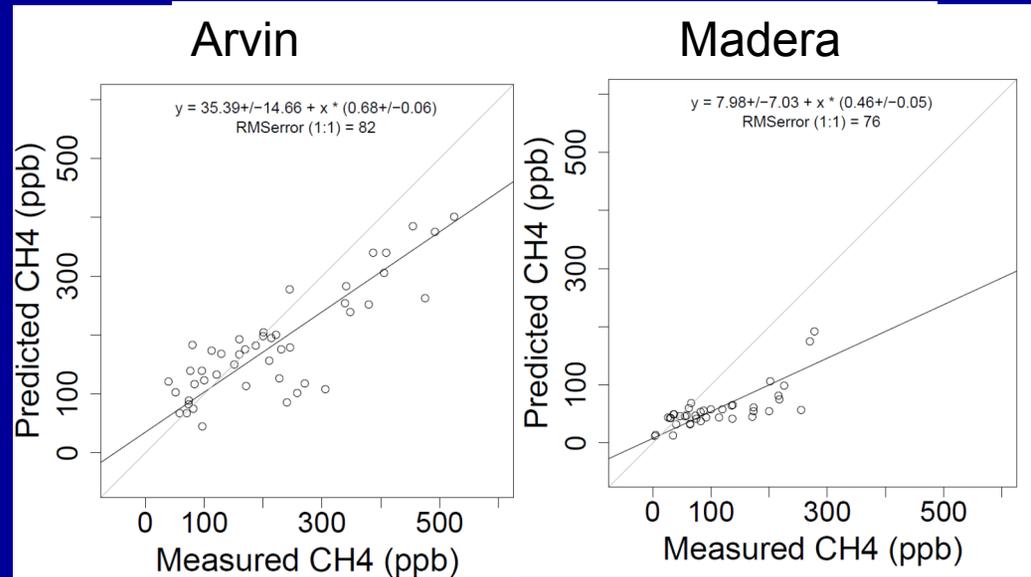
- New land surface model improves energy budget by incorporating irrigation practices
- PBL estimates are improved in native model
- Apply to CARB Central Valley CH₄ network

Predicted monthly mean winds and PBL heights for June, 2010, 1000 local.



Measured and Predicted Midday CH₄ Signals for CARB CH₄ Network September, 2010

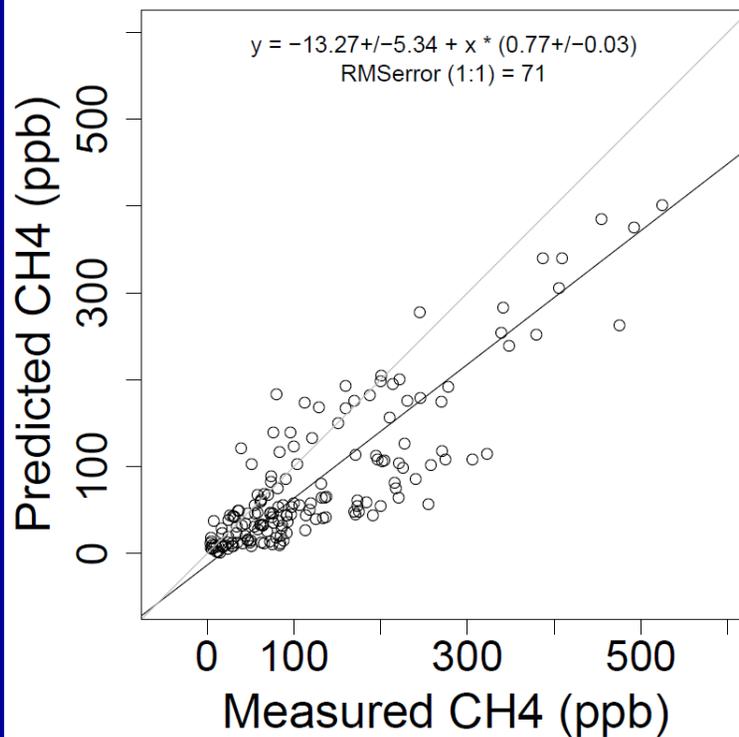
- N-S Gradient in San Joaquin Valley
- Significantly higher CH₄ signals at Arvin
- Madera and Trinity sites comparable
- Lower signals at Walnut Grove



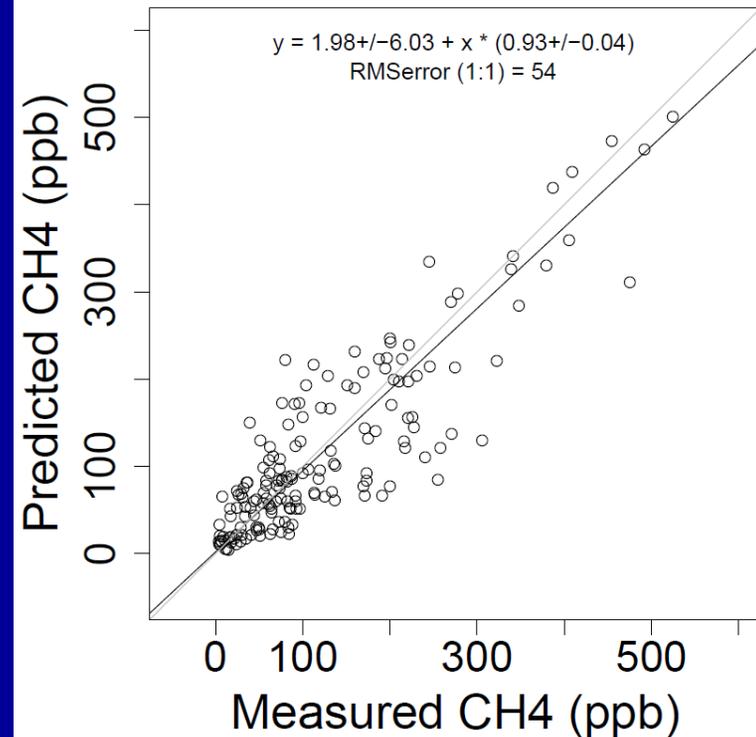
Prior and Optimized CH₄ Signals for Sept., 2110

- Composite data from all sites
- Scale prior emissions in 13 regions to obtain best posterior match

Prior Emissions (before inverse)

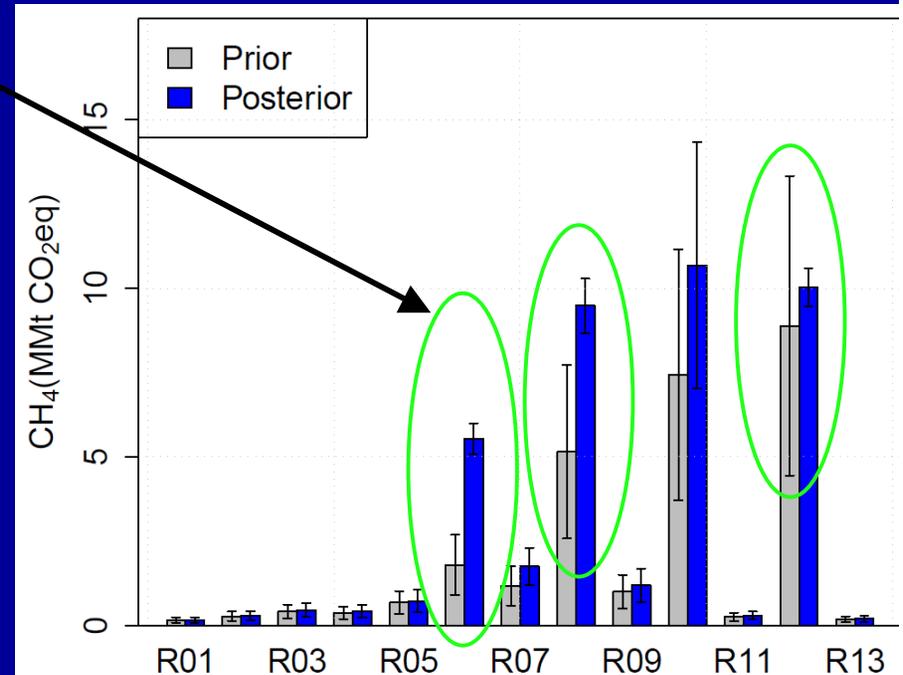
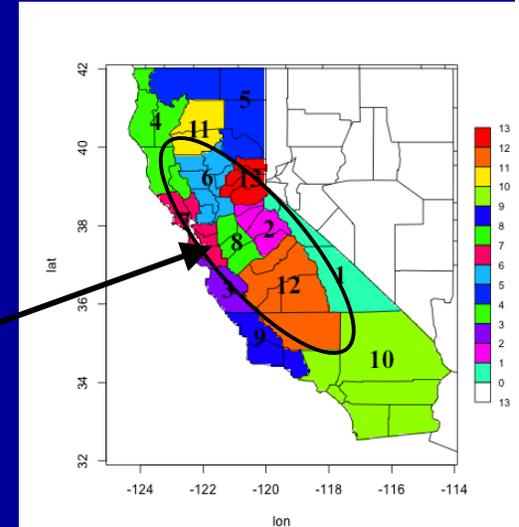


Posterior Emissions (after inverse)



Initial Results from CARB CH₄ Network

- CARB sites significantly reduce uncertainties for key regions
 - Capture most of Central and South San Joaquin valley
- Future towers planned:
 - Capture North Valley (rice agriculture)
 - Need for sampling in SoCal Air Basin



Conclusions

- Atmospheric measurements and inverse modeling provide a powerful independent constraint for emissions inventory validation
- Careful attention to uncertainties essential for quantitative emission inventory assessment
- Network of towers are effective in capturing regional CH₄ emissions from much of California's Central Valley
 - CH₄ emissions are modestly underestimated
 - N₂O emissions significantly underestimated

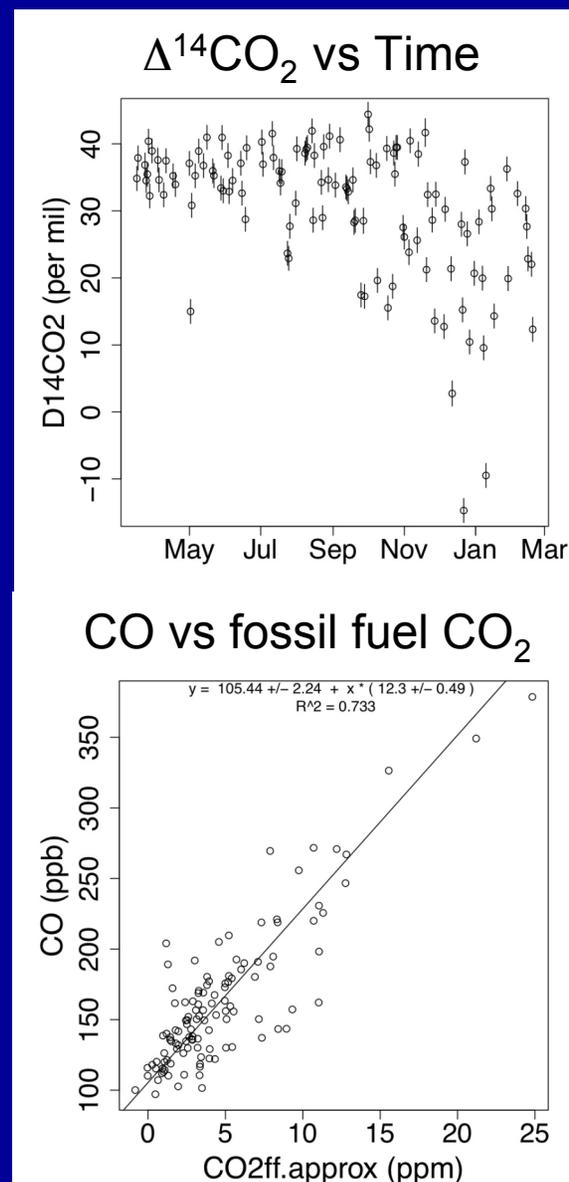
Suggested Research to Address CARB needs for emission inventory validation:

- Extend measurement records and capabilities at existing sites
 - Multi-species approach provides source attribution
- Improve inverse modeling framework
 - Hierarchical Bayesian models to identify unexpected emissions
 - Assimilation of network meteorology to reduce transport uncertainty
- New key sites and measurements for CARB network
 - Close gaps in spatial coverage to cover important regions

Extending capabilities at key sites

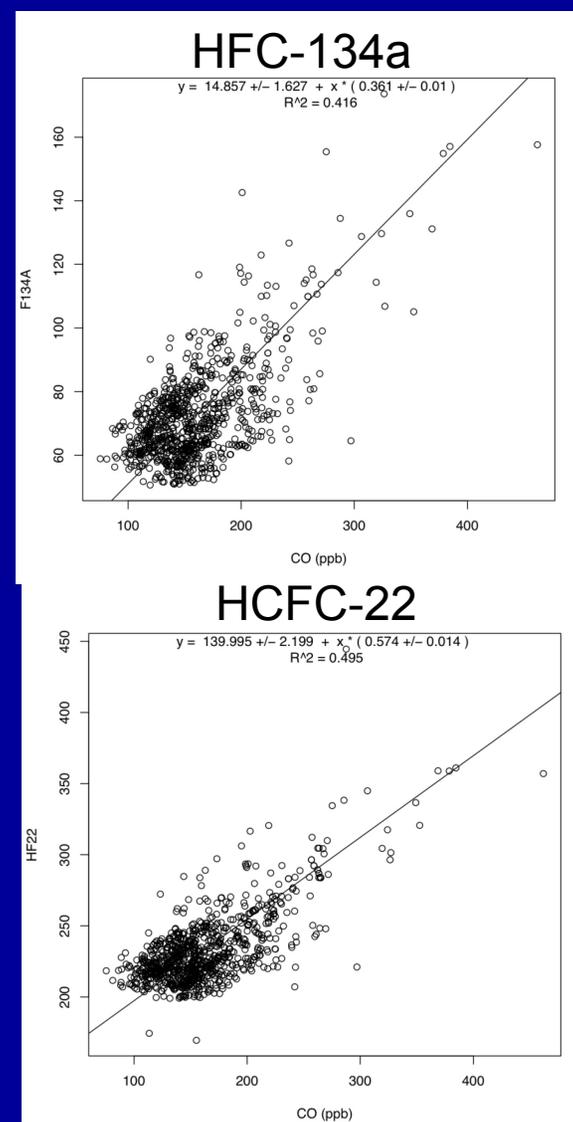
Radiocarbon $^{14}\text{CO}_2$ at Walnut Grove

- Daily flask sampling
- 2-3 weekly sub-samples for $^{14}\text{CO}_2$ 2009 - present
- First year analysis shows large depletion in winter, likely due to decreased mixing
- Excellent correlation of CO to fossil fuel CO_2
 - Use CO as continuous proxy for fossil CO_2
 - Need additional tracers of biomass combustion



Extending Key Measurements: High Global Warming Potential Measurements at Walnut Grove

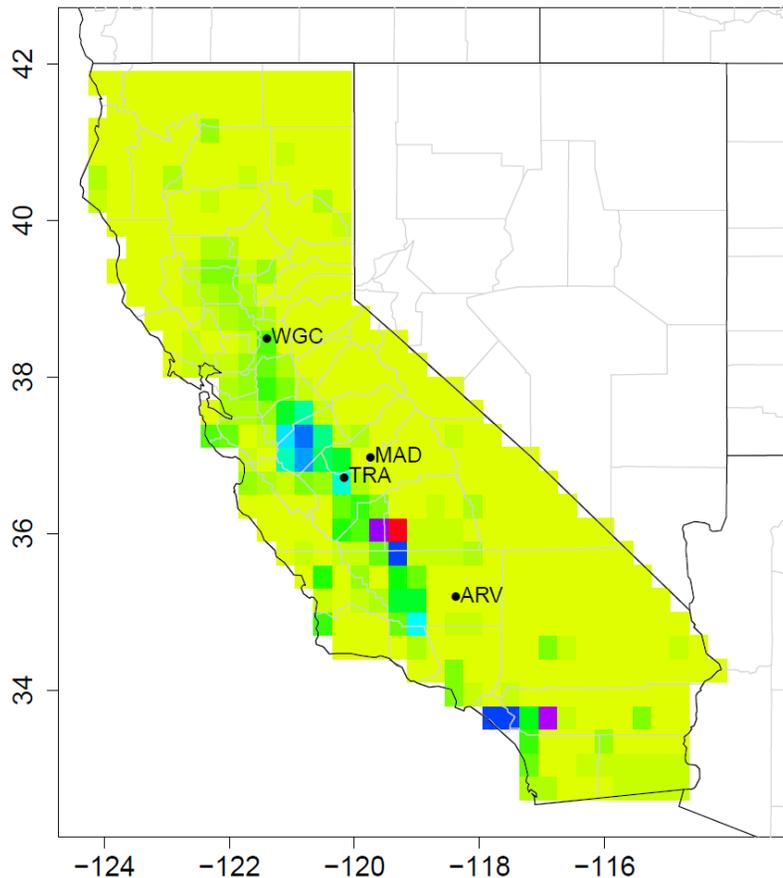
- All major warming gases accurately measured
- Inter-annual trends due to global background
- Synoptic variations reveal regional sources
- High correlation of some HWGP to CO or fossil fuel CO₂ may allow emission factor analysis of emissions (e.g., Hsu et al., 2009)
- Other gases may yield inverse model estimates of HWGP emissions



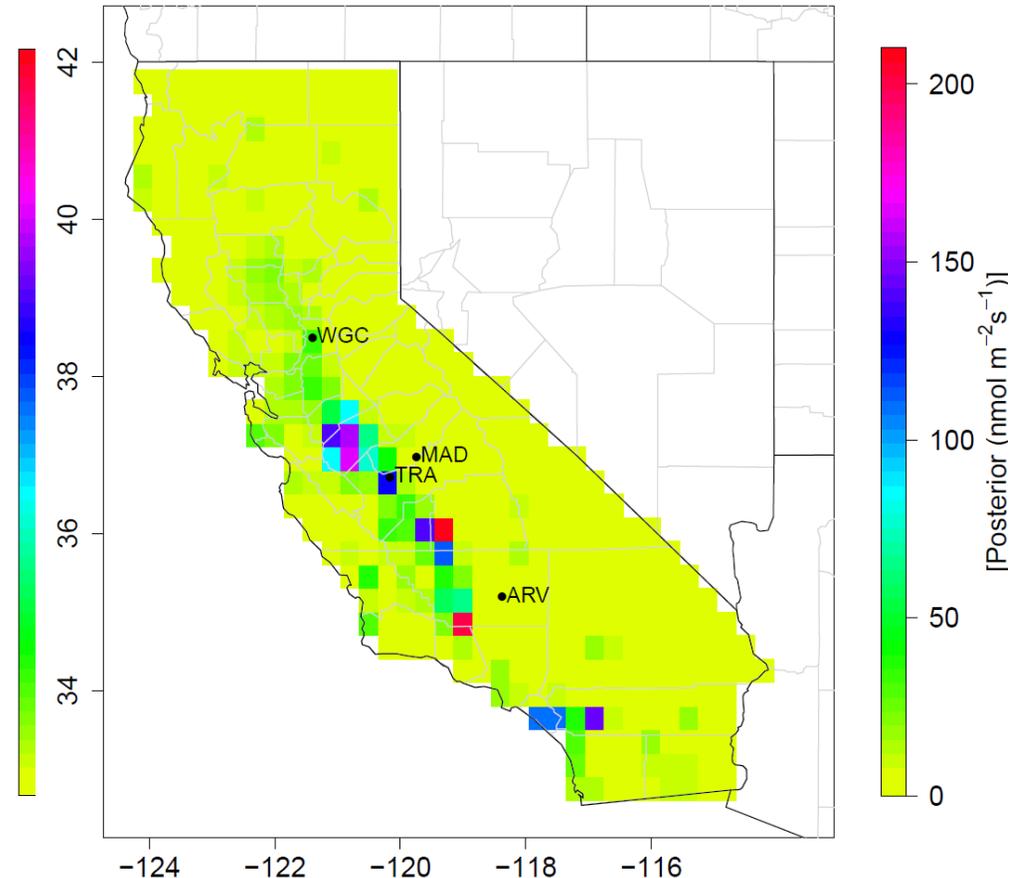
Improving Inverse Model Framework: Hybrid Model Maps Prior and Pixel-Based CH₄

- Identifies spatial features of CH₄ emissions beyond prior
- Requires larger data sets (e.g., aircraft, satellite)

Prior Emissions (before inverse)

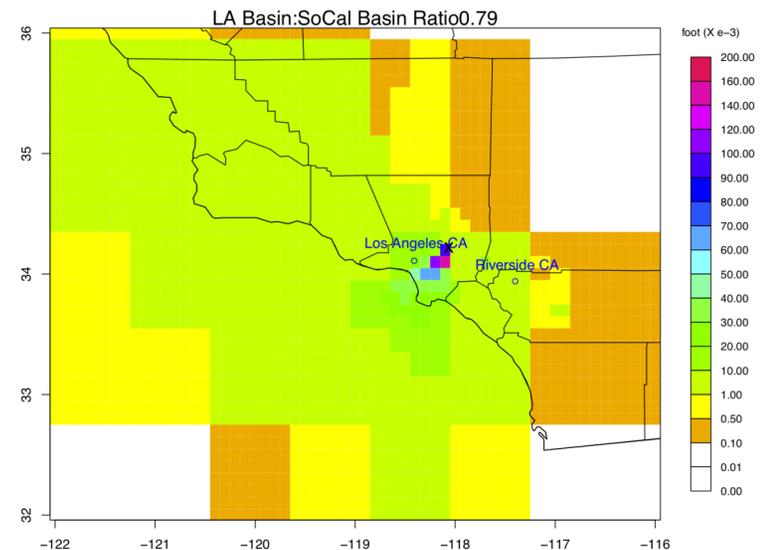
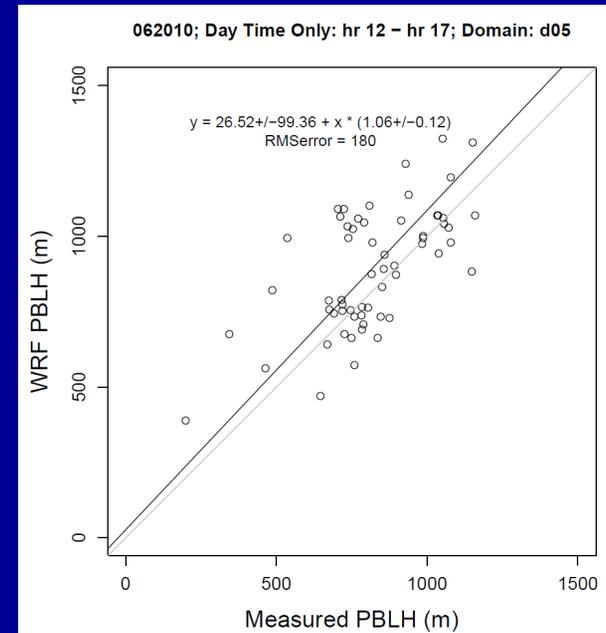


Posterior Emissions (after inverse)



Expand CARB Network to South Coast

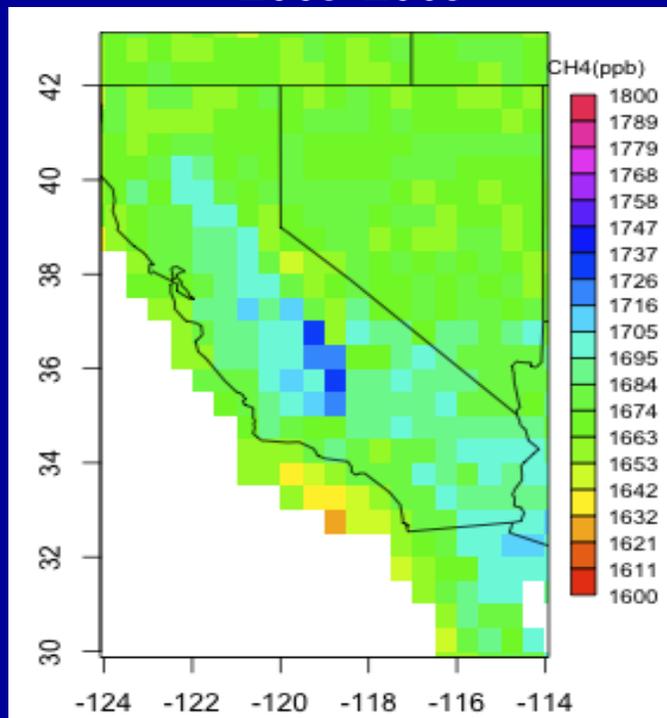
- Caltech/U-Houston Lidar measurements confirm accurate daytime PBL in WRF model
- Footprint model shows 80% of SoCal influence at Mt Wilson limited to Los Angeles County
- Proposed site near UC Riverside (San Bernadino) will provide coverage of larger area containing CH₄ source



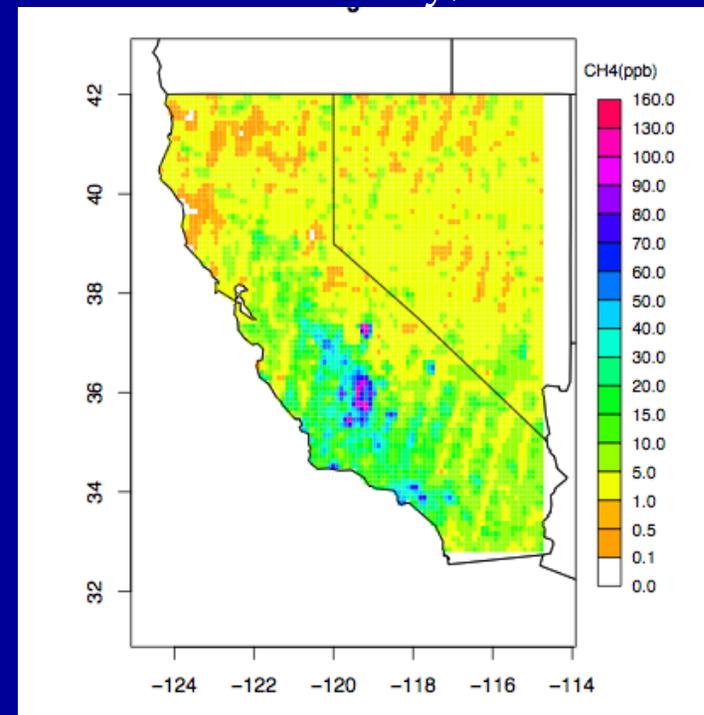
Including Key Measurements: Initial Comparison of SCIAMACHY Column Mean CH₄ Retrievals and Prior Model

- Satellite tracks cover CA require long integrations
- Clear spatial correlation of measurement and prediction

SCIAMACHY CH₄
2003-2005



Predicted Column CH₄
March-May, 2005





Thank You

Fossil Fuel CO₂ Observed with Radiocarbon

Measure $\Delta^{14}\text{CO}_2$ & $^{12}\text{CO}_2$ in local and background air

$$\Delta^{14}\text{C} = [({}^{14}\text{C}/{}^{12}\text{C})_{\text{air}}/({}^{14}\text{C}/{}^{12}\text{C})_{\text{std}} - 1] * 1000 \text{ ‰}$$

$$\Delta_{\text{ff}} = -1000 \text{ ‰ (fossil is radiocarbon free)}, \Delta_{\text{bck}} \text{ \& } \Delta_{\text{eco}} \sim 60 \text{ ‰}$$

Apply mixing model to determine fossil fuel CO₂, C_{ff}

1) $C_{\text{obs}} = C_{\text{bck}} + C_{\text{ff}} + C_{\text{eco}}$ --- total carbon, and

2) $\Delta_{\text{obs}} C_{\text{obs}} = \Delta_{\text{bck}} C_{\text{bck}} + \Delta_{\text{ff}} C_{\text{ff}} + \Delta_{\text{eco}} C_{\text{eco}}$ --- radiocarbon

⇒ Accurate radiocarbon measurements allow accurate estimate of fossil fuel CO₂

$$\Delta^{14}\text{C} \sim 2.8 \text{ ‰} \Rightarrow C_{\text{ff}} = 1 \text{ ppm}$$