Errors in top-down estimates of emissions using a known source

Wayne M. Angevine *CIRES, University of Colorado NOAA ESRL* with contributions from many colleagues

Objectives

Examine top-down estimates using aircraft data and forward transport modeling

Identify and classify errors

Estimate uncertainty from remaining unknowns

Try to understand impact of errors on more complex inversions/retrieval methods

Method

Power plant emissions are well known (Location, Amount (hourly)) Some uncertainties remain (Plume rise, Transport time)

Use Martin Lake power plant (east Texas) Sampled several times over 16 years

Met data from ERA5 reanalysis Provides uniform quality over the time span 10-member ensemble plus one control run

HYSPLIT Lagrangian dispersion model Forward mode Full time resolution of emissions (hourly)

Retrieval methods

Plume scaling Mass balance with observed winds & BLH Mass balance with modeled winds & BLH Full plane integration mass balance



25 June 2013 Time series

Four intercepts of SO2 plume Obs in red 10 ensemble members plus control run shown (blue)

First intercept is well aligned, others displaced Simulated plumes are wider than observed Little spread in ensemble Location Width

25 June 2013 Location and magnitude

Integrated concentrations match rather well Control is within ensemble for all four intercepts Ensemble spans 1:1 for all four







Classes of error

Wind direction -> plume displacement
Wind speed -> magnitude, timing w.r.t. time varying emissions
Horizontal dispersion -> plume width
Vertical dispersion -> magnitude
Background -> magnitude
Discretization

Temporal variation of wind can result in "storage" of pollutant and other errors when winds are not updated often enough

Mass balance flux comparison (130625)

Mass balance by several methods: Observed plume, BLH, wind speed Simulated plume, BLH, wind speed -- and combinations

CEMS data (reality) also shown

- BLH is the dominant difference in deterministic runs
- Ensemble members differ substantially, but using different BLH estimates separates them for plumes 1 and 2

BLH range 27%

Wind speed range 21%

Small correlation (R = -0.23)

Neither ensemble spans reality for plume 2

Deterministic estimates range 40%



Why doesn't mass balance using all simulated data reproduce reality perfectly? Violations of MB assumptions (constant wind speed, perfect vertical mixing, known BLH)
Check by integrating entire plume? Flux=7500 (within range of det. estimates and within 11% of reality)







Vertical dispersion

Ensemble of vertical cross-sections showing substantial differences Mixing height estimates differ strongly: Obs o (one value) Model x Concentration profile +

These patterns are similar if model emissions are limited to 1200-1300 CST, or if constant emissions are used

2013 0625 Obs mass balance bias +12% Sim mass balance bias -9% Full plane integration bias +10% Ensemble ranges 19%, 12% (do not overlap CEMS value) Sim mixing height biased low, partially offset by high wind speed



2006 0916 Obs bias +59% Sim bias -22% Full plane bias +12% Ensemble ranges 51%, 41% (concentration BLH overlaps CEMS) Plume not well mixed



2000 0903

Obs bias +0.9% (nearly perfect)

Sim bias +100% (not nearly perfect)

Full plane bias +28%

Ensemble ranges 75%, 122% (both overlap CEMS)

ERA5 mixing heights unrealistically large

HYSPLIT plumes not well mixed

Observed profiles show large change in mixing height during the flight

Plume rise is important and variable on this day



2000 0903

Obs bias +0.9% (nearly perfect) Sim bias +100% (not nearly perfect) Full plane bias +28% Ensemble ranges 75%, 122% (both overlap CEMS) ERA5 mixing heights unrealistically large

HYSPLIT plumes not well mixed

Observed profiles show large change in mixing height during the flight Plume rise is important and variable on this day









2015 0425 Obs bias +5% Sim bias -2% Full plane bias -20% Ensemble ranges 146%, 105% (both overlap CEMS) Simulated plumes not well mixed, variable Not accounting for background uncertainty



Conclusions

No general uncertainty estimate emerges from this study Peischl et al. (2015) estimate 35% uncertainty for methane from this area, which seems realistic

Vertical mixing is the largest source of uncertainty: Plumes not well-mixed (possibly in reality, no data to compare) All retrievals using single-level observations are sensitive to vertical dimension Full-plane integration sometimes gets closer to reality

All retrievals (period) are sensitive to time-varying winds (storage)

Background uncertainty matters (to observations) when plume-to-background ratio is low

Ensemble:

Spread not always sufficient Doesn't always span control run (time discretization?) Adds value by flagging more uncertain situations

Remaining issues: CEMS uncertainty (can you help?)