



Assessment of Discrete Time Filters of Mountaintop CO₂ Observations for Application to Inverse Models

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- Complex terrain poses considerable challenges for our ability to measure and predict exchanges of carbon dioxide between the ecosystem and atmosphere
- U.S. Mountain West is undergoing measurable climatic shifts impacting drought, fire, insect outbreaks and the ability of the region to serve as a sink of atmospheric CO₂
- Approaches for quantifying the carbon exchange differ in seasonal magnitudes and timing of peak uptake over various scales.



Motivation

Bark beetle infestation of lodgepole pine (August 2010, Keystone, Colorado)



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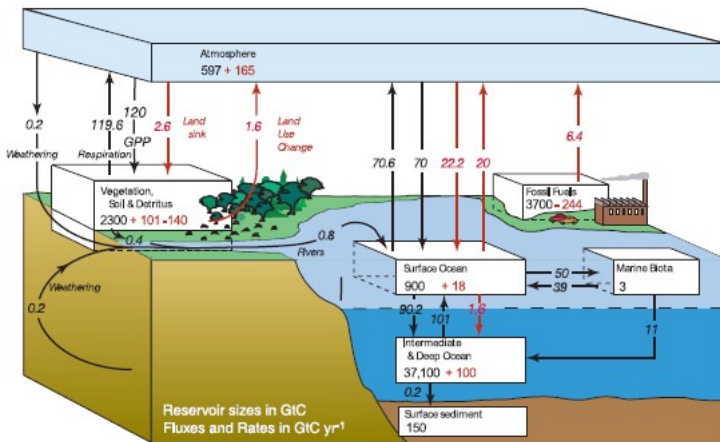
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The global carbon cycle for the 1990s, showing principal fluxes in GtC yr^{-1}
 (taken from IPCC AR4)



- Direct measurement of net ecosystem exchange (NEE) using the eddy covariance technique is preferred over flat landscapes with relatively homogeneous landcover (e.g. grasslands, wetlands, tree canopies)
- NEE responds strongly to changes in temperature and water availability, making it a useful indicator of carbon uptake

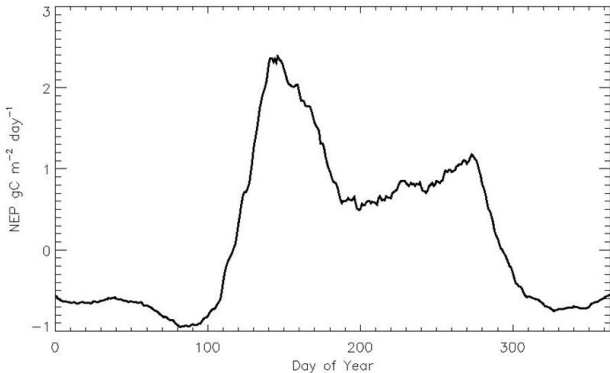
Wisconsin's tall tower at Park Falls





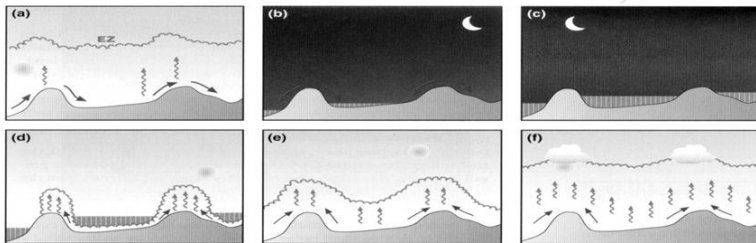
Measuring exchanges of carbon

- A derivative of NEE, net ecosystem production (NEP), showing net carbon gained within the tower's footprint over 1 year



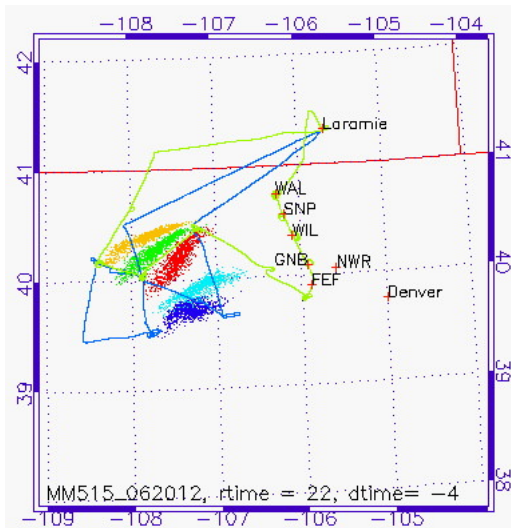
AmeriFlux Niwot Ridge (Colorado) tower

- ① Typically land cover is quite variable
- ② Frequent terrain and thermally driven (**upslope/downslope**) flows
- ③ Models are not good at simulating carbon exchange over terrain



Airborne Campaign in the Mountains Experiment '07 (ACME-07)

- Apprx. footprint: north cen. Colorado
- Cost & Not real-time





Alternatives: Process based (bottom-up) Modeling

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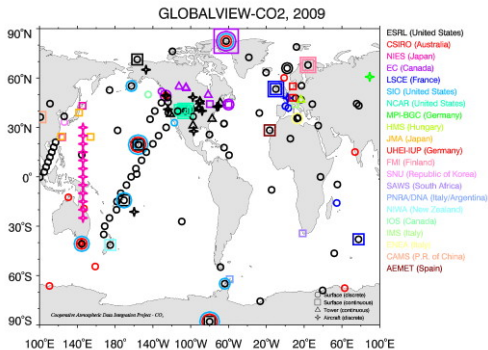
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- Bottom-up models such as NCAR's Community Land Model (CLM) and NOAA's LM3V use remotely sensed products (fPAR, LAI) to drive reaction rates and exchanges.
- Remotely sensed observations can be problematic, particularly in terrain that is subject to frequent disturbance

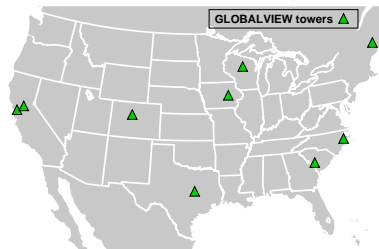
Advantages of this approach for modeling over complex terrain

① Extensive CO₂ concentration monitoring network. Isolated surface stations from the 1950's evolved into sophisticated satellite and *in situ* **observing systems**

② Infrastructure. Mountaintop monitoring of CO₂ has provided both motivation and infrastructural engineering required to study land-atmosphere carbon exchange



- Over complex terrain flux retrieval is preferred **but** results can be confounded by unresolvable flows that skew CO₂ measurements
- Our ability to obtain accurate CO₂ fluxes is limited by a large gap in the monitoring network over Western US





Regional Atmospheric Continuous CO₂ Network (RACCOON) Domain & Seasonal Cycle

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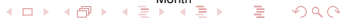
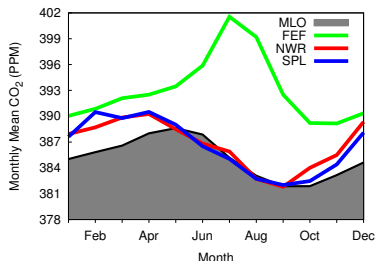
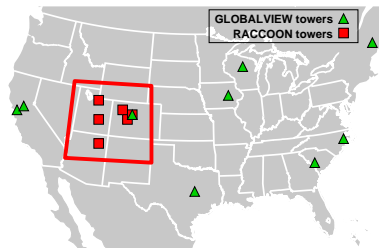
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- RACCOON covers a missing portion of the CO₂ monitoring network in the Mountain West
- Note the difference in seasonal cycles between Mauna Loa (MLO) and RACCOON towers





Example poorly mixed air (left) from Vertical CO₂ Profile over RACCOON (ACME '07 Campaign)

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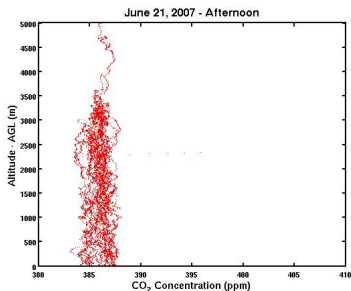
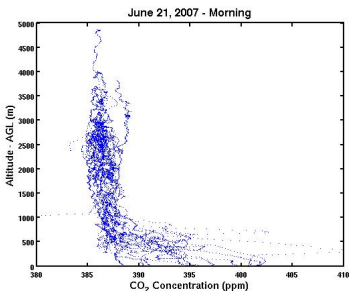
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CO₂ concentration: horizontal axis
Altitude: vertical axis



AIRCOA Design & Network Coverage

- Six RACCOON towers are outfitted with autonomous, inexpensive, robust CO₂ analyzer (AIRCOA) platforms
- Observed CO₂ mole fractions are sampled across multiple inlet heights (3-40 m)

Location and characteristics of the RACCOON towers.

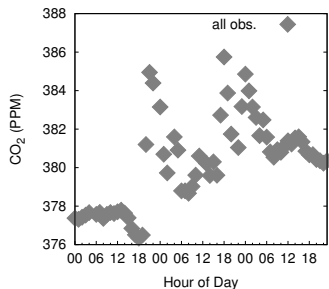
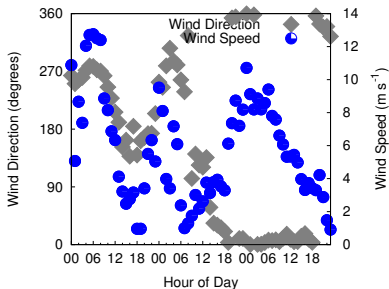
Tower	lat., lon.		Setting
EFS	38.80 N, 109.21 W		marginal plateau
FEF	39.91 N, 105.88 W		alpine valley
RBA	36.46 N, 109.10 W		mountaintop
SPL	40.45 N, 106.73 W		mountaintop
HDP	40.56 N, 111.65 W		mountaintop
NWR	40.05 N, 105.58 W		mountaintop

	Elevation (msl)	tower ht. (m)	Year Installed
EFS	1280	39	2007
FEF	2745	18	2005
RBA	2982	22	2007
SPL	3210	9	2005
HDP	3351	18	2006
NWR	3523	5	2005



Example of Scatter in Tower Data from Niwot Ridge

- 1 Need for rejecting data representative of upslope/downslope flows, nocturnal spikes, and exaggerated afternoon dips caused by local vegetation
- 2 Reducing “sampling error” improves flux retrieval representativeness





Motivating Questions

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- Can **spatial coherence** and simple evaluation of CO₂ time series be used to identify local or regional air mass?
- How do **statistical CO₂ filters** compare with respect to variances, growth rates, and 'flagged data'?
- What do these filters suggest about the use of mountaintop CO₂ observations in inverse models to estimate regional fluxes?



Discrete Time CO₂ Data Filters

- Detection error (**DE**) filters have effective uses as 'first pass' filters of atmospheric CO₂ data. Used by Gillette and Steele (1983) as a basic filter of flask sample data from Niwot Ridge, and Keeling and others (1976) for South Pole data. Filtered subsets are formed:

$$\{X_{i,h} \in x(t) : \sigma_{X_{i,h}} < 1 \text{ and } |X_{i,h} - X_{i,h-1}| < 0.5\} \quad (1)$$

- Statistical interpolation (**SI**) filters are useful for sophisticated signal processing, including NOAA's MLO tower (Thoning and others 1989). Data points from complete set X_j are used to fit a spline $S(X_i)$ curve. Outliers then rejected during iterative passes. Formation of final subset $x(n)$:

$$\{X_j \in x(n) : |S(X_i) - X_j| < 0.5\} \quad (2)$$



Discrete Time CO₂ Data Filters

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- Weighted median (**WM**) filter is designed to adjust for synoptic variability via a dynamically calculated limit l , that is used to form the subset:

$$\{X_i \in x(n) : |\tilde{X}(n) - X_i| < l\} \quad (3)$$

- Spatial **coherence** is designed to identify periods when observations from proximal towers (X_i, X_j) are sampling the same air mass:

$$\{X_i \in x(t) : |X_i - X_j| < 1\} \quad (4)$$



Spatial Coherence

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$$\{X_i \in x(t) : \\ |X_i - X_j| < 1\}$$

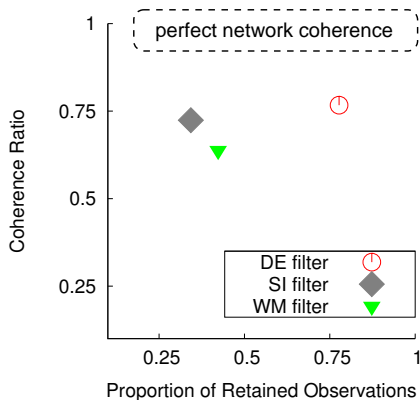
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Basic Statistics

Table: Basic statistics for the complete set of observations (**CS**), and filtered subsets (**DE**, **WM**, **SI**) across years 2005-2010. Normalized sum of squares (SS_N) represents average space-time variability.

	Retained Fraction	Grand Mean	SS_N
CS	1.000	388.5	50.2
DE	0.777	387.1	17.1
WM	0.423	387.5	9.2
SI	0.343	387.5	8.2

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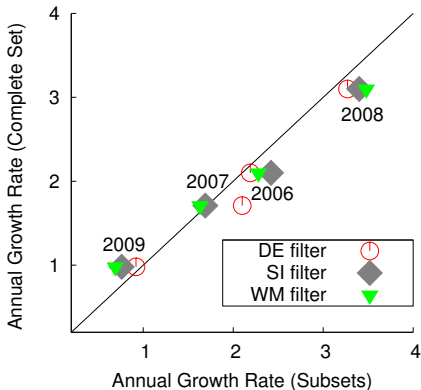
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Mean Annual Growth Rates

Figure: Mean annual CO₂ growth rates by year. Note that growth rates along the vertical axis greater than 2 PPM/year lie below the 1:1 line.



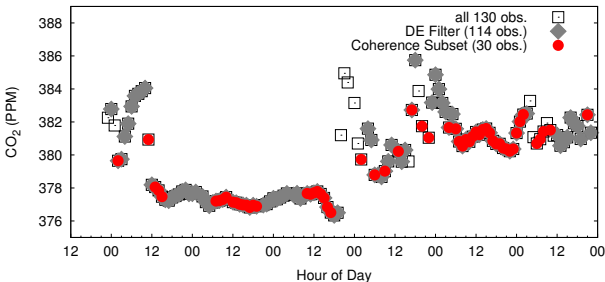
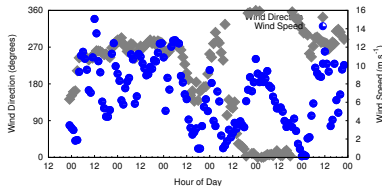
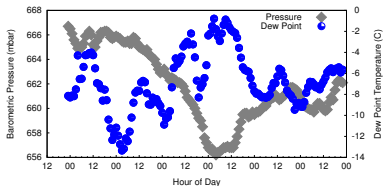


Synoptic Case Studies: DE & Coherence

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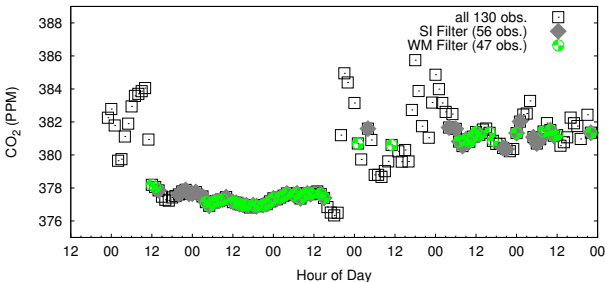
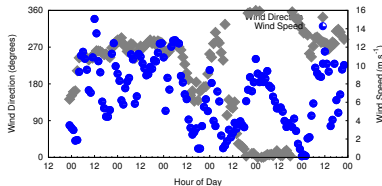
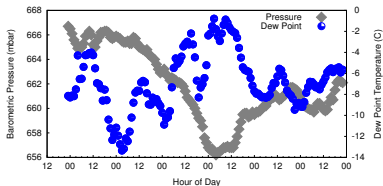


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Relationship between vertical CO₂ gradient and standard deviation

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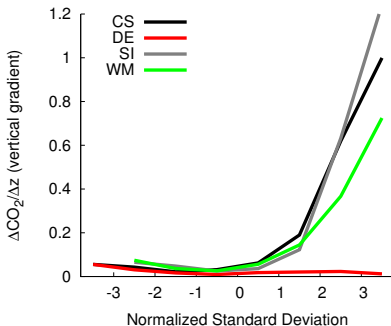
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Relationship between vertical CO₂ gradient and standard deviation. The $\Delta\text{CO}_2/\Delta z$ vertical gradient is plotted for each half-open range of standard deviations ($[-4, -3)$, $[-3, -2)$, ...)





Summary

- When used on nearby towers spatial **coherence** filters spurious observations reasonably well, although output can be sporadic. However, **coherence does not distinguish between well-mixed and poorly-mixed CO₂ samples.**
- The **DE** filter is the least stringent, permits the most spurious observations, largely because it **can only account for detection and sampling error across the height of the tower.**
- **SI & WM** filters reject more than half of the complete set of observations.
- The **SI** and **WM** filters remove the most spurious (non-synoptic) observations.

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Future Work: Experimental Design for Estimating Regional Fluxes

- CarbonTracker (CT) is an inverse tracer transport model that optimizes flux estimates using observed CO₂ mole fractions (www.esrl.noaa.gov/gmd/ccgg/carbontracker/)
- CT branched-run simulations assimilating various collections of RACCOON observations:
 - **Control.** Standard CT simulation assimilating data from 2 RACCOON towers
 - A. **Complete set** from 4 RACCOON towers assimilated
 - B-D. **Filtered subsets** from 4 RACCOON towers
 - 1 Nocturnal observations only
 - 2 Nocturnal and daytime observations
 - 3 Tower elevation adjusted to match model orography (nocturnal and daytime)



Future Work: Multi-model Regional Calibration Study

- In line with AmeriFlux and NACP objectives, carry out a diagnostic evaluation of carbon budgets over US Mountain West from forward and inverse models
 - How much do forward and inverse models differ with respect to time averaged carbon budgets and seasonal cycles over the US Mountain West?
 - Does the mismatch between carbon exchange estimates exceed model error estimates?
- 'Apples-apples' comparison of process based models using the same transport model

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This talk:

<http://www.climatemodeling.org/~bjorn/talks/2010-iastate.pdf>

RACCOON homepage: <http://raccoon.ucar.edu>

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