

The Utility of Continuous Atmospheric Measurements for Identifying Biospheric CO₂ Flux Variability

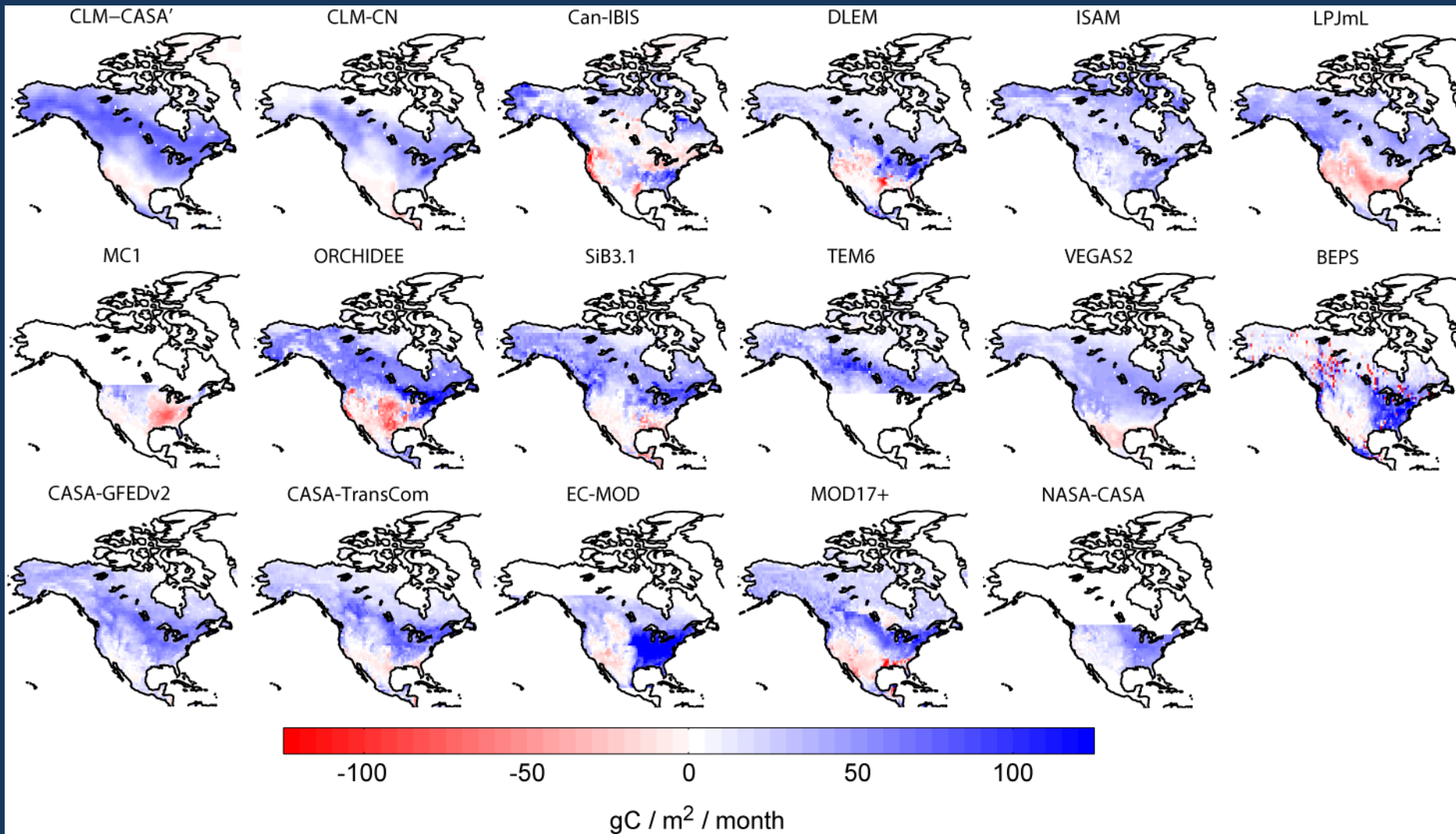
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Sharon Gourджи, Kim Mueller, Anna Michalak

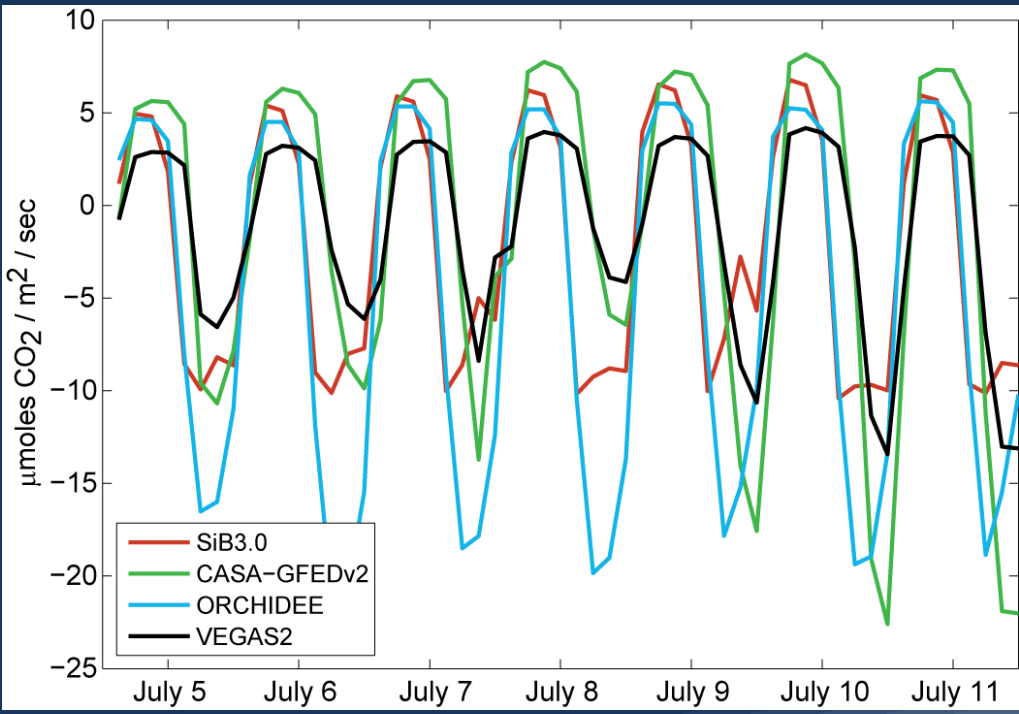
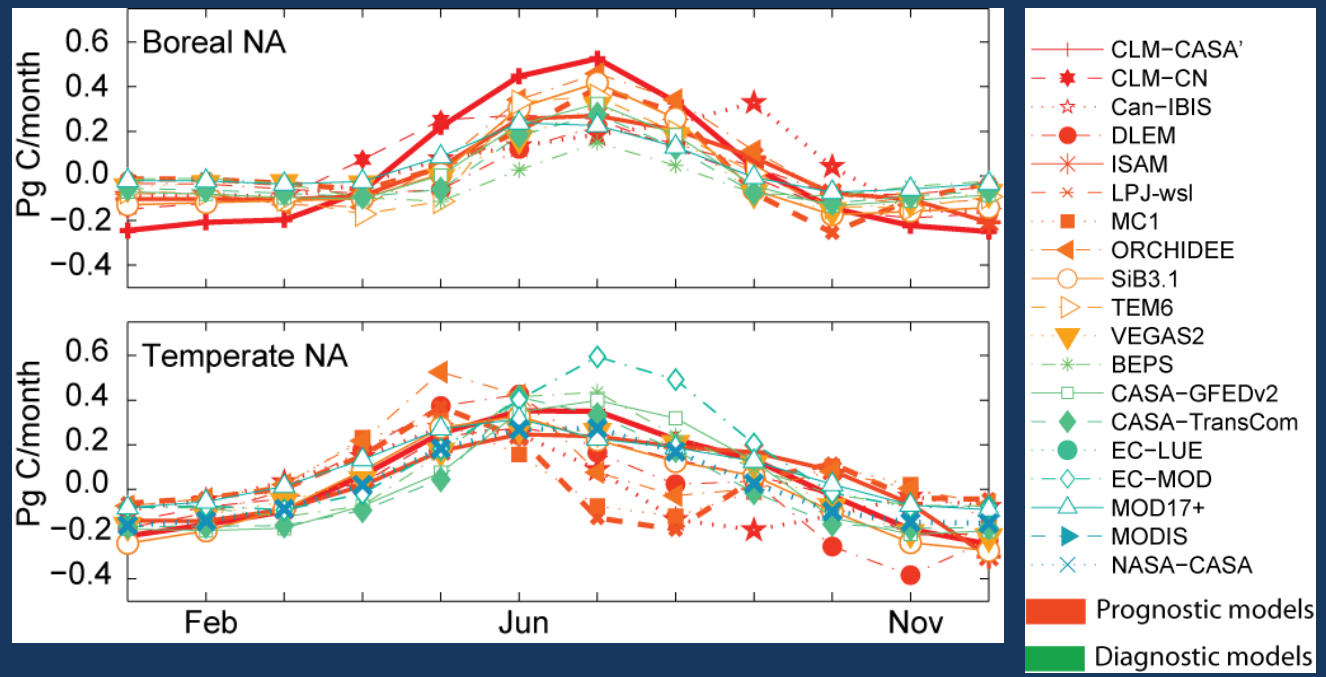
University of Michigan

Fall AGU, San Francisco, December 17th, 2010

Long-Term Mean (2000-2005) Summer (June, July, August) Net Ecosystem Productivity



Seasonal Cycle of NEP



Diurnal Cycle of NEE at WLEF

Which Model is “Better”?

- Is one model better than another?
 - Compare models to data / observations
- Can atmospheric observations be used as a data constraint?
 - Are the fluxes predicted by a given model compatible with the atmospheric observations?

Depends on whether the atmospheric data can detect differences among competing flux distributions.

Objective

Determine how much information atmospheric CO₂ observations can provide in either:

- (1) Evaluating pre-existing sets of surface flux estimates (e.g., from TBMs) across North America.
- (2) Estimating surface flux distributions at regional scales (e.g., from inversions).

Objective

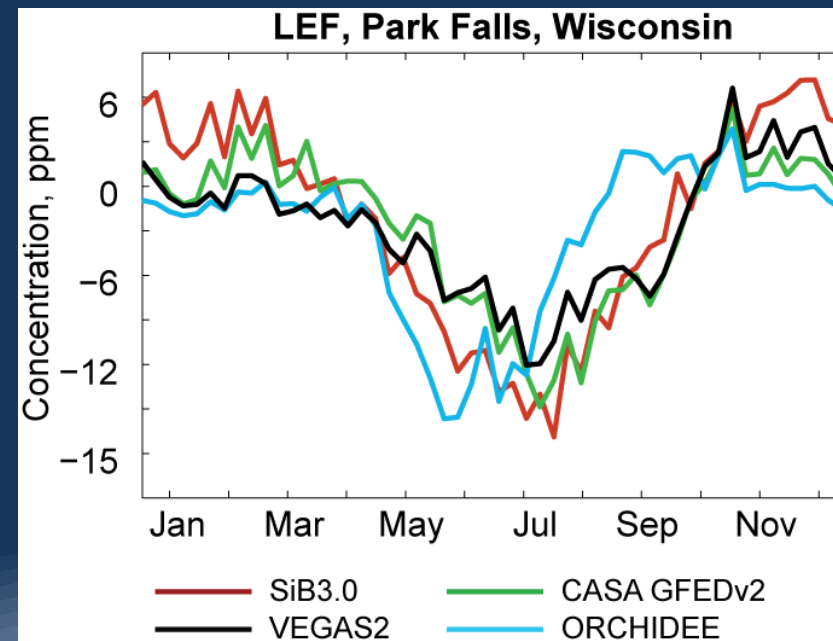
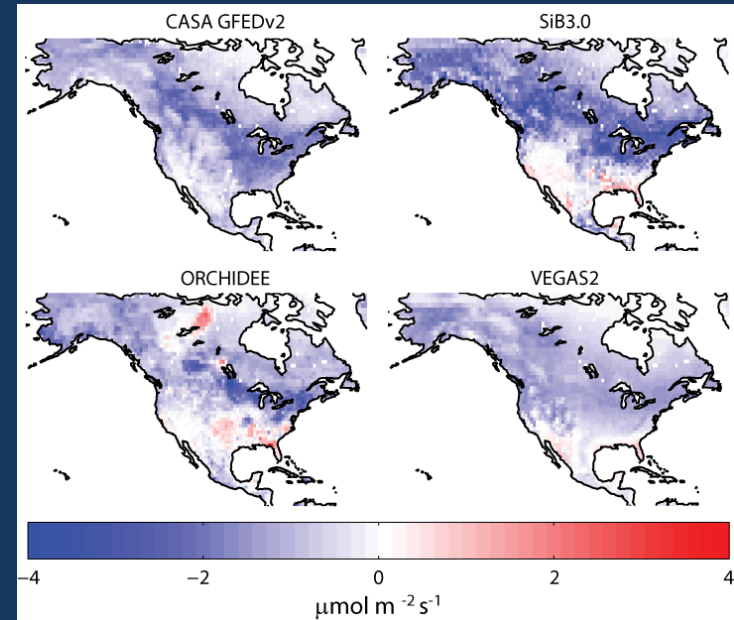
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Approach

- NEE estimates of 4 terrestrial biospheric models (TBMs) are used to represent **plausible scenarios** of surface flux distributions.
- TBMs coupled with the atmospheric transport model, WRF-STLT.
- Resulting atmospheric signals are compared at 9 towers in the continuous observation network (2004).

Mean 2004 summer (June, July, August)
net ecosystem exchange (NEE)



How different do signals have to be?

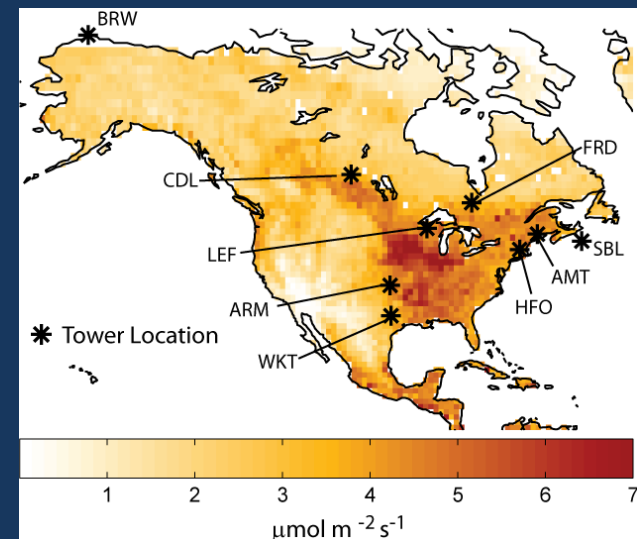
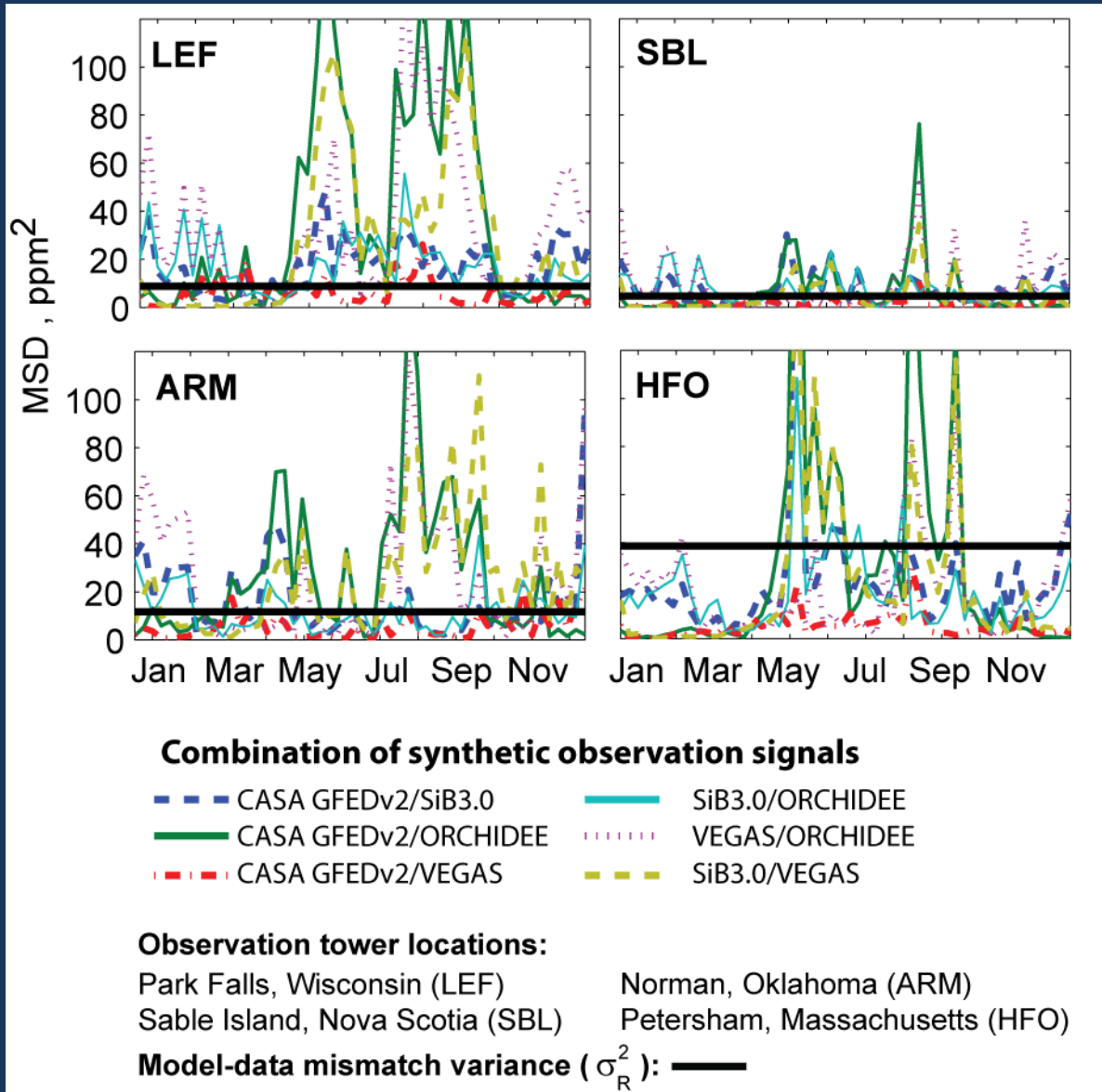
- Mean squared difference (MSD) is used to quantify the differences among pairs of synthetic observation signals from different TBMs.

The differences between the synthetic signals are compared within the context of expected or **estimated model-data mismatch error** (σ^2_R).

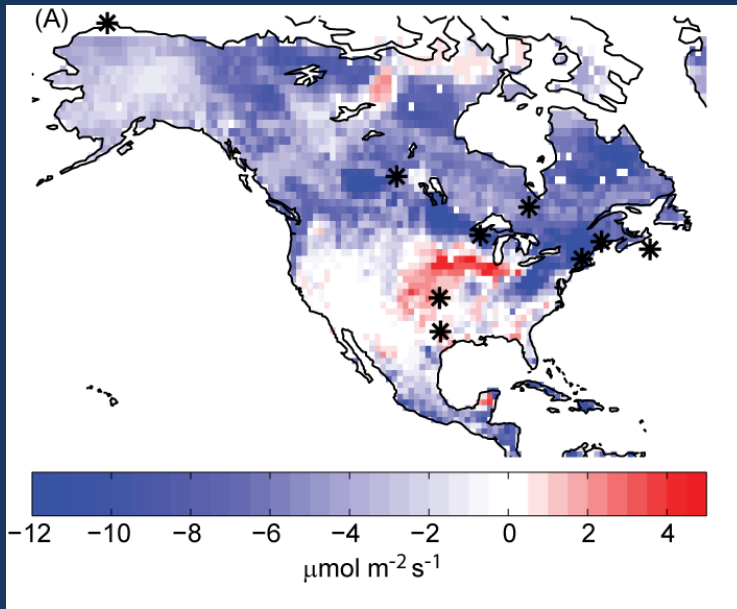
Aggregation, representation, and transport model error

Tower Name	Location	Height (m)	Tower Type	σ^2_R (ppm ²)
LEF	Park Falls, WI	396	Tall	8.8
WKT	Moody, TX	457	Tall	5.9
SBL	Sable Island, Nova Scotia	25	Marine boundary	4.7
BRW	Barrow, AK	10	Marine boundary	1.7
ARM	Norman, OK	60	Short	11.7
HFO	Petersham, MA	30	Short	38.6
AMT	Argyle, ME	107	Short	19.6
FRD	Fraserdale, Ontario	40	Short	7.9
CDL	Candle Lake, Saskatchewan	30	Short	4.0

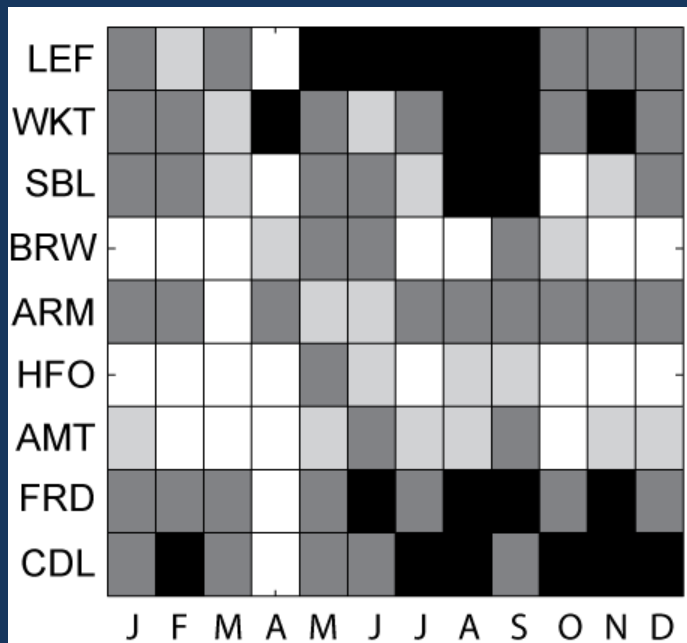
Compare differences in synthetic signals with estimated model-data mismatch variance at the tower.



Overall combined influence of differences in both the spatial distribution and magnitude of fluxes:



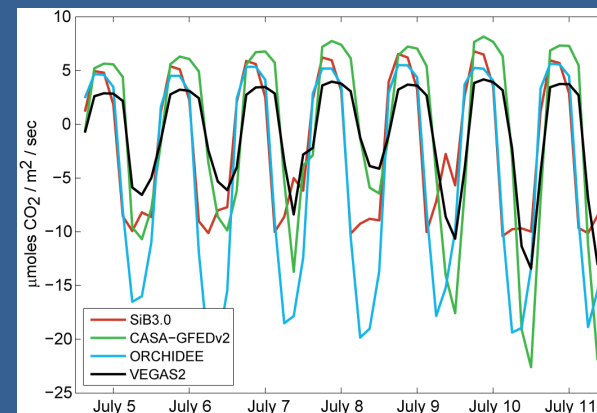
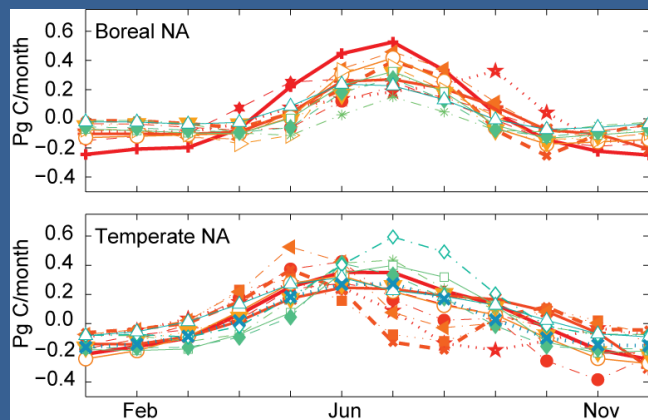
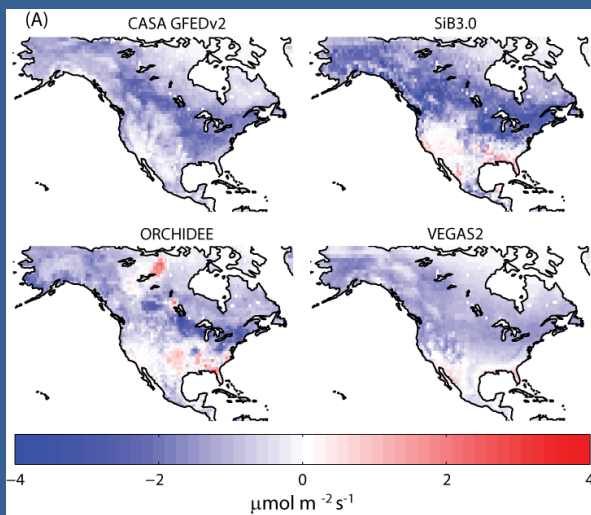
Large-Scale Spatial	Seasonal Cycle	Fine-Scale Spatial	Diurnal Cycle
✓	✓	✓	✓



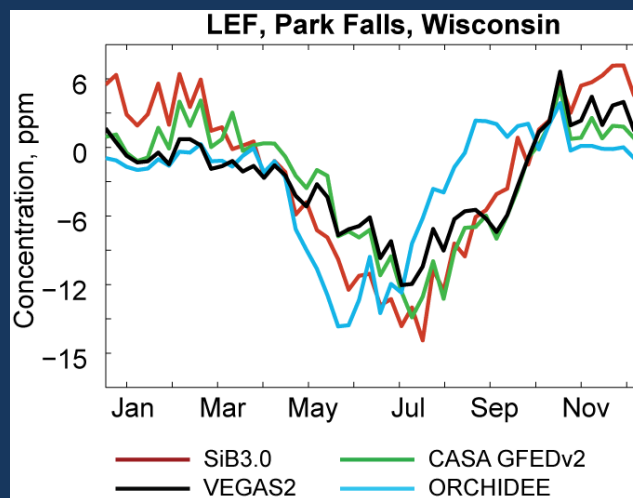
- 4 TBMs generate statistically different 3-hourly CO₂ time series during most months of the year.
- Differences are less detectable at towers with a higher model-data mismatch variance.
- Measurements **can detect overall differences** in CO₂ concentrations resulting from competing flux distributions

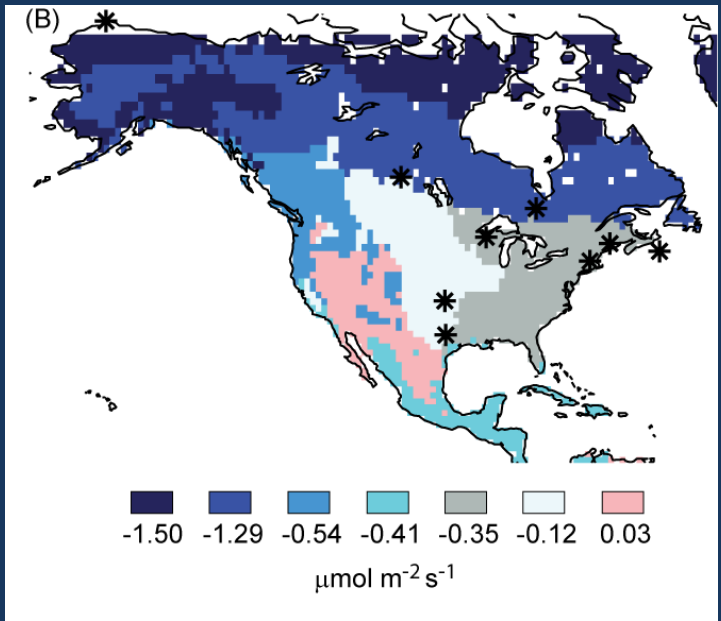
At what scale (spatial & temporal) is atmospheric data most informative?

How does the **spatial** and **temporal** variability in surface flux



translate into the variability observed in **synthetic CO₂ concentrations?**





Remove sub-ecoregion scale variability

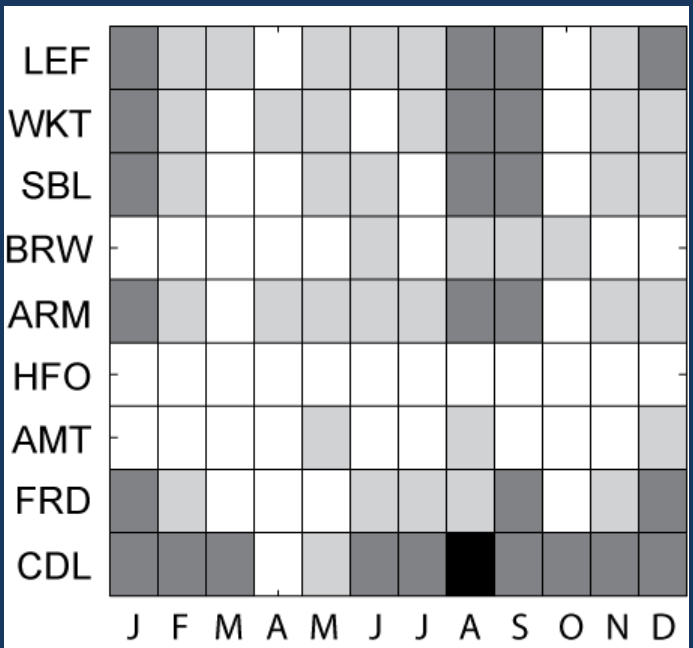
Isolate the influence of differences in regional flux magnitude on the generated CO₂ signals.

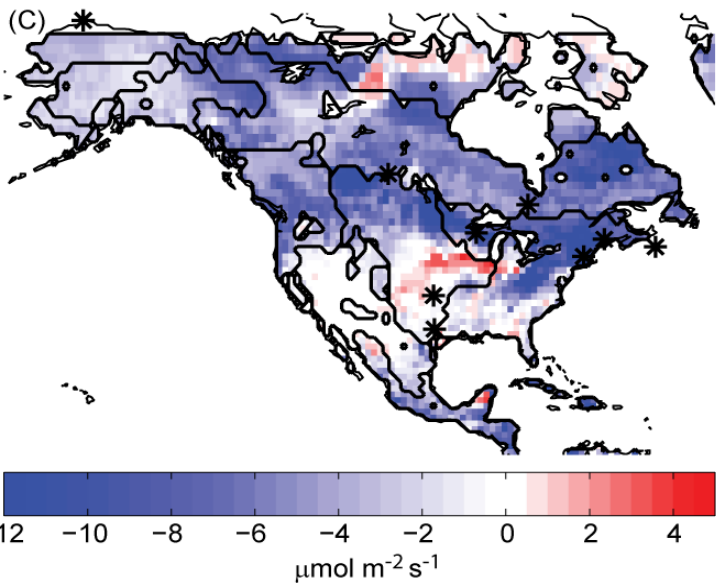
Large-Scale Spatial	Seasonal Cycle	Fine-Scale Spatial	Diurnal Cycle
✓	✓		

Overall, the towers are able to detect differences in flux magnitude over large regions.

Encouraging for TBM evaluation

Observations can be used to discriminate among large-scale fluxes as predicted by different TBMs.





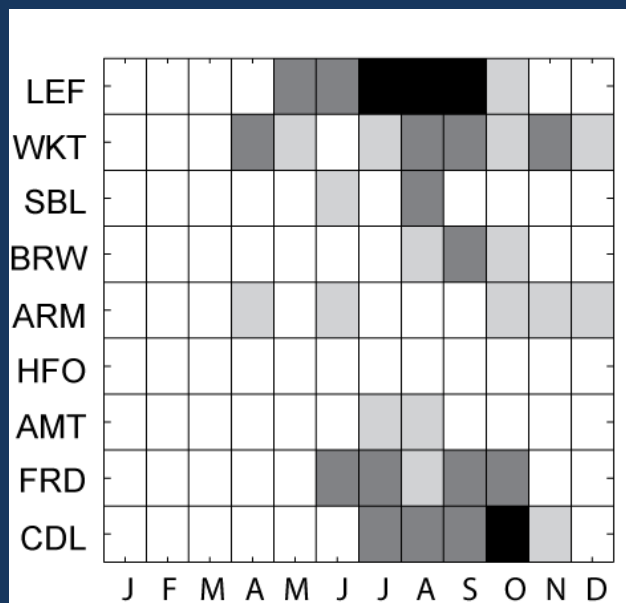
Normalized net ecoregion scale flux

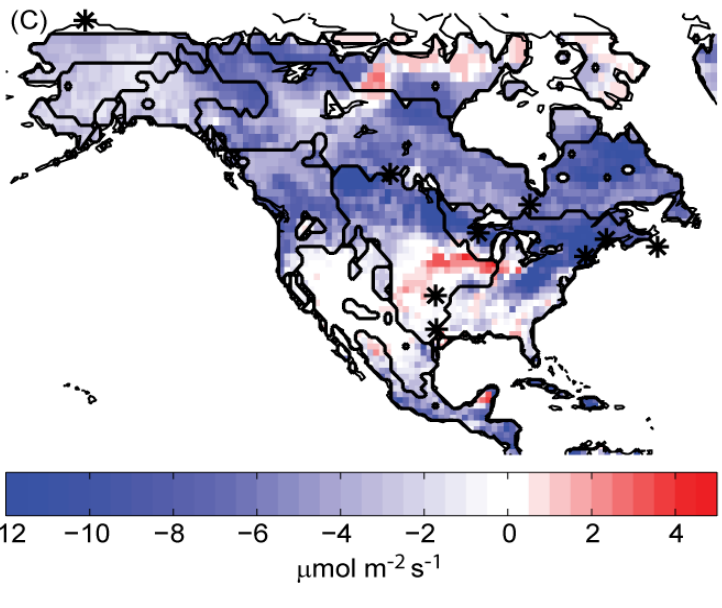
Examine how the distribution of fluxes within ecoregions influences CO₂ concentrations

Large-Scale Spatial	Seasonal Cycle	Fine-Scale Spatial	Diurnal Cycle
		√	√

Atmospheric measurements detect fine-scale (spatial, temporal) flux differences during the growing season.

What is the **relative importance** of spatial distribution of fluxes compared to differences in their diurnal cycle (e.g., timing, strength)?





Normalized net ecoregion scale flux

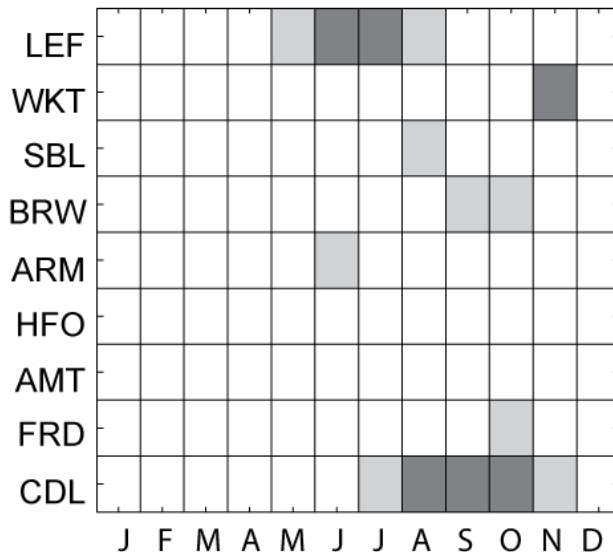
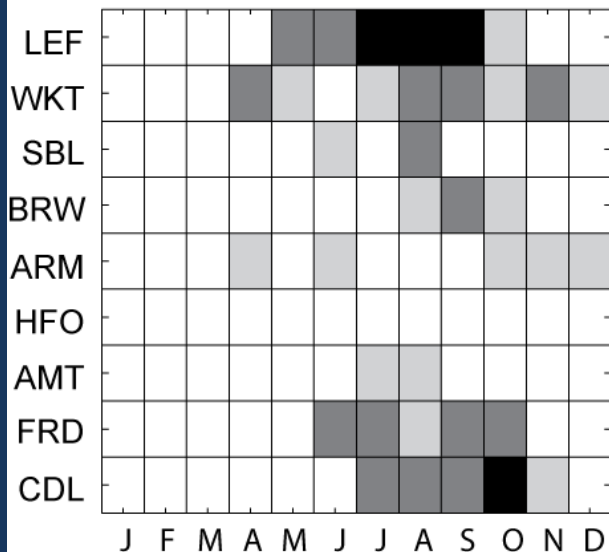
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Large-Scale Spatial	Seasonal Cycle	Fine-Scale Spatial	Diurnal Cycle
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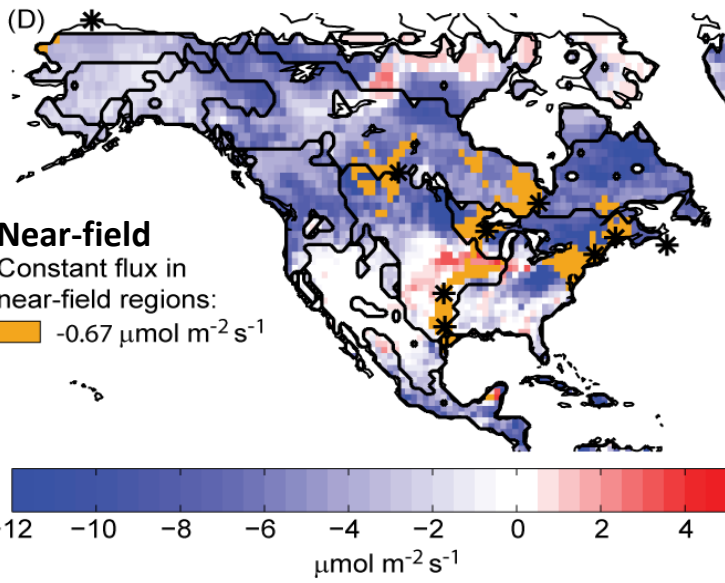
√

Remove diurnal cycle

No diurnal cycle



Differences in signals primarily driven by differences in diurnal cycle of fluxes between the TBMs

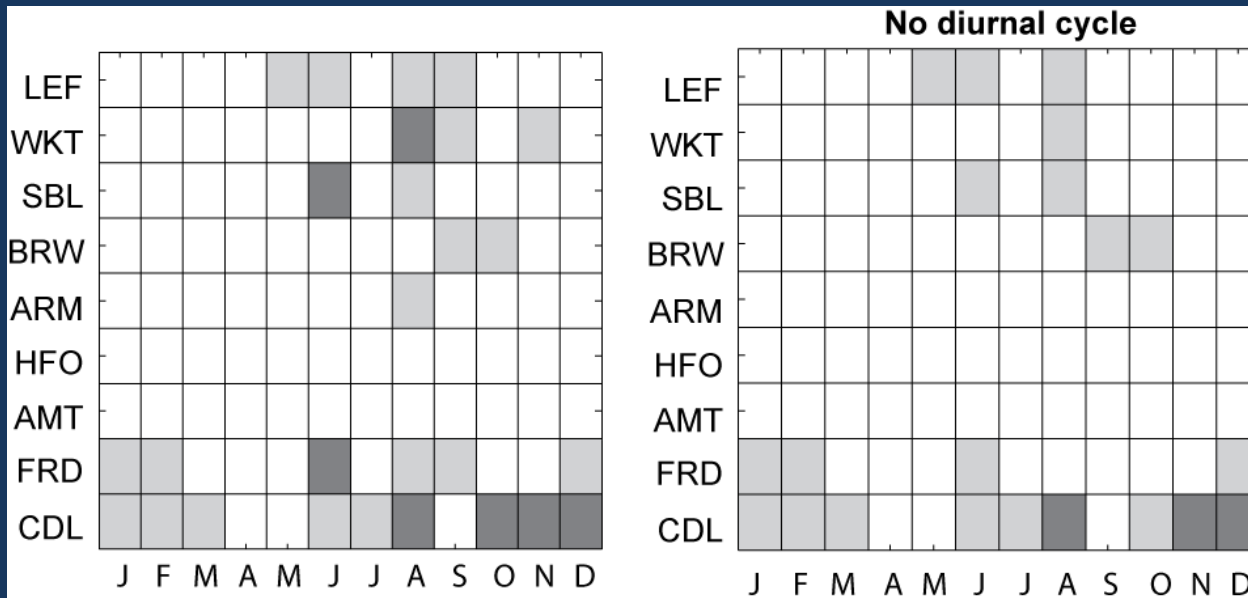


Isolating influence of far-field fluxes

Assess the impact of the fine-scale variability beyond the near-field of the towers

Large-Scale Spatial	Seasonal Cycle	Fine-Scale Spatial	Diurnal Cycle
		✓	✓

Only in far-field



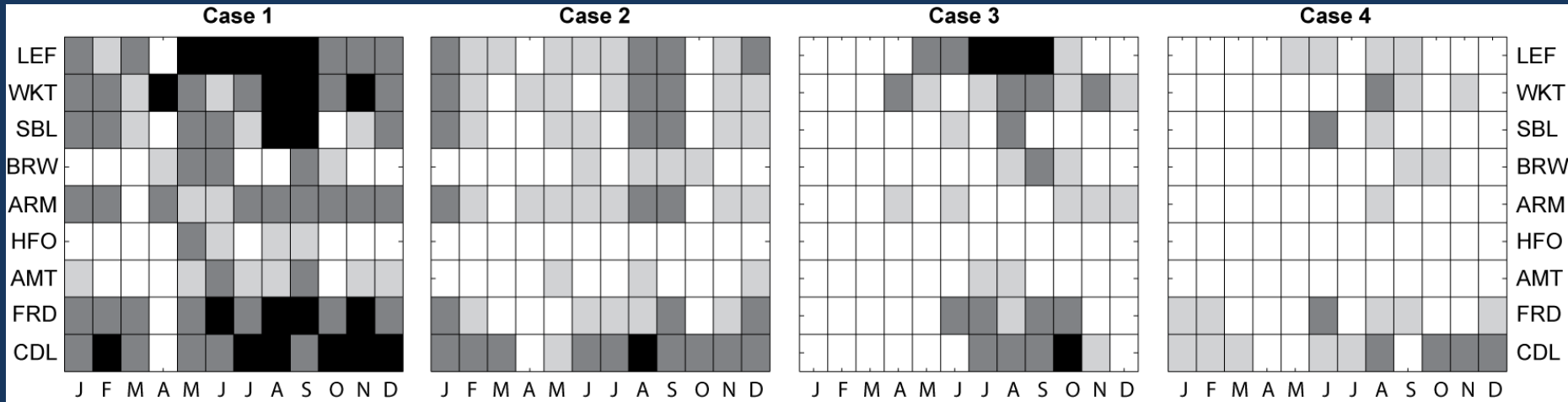
When remove diurnal cycle results, are largely unchanged.

Both large and small scale variability

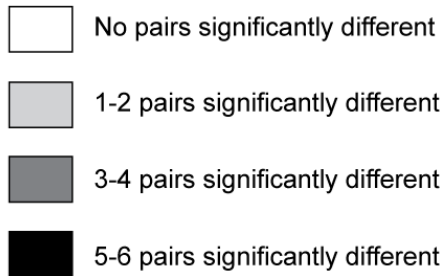
Large scale variability only

Small scale variability only

Small scale variability in far-field

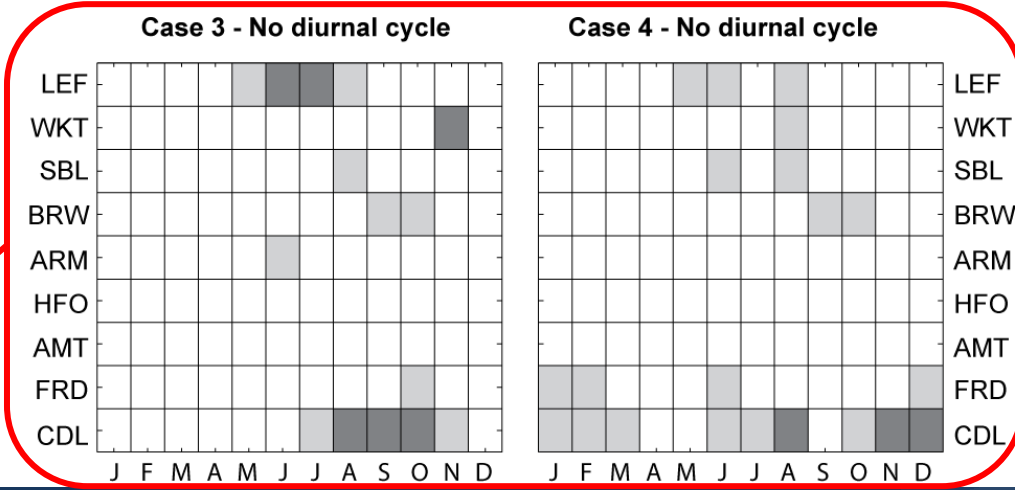


At a 0.05 significance level:



Case 3 - No diurnal cycle

Case 4 - No diurnal cycle



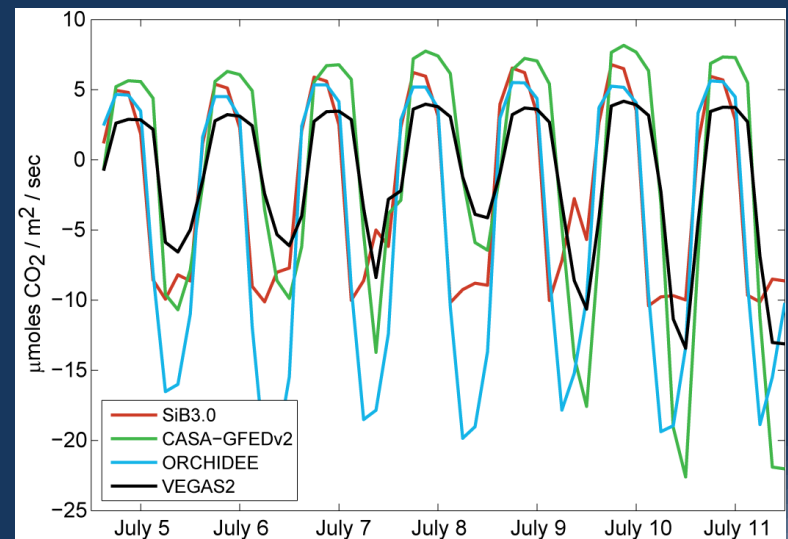
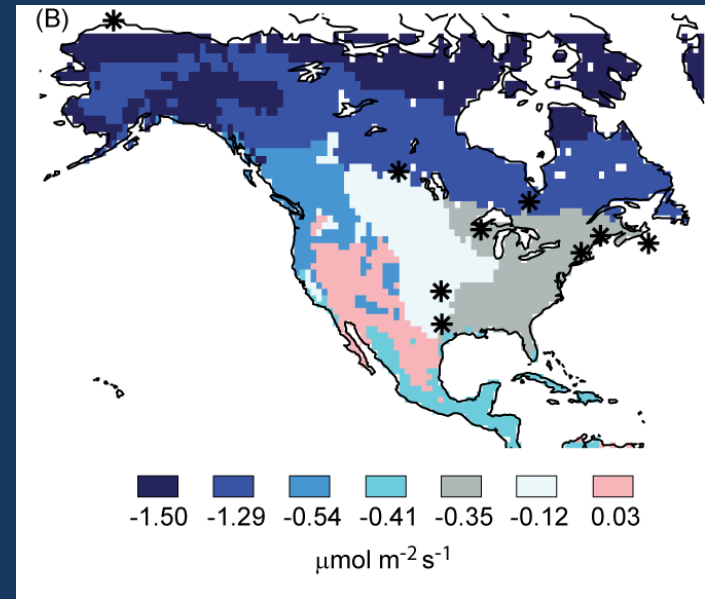
Diurnal cycle removed

Conclusions

Atmosphere data can detect **large scale** differences in flux magnitude among competing TBMs.

Atmospheric signal is **very sensitive** to slight differences in the **diurnal cycle** of fluxes represented by the models.

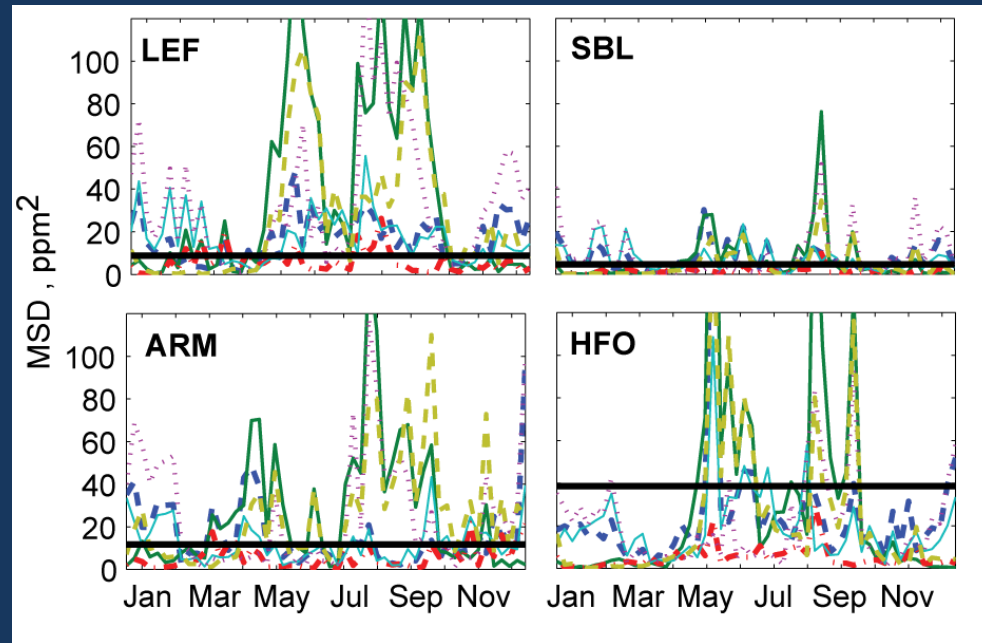
Important for both in inversions and process-oriented TBMs.



Conclusions

Magnitude of the **model-data mismatch error or variance** has a large impact on results.

As we **improve atmospheric transport** modeling (i.e., reduce uncertainties), we will be able to **detect more differences** among competing TBMs.



Acknowledgements

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