

# List of Papers

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Today

## First paper ‘Breathing of the terrestrial biosphere’ [1]

- Eddy covariance - measures the covariance between fluctuations in vertical velocity and the mixing ratio of trace gases of interest. Seasonal C flux.
- -ve values of  $F_n$  represent a loss of  $CO_2$  from the atmosphere and a gain by the surface.  $F_n$  - net ecosystem exchange,  $F_a$  - Gross canopy assimilation,  $F_r$  - ecosystem respiration.
- Hard to find natural stands of vegetation that meet the right standards for good eddy covariance measurement. Problems with night time measurement. EC not able to measure fluxes of complex terrain.
- Disturbance (fire, drought) needs to be incorporated into models. Annual variability in carbon fluxes.
- Mid aged forests (50-100 years) are best carbon sinks.
- Network now contains more varieties of forest and vegetation so PD has shifted towards zero (fig. 4).

## Second paper ‘An Improved analysis of forest carbon dynamics using data assimilation’ [13]

- Measurements from young pine stand in ponderosa oregon (warm dry summers wet cool winters).
- DA combining measurements of C stocks made over time and measurements of C flux made directly (EC), along with model.
- Ensemble Kalman filter being used.
- Results show significantly more accurate analysis when using DA than model alone.
- Monte-Carlo technique?
- dynamics of soil C and WD pools an area of uncertainty.
- No energy balance? lack of models to describe photosynthesis and plant growth.

## **Thesis Laura ‘Correlated observation errors in data assimilation’ [10]**

- Chapter 2: Different methods of DA and approximations. Error covariances and correlations, Desroziers method?
- Chapter 3: Matrix approximations to include some correlations for  $R$  and avoid the need to explicitly calculate  $R^{-1}$  (computationally taxing). Calculate trace of analysis error covariance matrix,  $S_a$ , to understand how useful an observation set is to the DA scheme (smaller trace  $S_a \implies$  better reduction of error variance). May have to calculate  $S_a^*$  if  $R$  known to be specified incorrectly. Look at vector and matrix norms.

## **The REFLEX project: comparing different algorithms and implementations for the inversion of a terrestrial ecosystem model against eddy covariance data [6]**

- Data given to participants to use with DALEC and DALEC-D and their choice of MDF algorithm. Tristan’s results not used?

## **Effect of correlated observation error on parameters, predictions, and uncertainty [12]**

- Hydrology, Considering a denitrification model with diagonal and then full weight  $\mathbf{R}$  matrix.

## **A regularization of the carbon cycle data-fusion problem [4]**

- Sensitivity testing of DALEC model with 4D-VAR.

## **Diagnosis of obs, background and analysis-error statistics in obs space [5]**

- Looking at  $d_o^b = y^o - H(x^b)$  and  $d_o^a = y^o - H(x^a)$  for finding  $R$  and  $B$  with covariances.

## **Dynamic Data Assimilation: a least squares approach [7]**

- Model adjoint Linear 4DVAR observation operators good fundamental DA reference.

## **SiPNET paper Braswell 2005 [2]**

- More complicated Carbon balance model to look at and understand!

## **Estimating parameters of a forest ecosystem C model with measurements of stocks and fluxes as joint constraints [8]**

- Reason Tristan wrote this project, can do better?
- Parameter estimation for DALEC using MC procedures.
- Woody biomass increment and to a lesser degree soil respiration measurements give marked reductions in uncertainties in parameter estimates and model predictions as they are ‘orthogonal’ constraints to NEE measurements.
- None of the data good at constraining fine root or soil C pool dynamics, need new measurements?

## Observation error correlations in IASI radiance data [11]

- Satellite takes same obs. every time step, Flux tower variable? Mainly temporal correlations possible!

## Data assimilation with correlated observation errors: experiments with a 1-D shallow water model [9]

- Use an idealised perfect model and approximate forms of R to include correlations.

## Where does the carbon go? A model–data intercomparison of vegetation carbon allocation and turnover processes at two temperate forest free-air CO<sub>2</sub> enrichment sites [3]

- Use temperate forest free-air CO<sub>2</sub> enrichment (FACE) to test. How better to control allocation parameters, allocation should be controlled by functions rather than parameters.

## References

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