

Scaling carbon dioxide exchange from sites to

regions

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Challenges in scaling terrestrial carbon fluxes

- A vast, complex system operating with poorly known feedbacks, periodic/stochastic behaviour...now in uncharted territory
- Eddy flux and biometric data 'effective samples?'
- Earth observations -'shadows on the wall?'
- Modelling 'reality?'





CARDAMOM - CARbon DAta MOdel fraMework



DATA

MODIS LAI time series, **Biometric Satellite data, Eddy** flux tower data, Plant trait data.

OPTIMIZATION

Metropolis-Hastings Markov Chain Monte Carlo **Ecological & dynamic** constraints





DALEC: Data Assimilation Linked Ecosystem Carbon model



phenology and turnover rates in ecosystem carbon cycling.

DALEC Parameter vector = $[M_r, f_a, f_f, f_r, L_L, t_w, t_r, t_{lit}, t_{SOM} T_{rate}, C_{eff}, B_{day}, f_{lab}, R_r, F_{day}, R_f, LMA, C_{LA}, C_{FO}, C_{RO}, C_{WO}, C_{LI}, C_{SO}]$ National Centre for Earth Observation Natural Environment research council Natural Environment research council

Model Data Fusion (MDF)



Bayes' Theorem

 $p(x|c) \propto p(c|x) p(x)$ Observation Posterior Prior

parameter probability likelihood, Parameter Probability parameters

Method = Metropolis Hastings MCMC

given

(1) Parameter value priors span across multiple orders of magnitude, BUT

(2) Only a subset of parameter space can be considered "ecologically consistent"





Ecological and Dynamic Constraints (EDCs)

 $\begin{array}{l} \mathsf{DALEC Parameter vector} = [\mathsf{M}_{\mathsf{r}},\,\mathsf{f}_{\mathsf{a}},\,\mathsf{f}_{\mathsf{f}},\,\mathsf{f}_{\mathsf{r}},\,\mathsf{L}_{\mathsf{L}},\,\mathsf{t}_{\mathsf{w}},\,\mathsf{t}_{\mathsf{r}},\,\mathsf{t}_{\mathsf{SOM}}\,\mathsf{T}_{\mathsf{rate}},\,\mathsf{C}_{\mathsf{eff}},\,\mathsf{B}_{\mathsf{day}},\,\mathsf{f}_{\mathsf{lab}},\,\mathsf{R}_{\mathsf{r}},\,\mathsf{F}_{\mathsf{day}},\,\mathsf{R}_{\mathsf{f}},\,\mathsf{LMA},\,\mathsf{C}_{\mathsf{LA}},\!\mathsf{C}_{\mathsf{FO}},\,\mathsf{C}_{\mathsf{RO}},\,\mathsf{C}_{\mathsf{WO}},\,\mathsf{C}_{\mathsf{LI}},\,\mathsf{C}_{\mathsf{SO}}] \end{array}$



Analytical dynamic constraints

order of magnitude constraint on proximity of steady state C pool to initial value (wood, roots, litter, SOM)



In total: 12 checks



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EDCs reduce flux bias and assist parameter constraint



Synthetic studies – 40 synthetic deciduous forests Assimilate: LAI time series, single soil carbon estimate





 xs_{EDC} (median NPSE= 0.281)

Independent tests at flux sites indicate value of EDCs



Independent tests of cumulative NEE at Ameriflux sites



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Mean monthly NEE at 1° x 1°

2001-2010: global terrestrial carbon cycle analysis.









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CARDAMOM fluxes 2001-2010: key results



(A) Global carbon flux estimates

Gross Primary Production GPP₀₁₋₁₀ = 123.2 \pm 7.5 PgC yr⁻¹

Net Ecosystem Exchange NEE₀₁₋₁₀ = -1.8 \pm 2.7 PgC yr⁻¹

Global carbon pool totals Labile = 4 ± 2 Pg C Foliar = 11 ± 7 Pg C Fine Roots = 16 ± 12 Pg C Wood = 535 ± 298 Pg C Litter = 17 ± 14 Pg C SOM = 1415 ± 735 Pg C





CARDAMOM DALEC – Posterior parameters







Next steps – perturbed systems

- Assimilating burned area data (MODIS) and deforestation time series (LandSat...)
- Assimilating sequential biomass maps (ALOS, BIOMASS...)
- Including croplands and other human managed systems







Conclusions

- CARDAMOM provides a data-consistent approach to terrestrial C analysis across scales
- Multiple data-streams combined with constrained mass balance modelling allow estimation of C dynamics from local to global
- Coupling to N and water cycles to allows link to further EO/field data constraints and process interaction







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