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# **Ecological Synthesis**

The art of scalable science

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www.keenangroup.info

#### FLUXNET: Global measurements

#### of earth-atmosphere exchange



#### FLUXNET: Global measurements

#### of earth-atmosphere exchange



### The Terrestrial Carbon Cycle

AIRS Mid Troposphere Carbon Dioxide 365 368 370 373 375 378 380 383 0.0 0.3 0.6 16 Day Composite MYD13C1 Jan

NASA Scientific Visualization Studio

http://svs.gsfc.nasa.gov/goto?3947

#### 'Ecosystem Science' joins the league of big data.



#### Farley et al., 2018

Today's big data is tomorrow's drop in the hard drive

Worldwide data volume doubled nine times between 2006 and 2011, with exponential growth continuing this decade (Chen et al. 2014)

Growth has outpaced the annual doubling in computing power predicted by Moore's law (Olofson and Eastwood 2014)



#### FLUXNET: Global measurements

#### of earth-atmosphere exchange



#### Today's big data is tomorrow's drop in the hard drive

#### FORUM

# The "Data-rich but Information-poor" Syndrome in Water Quality Monitoring

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ABSTRACT / Water quality monitoring conducted routinely over time at fixed sites has been a part of most water quality management efforts for many years. It has been assumed that such monitoring plays a major role in management. However, the lack of routine data analysis, and reporting of information derived from such analysis, points up the fact that the exact nature of the role of routine, fixed-station monitoring is poorly defined.

There is a need to very clearly define this role in the design of such systems if routine monitoring is to efficiently and effectively meet the information expectations placed on it. Design of routine monitoring systems will therefore have to consider not only the where, what, and when of sampling, but also why. A framework for including the "why" of monitoring in the design process is proposed and experience with using the framework in New Zealand is discussed.

#### Ward et al. 1986. Environmental Management

A post-science vision of the future

- Fishing expeditions through data-mining

- Hypothesis-free science



- Engage scientists with data gathering
- Foster a broad and deep understanding of the field
- Develop hypothesis guided approaches
- Develop the computational skills needed to test them



#### Lindenmayer, Likens & Franklin, 2017

Approach & Philosophy



"....forget about the numbers! We must <u>simplify and idealize</u> to develop robust understanding."



Approach & Philosophy



"....forget about the numbers! We must <u>simplify and idealize</u> to develop robust understanding."



"we need accurate and believable predictions to guide **policy**, which **requires complex models**...."

Policy

Theory

**Scalable Understanding** 

Approach & Philosophy





Understanding



"....forget about the numbers! We must <u>simplify and idealize</u> to develop robust understanding." "we need accurate and believable predictions to guide **policy**, which **requires complex models**...."

Policy

Theory

**Scalable Understanding** 

Approach & Philosophy

# Theoretical framework for synthesis?

# Ingredients of an effective synthesis study?

# Theoretical framework for synthesis?



### Ingredients of an effective synthesis study?

#### A theoretical framework for synthesis design



Adapted from Lindenmayer and Likens, 2010



# A question!

**Example 1:** 

"Examine variability in light use efficiency between sites"

# A question!

**Example 1:** 

"Examine variability in mortality between sites."

#### Example 2:

"What is the relative magnitude of environmental vs. PFT control of variability in mortality?"

#### Example 3:

"Variability in mortality is driven more by environmental forcings than PFTs."

A question!Formulated theory

A question!

Formulated theory

E.g.,

LUE: Is there physiological basis that allows us to predict changes in LUE?

Phenology: What theories exists, how do they differ, and can they be tested?

A question!Formulated theoryA refined target

- A question!
- Formulated theory
- A refined target
  - More data does not mean better data
  - LUE: cloudy, clear, solar zenith angle, etc.
  - Phenology: Summer, winter, night-time vs day-time
  - WUE: rain, VPD, soil moisture, canopy closure

- Formulated theory
- A refined target
- Continual refinement
  - The more you can refine at the small scale, the easier scaling will be

- Formulated theory
- A refined target
- Continual refinement
- Start local
  - Start small a site and year you trust
  - Develop the analysis as fully as possible
  - Think about how the question might scale

- Formulated theory
- A refined target
- Continual refinement
- Start local
- Modularly scaled
  - Think about workflow, function design
  - The utility of intermediates

- Formulated theory
- A refined target
- Continual refinement
- Start local
- Modularly scaled
- Ancillary measurements















#### Ancillary Measurements



#### Long-term changes in ecosystem function at Harvard Forest



Keenan et al. (2012)



#### Long-term changes in ecosystem function at Harvard Forest





#### Long-term changes in ecosystem function at Harvard Forest



# Flux-driven modeling at Harvard Forest



**Ancillary Measurements** 

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Keenan et al. (2012) Global Change Biology



**Ancillary Measurements**
# Data available at Harvard Forest

- Net Ecosystem Exchange (hourly carbon & water fluxes)
- Soil Respiration
- Leaf area
- Leaf litterfall
- Carbon in roots
- Carbon in wood
- Woody litterfall
- Phenology
- Soil carbon turnover rates









#### **Model Performance**



#### Rate my data:

How much data do we need to measure?

Identifying the most valuable data.

Ancillary Measurements

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Keenan et al. (2013) Ecological Applications

#### **Model Performance**

Lfwood

Lfleaf

Leaf In

Croot

Clit

Cw

Soil C

All

0



#### NEEday Soil C Turnover NEE monthly Rsoil + iRsoil NEE annual Rsoil night Leaf Out Root Resp in Soil Resp Litter Turnover NEEnight 67 2 3 4 5 1 Error (log) Parameters (1-40)

## Rate my data:

How much data do we need to measure?

Identifying the most valuable data.

Ancillary Measurements

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Keenan et al. (2013) Ecological Applications

# Ingredients of an effective synthesis study

# A question!

- Formulated theory
- A refined target
- Continual refinement
- Start local
- Modularly scaled
- Ancillary measurements

# Ingredients of an effective synthesis study

## A question!

- Formulated theory
- A refined target
- Continual refinement
- Start local
- Modularly scaled
- Ancillary measurements
- Models to confront

# Hypothesis driven / process based:



#### Hypothesis driven / process based: Hypothetic-deductive modeling approach Ecosystem Parameterization model with **Hypotheses** set of and evaluation fixed equations Inductive modeling approach Data mining: ? ? Data ð Purely Extraction of the **Hypotheses** empirical functional model relationships Characterization and evaluation

Models to confront

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Moffat et al. (2010)





- NACP Interim synthesis models (freely available)
  30+ models run at ~40 sites across US and Canada.
- Plumber2 (freely available)

10s of models run at ~100 flux sites globally.

• MsTMiP (freely available)

22 models run globally under different scenarios

• TRENDY (freely available)

~20 DGVMs run globally under different scenarios

• PEcAn project

# Ingredients of a good synthesis study

## A question!

- Formulated theory
- A refined target
- Continual refinement
- Start local
- Modularly scaled
- Ancillary measurements
- Models to confront (last!)

#### plants -> landscapes -> the globe







#### Changes in Phenology

#### The timing of phenology is changing as the climate warms

#### 160 155 150 145 140 135 130 125 130 125 120 1990 1995 2000 2005 2010 Year

165





# 1. Spring is getting earlier

2. Autumn is

getting later

#### Keenan et al. (2014)









#### And the performance of land surface models



# Synthesis













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# Thank you!









