

Ecological resilience in messy time series

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Memorial University • Biology Seminar • September 28th 2018

Acknowledgements



**NSERC
CRSNG**

Slides: bit.ly/muntalk2018

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Community response to environmental change



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Community response to environmental change



Credit: NOAA George E. Marsh Album [Public domain], via Wikimedia Commons

Community response to environmental change

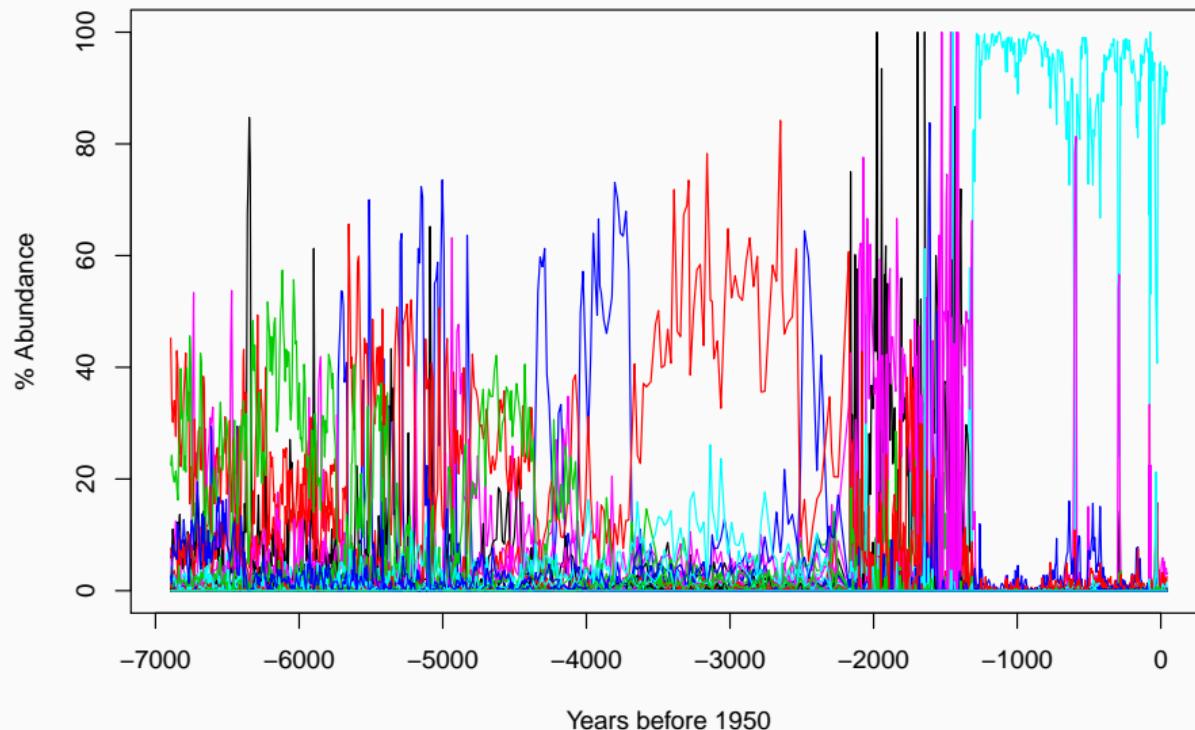


Credit: Michael Knall [Public domain], via Wikimedia Commons

How do communities respond to rapid change?

How does resilience change over time?

Messy time series



Messy time series

Most long ecological & environmental time series are messy

- Collected irregularly
- Missing data — lab or field widget breaks, weather
- Multi-year gaps — loss of funding
- Methodological changes

They might be messy, but they're still useful!

Complexity & Resilience

Logistic map

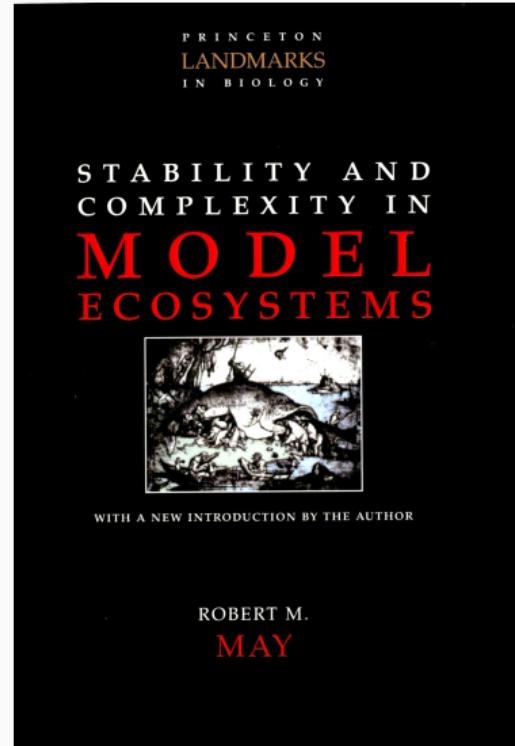
Complex behaviour from simple equations

$$x_{n+1} = rx_n(1 - x_n)$$

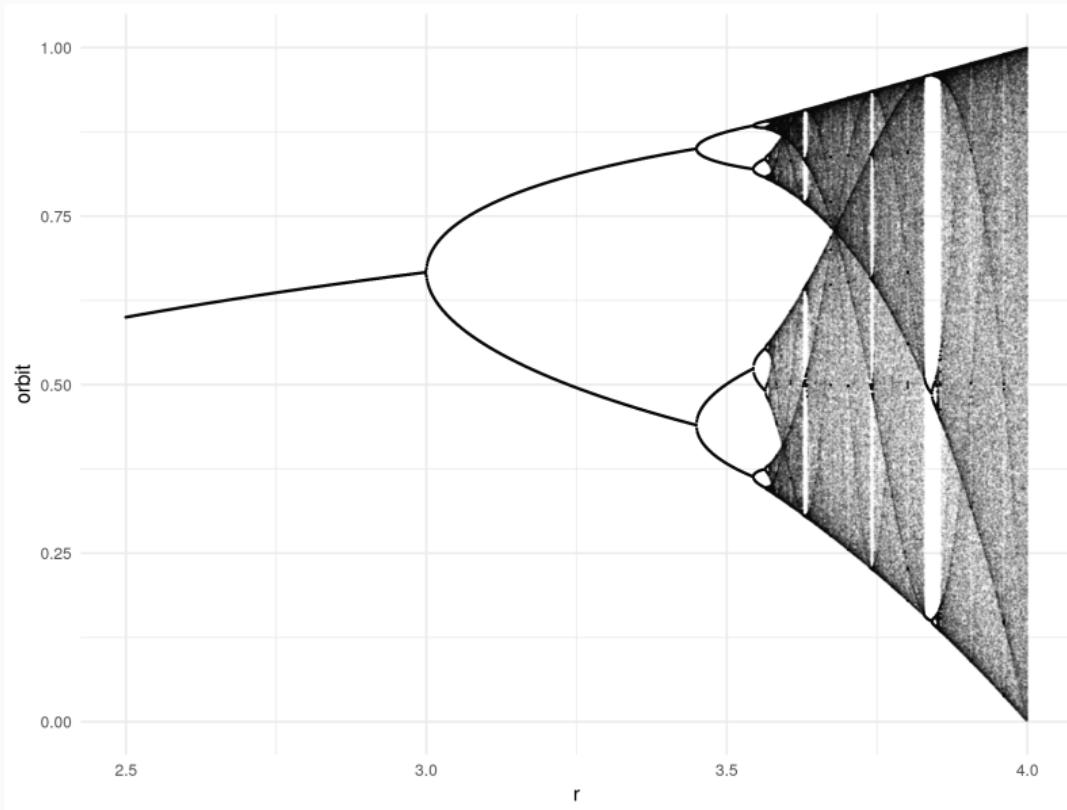
$0 < x_n < 1$ is the *fraction* of the **carrying capacity** of the ecosystem

r combines the birth and death rates

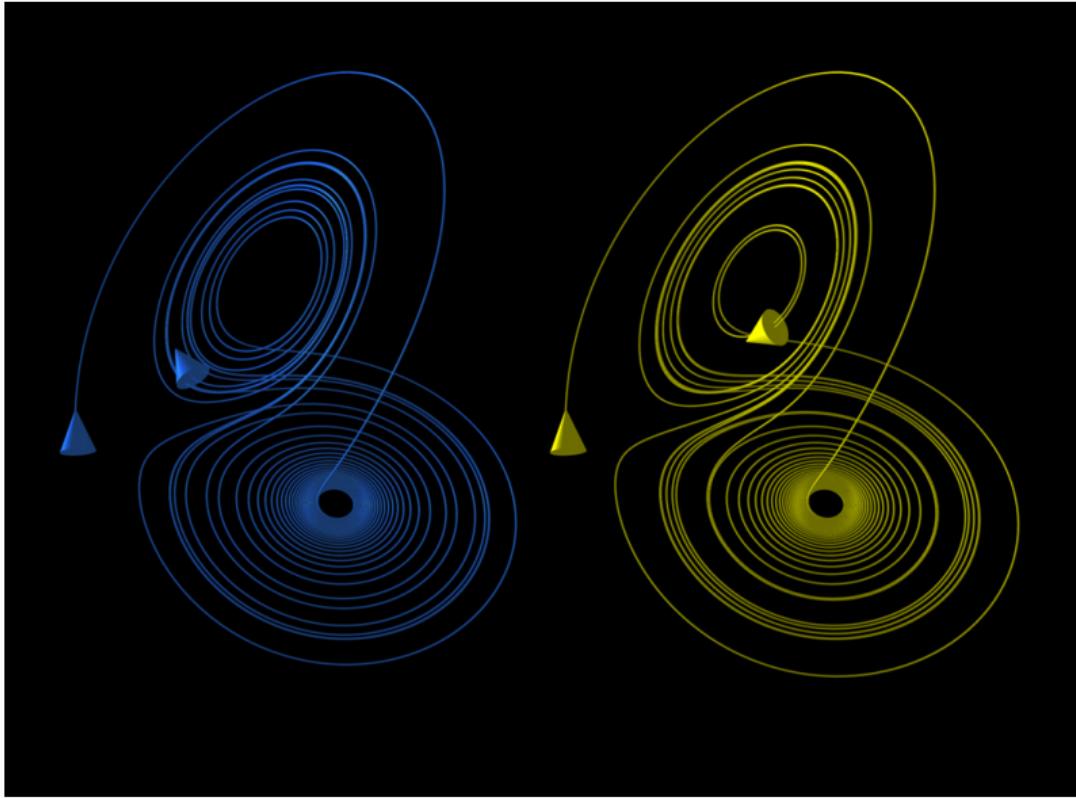
Interest is for values of r in the interval $[0, 4]$



Logistic map



Butterfly effect — Ed Lorenz

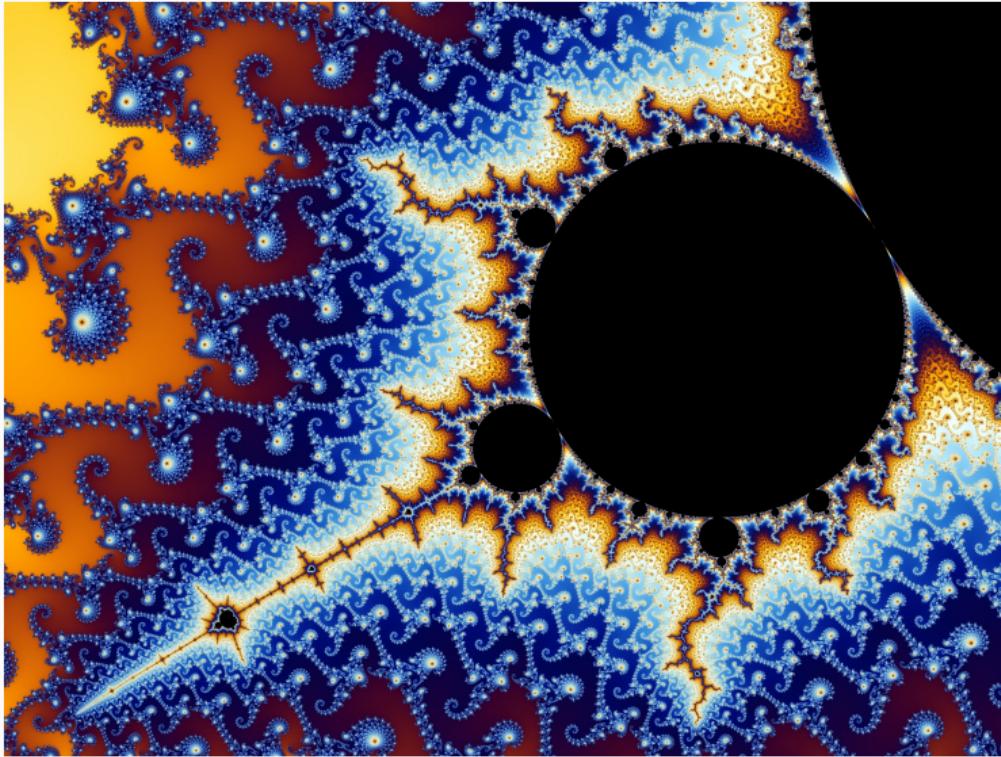


Turbulent flow — Lev Landau



Credit: NASA [Public domain], via Wikimedia Commons

Fractal geometry – Benoit Mandelbrot



Credit: Wolfgangbeyer [CC BY-SA], via Wikimedia Commons

Alternative stable states – Marten Scheffer

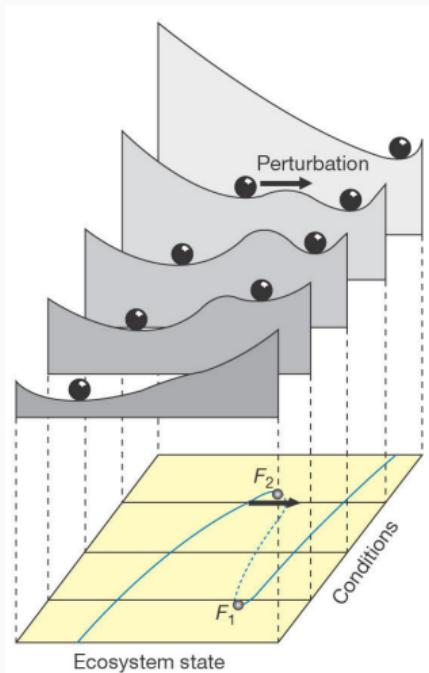
Under range of conditions multiple ecosystem states may exist

Stability landscapes depict equilibria and basins of attraction

Valleys are stable states

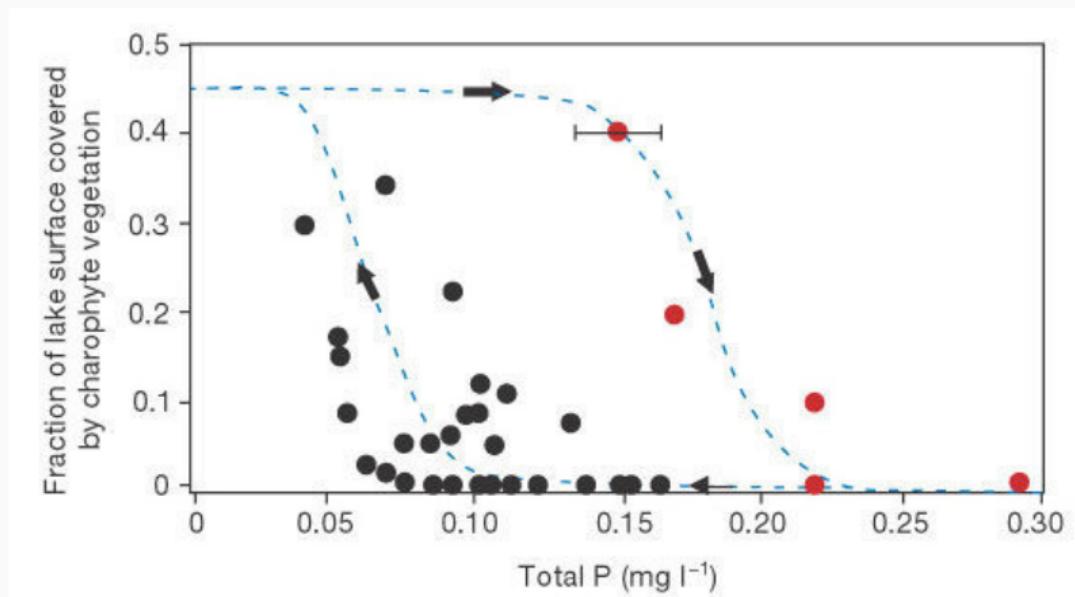
External conditions alter the stability landscape

Disturbances can kick a system between states



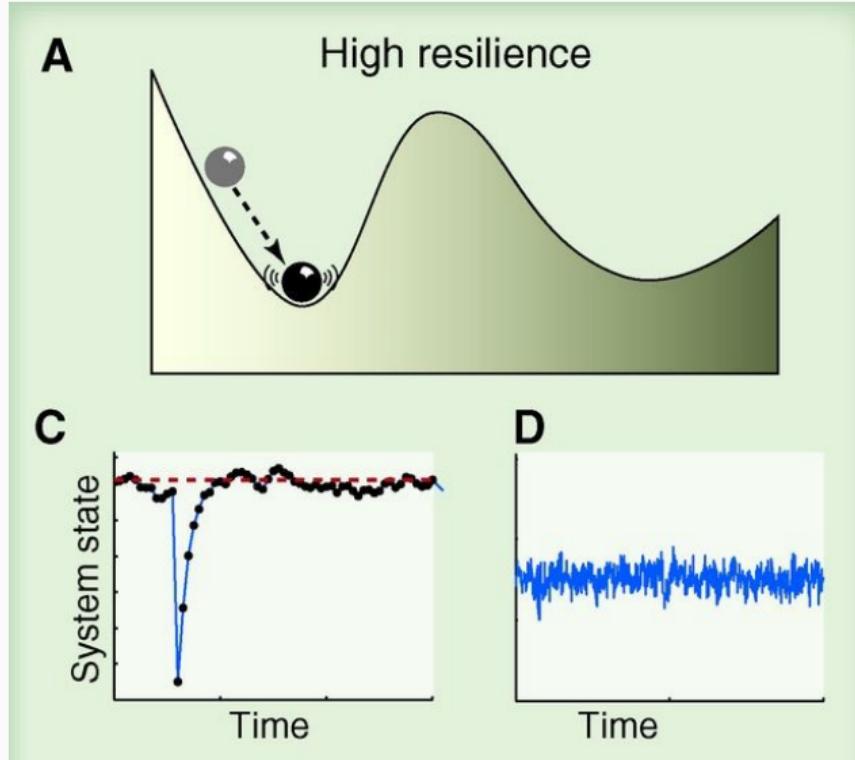
Source: Scheffer et al *Nature* 2001

Hysteresis

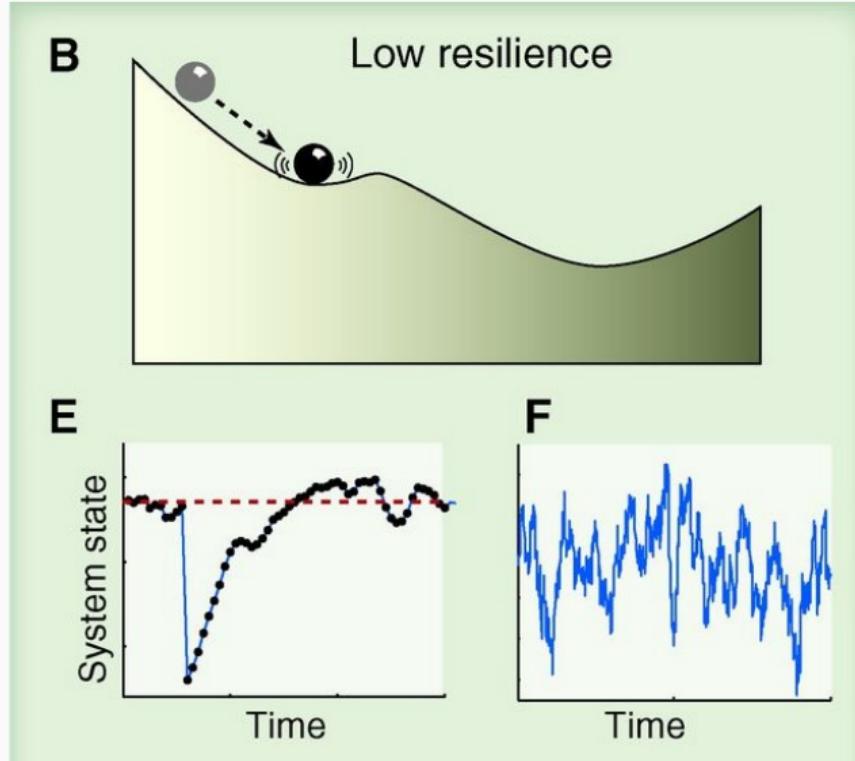


Source: Scheffer et al *Nature* 2001

Early warnings & Resilience to change

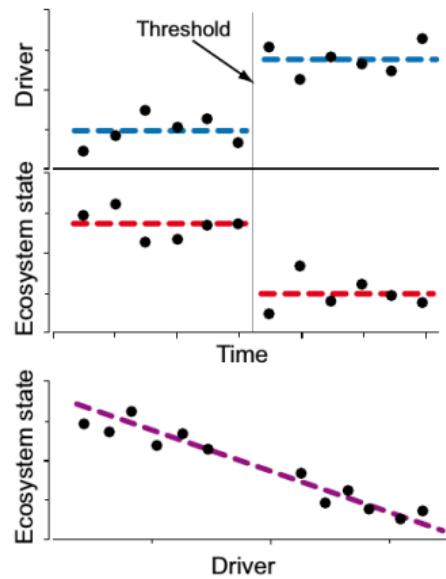


Early warnings & Resilience to change

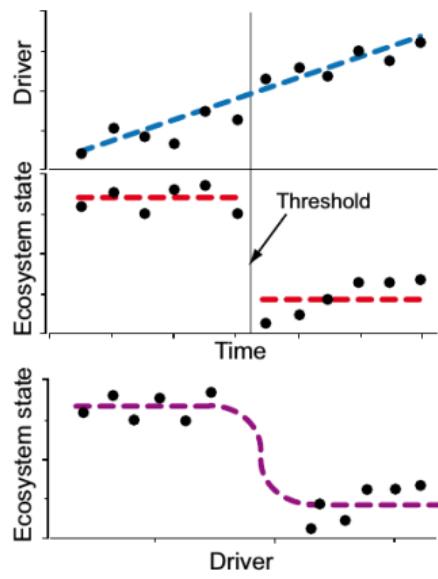


Regime shifts

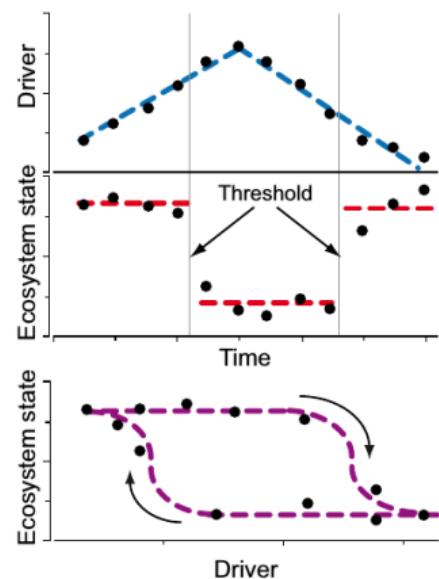
(A) Type I, driver threshold



(B) Type II, state threshold



(C) Type III, driver-state hysteresis



Source: Randsalu-Wendrup et al *J. Paleolimnology* 2016

Palaeoecology — Palaeolimnology

Palaeoecological data: valuable & unique source of data to study effects of environmental change

Most lakes not annually laminated

- mixing as a form of temporal averaging
- also age/dating uncertainty
- limits temporal resolution of data
- compaction of older sediments → varying time per sample

How do dynamics of species composition vary with environmental change?

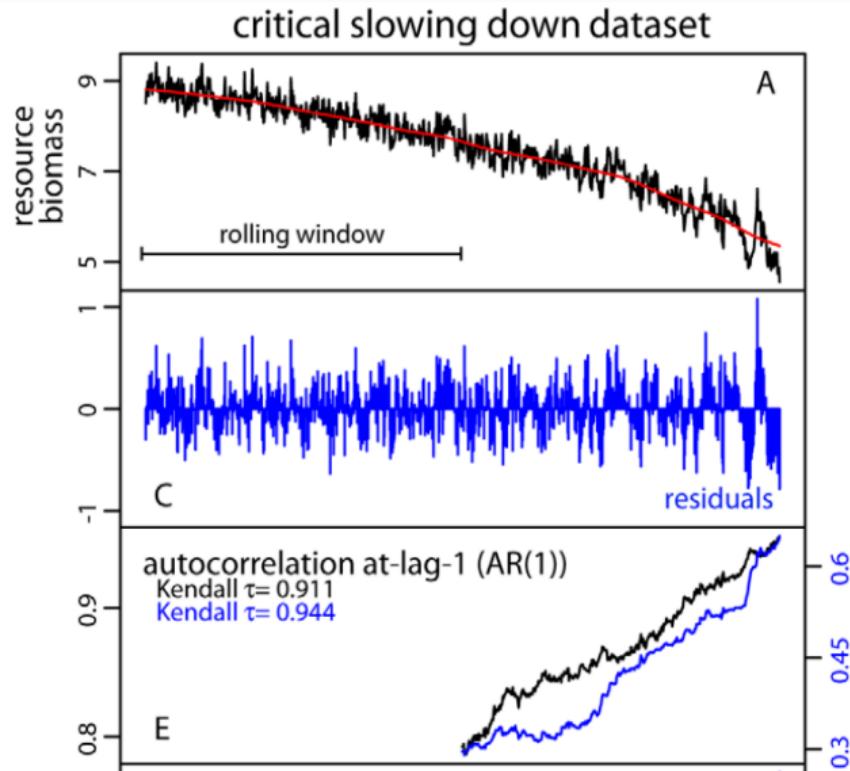
- Trend (mean), Variance, Higher moments...?



Source: © Ewan Shilland

Modelling variance

Moving windows



Problems

- Ad hoc
 - What window width?
 - How to detrend (method, complexity of trend, ...)
- Ideally regularly spaced data
 - What to do about gaps, missing data?
- Statistic testing hard
 - τ assumes independence
 - Surrogate time series assumes regular sampling, sensitive to choice of ARMA
- Results are incomplete time series



WIGGLY -THINGS-

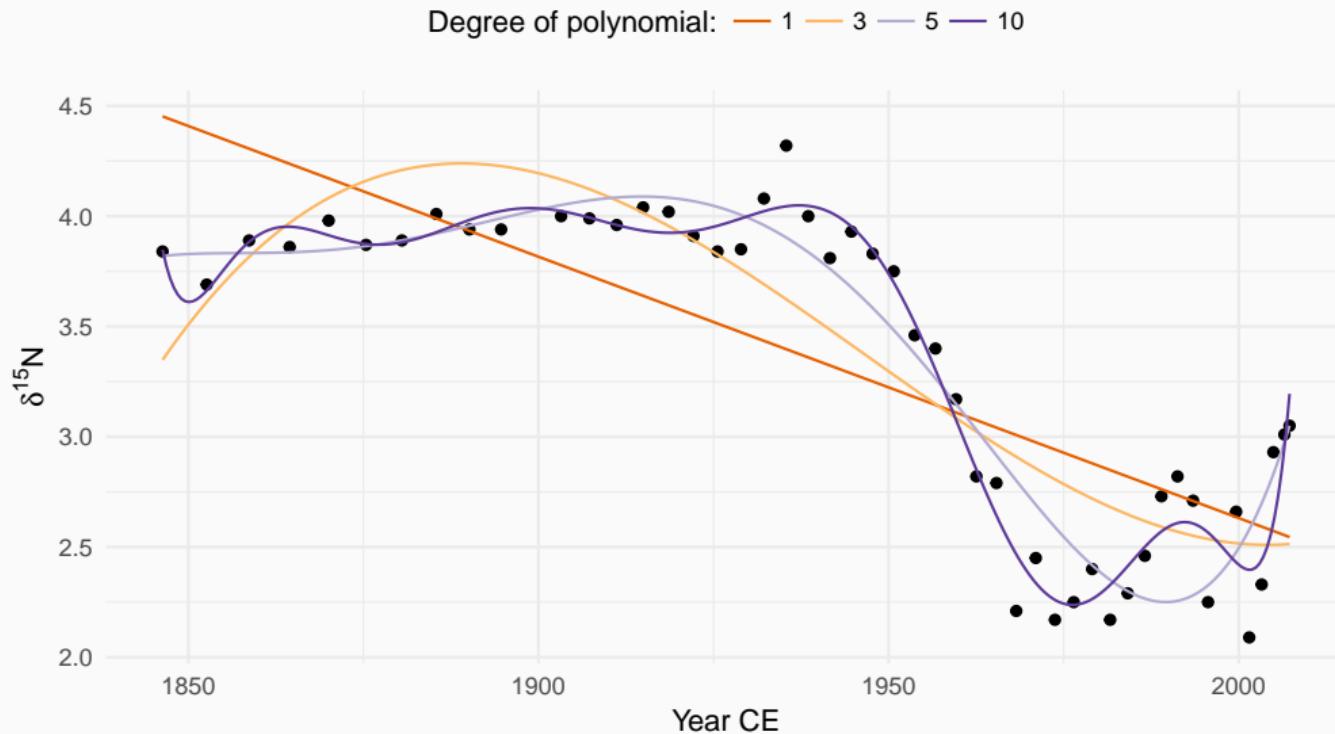
Linear models

{

$$y = \beta_0 + \beta_1 \text{time} + \varepsilon$$

}

Linear models don't fit well



Generalized additive models (GAMs)

{

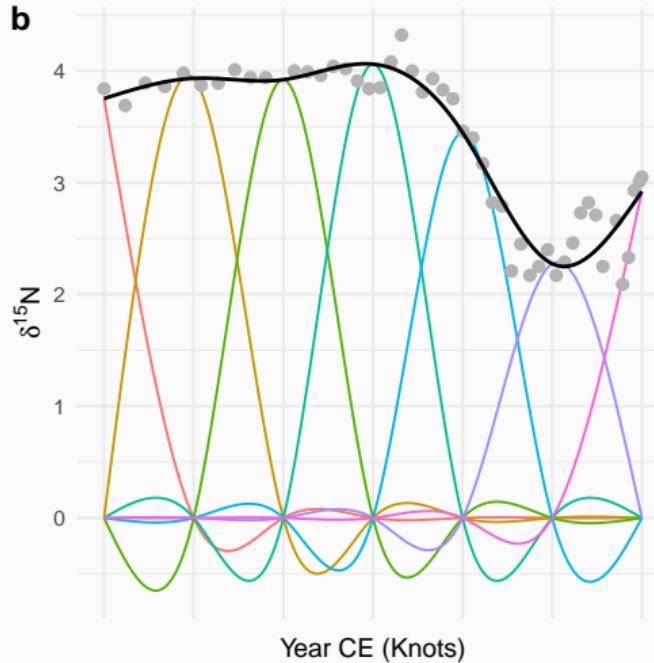
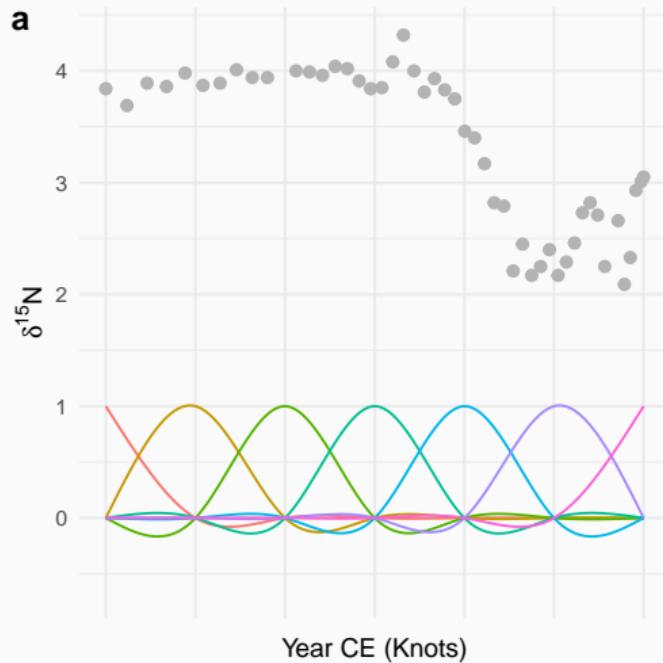
$$y = \beta_0 + f(\text{time}) + \varepsilon$$

}

f is a smooth function

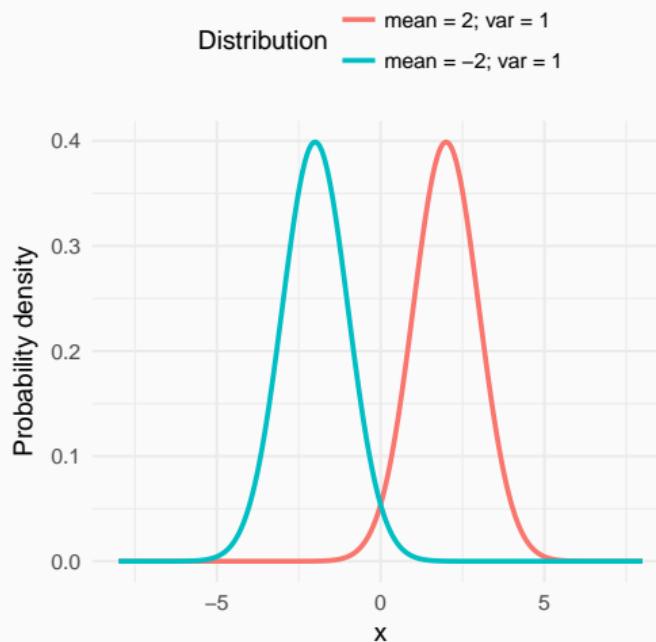
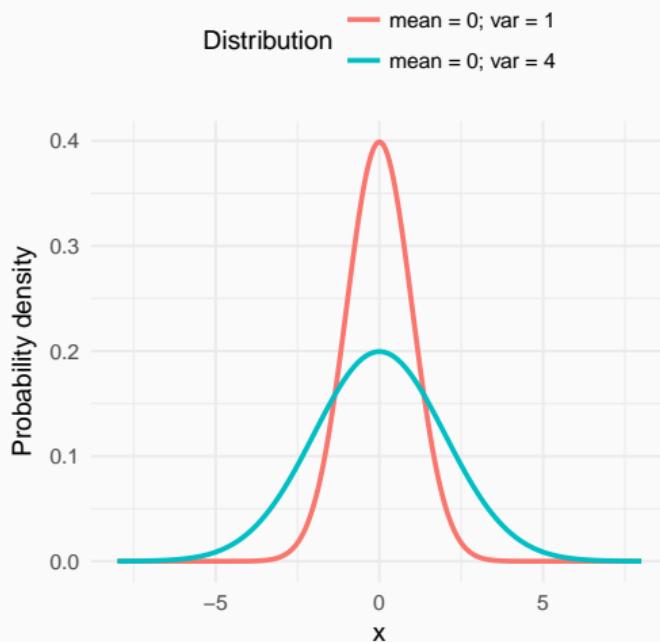
Learn about the shape of f from the data

Generalized additive models



Gaussian distribution – defined by 2 parameters

- mean – μ
- variance – σ^2



Generalized additive model for location, scale, and shape

GAMLSS allows, for certain suitable distributions, the ability to model one or more of the mean, variance, skewness, and kurtosis.

- GAMLSS models of Rigby and Stasinopoulos (2005)
- Vector GAM (VGAM) models of Yee (e.g. Yee and Mackenzie (2002))
- General smooth models of Wood/mgcv
- E.g.: y_i are Gaussian with mean μ_i and variance σ_i^2 , each of which are modelled via a linear predictor η of smooth functions of covariates x_j and z_j

$$y_i \sim \mathcal{N}(\mu = \eta_{1,i}, \sigma^2 = \eta_{2,i})$$

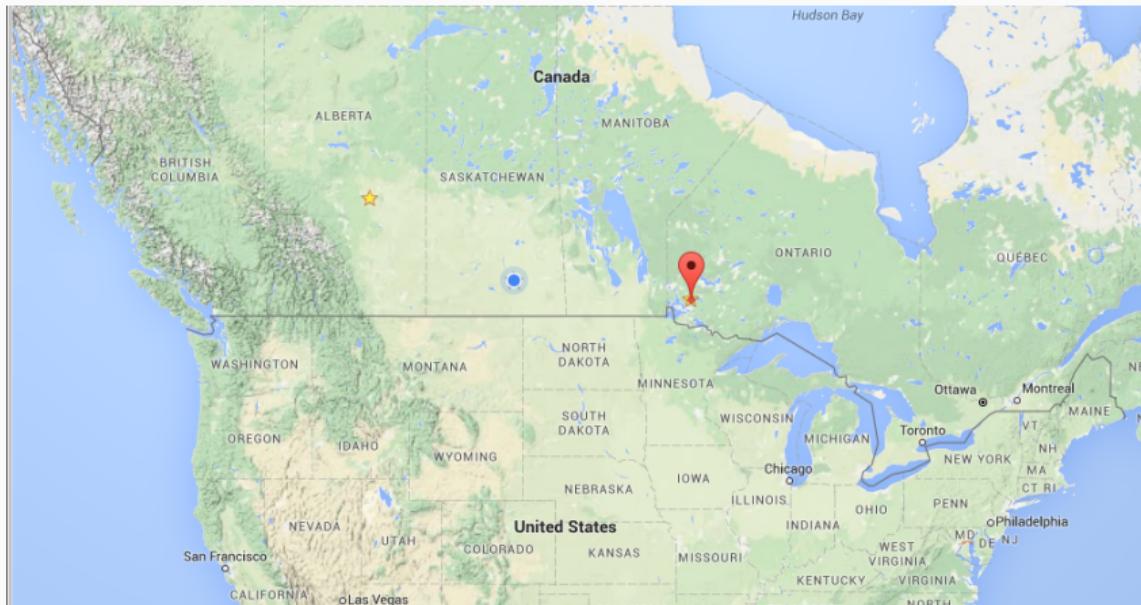
$$\eta_{1,i} = g \left(\sum_{j=1}^k f_j(x_{ij}) \right)^{-1}$$

$$\eta_{2,i} = g \left(\sum_{j=1}^k f_j(z_{ij}) \right)^{-1}$$

Lake 227

Lake 227

Experimental Lakes Area, NW Ontario, Canada — Dilute headwater lake (5 ha, 10 m deep)



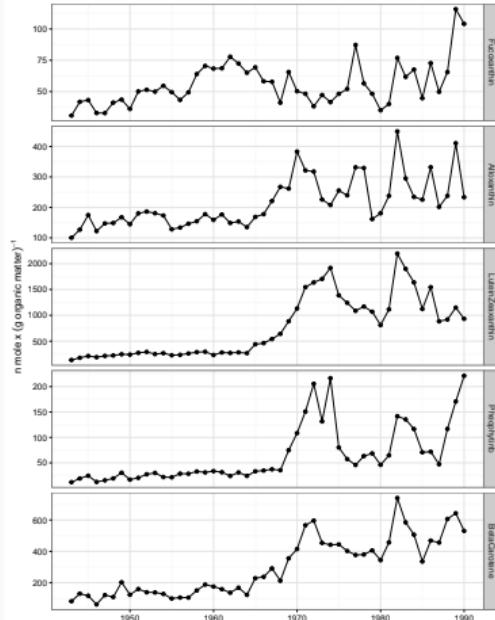
Lake 227

Experimental Lakes Area, NW Ontario, Canada — Dilute headwater lake (5 ha, 10 m deep)



Lake 227 sedimentary pigments

- Experimentally manipulated
 - 1969– :~10x increase in P load
 - 1969–74: N added in 14:1 ratio
 - 1975–82: N load reduced to 5:1
- Annual sediment samples 1943–1990
- Analysed for fossil pigments
 - *Fucoxanthin*; diatoms, chrysophytes
 - *Alloxanthin*; cryptophytes
 - *Lutein-zeaxanthin*; cyanobacteria, chlorophytes
 - *Pheophytin b*; chlorophytes
 - β carotene; total algae
- Cottingham, Rusak, and Leavitt (2000)



Lake 227 sedimentary pigments

Fitted Gaussian location, scale additive model using **mgcv**

$$y_t \sim \mathcal{N}(\mu = \eta_{1,t}, \sigma^2 = \eta_{2,t})$$

$$\eta_{1,t} = \exp(f_1(\text{Year}_t) + f_{1,\text{pigment}}(\text{Year}_t))$$

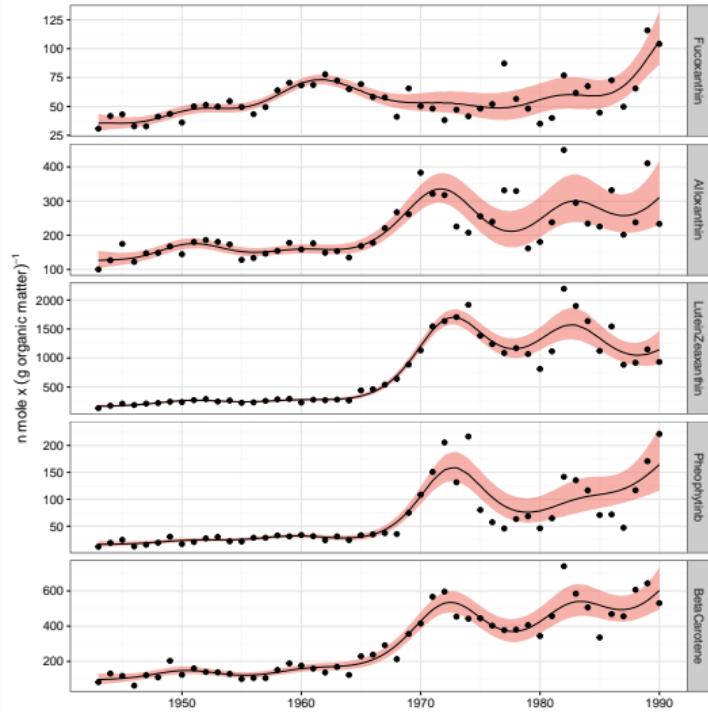
$$\eta_{2,t} = g(f_2(\text{Year}_t) + f_{2,\text{pigment}}(\text{Year}_t))^{-1}$$

Mean & variance modelled with:

1. Global function of Year, plus
2. “Random effect” spline (spline-factor interaction **bs** = "fs") with penalty on first derivative — **penalizes departures from global function**
3. $g()$ is a log link function modified to avoid $\hat{\sigma}^2 = 0$

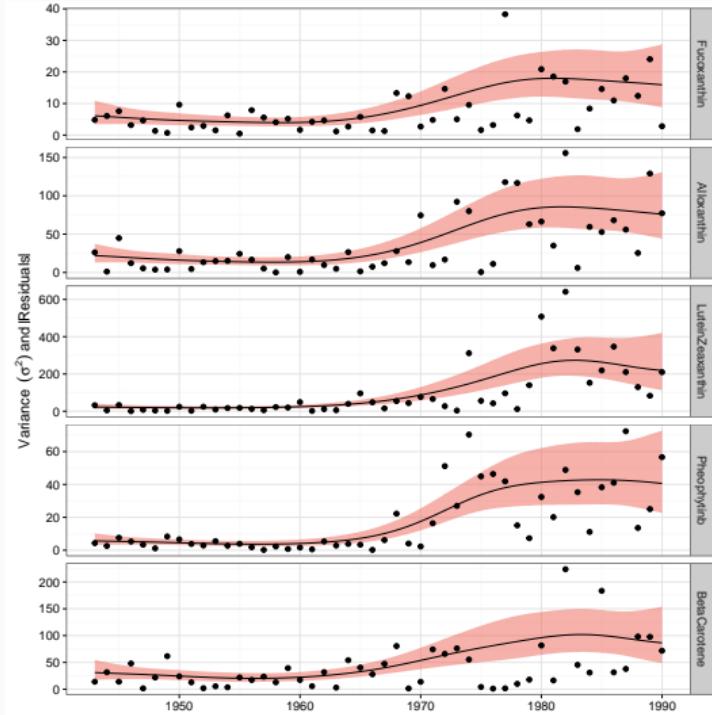
Lake 227 sedimentary pigments — fitted mean functions

- Random effect spline provides better fit of trend (mean)
- Undersmoothing of several pigments if separate models fit
- Increase in most algal groups following fertilization



Lake 227 sedimentary pigments — fitted variance functions

- Increasing variance in pigment concentration following manipulation
- Stabilisation post manipulation
- Consistent with median-log Levene's test (Cottingham, Rusak, and Leavitt 2000)



Baldeggeree

Baldeggsee

- Nutrient rich, hardwater lake, located on the central Swiss Plateau
- Developed anoxic bottom waters in 1885
- Intensive farming & agriculture in catchment
- Artificial oxygenation & circulation as remediation from 1982
- Cored in 1993 • 91 samples • 75 diatom taxa
- Source: Lotter (1998) *The Holocene* 8(4), 395–405



Source: © Andy Lotter

Baldeggeree

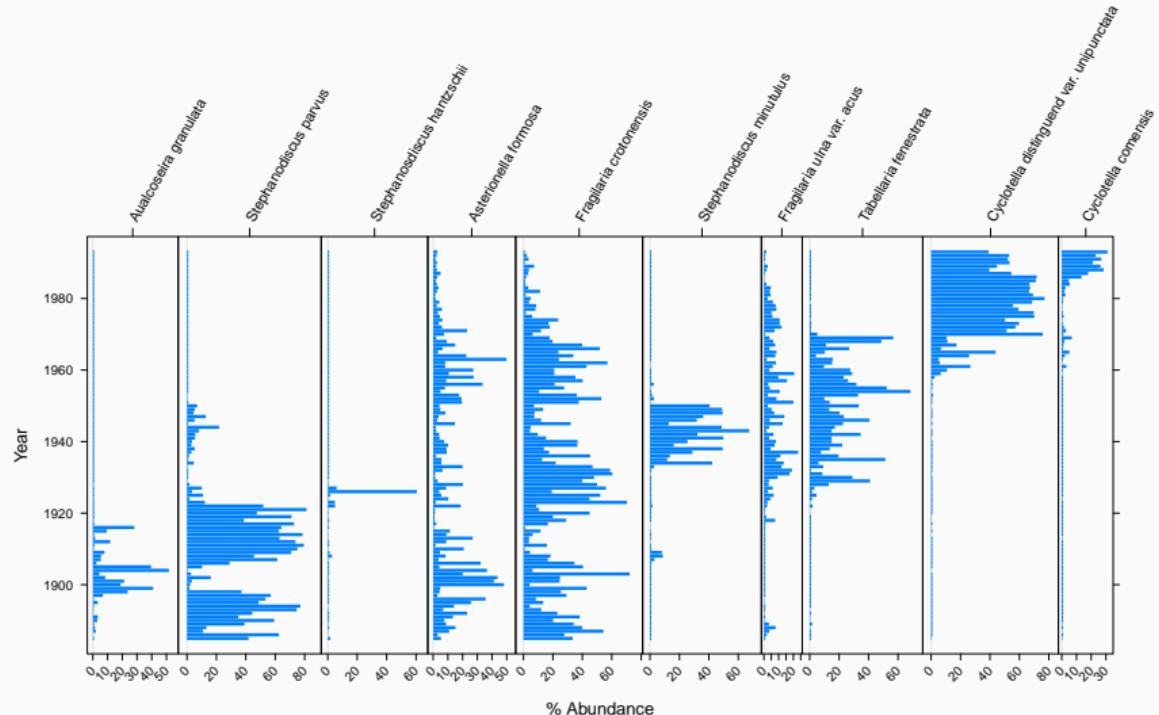
- Iconic example of a **critical transition**
- Deep lakes can exhibit bistable conditions
 - Nutrients increase phytoplankton
 - Decomposition can increase anoxia
 - Stimulates P release from sediments
- Marl lake
- **Flickering** (?)

Are early warning signals (increased variance) present?

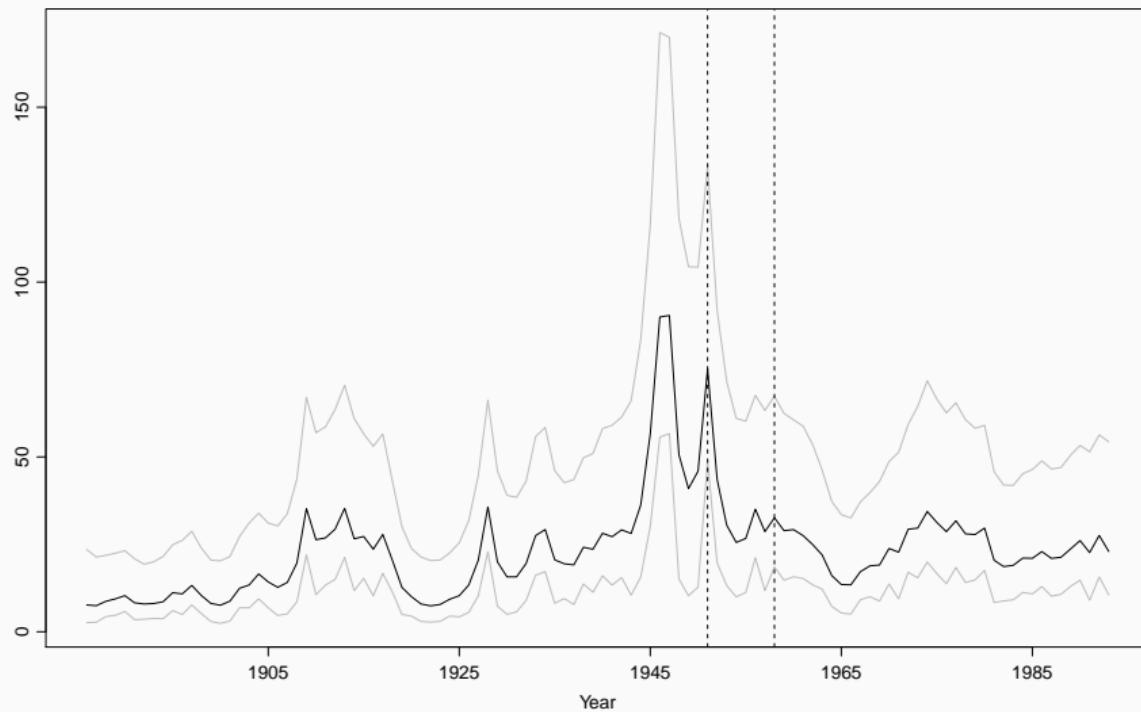


Source: © Andy Lotter

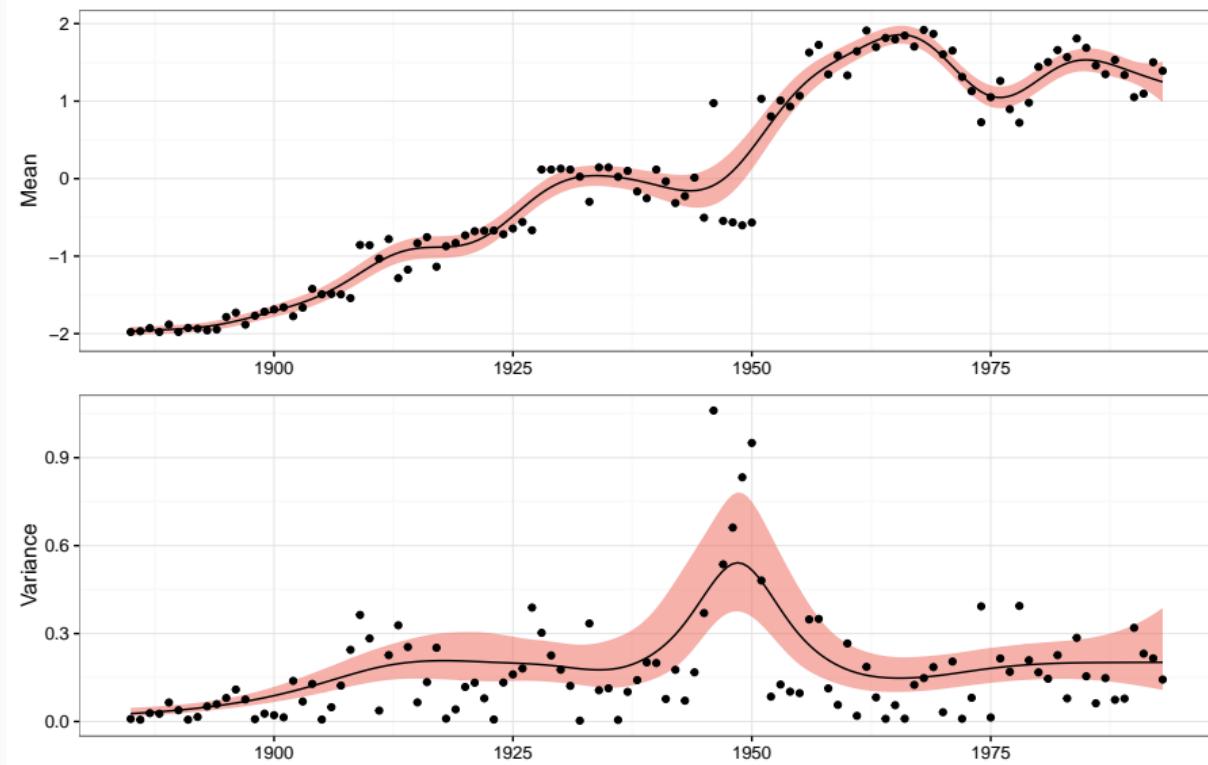
Baldeggersee – Stratigraphy



Baldeggsee – Stochastic Volatility Model



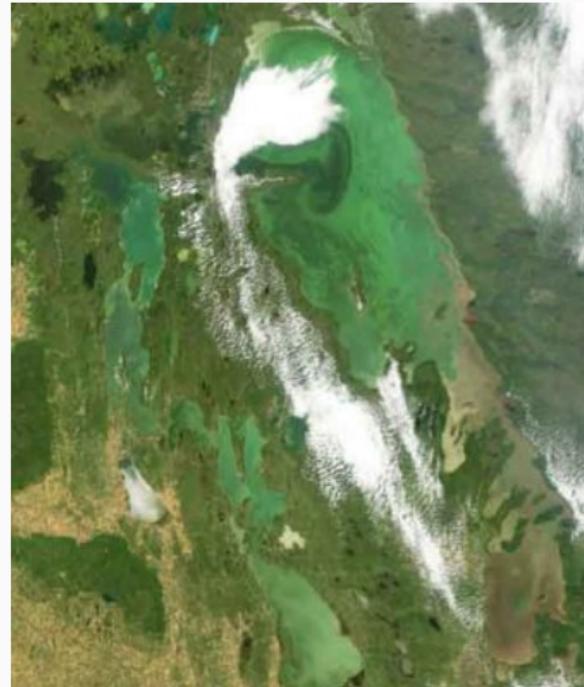
Baldeggersee – GAM Location and Scale fits



Lake Winnipeg

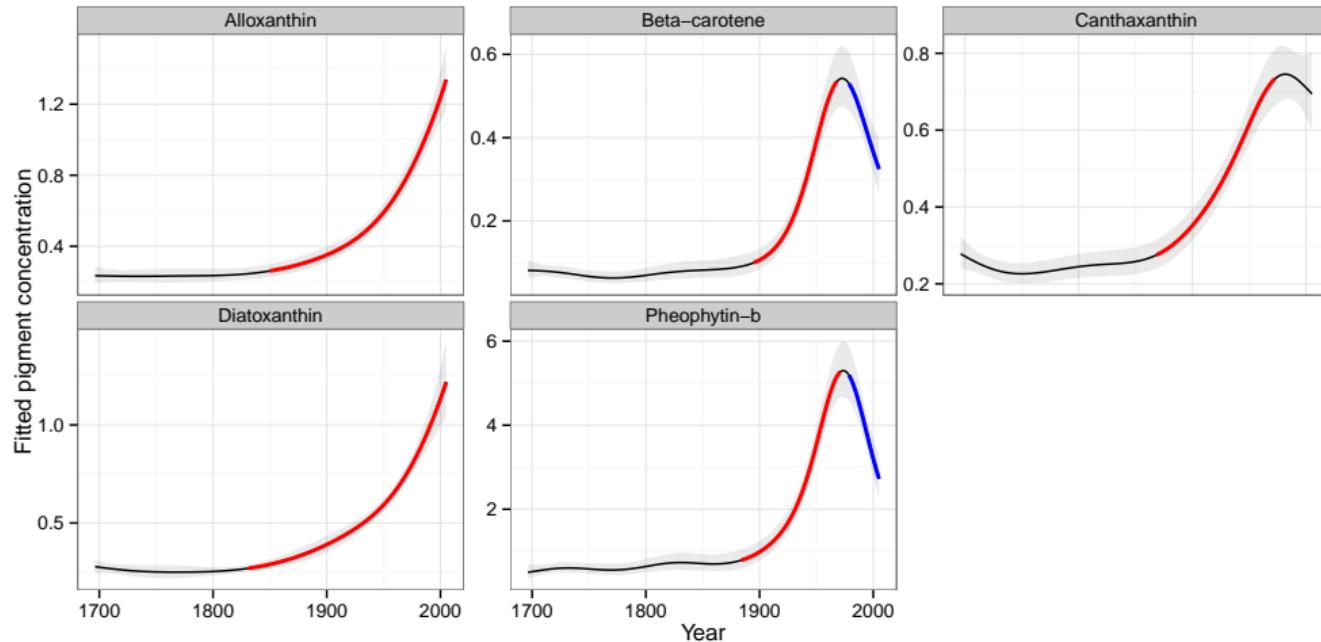
Has Lake Winnipeg suffered a catastrophic regime shift?

- Large, shallow, polymictic, lake
- Eutrophic
- Algal blooms
- Regime shift / critical transition?
- Increased destabilization with increased N & P?



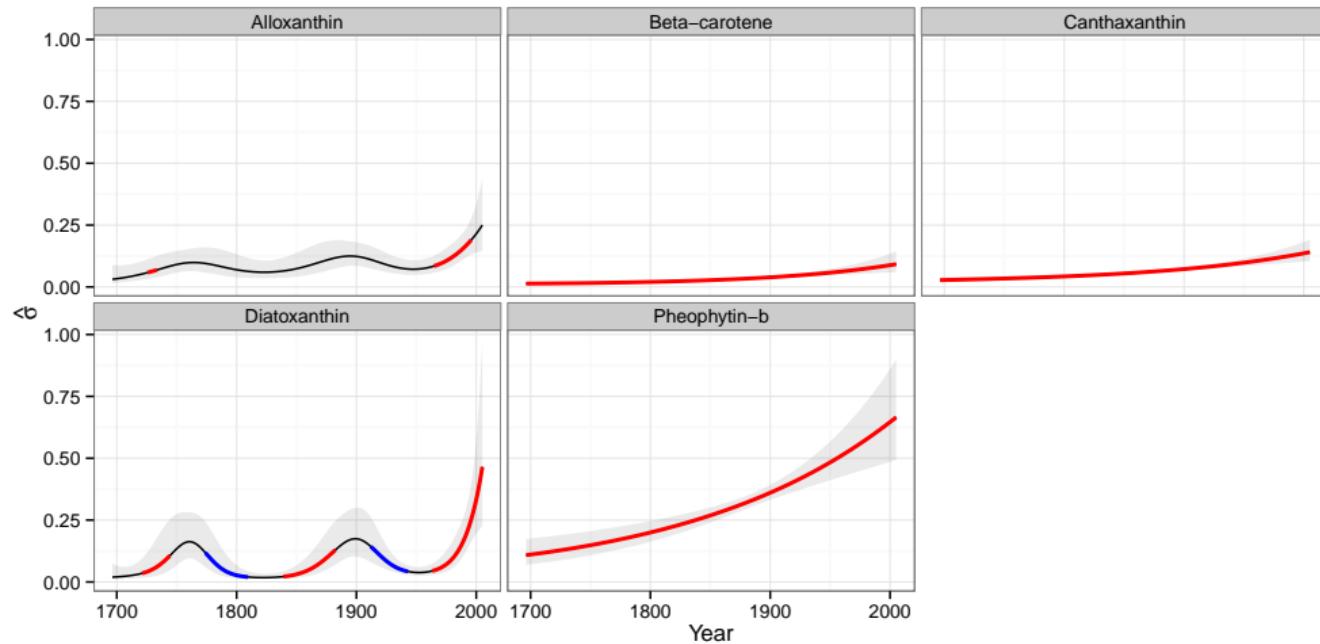
Source: Public domain

Has Lake Winnipeg suffered a catastrophic regime shift?



Source: Bunting et al *Limnology & Oceanography* 2016

Has Lake Winnipeg suffered a catastrophic regime shift?



Source: Bunting et al *Limnology & Oceanography* 2016

Summary

What's next...?

GAM-based models for other EWS?

- Autocorrelation
- Skewness & kurtosis

Distributional models & recently-developed quantile GAM approaches might work

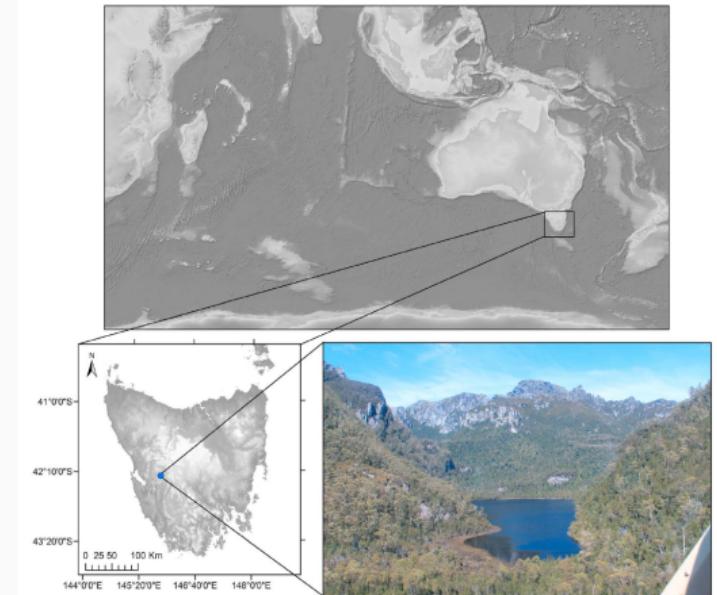
Extra examples / slides

Extra Slides

Lake Vera, Tasmania

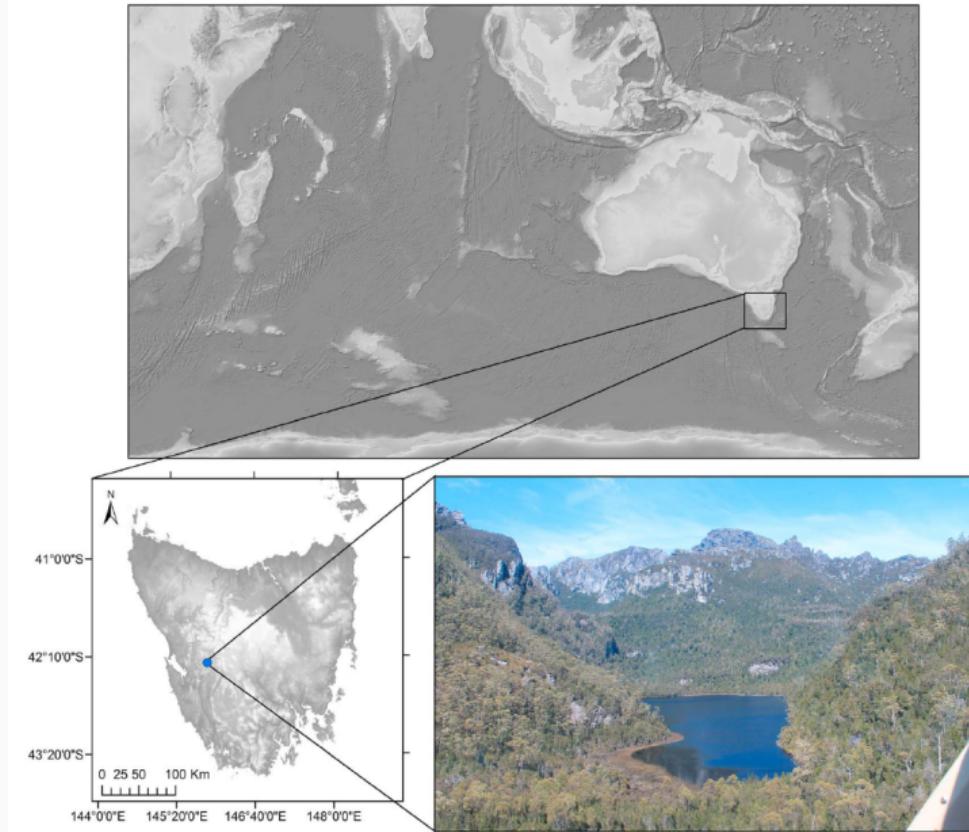
Lake Vera

- Small, moraine-bound lake
- Deep (48 m)
- Oligotrophic, acidic
- Fire is major driver of terrestrial change
- What affect on the lake?



Source: Beck et al *JGR: Biogeoscience* 2018

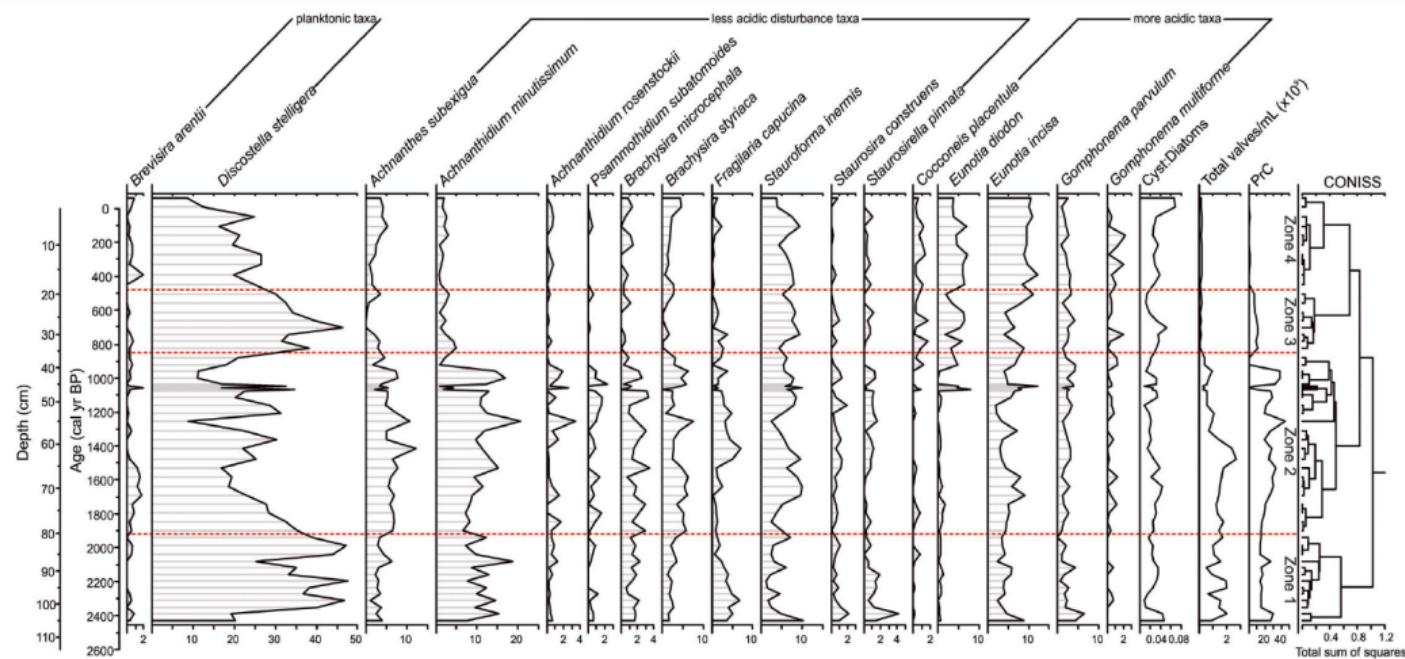
Lake Vera



Lake Vera

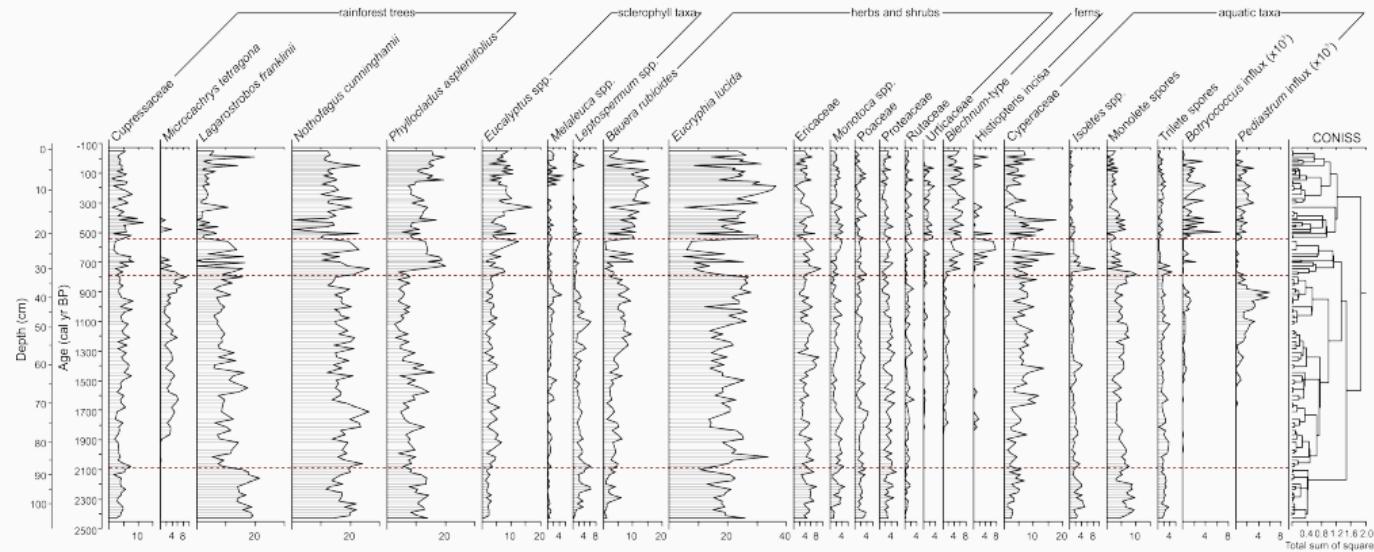


Lake Vera



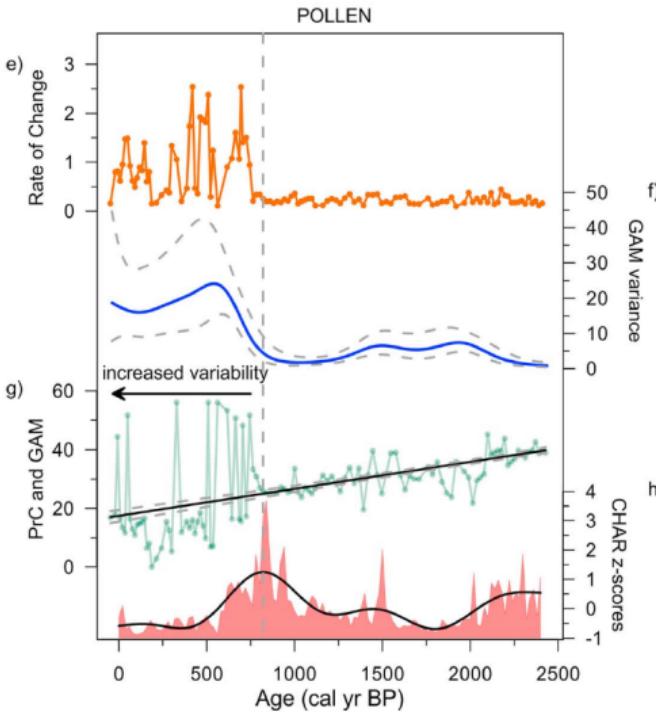
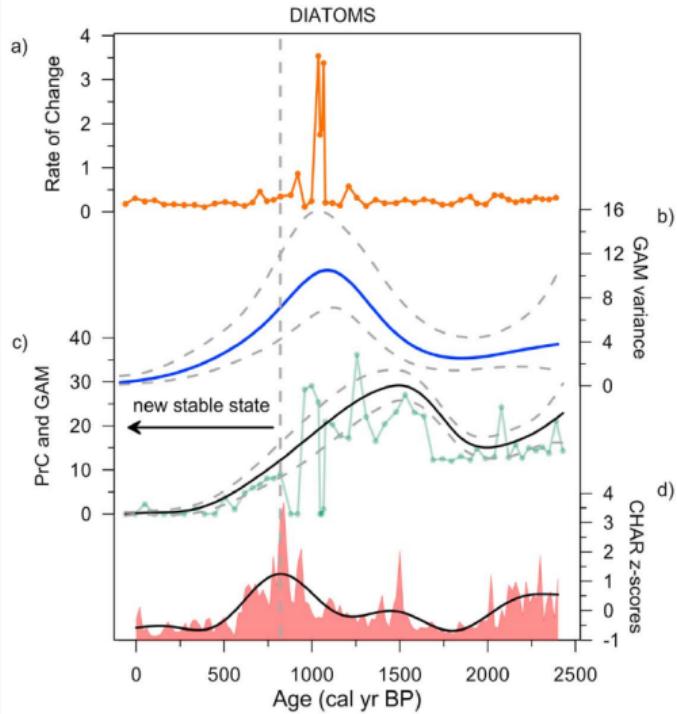
Source: Beck et al J Geophysical Research: Biogeoscience 2018

Lake Vera



Source: Beck et al J Geophysical Research: Biogeoscience 2018

Lake Vera



Source: Beck et al *J Geophysical Research: Biogeosciences* 2018

References i

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