

The study of uncertainty in palaeoclimate reconstruction

John Haslett, TCD, Ireland

...and many others

Supported by Science Foundation Ireland

IPCC Inter Governmental Panel on Climate Change

Abrupt climate change

IPCC WG1 2007

“During the last glacial period, abrupt regional warmings (probably up to 16°C within decades over Greenland) occurred repeatedly over the North Atlantic region”

(Fifth Assessment Report due 2014)

What do we know about abrupt changes?

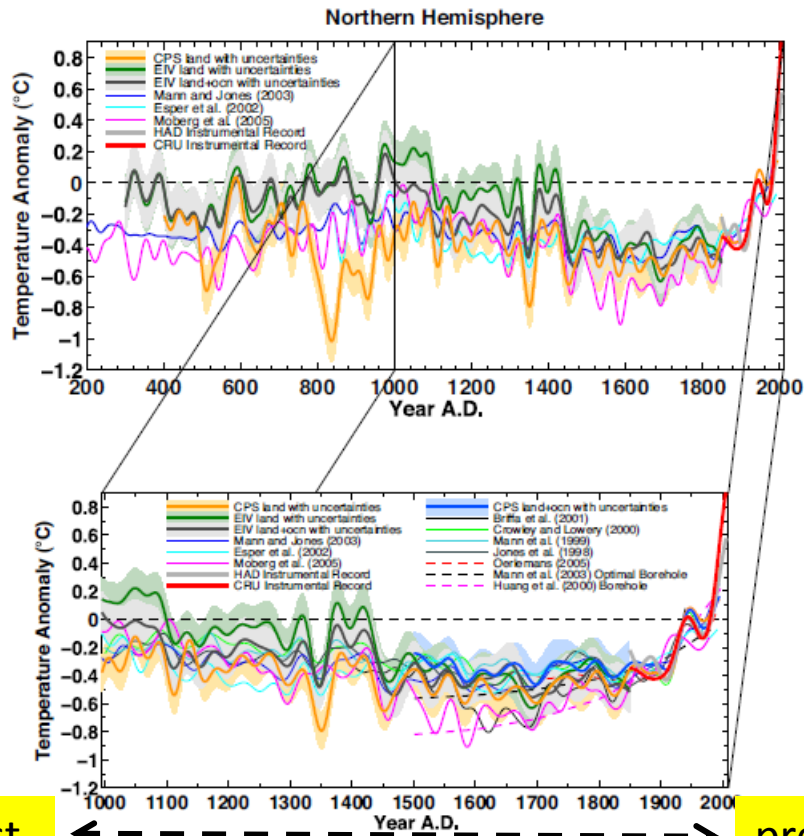
Pitifully little



N Hemisphere Air Temp multi-proxy past 2000 years

What do we know about past millenium?

Quite a lot



Classic study

Mann et al (2008)

2008 update of 1998

Wiki: [Hockey Stick Controversy](#)

Palaeoclimate: **What, How** and Why?

- What? Past climate

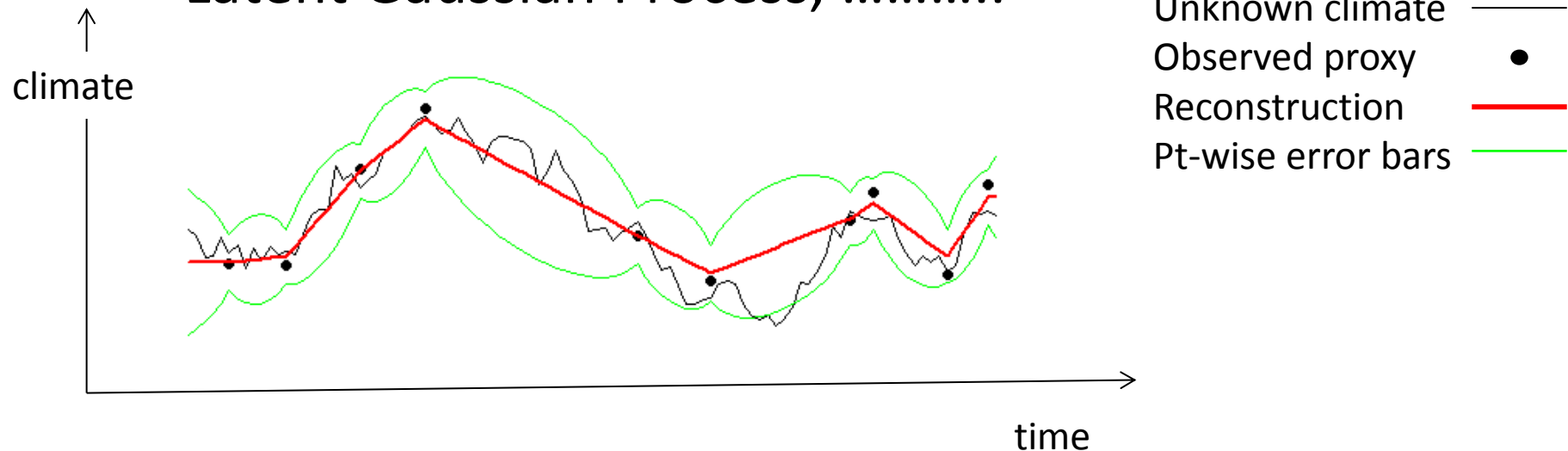
- Temporal scale *1000 years, 100,000 years*
- Spatial scale *Hemisphere; China; Italian/Irish lakes*
- Physical Process *Ocean, Atmosphere, Ice, Biosphere*
- Statistical summary *Avg Atmos temp, Avg Ocean Summer temp
Atmos moisture extrema, Abrupt change*

- How

- Computer Models *General Circulation Models (GCM)*
- Proxy Data
- Statistical Methods

Statistical Methods in Palaeoclimate Reconstruction

- Partially observed stochastic processes
 - Simple case
 - BLUP, Kriging, State-space, Latent Gaussian Process,



Palaeoclimate: What, How and **Why**?

- Why at YSI2013?
 - Topical, Policy, Challenging, Defined
 - Never-ending source of statistical problems (& grants)
 - Computational
 - Conceptual
 - **Communication**
 - **Uncertainty central**
 - “**probably** up to 16°C within decades over Greenland”

**Statisticians specialists in uncertainty.
How do we communicate it?**

Palaeoclimate: What, How and **Why?**

System	Climate	Economy	Human Body	?
Subsystem	Atmosphere, Ocean...	Domestic, International...	Kidney, Heart...	
Math Theory, Comp Models				
Proxy Systems				
Data				
Combining Uncertainty				
Communicating Uncertainty				
Policy				

Monte Carlo methods

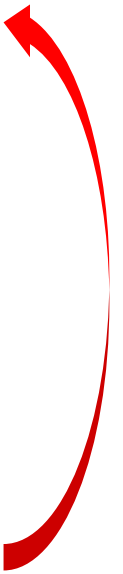
Experienced Statistician

The *process* of modelling

Adopt simplified math representation

Solve to get useful information

Too hard? stronger and stronger assumptions



Experienced Statistician

The *process* of modelling

Adopt simplified math representation

Solve to get useful information

Too hard? stranger and stranger assumptions



Modelling has got easier

- Linear regression; Normal errors
- glm, gam
- Hierarchical models
- Bayesian methods
- Stochastic processes
- MCMC
- Data-bases, Computers, Algorithms

*New
challenges*

IPS006

Statistical methods in the study of palaeoclimate

09:00 Monday

- *Preliminary multiproxy surface air temperature field reconstruction for China over the past millennium*
Feng Shi; China
- Paleoclimate reconstruction using statistical nonlinear forward models
Peter F. Cragmle; United Kingdom
- Bayesian palaeoclimate inference from pollen in Southern Italy
Andrew Parnell; Ireland
- Discussion
Paul Switzer; USA

Palaeoclimate Reconstruction

What: Examples

How: Uncertainty as Stat Inference on
Latent Climate

Challenges in Communicating Uncertainty

China Air temp “anomalies” wrt present

Multi-proxy – last 1000 years

Shi Feng IPS006

Focus: Air Temp (spatial avg, annually)

Proxies:

Tree Rings	373 cores
Ice Core	10 cores
Speleothem	6 cores
Historical	15
Other	11

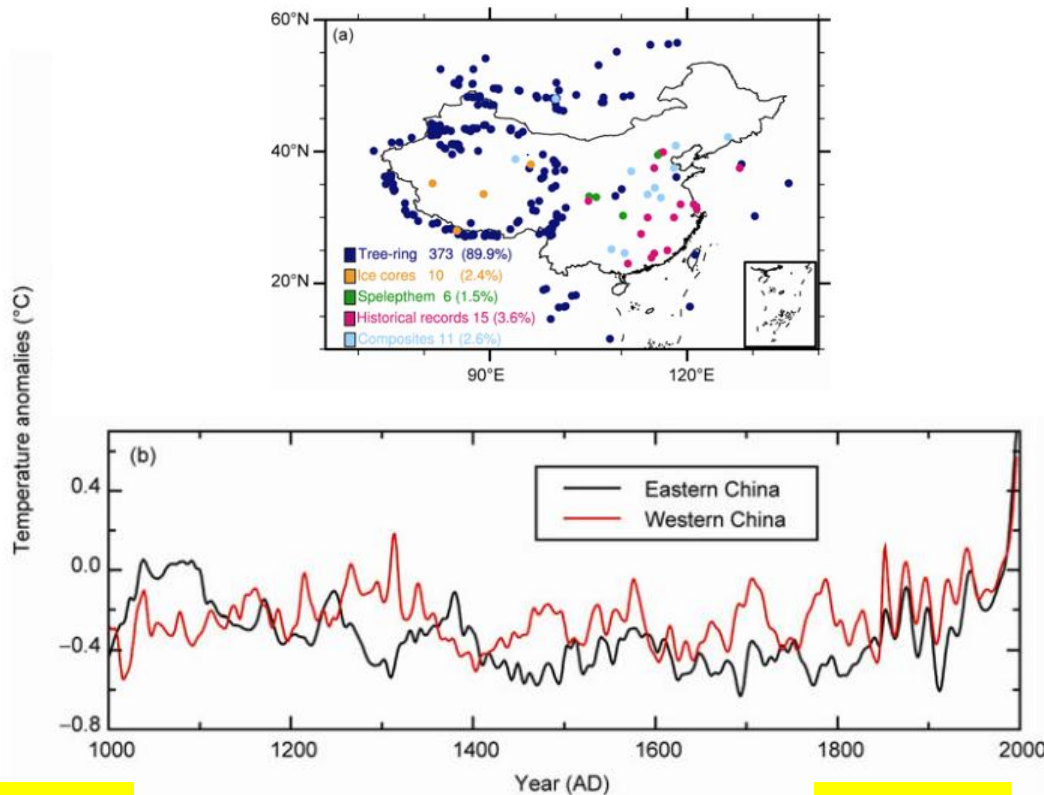
Heavily pre-processed data

No dating uncertainty issues

Linear methods - ‘Regression’

Assumption:

Temp linearly related to proxy

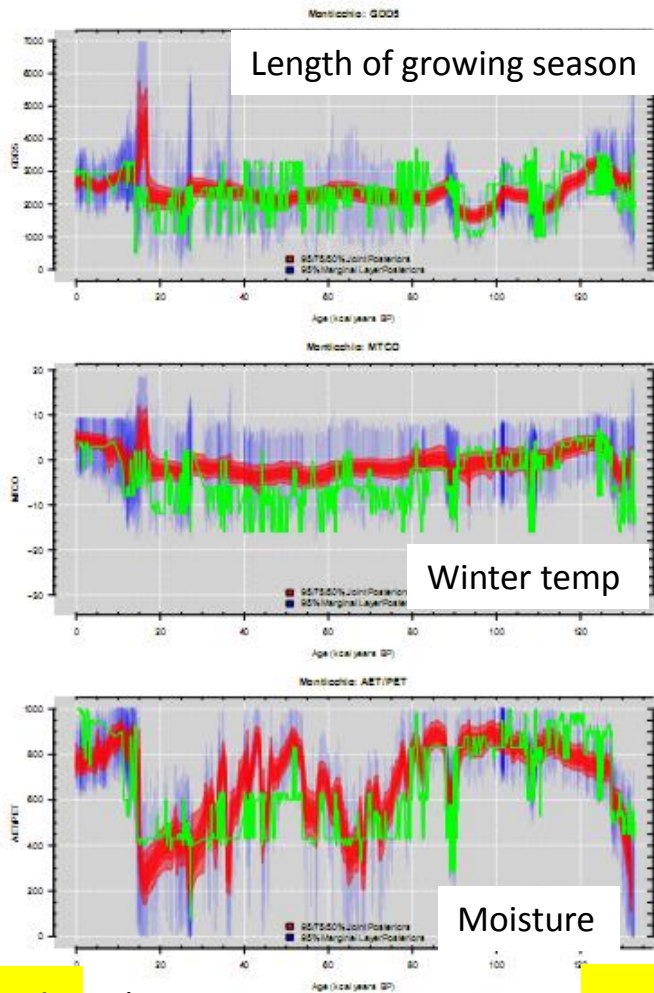


past



present

Italy – Multiple aspects of climate Pollen, last 100,000 years



Parnell IPS006

Focus – Multiple aspects of climate

Uncertainty

Proxy One core

924 samples; 28 taxa

Raw data

No dating uncertainty (unusual for pollen)

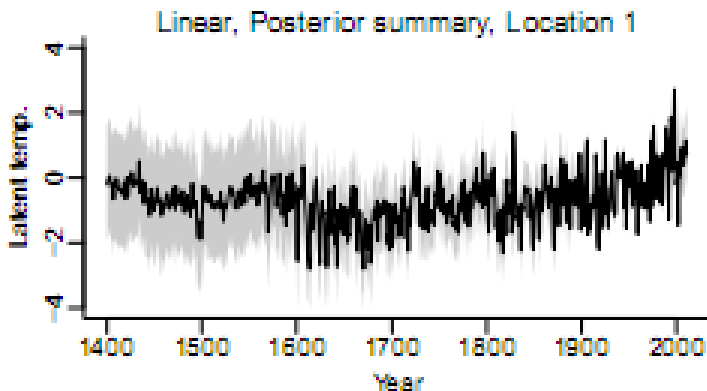
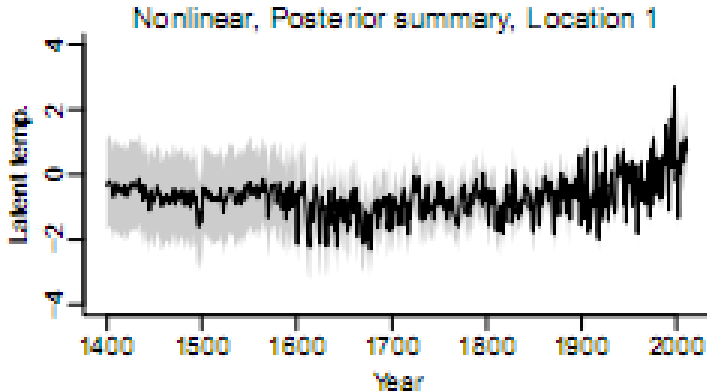
Non-linear, non-Gaussian

Bayesian process modelling

Assumption:

**Non-monotone multivariate relationship
between proxy and climate**

N. America – Air Temp Tree Rings, past 600 years



past



present

Craigmile IPS006

Focus – Air temp

Uncertainty

Proxy Two cores

Tree ring

Heavily processed data

No dating uncertainty issues

Non-Linear, non-Gaussian

Bayesian - process modelling

Assumption:

**monotone relationship between
(processed) proxy and temp**

IPCC Inter Governmental Panel on Climate Change

Abrupt climate change

IPCC WG1 2007

“During the last glacial period, abrupt regional warmings (probably up to 16°C within decades over Greenland) occurred repeatedly over the North Atlantic region”

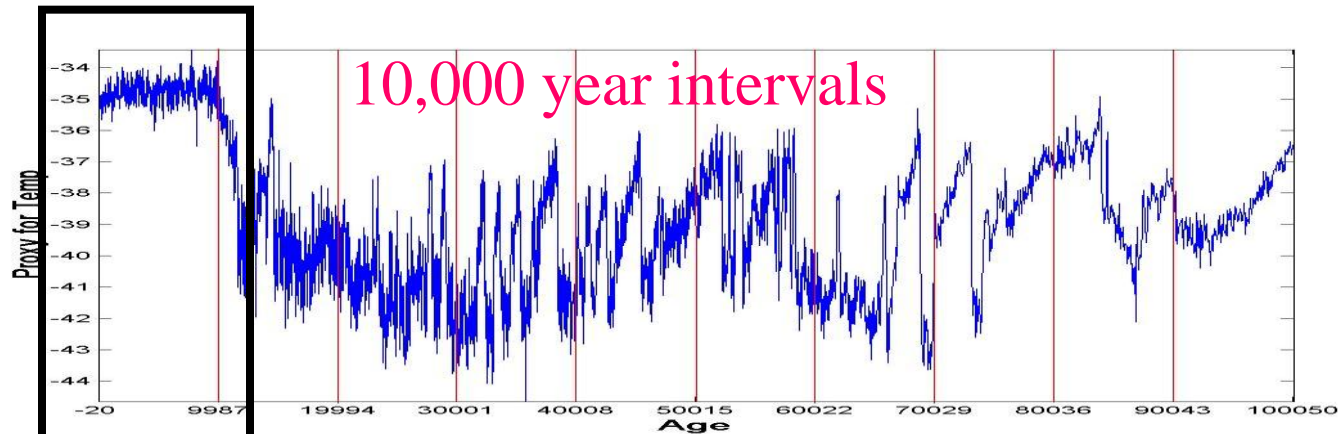
How do we know
anything?



Greenland summit – Air Temp Oxygen Isotope ice last 100,000 years

Modern

Ancient



Irish study period:
next slides

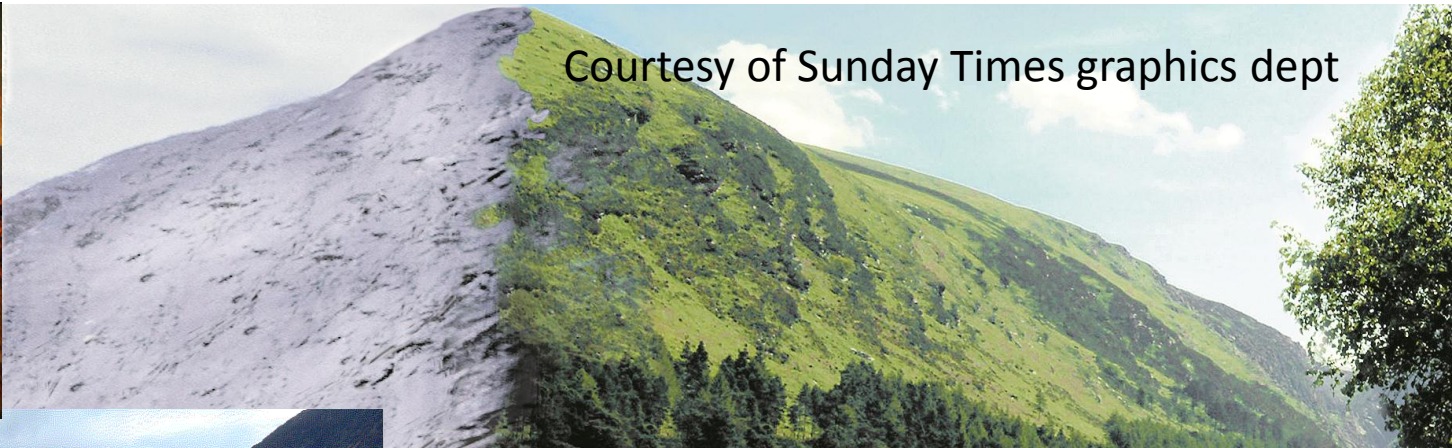
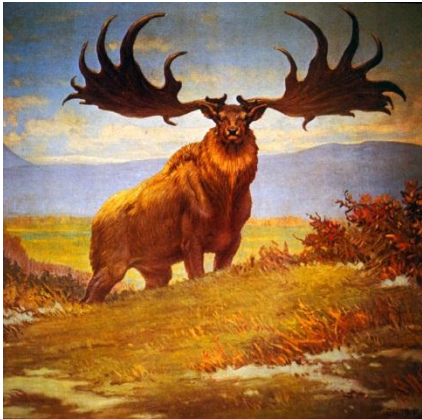
Proxy: Oxygen isotope – proxy for temp

Temporal structure

Frequent small changes, occasional large changes

Roughly (*at this scale*) : long tailed random walk

Glendalough Ireland

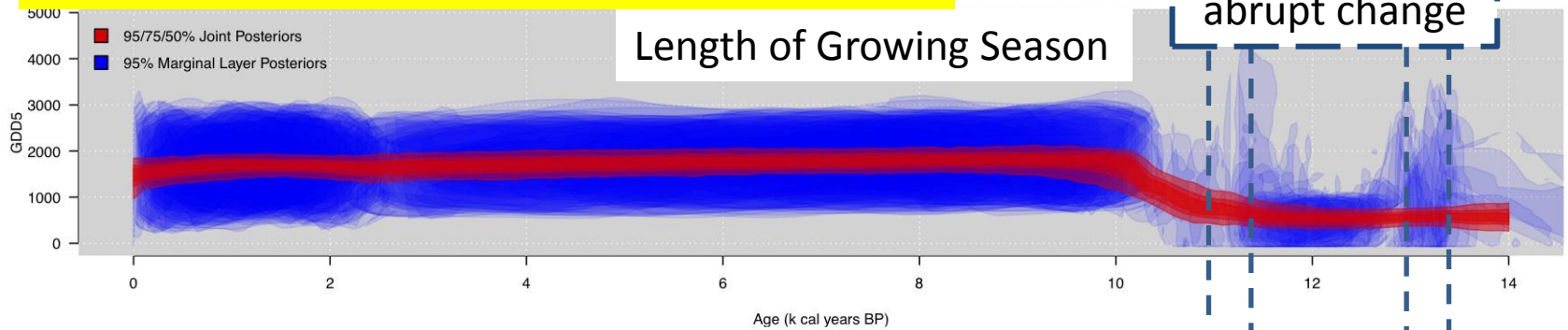


Courtesy of Sunday Times graphics dept

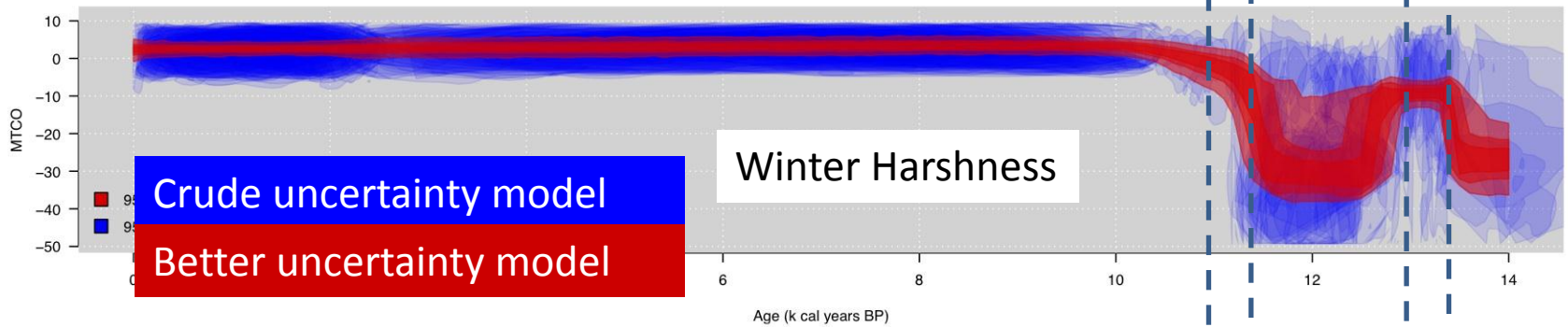


Ireland – Multiple aspects of climate Pollen, last 15,000 years

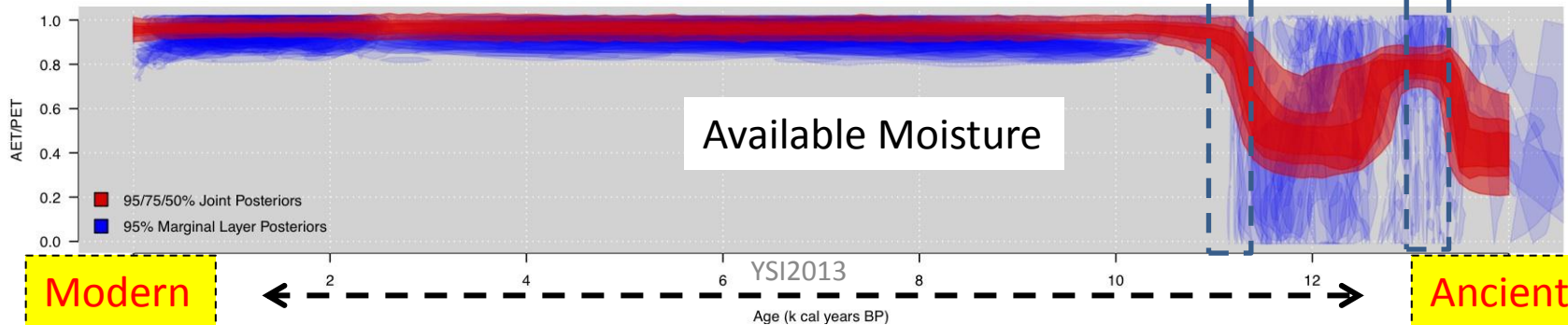
Great dating uncertainty: ± 200 to ± 2000 years



Glendalough: MTCO



Glendalough: AET/PET



Modern



2

4

6

8

10

12

14

YSI2013

Age (k cal years BP)

→

Ancient

Palaeoclimate Reconstruction

What: Examples

How: Uncertainty as Stat Inference on
Latent Climate

Challenges in Communicating Uncertainty

Uncertainty Challenges

- Simple but adequate statistical models of multiple forward stochastic processes
- Data collection and processing
- Inversion – statistical inference, latent processes
- **Communicating uncertainty**
 - How? To whom?
 - About what?
- **Combining uncertainties**

“probably up to 16°C within decades over Greenland”

“Uncertainty *Fusion*”

Statistics: Inverting Forward Processes

Science

Forcing
Insolation
Volcanoes
CO₂

Feedback
Ocean/Ice
Bio-sphere
Anthro-CO₂

Climate processes $C(s,t)$



Proxy processes $P(s,t)$



Data collection $Y(s_i, d_i)$



Data processing $\{Z(s,t)\}$

Statistics

Models

Physical/Statistical

- Uncertainty
- Inference

Eg Depth

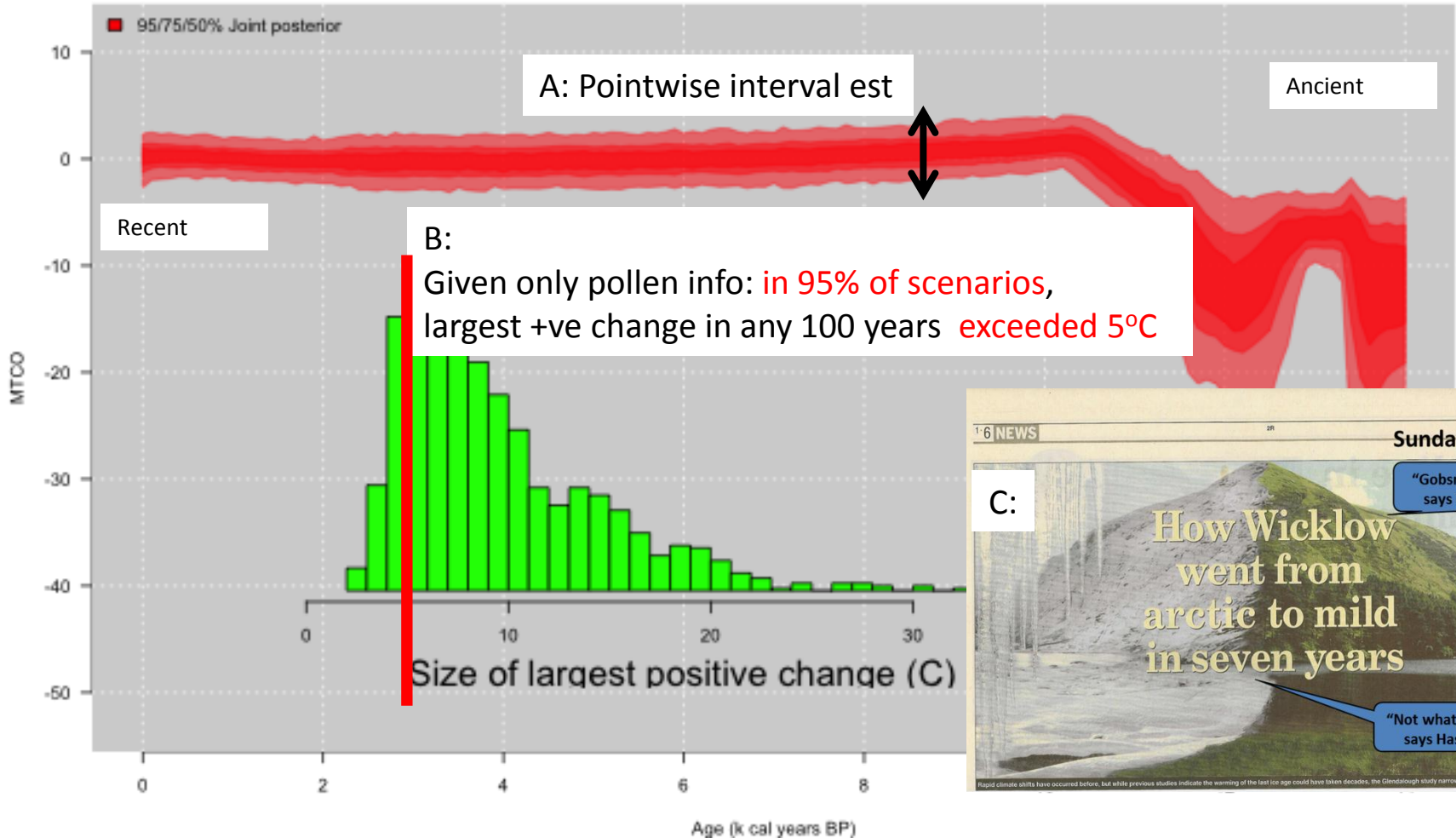
- Proxy for Age
- Uncertain time

Data

- Raw
- Processed

Communicating Uncertainty

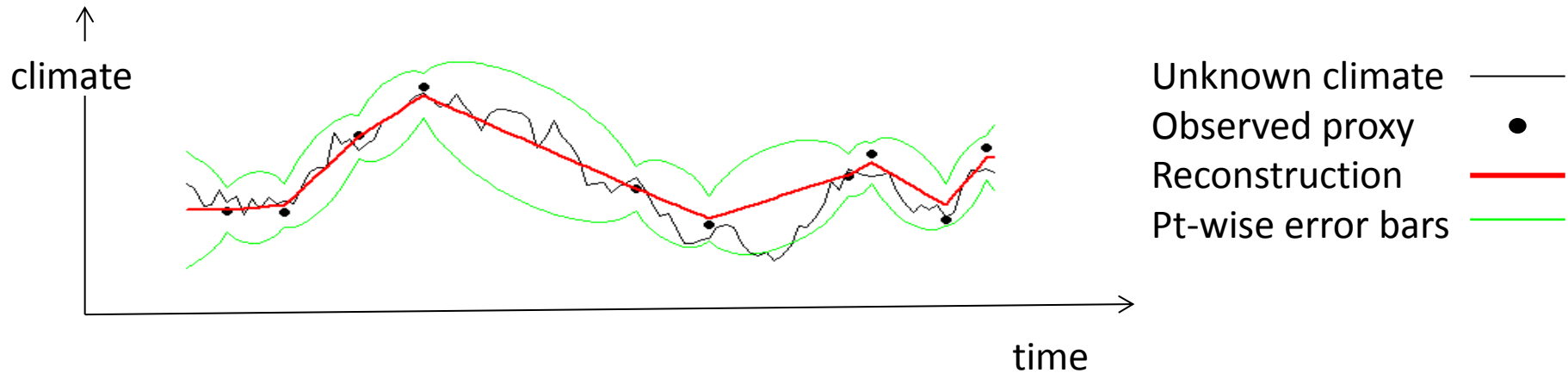
Glendalough: Harshness of Winter



Statistical Methods

- Partially observed stochastic processes

Simple case: single core: one location
underlying latent $c(t)$ univariate Gaussian
slices correspond to known irregular times
proxy process \Rightarrow additive univariate Gaussian errors
with simple variance structure
'Reconstruct' \Rightarrow make statistical inference

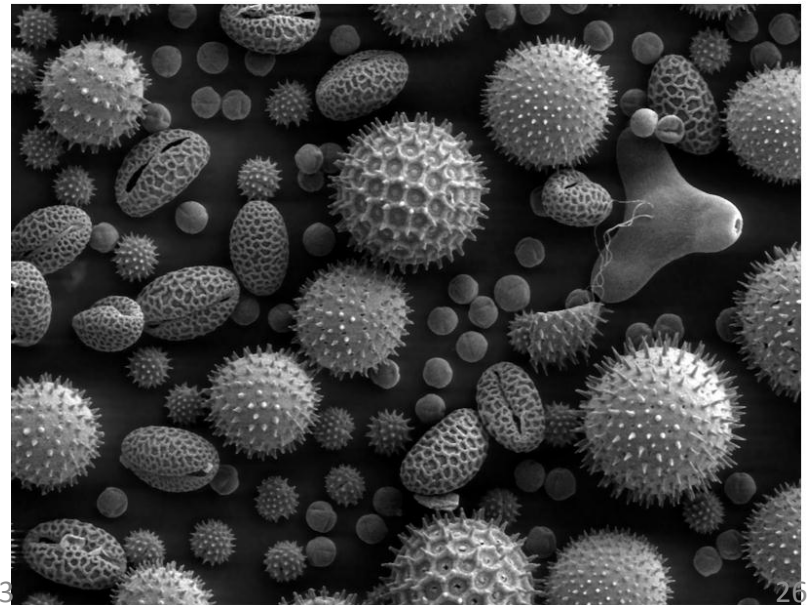


Statistical Methodologies

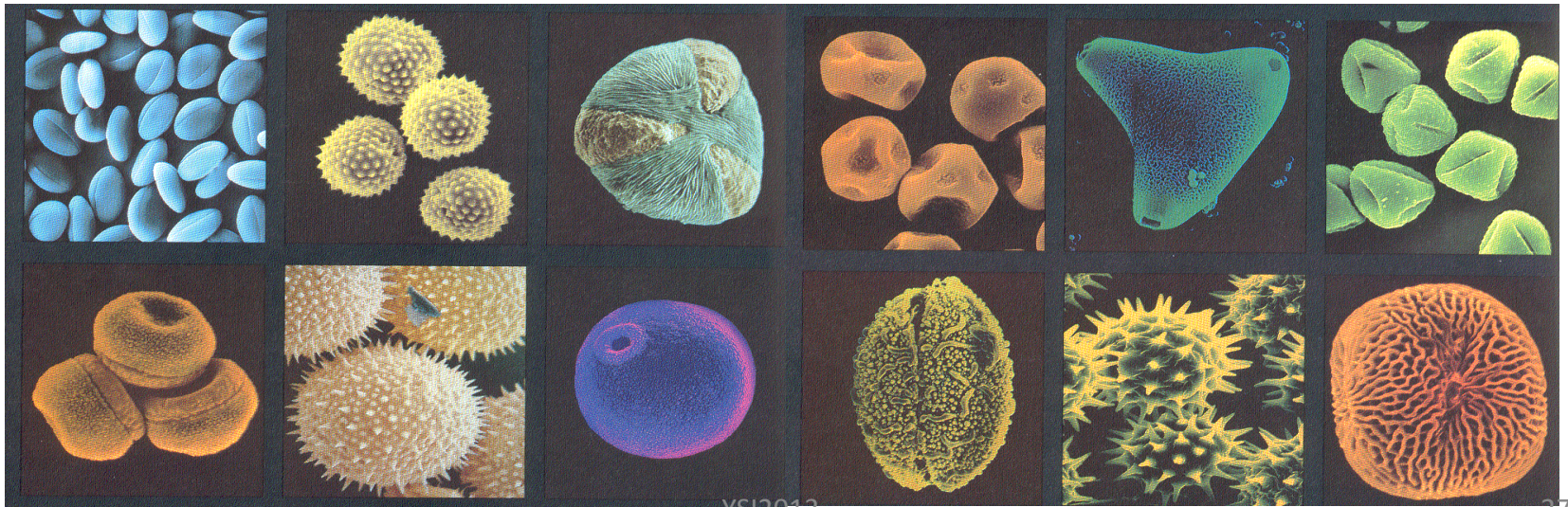
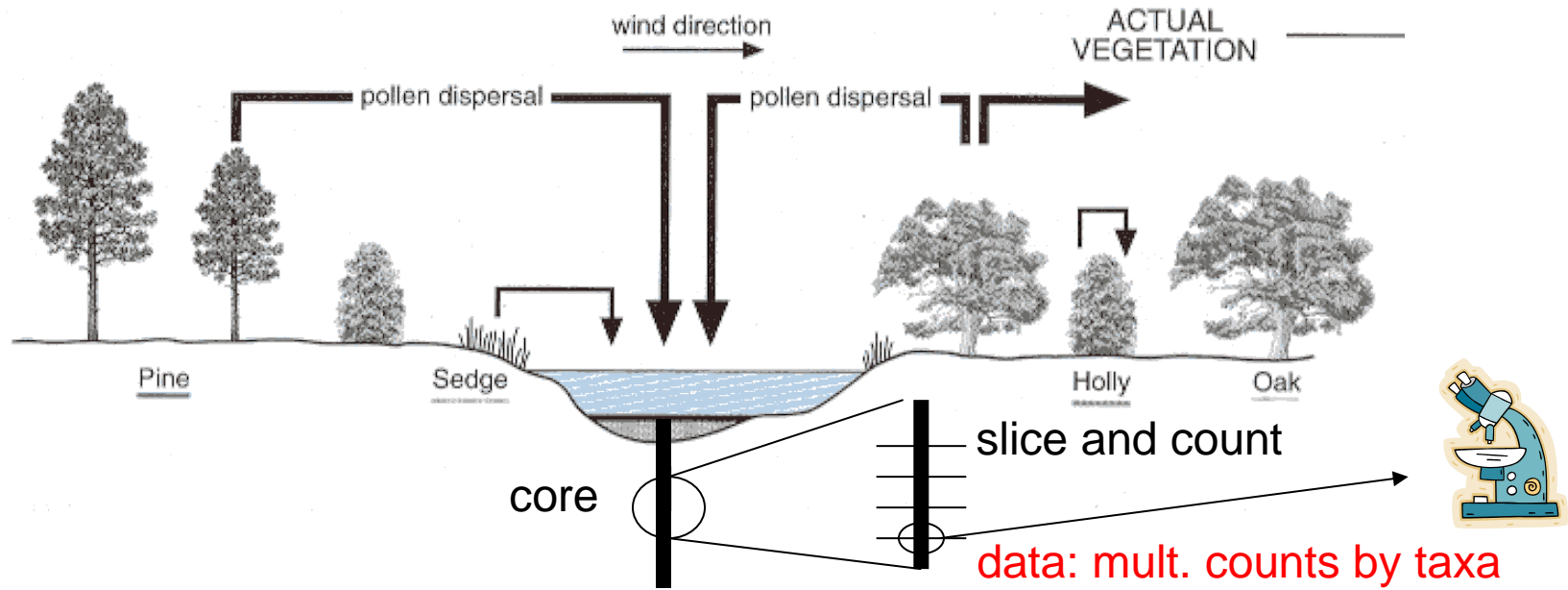
- Partially observed stochastic processes
 - State-space modelling *multivariate* climate
 - Space-time processes *long-tailed* climate inc's
 - Zero-inflated multivariate proxy processes *pollen*
- Multiple irregular time series
 - Time series with uncertain times *sediment*
- Interest in possibly *non-linear functionals*
- Monte Carlo methods in Bayesian Modelling
 - Integrated Nest Laplace Approximations INLA
 - Stochastic partial differential equations SPDE

probably
up to 16°C
within
decades
over
Greenland

Show me the data

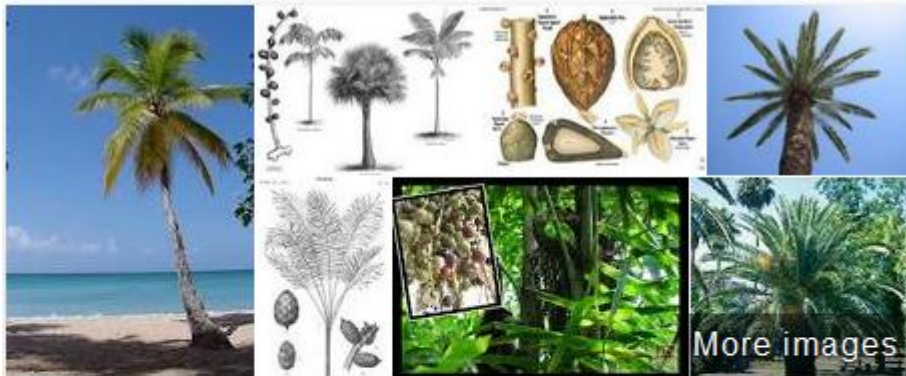


Pollen: widely available, but challenging



Asian pollen - palms

From Wikipedia



Areaceae

Plant

The Areaceae are a botanical family of perennial lianas, shrubs, and trees commonly known as palms. They are flowering plants, the only family in the monocot order Arecales. [Wikipedia](#)

Roughly 202 genera with around 2600 species are currently known, most of them restricted to tropical, subtropical, and warm temperate climates.

Morphologically diverse, inhabit nearly every type of habitat within their range, from rainforests to deserts

Distinctive pollen... but

*how many of 2600
can be distinguished by eye
from ancient pollen grains?*

Changing pollen composition

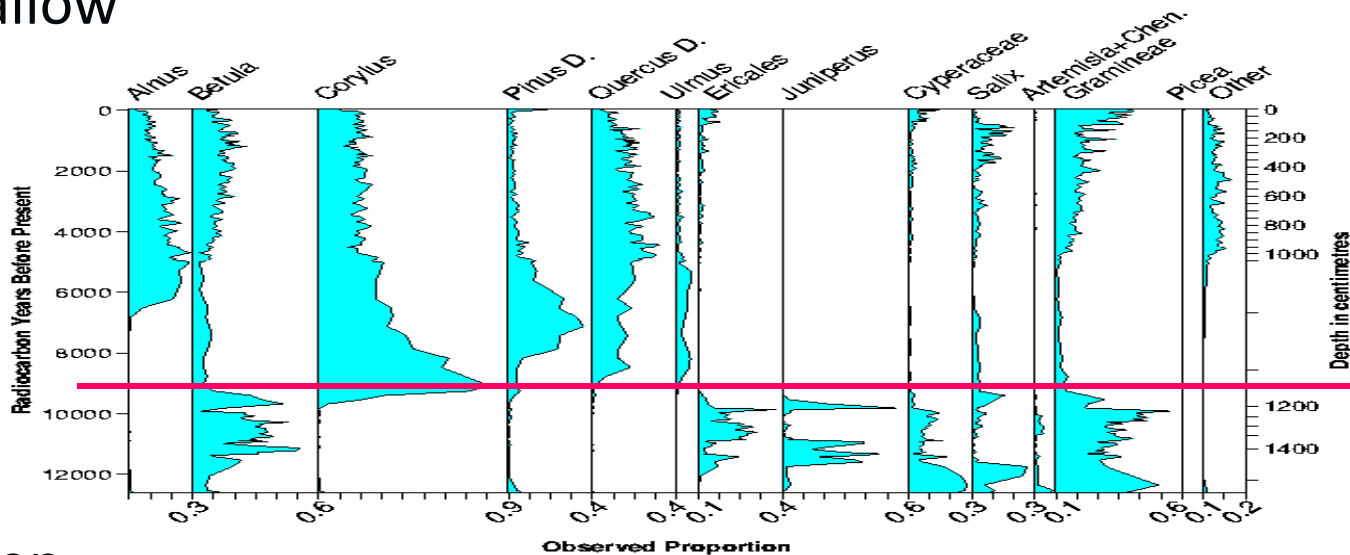
Data: compositional vector of observed pollen proportions

Proxy process: pollen rain,

Statistical issues discrete multivariate,
zero inflated, non Gaussian, non-linear

Pollen composition changes \Rightarrow Climate changes

Recent Shallow

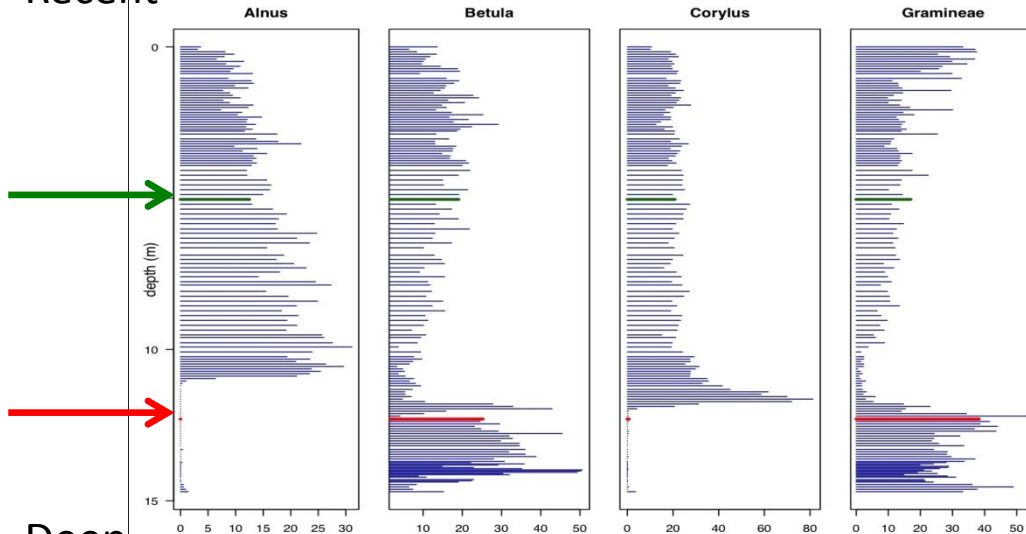


Ancient Deep

Climate Information from Pollen

Shallow
Recent

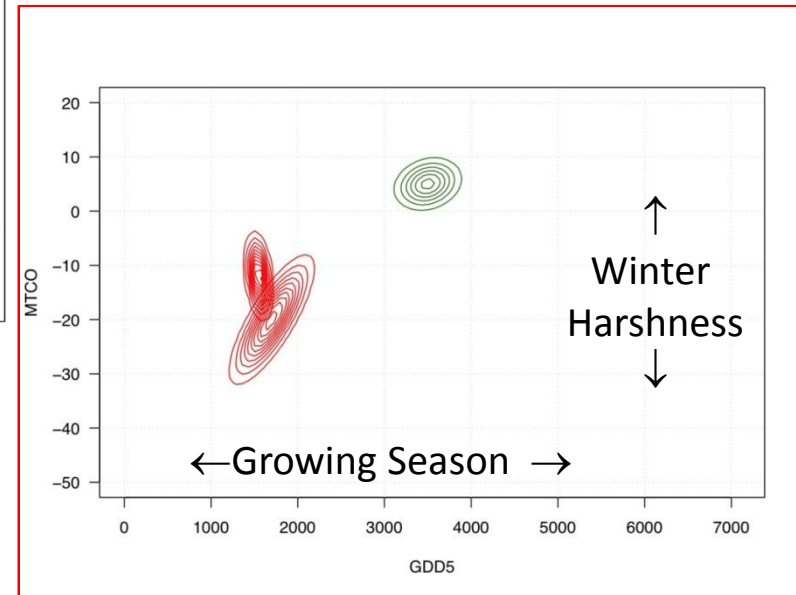
showing 4 of 28 taxa



Deep
Ancient

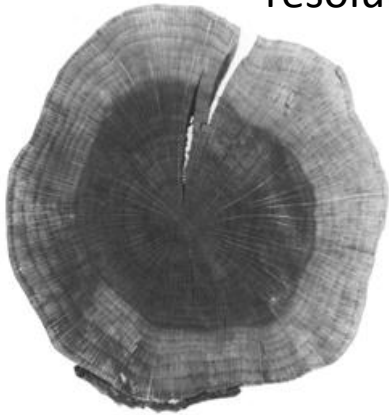
Depths known,
Times uncertain

Pollen expert \Rightarrow
Bivariate (poss bimodal) climate



Other Proxies

'Laminar' –
annual
high-temporal
resolution



Oak tree

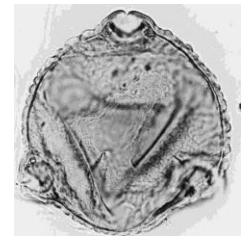
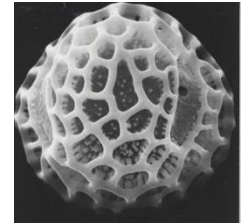
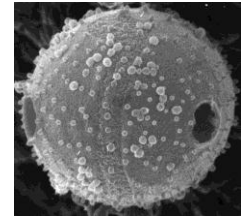


GISP ice



Sediment

low-
temporal
resolution

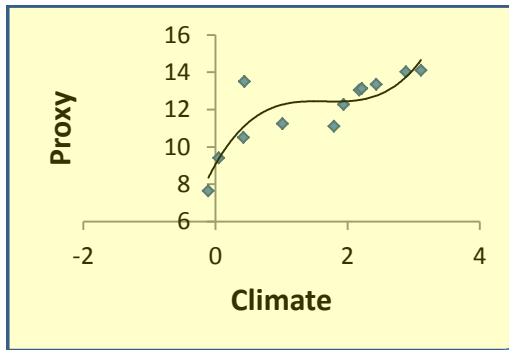


Pollen

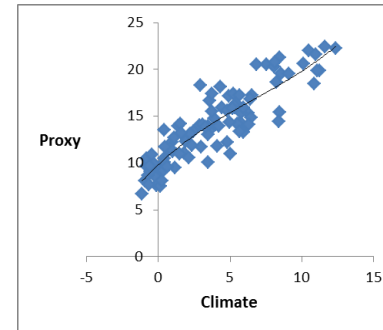
Calibration data

- Past ~ 150 years
 - Overlapping time series

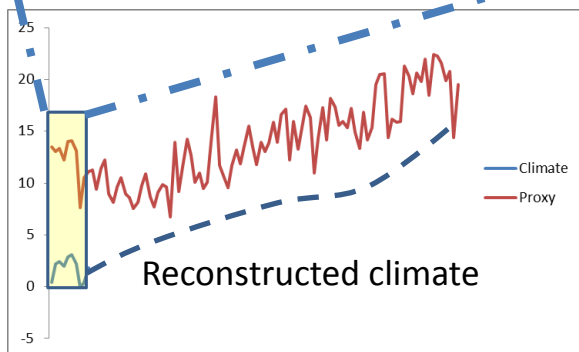
- Modern analogue
 - Space for time



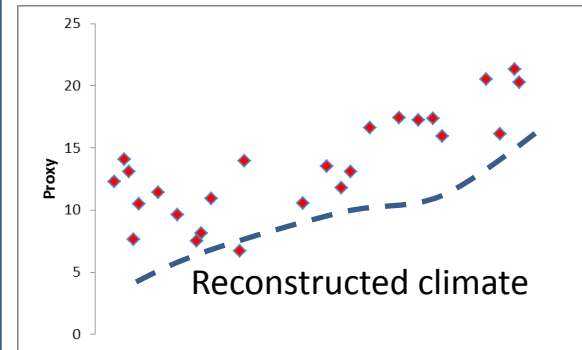
Proxy and known recent climate at different times at *that* location



Proxy and known *modern* climate at *many* locations across space



Common when proxy has good temporal resolution

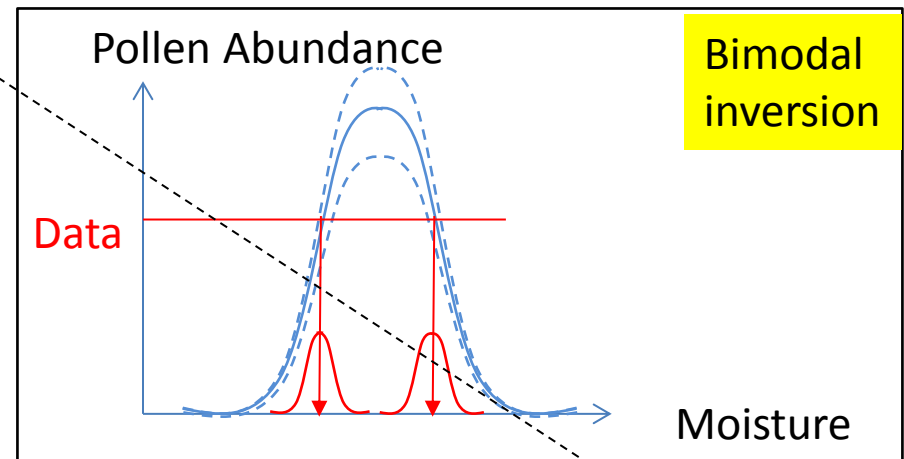
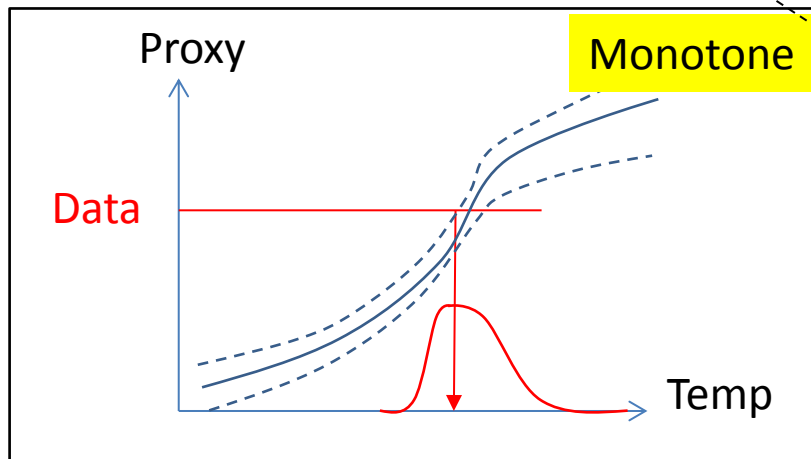
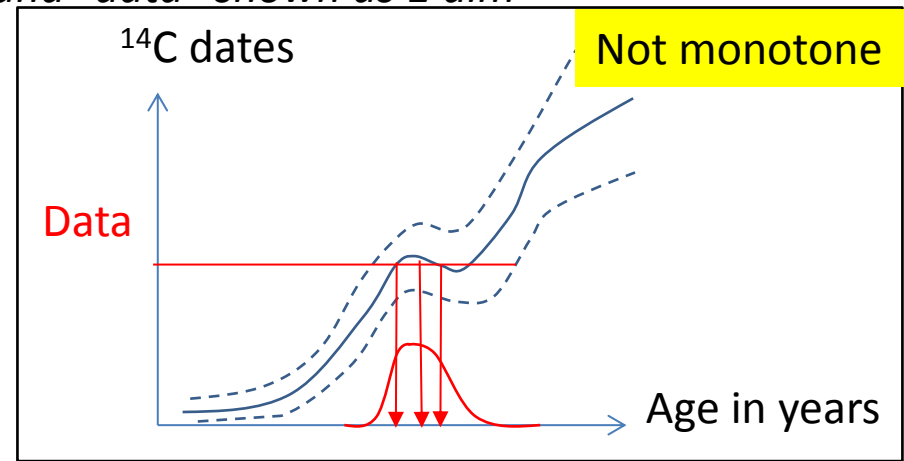
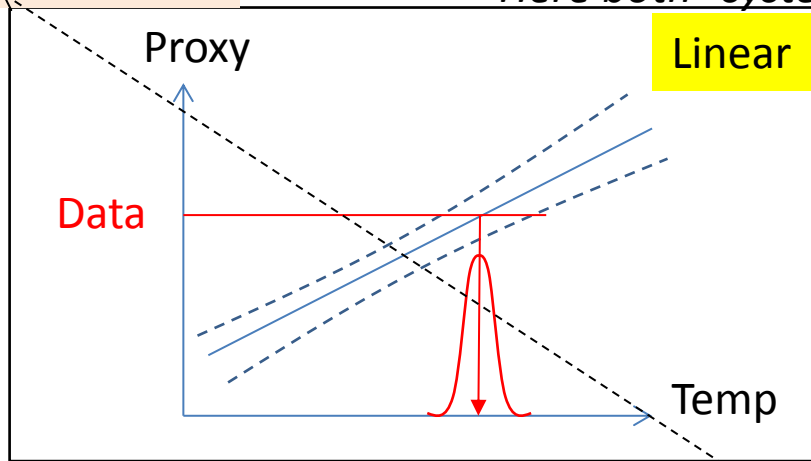


Common when proxy has poor temporal resolution

Inverting Simple Proxy Relationships

Easy proxies

Here both "system" and "data" shown as 1 dim



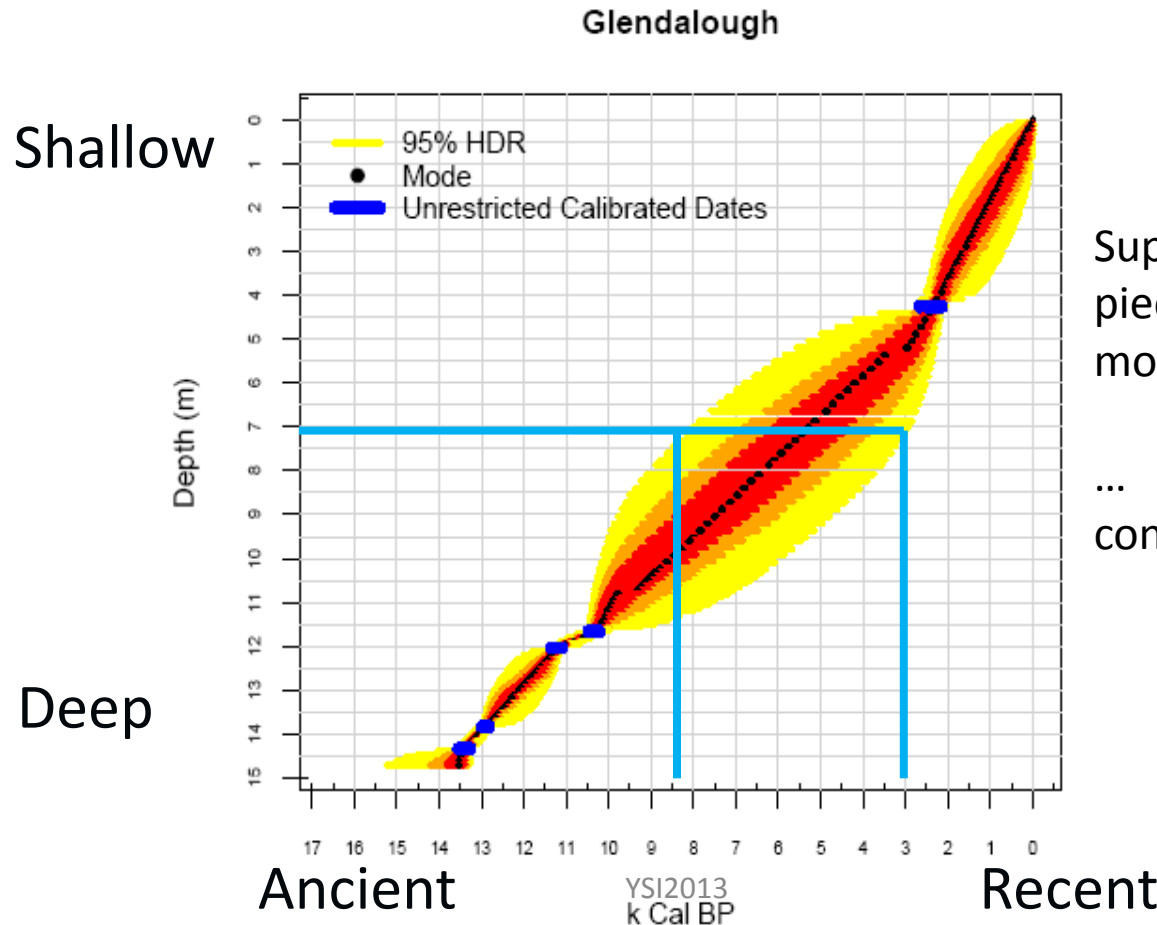
Forward: Climate → Proxy

Inverse: Proxy → Climate

Hard proxies

Temporal Uncertainty: Depth \Rightarrow Age?

Glendalough: 150 depths; 5 ^{14}C dates

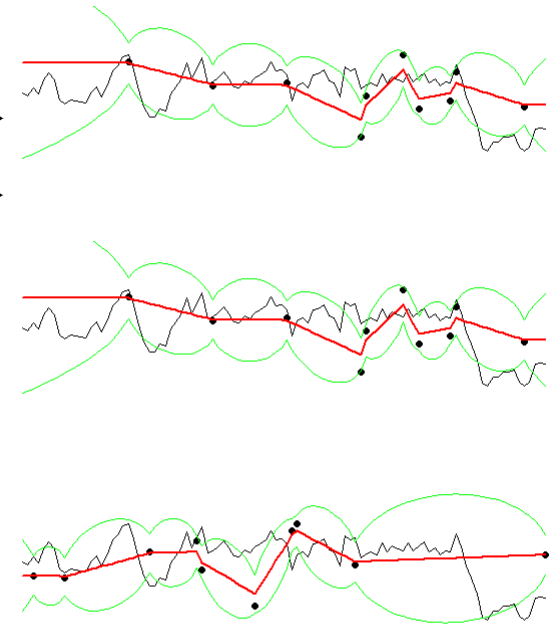


Random
Chronologies
Superimposed
piece wise linear
monotone functions
 $d(t)$
...
consistent with ^{14}C data

Working with uncertain times

Proxy values known $y = \{y_1, y_2, y_3, \dots\}$
 Depths known: $d = \{d_1, d_2, d_3, \dots\}$
Chronology unknown $t = \{t(d_1), t(d_2), t(d_3) \dots\}$
 Seek distribution of $c = \{c(t_1), c(t_2), c(t_3) \dots\}$

Same y values, different times



Algorithm

1. Sample chronology from $p(t | d)$
2. Compute $p(c | y, t)$
3. Repeat

and then Combine... $p(c | y, d)$

Combining Uncertainties

- (proxy, depth) pairs \Rightarrow random (climate, age)
- Ensemble of (climate, age) pairs \Rightarrow
random history
- Sample histories $c(t)$ for each set of times
 - Combine random histories
 - *Monte Carlo marginalising*

Reconstructing Palaeoclimate

- Data:
 - (climate proxy, age proxy) pairs
- What was the most likely palaeoclimate?
 - What are the associated uncertainties?
- Conceptual algorithm
 - Find palaeoclimate histories
that are consistent with the data
 - Summarise

Palaeoclimate Reconstruction

What: Examples

How: Uncertainty as Stat Inference on
Latent Climate

Challenges in Communicating Uncertainty

Communicating Uncertainty *scientist to scientist*

Scientist 1 Chronology expert

Generates random chronologies

Publishes random chronologies

(or algorithm)

Scientist 2 Pollen expert

Downloads random chronologies

Reconstructs random climate for each

Publishes random climates

(or algorithm)

Marginal and Joint Reconstruction given single core

- Marginal Data: (less info)
 - (climate proxy, age proxy) pair, each slice
 - What was the most likely palaeoclimate?
 - What are the associated uncertainties?

For each
slice?

- Joint Data: (more info)
 - (climate proxy, age proxy) pair, all slices
 - What was the most likely palaeoclimate?
 - What are the associated uncertainties?

For all
slices
taken
jointly?

Palaeoclimate Reconstruction

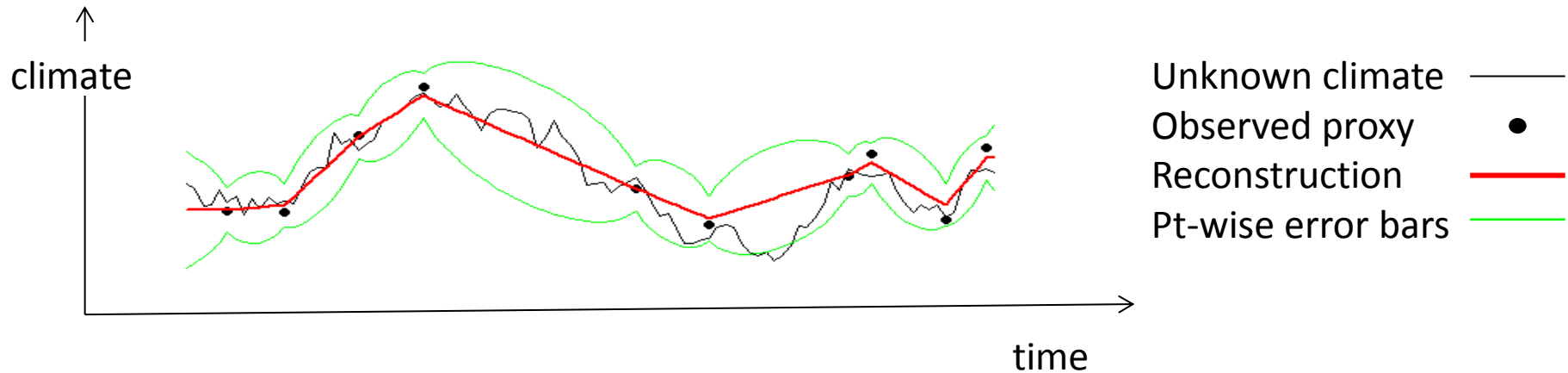
given multiple cores, multiple proxies

- Data:
 - (climate proxy, age proxy) pairs
 - Multiple cores, proxies, slices
 - What was the most likely palaeoclimate?
 - Aspects of climate
 - Space-time histories
 - Associated uncertainties
 - Define, compute, present

Statistical Methods

- Partially observed stochastic processes

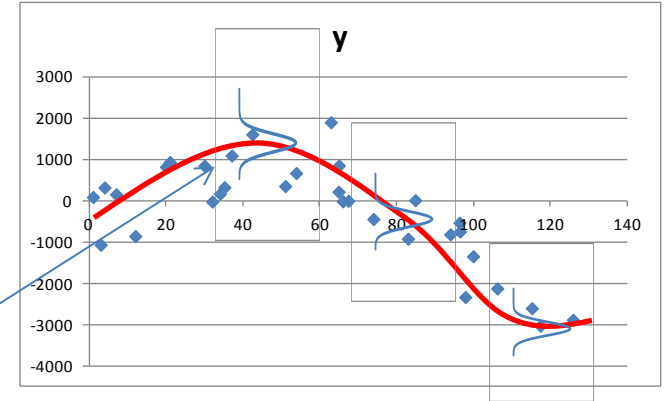
Simple case: single core: one location
underlying latent $c(t)$ univariate Gaussian
slices correspond to known irregular times
proxy process \Rightarrow additive univariate Gaussian errors
with simple variance structure
'Reconstruct' \Rightarrow make statistical inference



Statistics by Ski Slalom



- **Marginal** data \Rightarrow gates
 - Different “widths”
- Physics defines the smoothness
- *Challenge*
 - ‘Location’ of gates uncertain



Collect sample paths *only* for those that successfully pass thru *all* the gates?

Joint inference

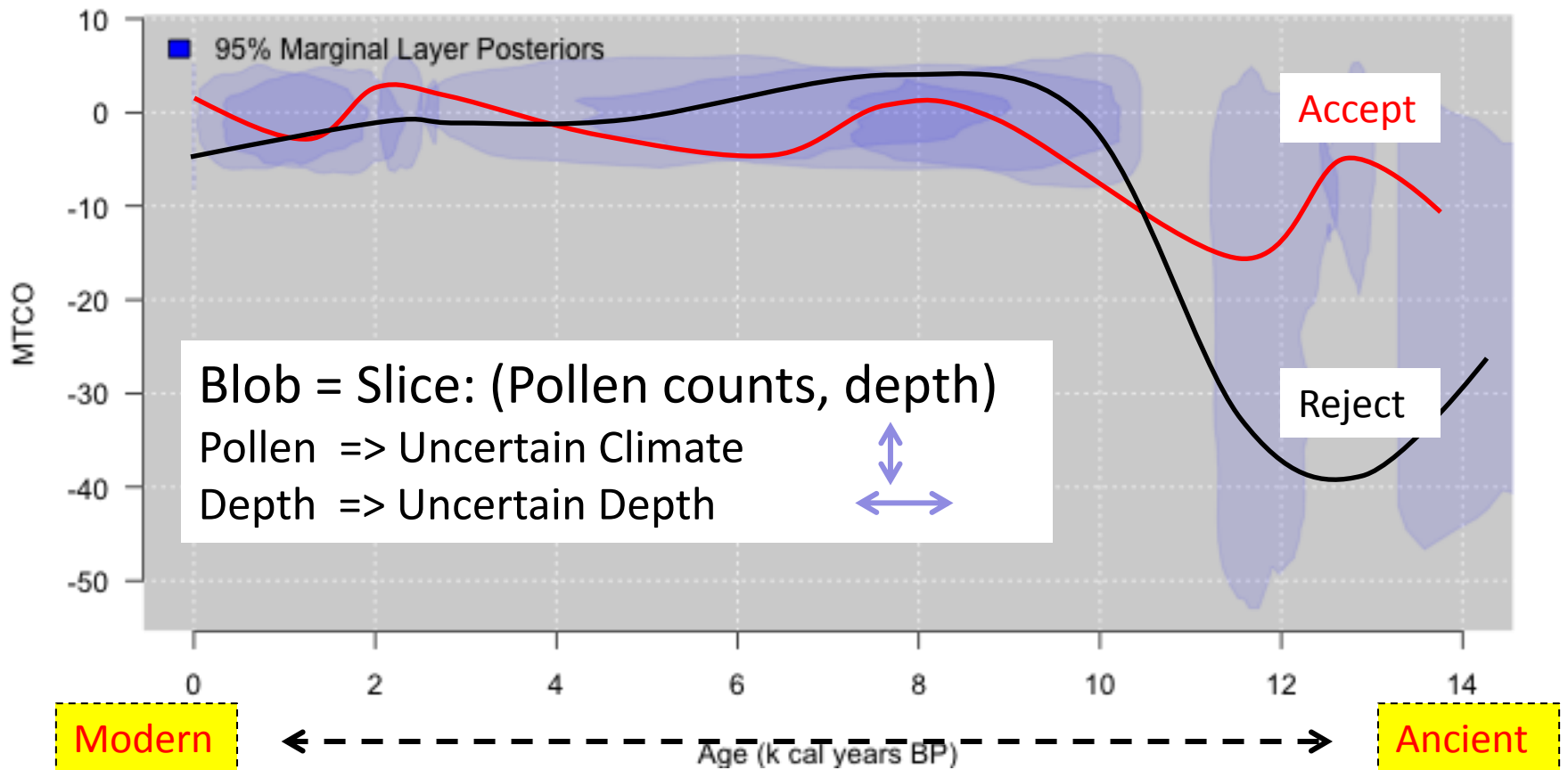
Inference by rejection sampling

- Model $y = f(x, z, \theta)$ statistical/physical
 - Observe data y , covariates x
 - Infer latent processes z , fixed parameters θ
- Conceptual algorithm
 - Propose z, θ
 - Statistically consistent with y, x ?
 - Yes No
 - \Rightarrow Reject with low prob \Rightarrow Reject with high prob

Algorithmically inefficient

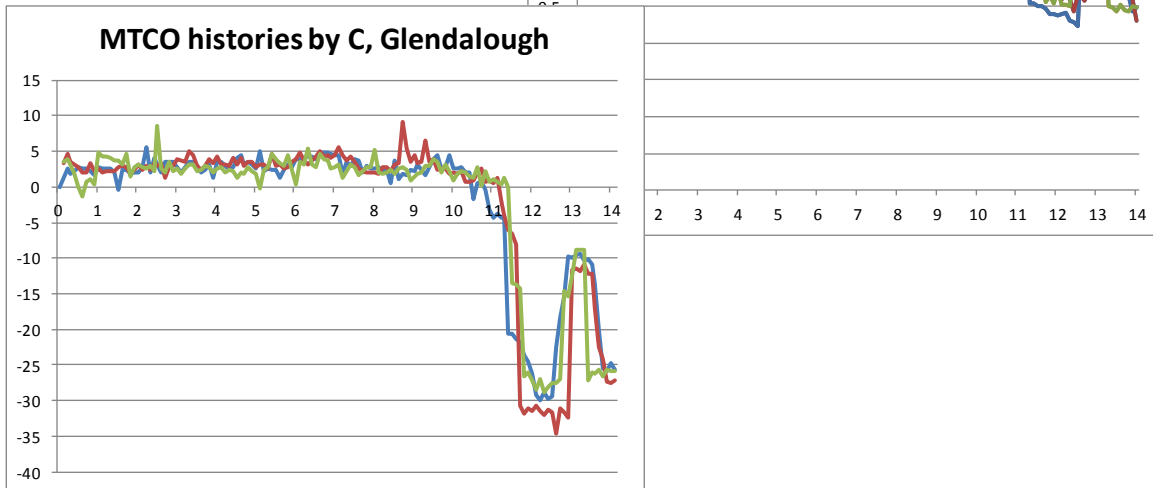
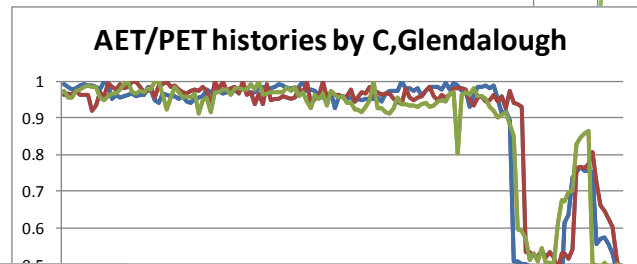
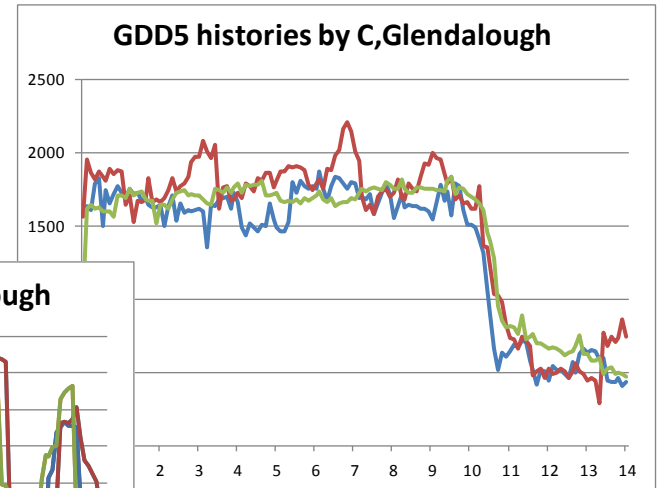
Data as gates with uncertain time

Winter harshness



Sampled histories at one location

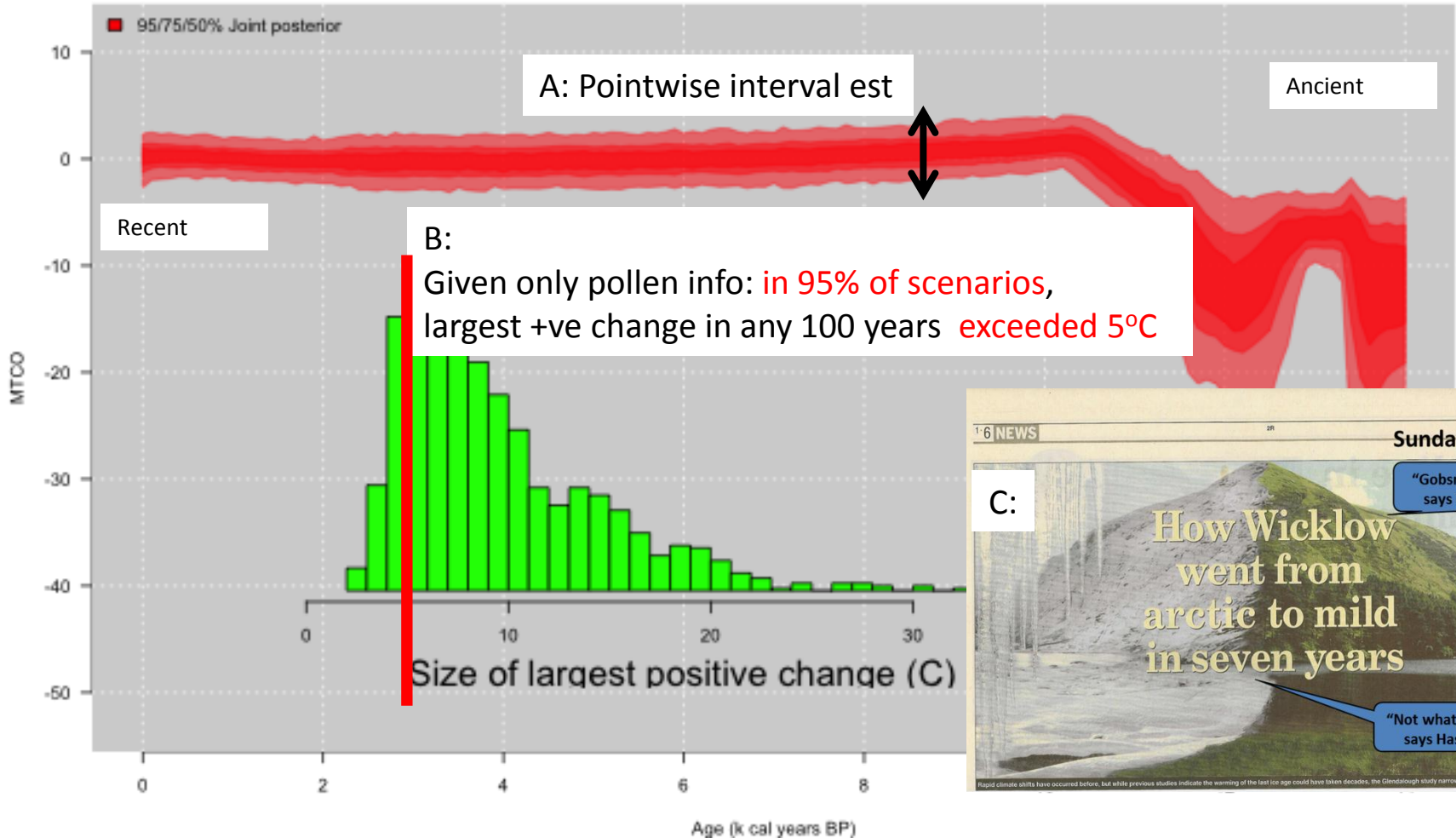
Three aspects of climate
One three-dim history



One site in Ireland

Communicating Uncertainty

Glendalough: Harshness of Winter



Palaeoclimate Reconstruction

Examples

Statistical Methods:

Uncertainty in Stat Inference on Climate

Challenges in Communicating Uncertainty

Uncertainty Challenges

- Communicating uncertainty
 - Scientist to scientist
 - Scientist to public
- Modularising uncertainty

Modularising Uncertainty

- Scientist 1 Pollen expert
 - publishes (samples from) $p^{pollen}(c_i | y_i)$ as $\{ c_i^{pollen} \}$
- Scientist 2 Tree Ring /Dendro expert
 - publishes (samples from) $p^{dendro}(c_i | y_i)$ as $\{ c_i^{dendro} \}$
- Scientist 3 SpaceTime Series expert
 - accepts (samples from dists) $\{ c_i^{pollen} \} \{ c_i^{dendro} \}$
 - accepts random chronologies $\{ t_i \}$
 - rejects ensembles of pairs $\{ c_i^{pollen}, c_i^{dendro} \}$
 - not jointly consistent
 - with each other
 - with ‘smoothness’ – in space
 - in time
 - publishes jointly *acceptable space/time histories* $\{ c^{pollen}, c^{dendro} \}$

Modularising Uncertainty

Not (quite?) regular Bayesian paradigm

Regular Bayes

Prior for $[c(t, s) | \theta_c]$

Cond Indep proxy processes

Likeli-
hoods

depth/time $[d_i | t_i, \theta_{chron}]$

pollen $[y_i^{pollen} | c(t_i, s_j), \theta_{pollen}]$

dendro $[y_i^{dendro} | c(t_i, s_j), \theta_{dendro}]$

Posteriors based on
partial data

depth/time $[t_i | d_i]$

pollen $[c(t_i, s_j) | y_i^{pollen}]$

dendro $[c(t_i, s_j) | y_i^{dendro}]$

$$\Rightarrow [\{c(t_i, s_j)\}, \theta | \{d_i, y_i^{pollen}, y_i^{dendro}\}]$$

$$\Rightarrow [c | \{d_i, y_i^{pollen}, y_i^{dendro}\}]$$

Uncertainty in palaeoclimate reconstruction

- Science
 - Climate multivariate (not just temperature)
 - Driven by physics, forcing (and computer models)
 - Multiple processes
 - Including multi-modal reconstruction
 - Uncertain times
- Challenges in Uncertainty
 - Modularising
 - Communicating

Conceptual
Algorithmic

Challenges

- Communicating uncertainty
 - Scientist to scientist
 - Scientist to public
- Modularising uncertainty
- Fast algorithms
 - Avoid MCMC (as much as possible)
- Inference for computer model parameters

Monte
Carlo
methods

Resources

Past 1000 yrs http://en.wikipedia.org/wiki/Temperature_record_of_the_past_1000_years

Past ~11000 years, Holocene <http://www.ipcc.ch/ipccreports/tar/wg1/073.htm>

Past 400ka http://en.wikipedia.org/wiki/Ice_core#Ice_core_data

Abrupt Change <http://www.ncdc.noaa.gov/paleo/abrupt/index.html>

Haslett, John, et al. "Bayesian palaeoclimate reconstruction." *Journal of the Royal Statistical Society: Series A (Statistics in Society)* 169.3 (2006): 395-438.

Li, Bo, Douglas W. Nychka, and Caspar M. Ammann. "The value of multiproxy reconstruction of past climate." *Journal of the American Statistical Association* 105.491 (2010).

Tingley, Martin P., et al. "Piecing together the past: Statistical insights into paleoclimatic reconstructions." *Quaternary Science Reviews* 35 (2012): 1-22.

See also IPS006

General

- The long summer : how climate changed civilization Fagan, Brian (2004), Granta
- Collapse : how societies choose to fail or survive, Diamond, Jared (2006) Penguin
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