

A chain of processes - from past climate variations to paleoclimate reconstructions

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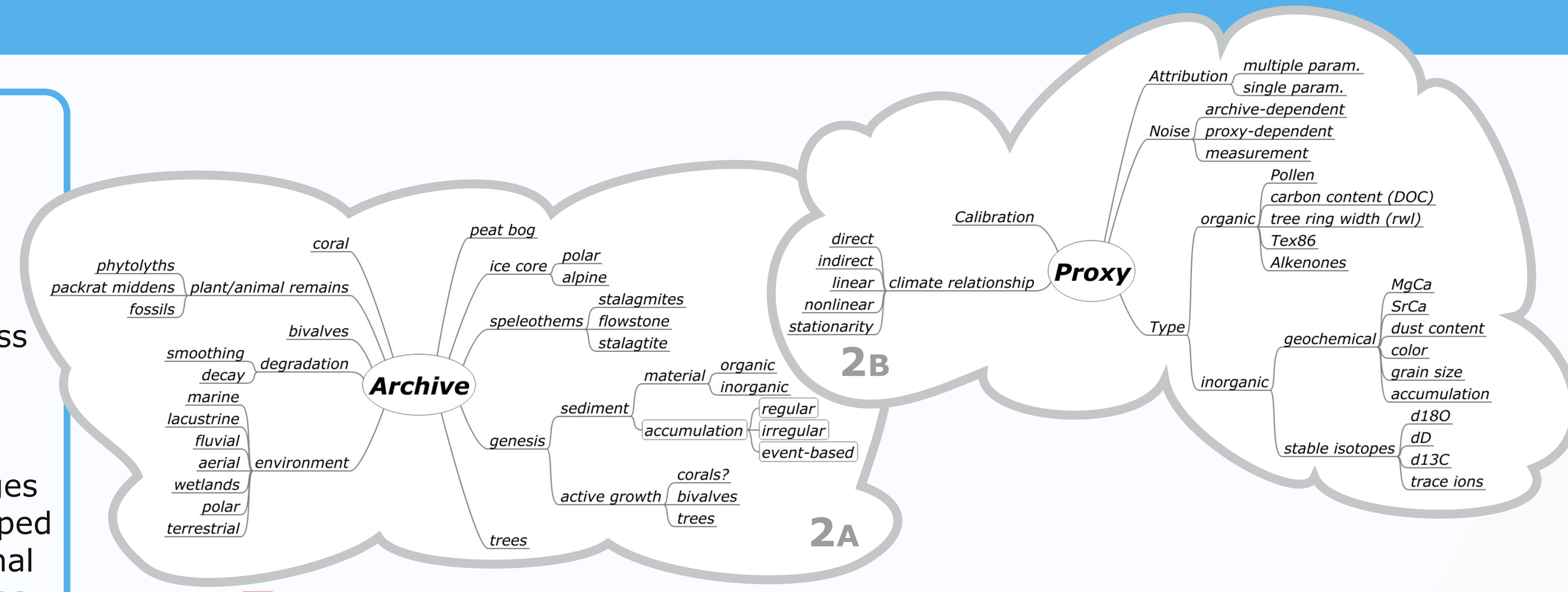
1 **Past climate variations**

are recorded, through nature's sampling, in paleoclimate archives. It is essential to quantify estimation uncertainties, to which each step along the way contributes. Integrating techniques and information across archives and proxies, allows to identify complimentary archive-proxy combinations which can improve past climate estimates across timescales. We find common challenges for many archives, where techniques developed in different communities could reduce the final uncertainty of **paleoclimate reconstructions**.

- Pre-deposition:** $\int dxdt$
Sampling the environment
- Deposition/Growth** into archive
- Post-deposition processes** during environmental storage
- Human sampling** the archive
- Measurement of proxy** $P(z)$
- Age reconstruction** $z(t) \rightarrow t^*(z)$
- Climate reconstruction** $P(z) \rightarrow C^*(t)$

$T(t), P(t), \rho(t)$

$T^*(t^*), P^*(t^*), \rho^*(t^*)$



2 The structure of the archive and proxy landscape is complex. Which processes affect archives (2A) - and which are proxy-dependent (2B)? Which proxies can be measured on the same archive sample?

3 **Common challenges across archives and proxies**

Challenge	Problematic aspect	Effect on the reconstruction	Approaches (additions & feedback welcome)	Comments (examples/ suggested references)
Attribution & fidelity	Nonstationarity of proxy ↔ climate response	False trends & nonstationarities	Multi-proxy reconstructions	
	Control by multiple climate parameters	Non-independent reconstructions, false trends & covariances	a) forward modeling (req.: good proxy understanding) b) test against modeled/observed parameter combinations c) use multiple proxies to find common response	
Recorder sampling → affects mean/ variance	Snapshot/Intermittent/Event-based	Intra-annual variance > inter-annual variance → snapshot-sampling increases the proxy variance	Simulate & test the effect of aliasing and intermittency (e.g. Persson et al., 2009). With known seasonality: Correct in spectral domain (Kirchner, 2009; Laepple & Huybers, 2013)	
	Quasi-continuous	Growth rate changes → time series irregularity	E.g. adapted time series estimators (e.g. Rehfeld & Kurths, 2014)	
	Organic growth	Bias effects?	?, Monitoring & sensitivity studies	
Seasonality	Spatial vs. temporal signal integration?	Residence/mixing times, changing sources	Monitoring & sensitivity studies	
	Seasonal recorder sampling	If unaccounted for: Biased annual mean, also potentially false trends over millennial timescales (insolation)	Test sensitivity on seasonal recording (Laepple et al., 2011)	
Smoothing/Filtering	Damping of the signal	Variance reduction – can be accounted for if smoothing depth is known	At low noise levels: Unfolding (Berger, Johnson & Killingley, 1977; Johnsen et al., 2000)	
			Variability estimates: Correction of the variability spectrum (Laepple & Huybers, 2013)	
Chronologies	Age modeling	Age tracer ≠ Proxy → time uncertainty leads to bias/variance & problem of cause vs. effect – difficult comparison	Transfer time uncertainty to proxy uncertainty (COPRA, Breitenbach et al., 2012) using age modeling with point-based uncertain (e.g. U/Th, C ¹⁴) and fixed dates (e.g. Volcanoes, Dust layers, known core top).	
	Layered archives	Missing layers, combination of layer counts with point-age estimates → time uncertainty	Layered archives: Cross-dating for multiple cores, combination of laminated sections with point-wise age estimates (COPRA) to reduce age uncertainty.	
Calibration	Spatial and temporal climate/proxy variance can be different	Under-/Overestimation of Low-/High-frequency components	Monitoring & sensitivity studies for proxy understanding; Multi-Proxy reconstructions.	
Time series properties	1) irregularity	biases/ variance increase in time series analyses and reconstructions	Adapted time series estimators (Rehfeld & Kurths, 2014)	
	2) age uncertainty	increasing uncertainty & variance	Quantification of uncertainties by using time series ensembles from age modeling (e.g., Rehfeld & Kurths, 2014); variability / spectral estimates are less sensitive to time uncertainty (Rhines & Huybers, 2011).	

4 **Cross-pollination potential**

Age-uncertainty can be evaluated in time series analyses using Monte-Carlo techniques (Fig. 4A, Rehfeld & Kurths, 2014). Some age-modeling approaches allow for the combination of layer counting and point-based ages. This, along with cross-dating as for tree-ring chronologies, could reduce age uncertainties. When the fundamental processes affecting the proxy signal are known, a **proxy error model** can be used to quantify the uncertainties if the essential parameters can be estimated. As an example, Fig. 4B shows a synthetic marine core resembling tropical MD03-2707, with proxy modelling as in Laepple & Huybers, 2013.

5 **References**

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