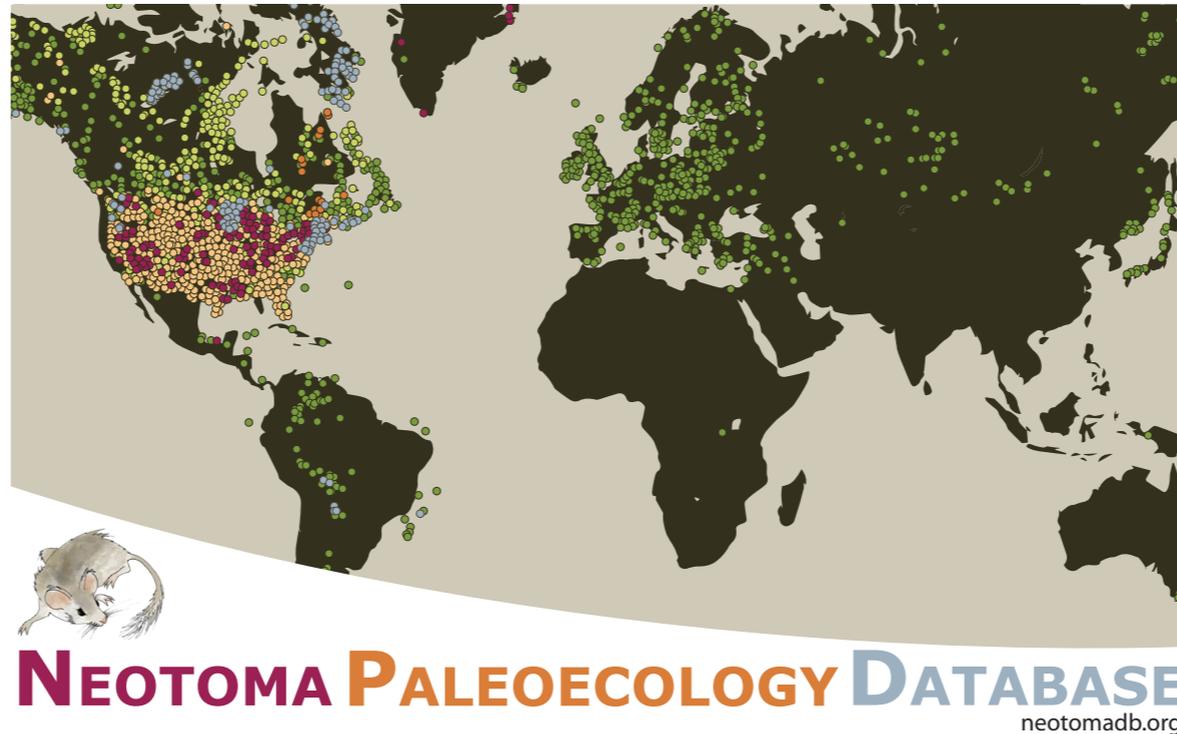


The Neotoma Paleoecology Database: Current Infrastructure, Ongoing Challenges, and Future Directions



Jessica Blois
University of California, Merced

Oh behalf of the Neotoma DB Consortium: John Williams, Eric Grimm, Don Charles, Ed Davis, Simon Goring, Russ Graham, Alison Smith, Mike Anderson, Allan Ashworth, Julio Betancourt, Brian Bills, Bob Booth, Phil Buckland, Brandon Curry, Thomas Giesecke, Sonja Hausmann, Steve Jackson, Claudio Latorre, Doug Miller, Jonathan Nichols, Timshel Purdum, Rob Roth, Hikaru Takahara, and many many others

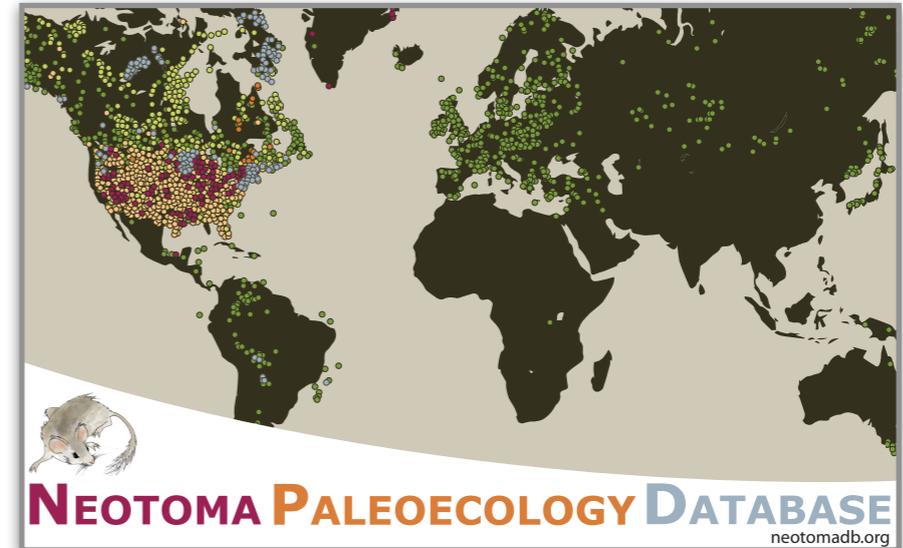
OVERVIEW

- ▶ Typical paleoecological data and data management approaches

- ▶ Neotoma: Current status

- ▶ Neotoma: Future directions

- ▶ Neotoma: Challenges and gaps



neotomadb.org

- ▶ See two resources for additional information:

- ▶ Williams et al. 2018, *Quaternary Research* 89: 156-177. *The Neotoma Paleocology Database, a multiproxy, international, community-curated data resource*
- ▶ Williams et al. whitepaper posted to *Authorea* (<https://goo.gl/ZopKco>). *Cyberinfrastructure in the Paleosciences: Mobilizing Long-Tail Data, Building Distributed Community Infrastructure, Empowering Individual Geoscientists*

KEY CHARACTERISTICS OF PALEOECOLOGICAL DATA

Bad news

Good news

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- ▶ **Useful**: Increasingly assimilated with Earth System Models and conservation biology

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- ▶ Potentially different versions floating around as revisions are made

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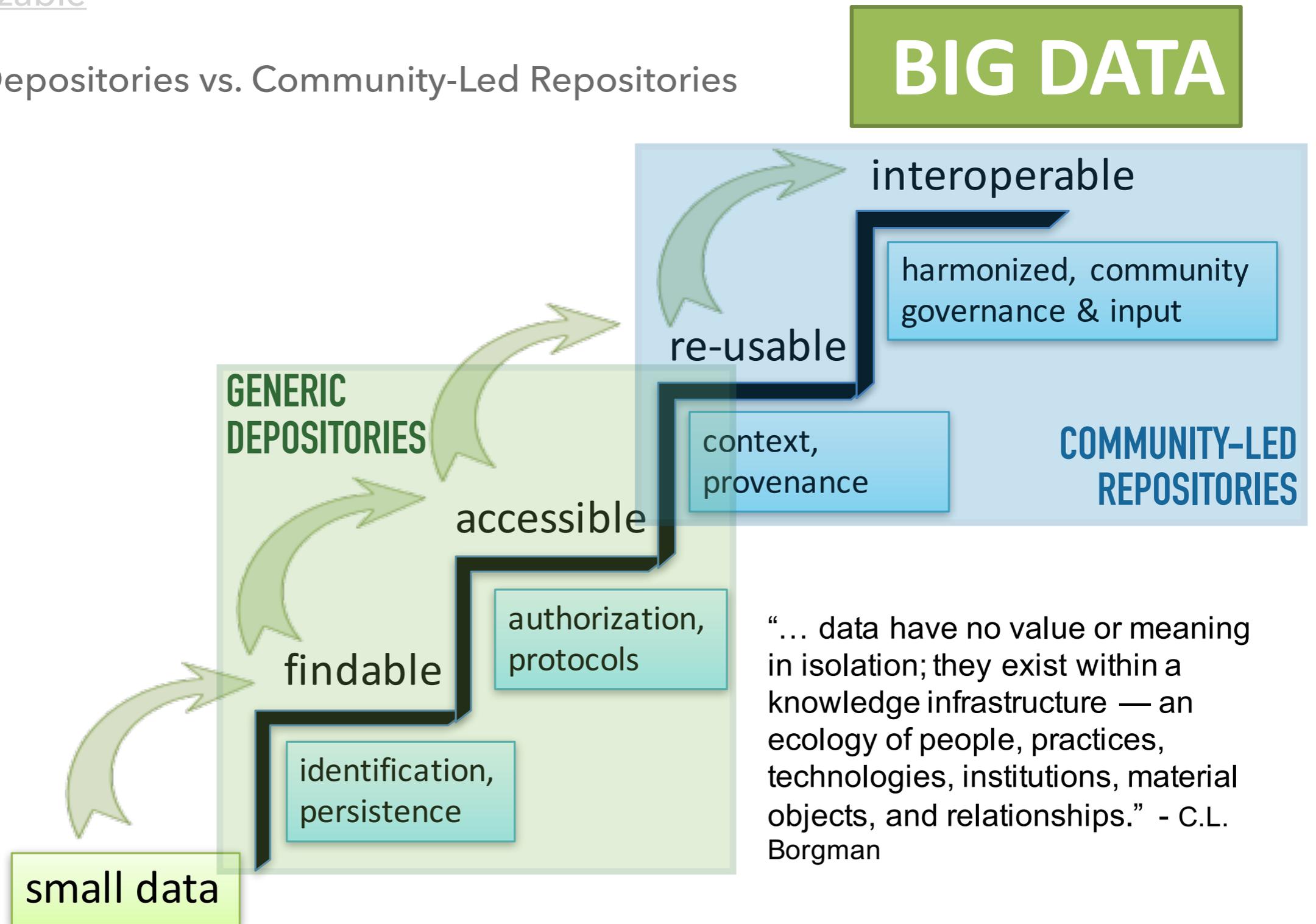
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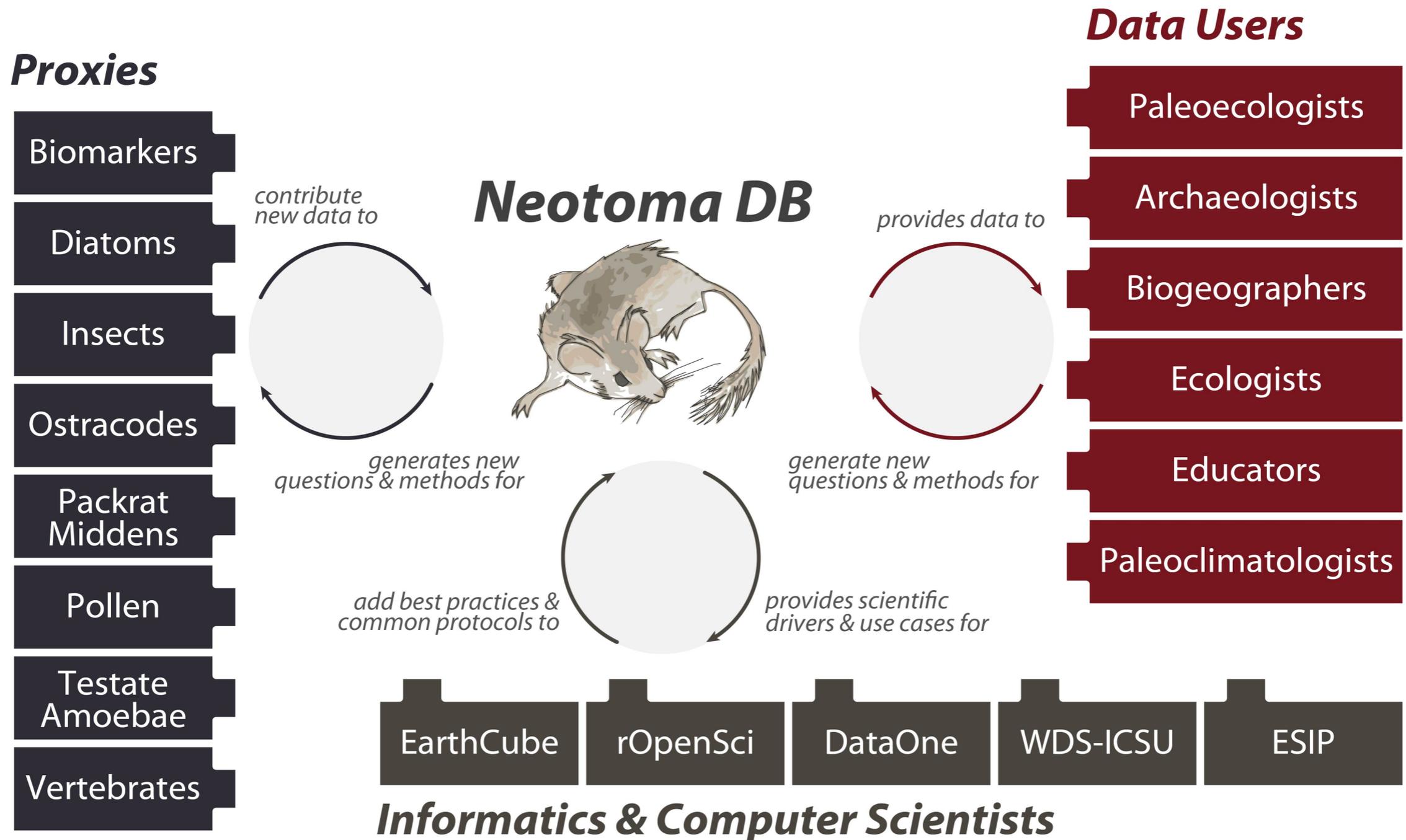
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NEOTOMA PALEOECOLOGY DATABASE: ECOSYSTEM

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- ▶ Community-curated database consortium focused on Pliocene to Quaternary data from around the world



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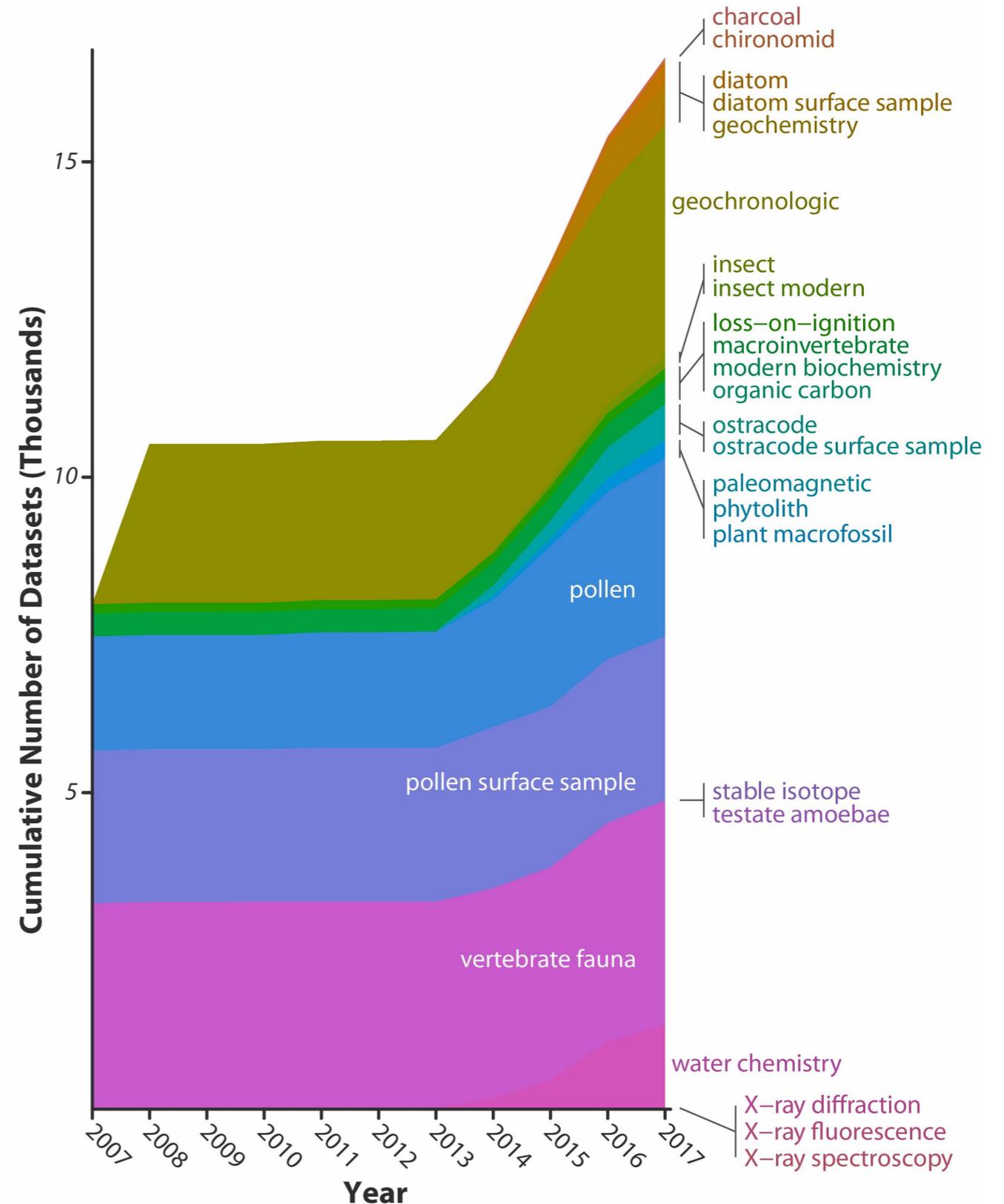
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- ▶ Broad community support and partnerships: with NOAA, PaleobiologyDB, Linked Earth, etc.

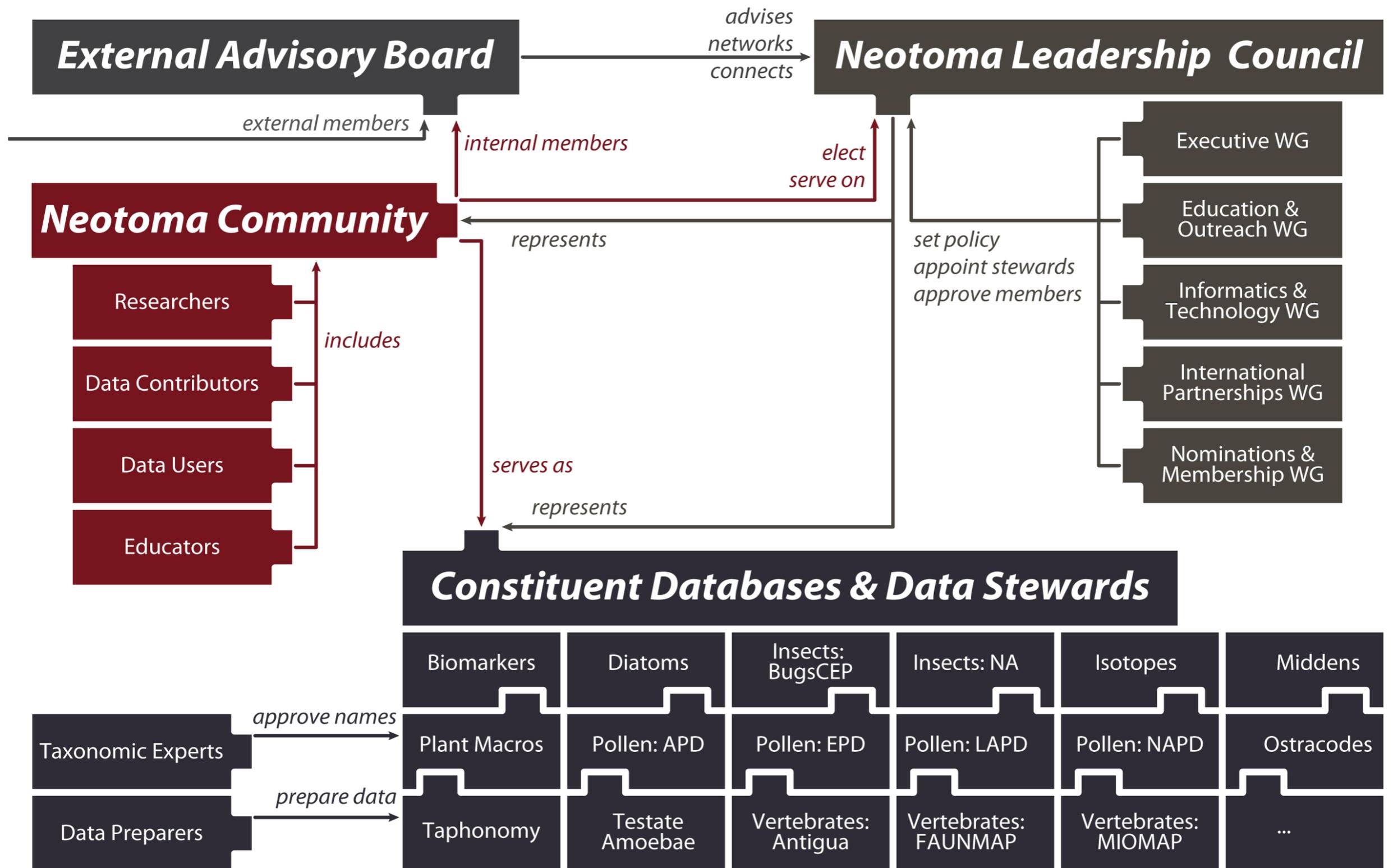
NEOTOMA: DATA MOBILIZATION CAMPAIGNS

- ▶ Pollen: NAPD, EPD, et al.
- ▶ Vertebrates: FAUNMAP2+, MIOMAP, ANTIGUA, MQMD
- ▶ Ostracodes: NANODE
- ▶ Diatoms: Drexel DB
- ▶ Testate Amoebae
- ▶ Packrat Middens
- ▶ Organic Biomarkers

- ▶ Current status: >3.8 million observations, >17,000 datasets, and >9,200 sites.



NEOTOMA: GOVERNANCE



NEOTOMA: NEW DEVELOPMENTS

- ▶ Recently finished or coming down the pipeline
 - ▶ Specimen-level data (BETA RELEASE)
 - ▶ Stable isotopes as new data type (BETA RELEASE)
 - ▶ DOI assignments to datasets (ALMOST DONE)
 - ▶ Ice Age Mapper (UNDERWAY)
 - ▶ Embargoes for unpublished data (STARTED)
 - ▶ Webform uploader to contribute data (A TWINKLE IN THE EYE)
- ▶ **Earth-Life Consortium** (<http://earthlifeconsortium.org/>): seeks to make all *paleobiological* data easily discoverable, accessible, and analyzable, with the larger goal of understanding the interactions between the Earth's biological and geophysical systems across all timescales of the Earth's history.



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 - ▶ Ability to query across different data repositories
- ▶ **Funding sustainability**, particularly related to supporting the geoinformaticists necessary for database maintenance and development

THANKS!

▶ Neotoma Executive Committee

- ▶ Chair: Jack Williams
- ▶ Associate Chair: Jessica Blois
- ▶ Alison Smith
- ▶ Eric Grimm

▶ Neotoma Leadership Council

- ▶ EC +
- ▶ Allan Ashworth, International Working Group Chair, Steward, Insects
- ▶ Suzanne Pilaar Birch, Steward, Isotopes
- ▶ Phil Buckland, Steward, Insects
- ▶ Don Charles, Steward, Diatoms
- ▶ Thomas Giesecke, International Working Group, Steward, European Pollen Database
- ▶ Simon Goring, IT Working Group Chair
- ▶ Claudio Latorre, International Working Group, Steward, Packrat Middens
- ▶ Hikaru Takahara, International Working Group, Steward, Japan Pollen Database

▶ Neotoma database contributors



▶ Funding

- ▶ NSF EAR 1550700
- ▶ NSF ICER 1540977



THE IDEAL!

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Best Practices for Scientific Computing

Greg Wilson , D. A. Aruliah, C. Titus Brown, Neil P. Chue Hong, Matt Davis, Richard T. Guy, Steven H. D. Haddock, Kathryn D. Huff, Ian M. Mitchell, Mark D. Plumbley, Ben Waugh, Ethan P. White, Paul Wilson

Published: January 7, 2014 • <https://doi.org/10.1371/journal.pbio.1001745>



Data Carpentry develops and teaches workshops on the fundamental data skills needed to conduct research. Our mission is to provide researchers high-quality, domain-specific training covering the full lifecycle of data-driven research. Data Carpentry is a sibling organization of Software Carpentry. Where Software Carpentry teaches best practices in software development, our focus is on the introductory computational skills needed for data management and analysis in all domains of research. Our lessons are domain specific, from life and physical sciences to social science and build on the existing knowledge of learners to enable them to quickly apply skills learned to their own research. Our initial target audience is learners who have little to no prior computational experience. We create a friendly environment for learning to empower researchers and enable data driven discovery.

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Our path to better science in less time using open data science tools

Julia S. Stewart Lowndes , Benjamin D. Best, Courtney Scarborough, Jamie C. Afflerbach, Melanie R. Frazier, Casey C. O'Hara, Ning Jiang & Benjamin S. Halpern

Nature Ecology & Evolution **1**,

Article number: 0160 (2017)

doi:10.1038/s41559-017-0160

Received: 25 November 2016

Accepted: 07 April 2017

Published online: 23 May 2017

<http://www.datacarpentry.org/>

NEOTOMA: ACCESSING DATA

▶ Finding, Exploring, Downloading Data

▶ Explorer

▶ <https://apps.neotomadb.org/Explorer/>

▶ APIs

▶ <https://api.neotomadb.org/>

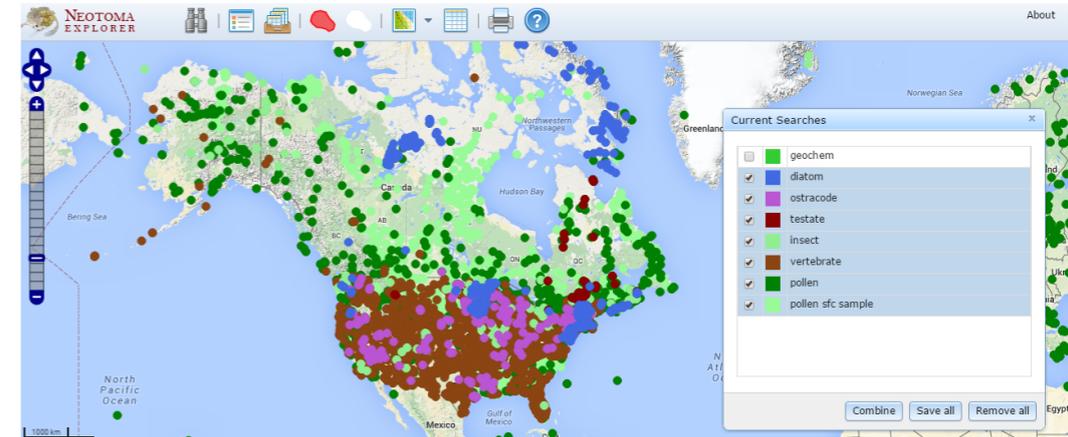
▶ R

▶ <https://cran.r-project.org/web/packages/neotoma/index.html>

▶ <https://github.com/ropensci/neotoma>

▶ DOIs & Landing Pages (coming soon)

▶ <http://data.neotomadb.org/datasets/1001/>



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More info:

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NEOTOMA: EDUCATION

- ▶ Teaching Resources

- ▶ SERC Carleton

- ▶ <http://serc.carleton.edu/neotoma/activities.html>

- ▶ Neotoma Webpage

- ▶ https://www.neotomadb.org/education/category/higher_ed/