Leveraging large datasets to discover protistan diversity across scales



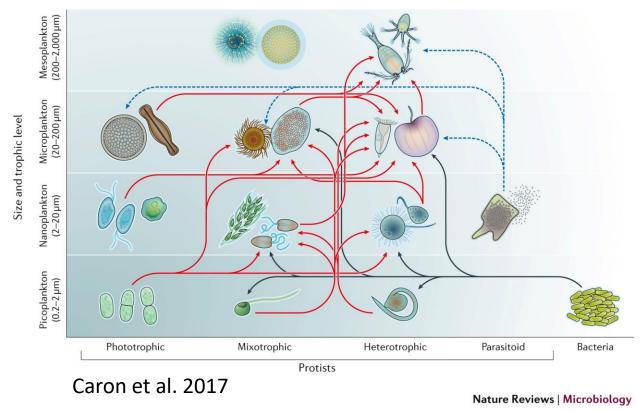
<u>Arianna I. Krinos</u>



Margaret Mars Brisbin, Natalie Cohen, Sarah Hu, Weixuan Li, Sara Shapiro, Stephanie Dutkiewicz, Frederik Schulz, Michael Follows, and Harriet Alexander

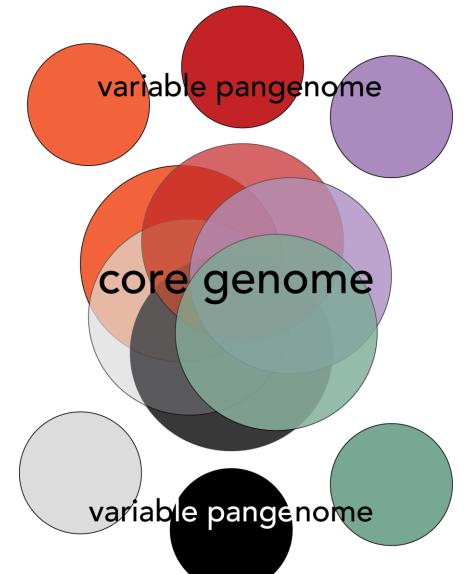
Protists are essential players in global biogeochemical cycles

- Single-celled microbial eukaryotes are an essential link in the food web between numerically-abundant bacteria & higher trophic levels
- Phytoplankton are photosynthesizers that contribute to primary production - they fill the role that plants fill on land and on coasts in the open ocean



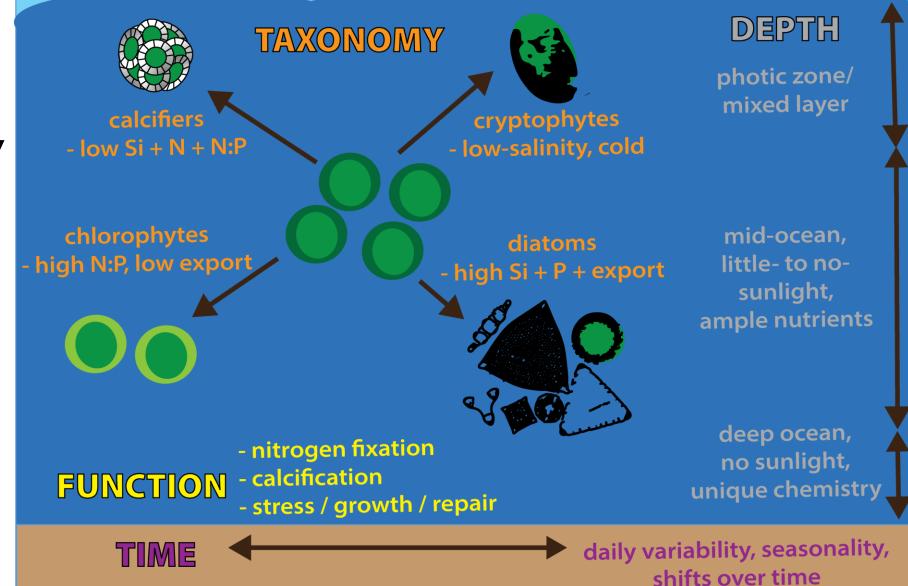
But their distribution, ecology, and genetics is <u>complicated</u>

- "Biological species" is complicated by the fact that a growing number of taxa have a variable "pangenome"
- Having many shared genes means that you're all part of the same group; having a set of genes you don't share defines you as a **strain**

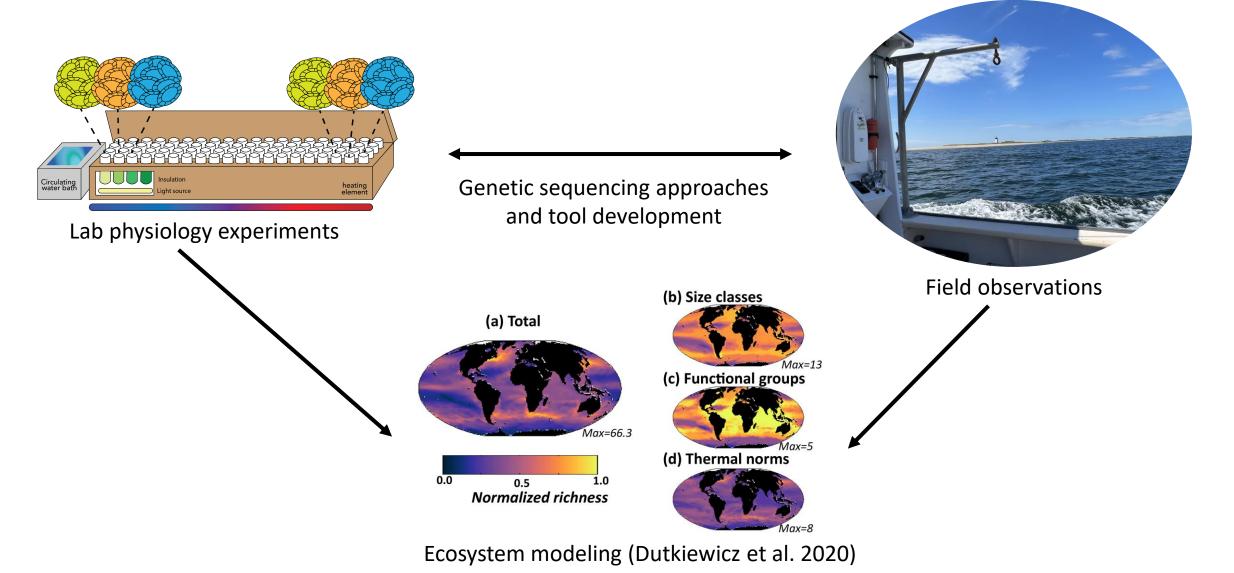


In situ, phytoplankton genetic diversity is one part of many axes of variability

Axes of variability in marine phytoplankton



To study protists, we'll need cross-scale tools.



And a lot of data **⊖ BD Accuri™ C6 Plus** Flow Cytometer Population-level Community-level Large sequencing datasets Daily measurements of cell geometry 3.0 ٦**١**. 11. rate (1/day) 2.0 rate (1/day) 2.0 and culture density via flow owth 1.0 Growth 1.0 cytometry 10 15 20 25 30 35 0 5 10 15 20 25 30 35 0 5 Temperature (°C) Temperature (°C) 3.0 **III**. IV. rate (1/day) 2.0 owth 1.0 10 M

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Temperature (°C)

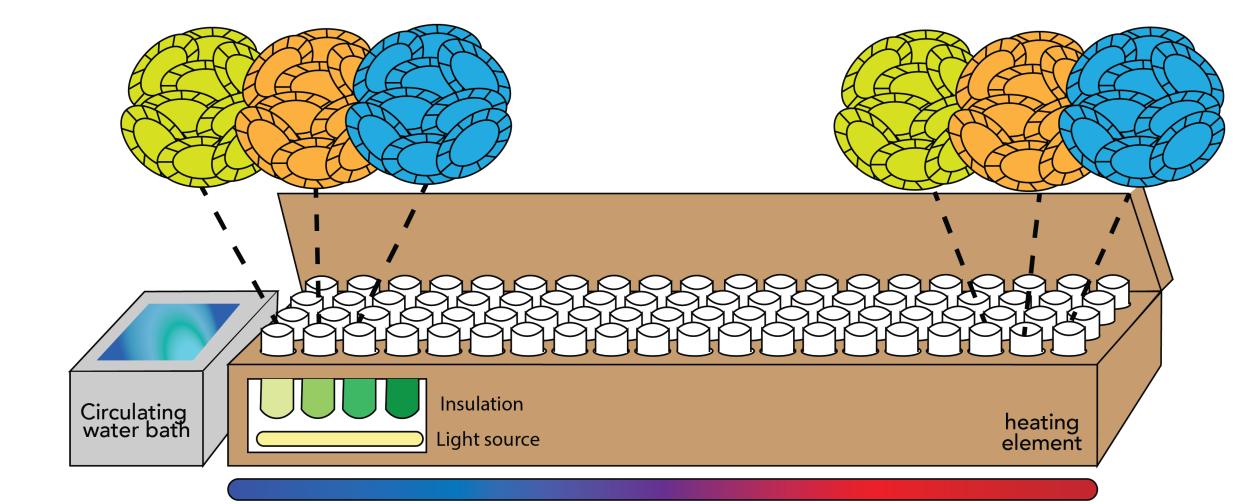
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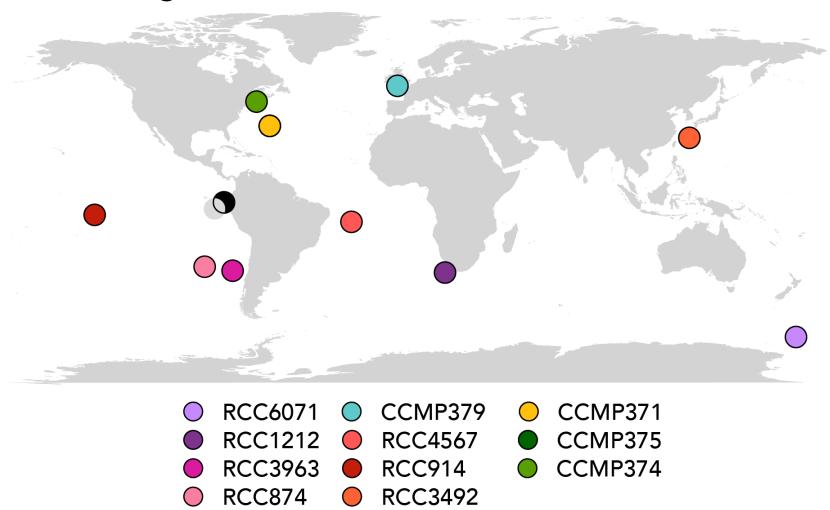
Temperature (°C)

Ecosystem model outputs

How do strains of *Emiliania huxleyi* acclimate to local environmental temperature?

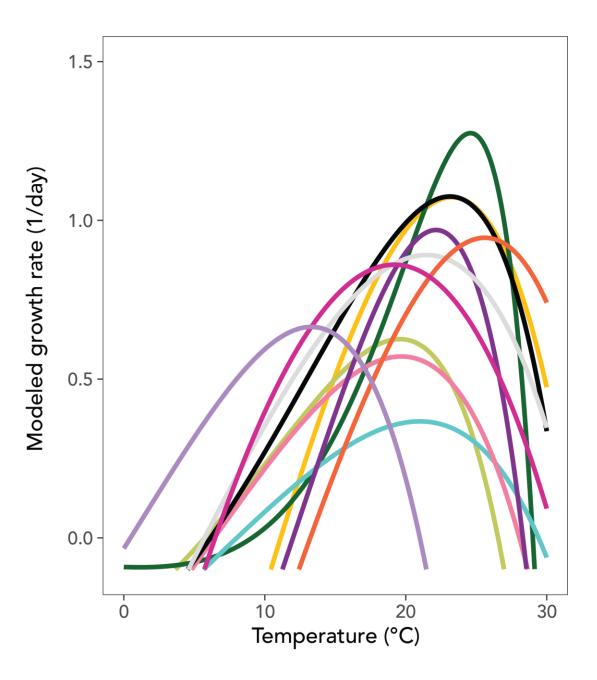


Strains of *Emiliania huxleyi* isolated from across the global ocean

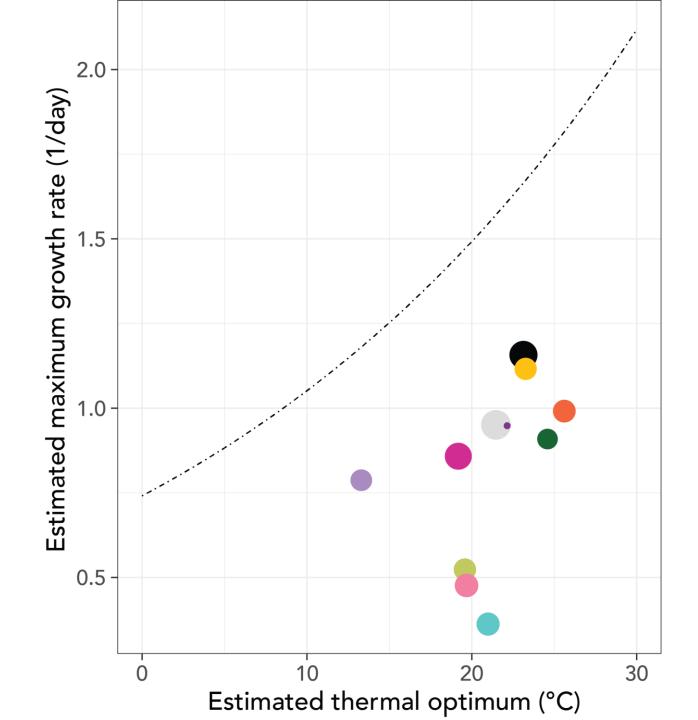


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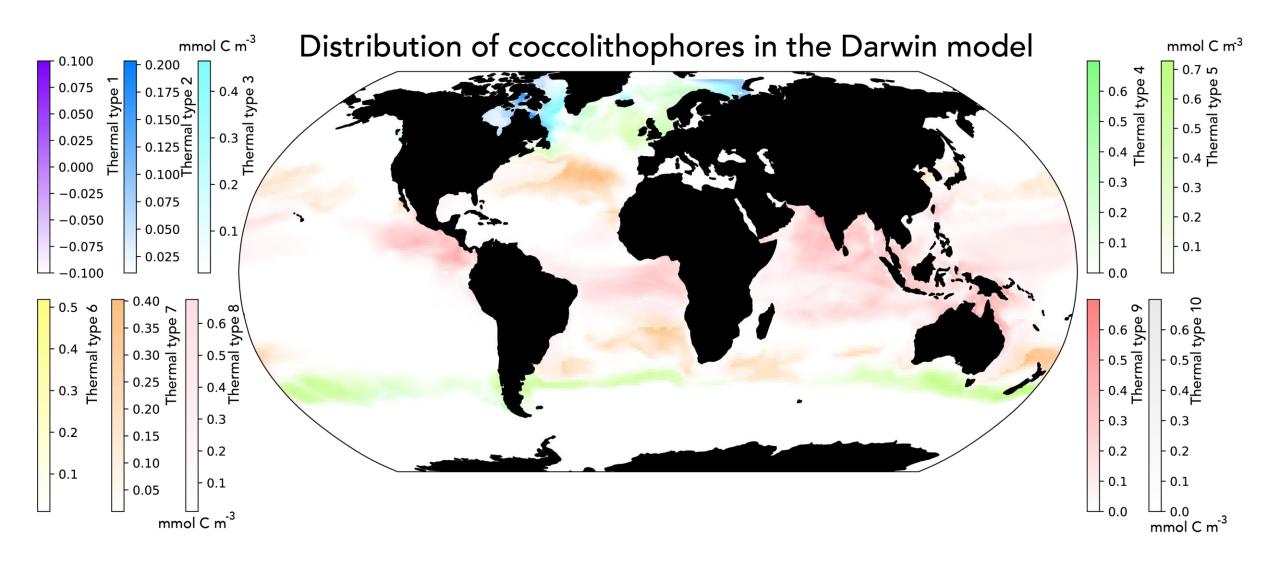
Thermal response curves vary in shape and according to original isolation latitude



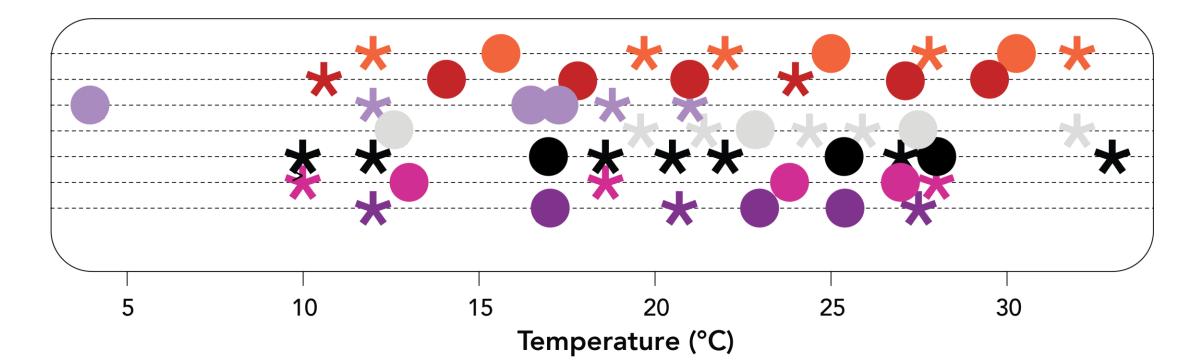
And differ from the growth rate – temperature scaling we expect from theory



We can encode these insights into ecosystem models that can predict coccolithophore distribution

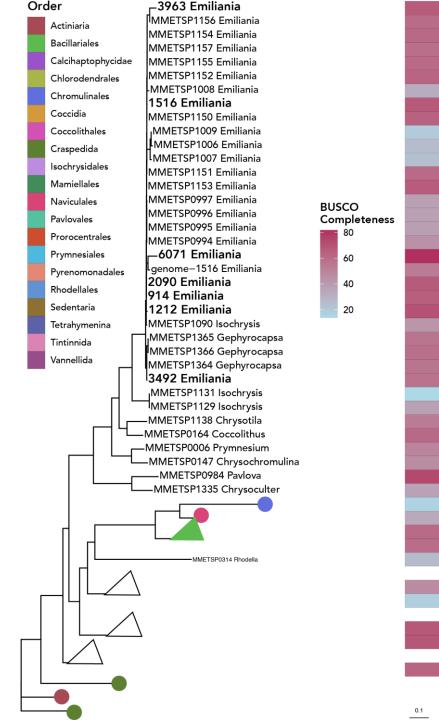


Sequencing transcriptomes can provide insight into the <u>mechanism</u> of thermal response

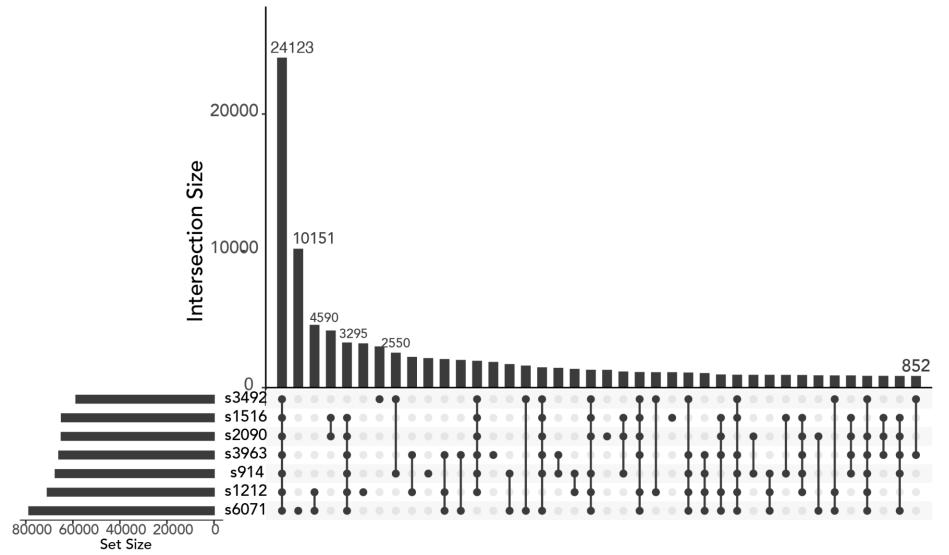


<u>Total number of raw read sequences</u>: >1,300,000,000 <u>Number of predicted transcripts</u>: 981,810; 1,161,365

Strains of *Emiliania huxleyi* with different expressed traits have similar core genes



But vastly different pools of overlapping gene content

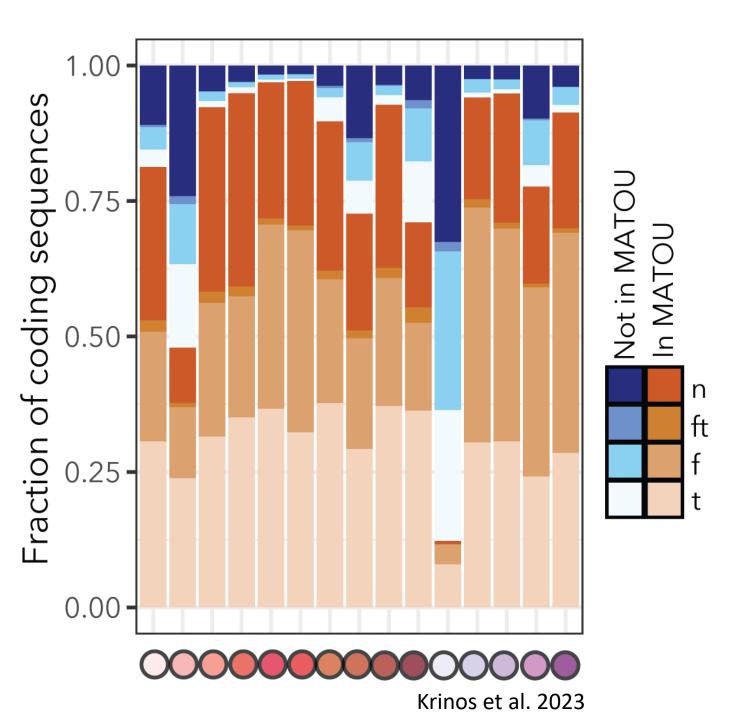


Tools are seldom built for eukaryotes

- To process metatranscriptomes, I built an open-source pipeline called <u>eukrhythmic</u>
- For taxonomic annotation of assembled sequences from metagenomeassembled genomes and metatranscriptomes, I built a Python package called <u>EUKulele</u>



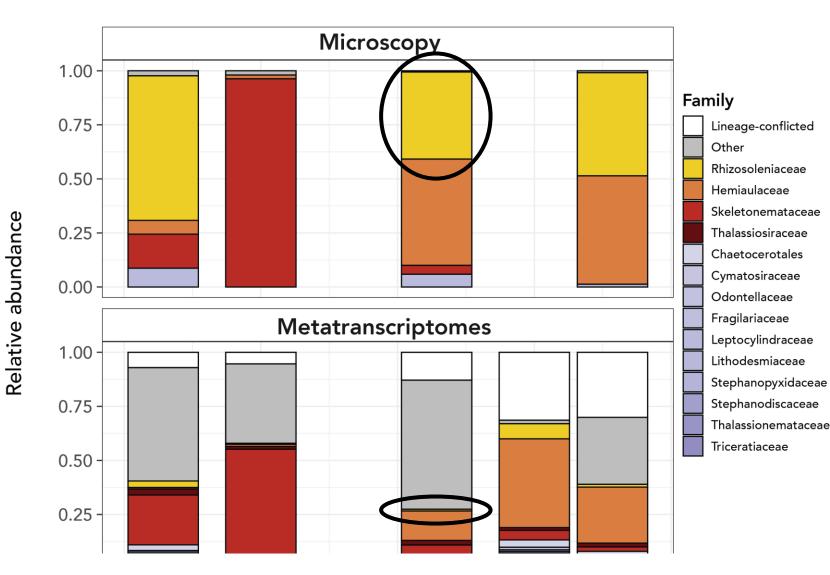
Reassembling global ocean samples using the eukrhythmic pipeline expands gene recovery in environmental datasets



Just these two samples from the Tara Oceans dataset require up-to-date computing resources

- Total size of raw sequencing reads: 94Gb
- CPU hours required for assembly using just one of the four assembly algorithms and one of the samples: 20 hours over 16 cores using 500GB of RAM
- Total number of predicted genes from one resulting assembly: 308,532
- Total number of predicted genes from all assemblies on previous plot: 15,241,002

Comparing traditional multiomic annotations to microscopic counts



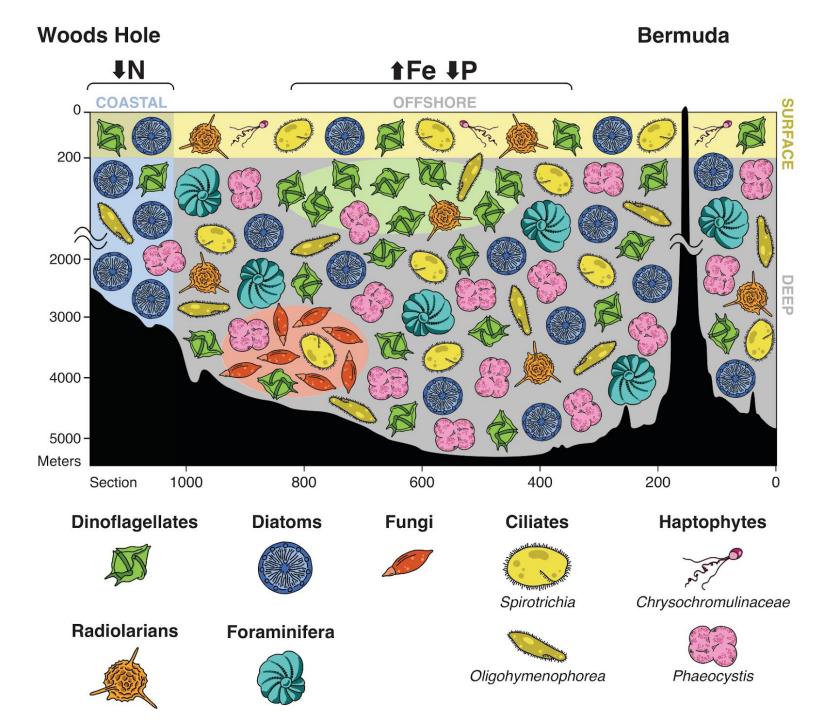
How can we develop new algorithms to process taxonomic annotations of protists in a high-throughput way?

database meta-omic sequences sequences tax-aliquots: related by sequence identifying alignment **DIAMOND** DeepClust to efficiently & permissively cluster proteins taxonomic identity of eukaryotic No database cluster: Clusters with Clusters with communities via protein annotation only one sequence: multiple sequences: is not possible; use percentage count overlapping return size of identity of alignment kAAmers in each homology-based clustering to assign taxonomy protein contained cluster & evaluate in homology group read recruitment Distance matrix **Hierarchical** related by of kAAmer overlap clustering with userbetween sequences subsequence defined distance-based in cluster patterns cutoff or cluster coherence Label assignment is based on (1) homology, (2) hierarchical clustering, and (3) taxonomic cluster coherence

High-performance computing enables this refinement

- Just in our database, we have 15,514,482 sequences that contain 194,481 unique 4-base amino acid kAAmers and 625,624,476 unique 7-base amino acid kAAmers
- Making comparisons between the training data <u>alone</u> requires **2.41e14** computations

High-performance computing enables important ecosystem insights about where protists are found and what ecosystem services they're providing



Thank you!

- Dr. Harriet Alexander, WHOI
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- Sara Shapiro, WHOI
- Weixuan Li, MIT
- Quinn Perian, MIT



