Towards a modern analysis of omics data of the Ocean

Mission Microbiome: CEODOS and AtlantEco expeditions

As genomics sciences meet data sciences and modeling

Olivier Jaillon Genoscope – CEA

Valparaiso. May 16th 2023.

Genoscope overview

• French National Sequencing Center lead by Patrick Wincker, created in 1997 and part of the CEA since 2007.

• Provide high-throughput sequencing data to the Academic community, and carry out in-house genomic projects

• Focus on biodiversity : *de novo* sequencing and metagenomic projects (TaraOceans)

Human

Paramecium Tetraurelia

' Vitis vinifera (grape vine)



Musa acuminata (banana)



http://www.genoscope.cns.fr



Quercus robur (oak)









Drosophila melanogaster



Arabidopsis thaliana



Tetraodon nigroviridis

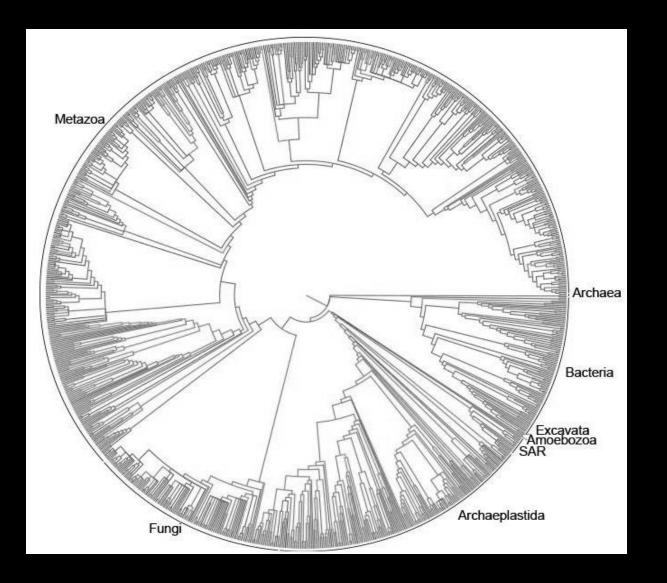




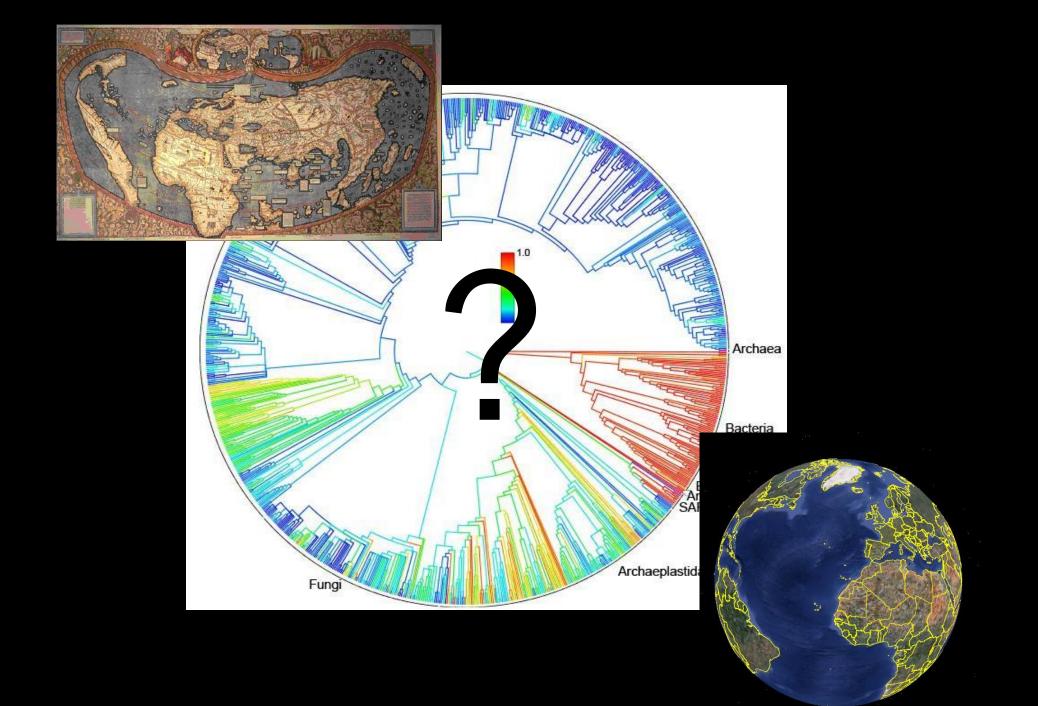


5 centuries

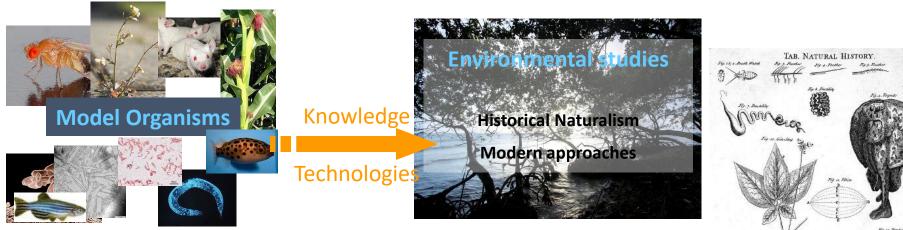




Hinchliff et al.. PNAS 2015



Biology may lead to 2 strategies :

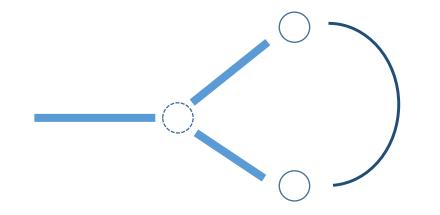


- Different models for different goals
 - •Evo/devo
 - •Genetics
 - •Genomics
 - •...
- Precise knowledge but restricted on examples.

- •Catalog of organisms
- •Biotic & Abiotic Connections
- •Recent interest motivated by
 - global warming
 - •Impact of human overpopulation
 - •Screening of chemicals, enzymes of
 - interest



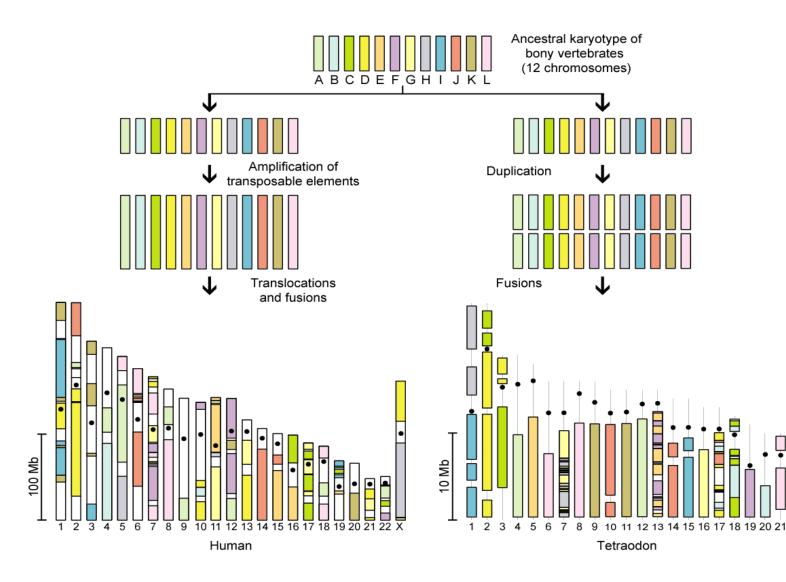
Sequence comparisons on Model Organisms



Comparing 2 sequences is always tracking back from last common ancestor.

Nothing in Biology Makes Sense Except in the Light of Evolution. Dhobzanski 1973.

GENOME SEQUENCES REVEAL MAJOR EVOLUTIONARY EVENTS





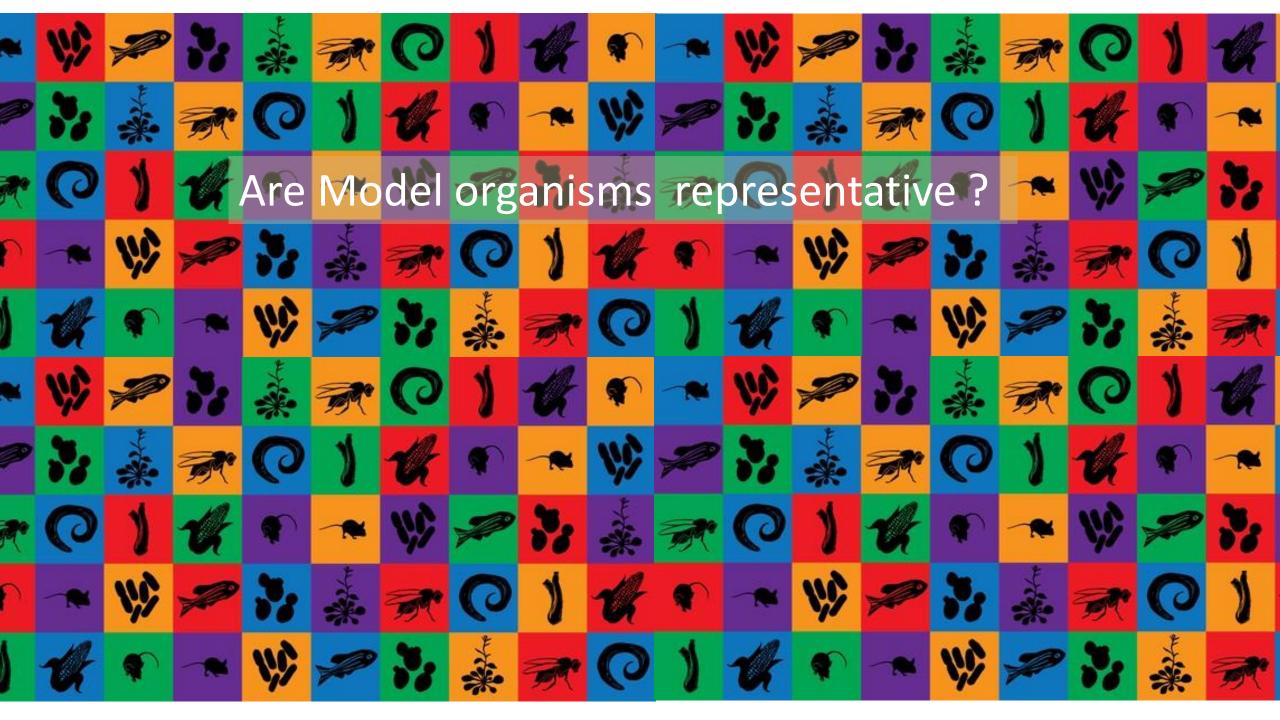
Unknown Genic diversity in Model Organisms:

Taxonomically Restricted Genes

duencea genome		
а	Species	Percentage of TRGs in the genome
es m		
S.	Saccharomyces cerevisiae	19
is	Caenorhabditis elegans	
/n	Drosophila melanogaster	²
te	Homo sapiens	7
of	The second s	
ie	Schizosaccharomyces pombe Mus musculus	
es	Ciona intestinalis	20
er,		
а	Anopheles gambiae	22.3
а		
es	Caenorhabditis briggsae	
st	Pattus parvasious	
si fy	Rattus norvegicus Pan troglodytes (chimpanzee)	n/d
re	Strongylocentrotus purpuratus	n/d
a-	Aedes aegypti	27.1
ed	Nematostella vectensis	n/d
By		20
nd	Brugia malayi D. simulans	
ed		20.6
)e	D. sechellia	23.4
ul	D. yakuba	19.6
ıy	D. erecta	24.1
y)	D. ananassae	
th	D. pseudoobscura	22.0
at	D. persimilis	30.3 26.9
ed	D. willistoni	
e-	D. virilis	22.1
a er	D. mojavensis	20.7
01	D. grimshawi	20.8
	Tribolium castaneum	
-	Ornithorhynchus anatinus	18
30	Pristionchus pacificus	30
ne of	Bos taurus	
te		5.5
os	E Fack black services and	
TP	Each block represents 2% of the predicted gene models	
off	Gene models without homology (putative TRGs)	
on on	Species where the number of TRGs was not calculated	
028	TRENDS in Constion	

quenced genome

TRENDS in Genetics Khalturin et al.



<<Scientists know a great deal about model organisms as diverse as Arabidopsis thaliana and the zebrafish, but there is still a lot to learn about their life and biology in the wild, including the way that their behaviors are shaped by habitat, their predatorprey relationships, their ability to adapt to different environmental conditions, and the

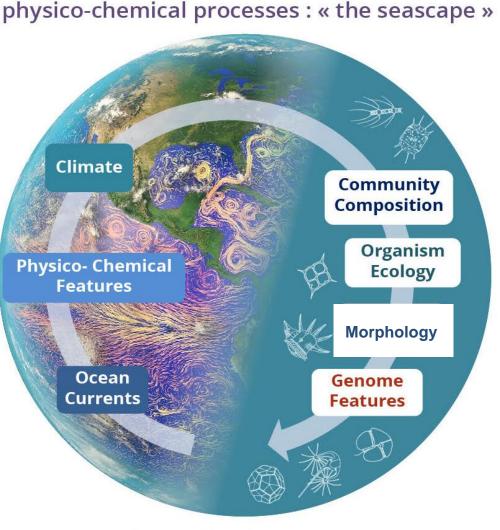
genetic basis for this adaptability.>>

Alfred & Baldwin elife 2015. Illustration : claudiastocker.com



How to scale up from model organisms to a more global vision ?

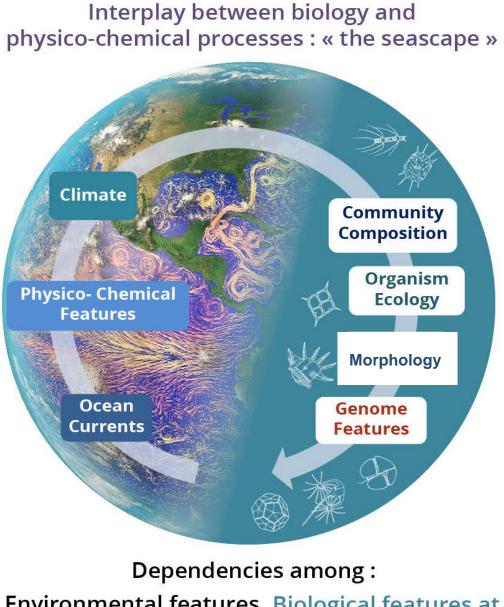




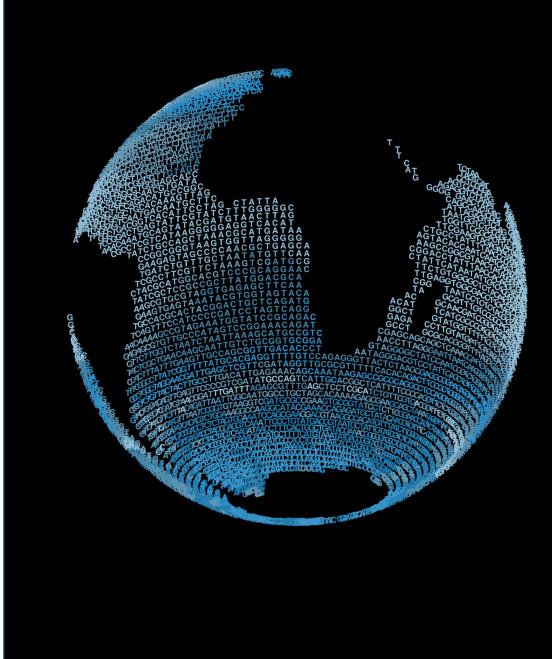
Interplay between biology and physico-chemical processes : « the seascape »

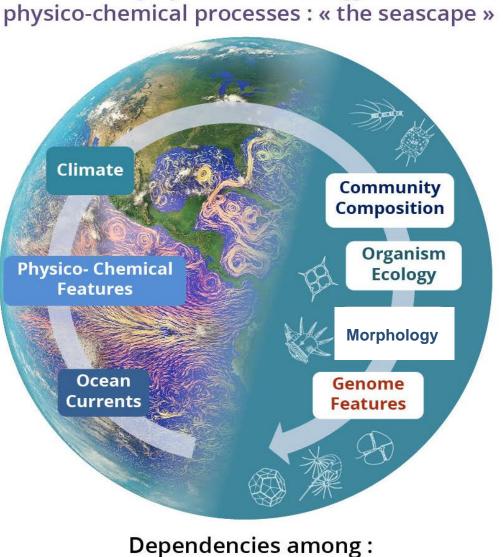
Dependencies among :

Environmental featuresBiological features atat different scalesincremental scales



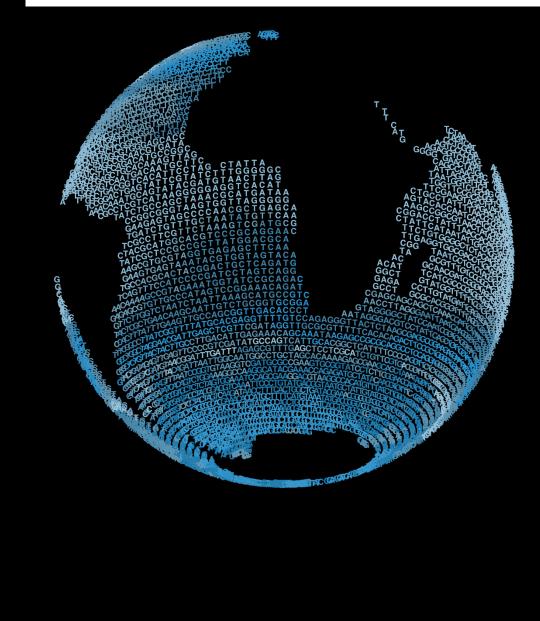
Environmental features Biological features at at different scales incremental scales





Interplay between biology and

Good data, bad data and ugly data Nat Microbiol 4, 209 (201



Environmental features Biological features at at different scales

incremental scales

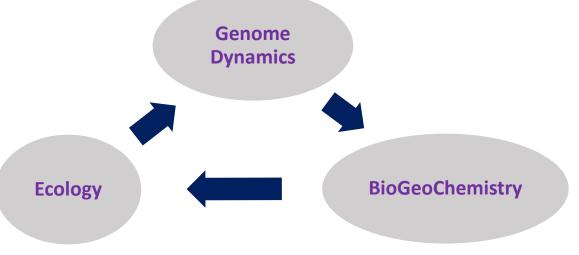
Seascape Genomics & Biogeography are linked

From Ramette A. & Tiedje J. M. Microbial Ecology 2006 :

In a broader sense, this discipline examines variation of microbial features (e.g., genetic, phenotypic, physiological) at different spatial scales, between distantly located sampling sites or along large environmental gradients.

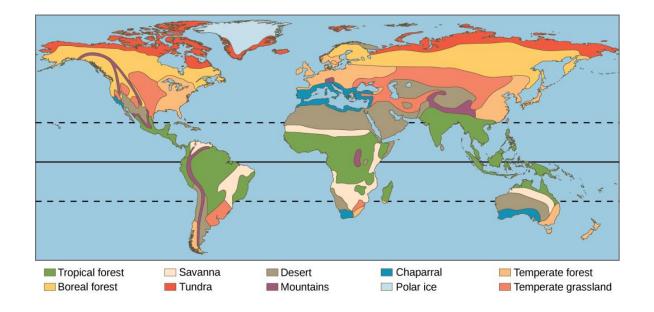
Its scope also encompasses the understanding of the processes generating and maintaining those distribution patterns.

The ultimate goals are to propose and evaluate theories regarding the creation and evolution of such diversity patterns in the environment.



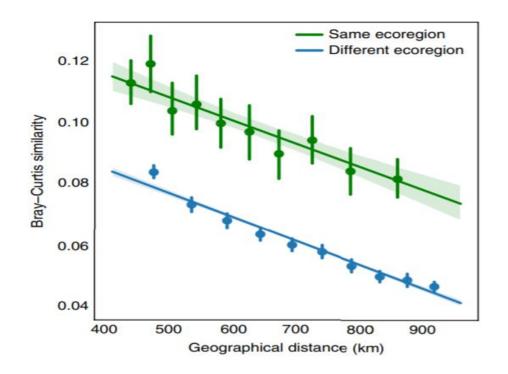
Terrestrial Biomes

Eight major terrestrial biomes defined by temperature and precipitation

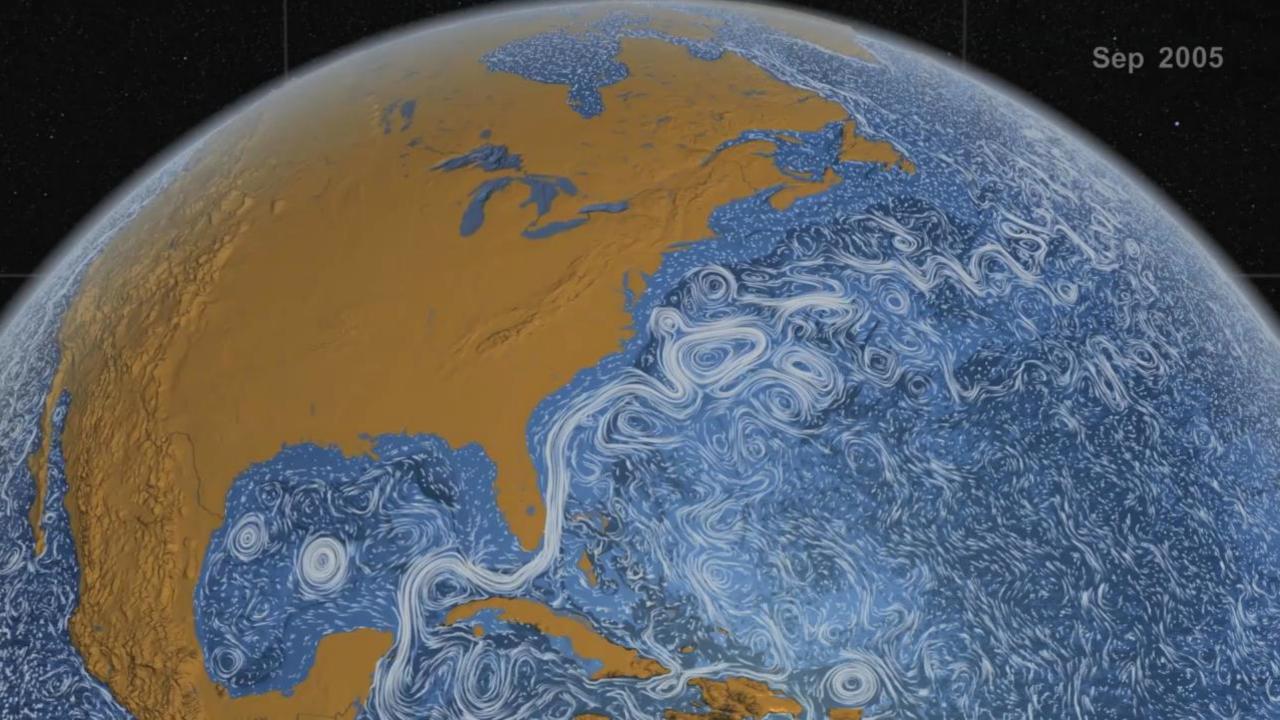


Credit: "Climate influence on terrestrial biome" by Navarras is in the Public Domain

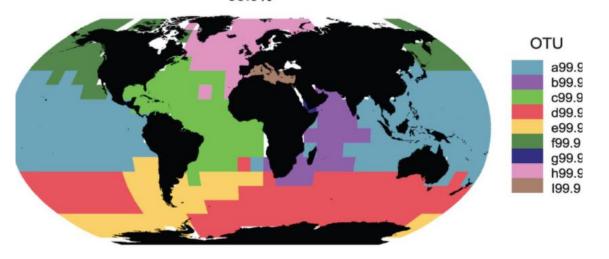
β-diversity is proportional to Geographical distance



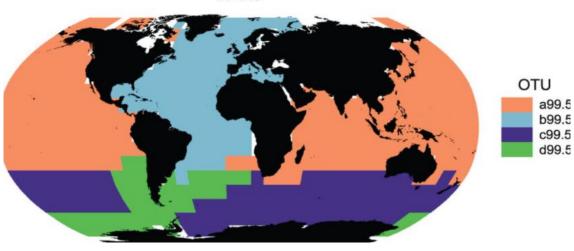
Smith et al. Nature Ecology & Evolution 2018



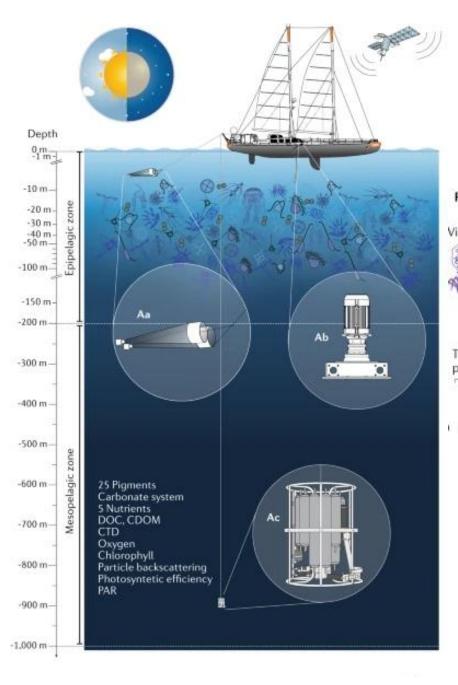
Combination of neutral evolution of DNA with Ocean Circulation

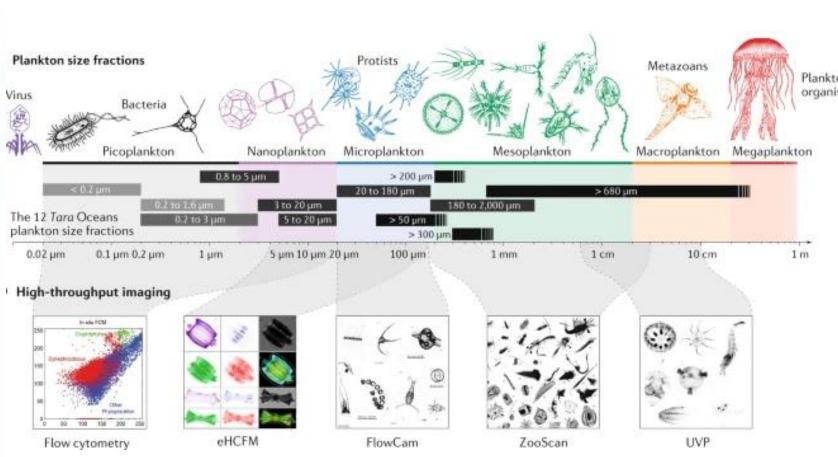


99.5%



Ferdi L. Hellweger Erik van Sebille, Neil D. Fredrick Science 2014

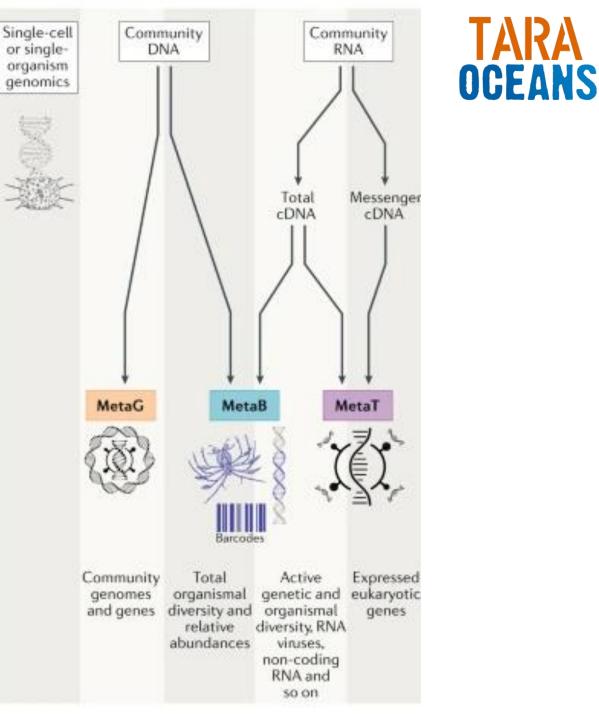




TARA Oceans

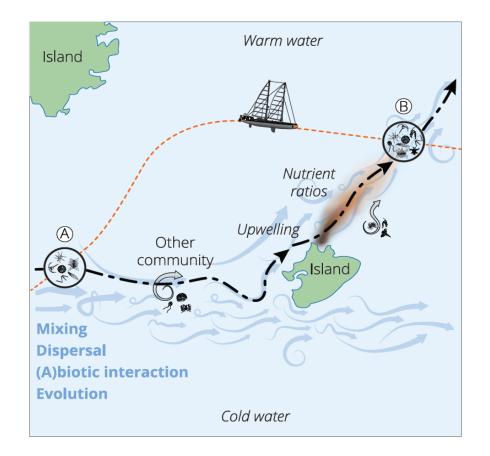
Differents types of "omics" data : different goals

Sunagawa, S., *et a*l. Tara Oceans: towards global ocean ecosystems biology. Nat Rev Microbiol 18, 428–445 (2020).



SEASCAPE GENOMICS WITH TARA OCEANS

Oceanographic variations and genomic variations are closely linked

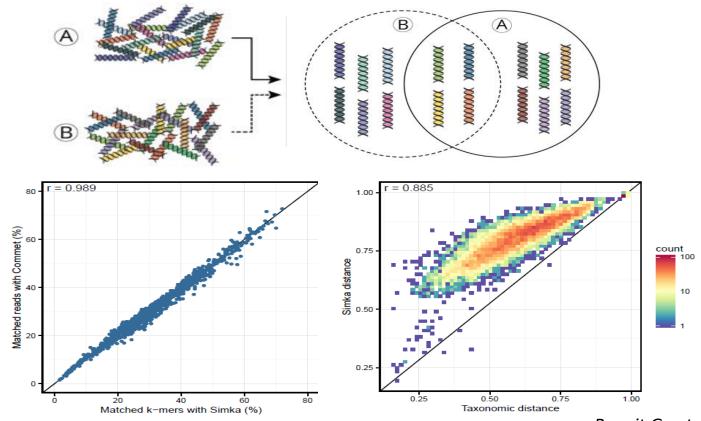




Genomic Variations (β diversity)

Rationale: Leverage Metagenomes in a Blind approach to reveal a structure of global ocean partitioning

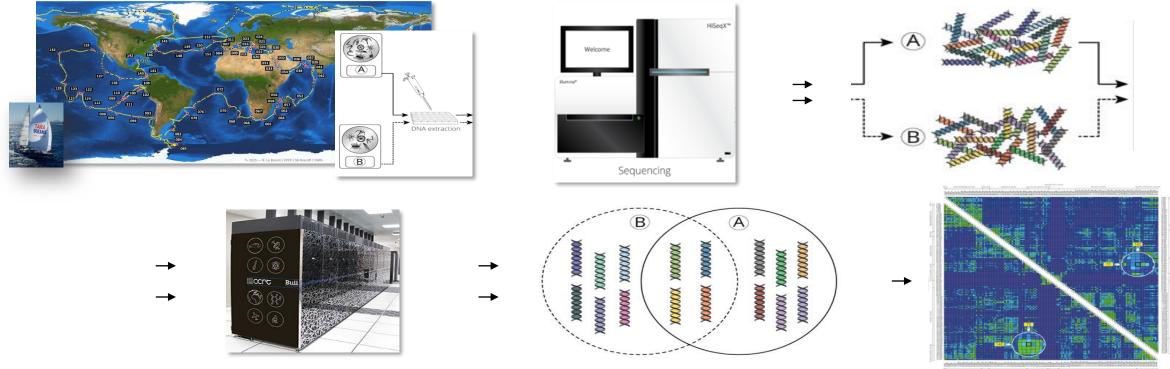
- Use DNA kmers from metagenomes as biological tracer of organisms and to approximate betadiversity
- Compute beta-diversity between metagenomic samples using DNA kmers. etag : Simka tool (Benoit, G. Et al. PeerJ Computer Science, 2016; Benoit, G. ET al. Bioinformatics, 2020)



Benoit, G. et al. PeerJ Computer Science, 2016

Deciphering a global Biogeography of plankton without reference

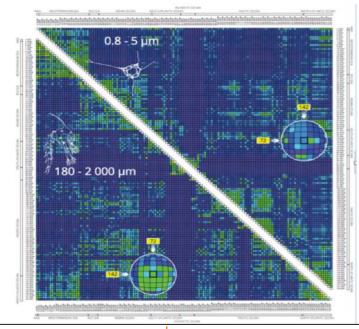
- Consider each set of metagenomic reads as representative of local plankton community.
- Compute the number of very similar reads between all pairs of samples => proxy for a "metagenomic distance" between communities



Simka tool. Benoit et al. Peer. J. Comput. Sci. 2016

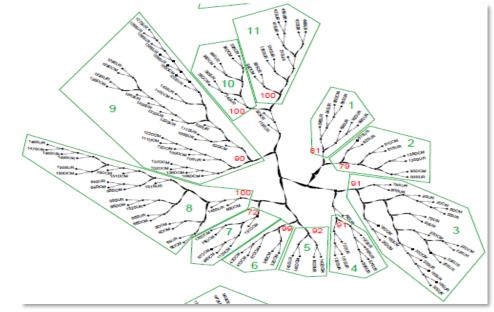
Deciphering a global Biogeography of plankton without reference

Simka Matrix



UPGMA clustering

Relational Trees of Samples



RGB Coloring of samples

PCA •

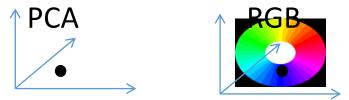


Samples colored in RGB values using 3 main axis of Principal Component analysis => differences in colors reflect betadiversity among samples.

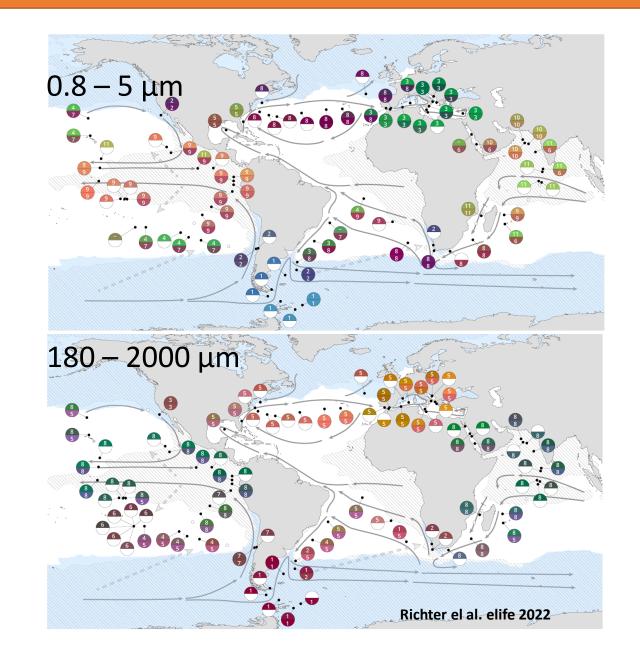
=> Display in a geographical map

A global genomic biogeography of plankton community

 Samples colored in RGB values using PCA 3 main axis component values => differences in colors reflect beta-diversity among samples

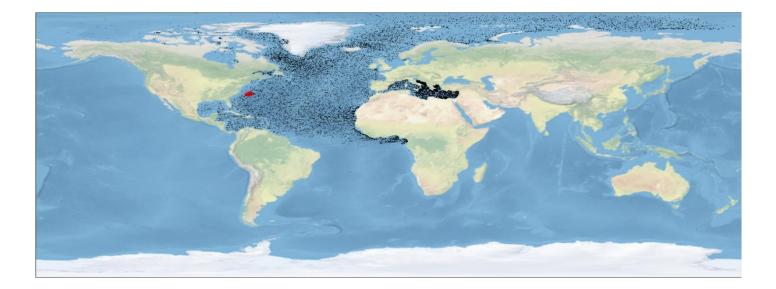


- Spatial organization in *genomic provinces*.
- Different scale of Spatial Distribution among organism size fractions.
 - Smaller size fraction, smaller provinces
 - Reflect different ecology, behavior (reproduction time).
 - Plasticity of biotic interactions ??

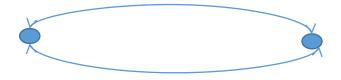


Lagrangian distances

- Could we correlate dissimilarity of community composition between 2 stations with transport time ?
- Rationale : Compute Minimum Transport Time (Tmin) between pairs of samples.

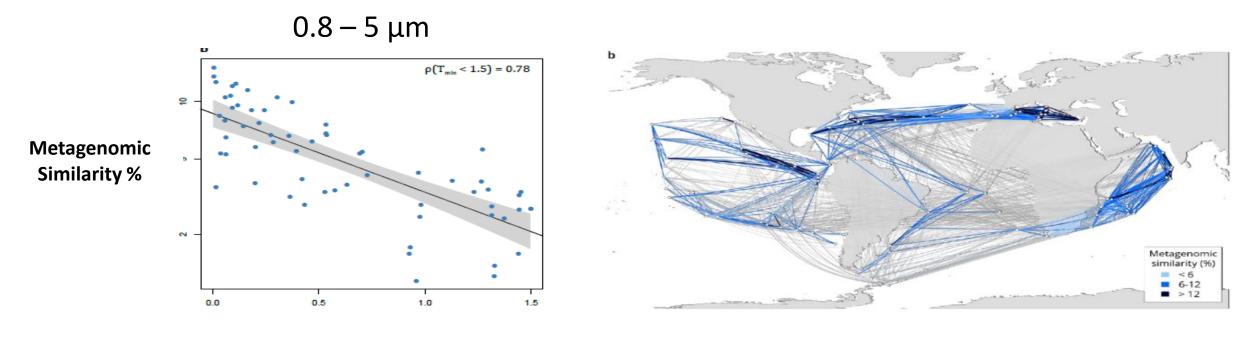


Lagrangian distance (Tmin)



Metagenomic distance (Simka)

Influence of Transport Time : zoom in North Atlantic

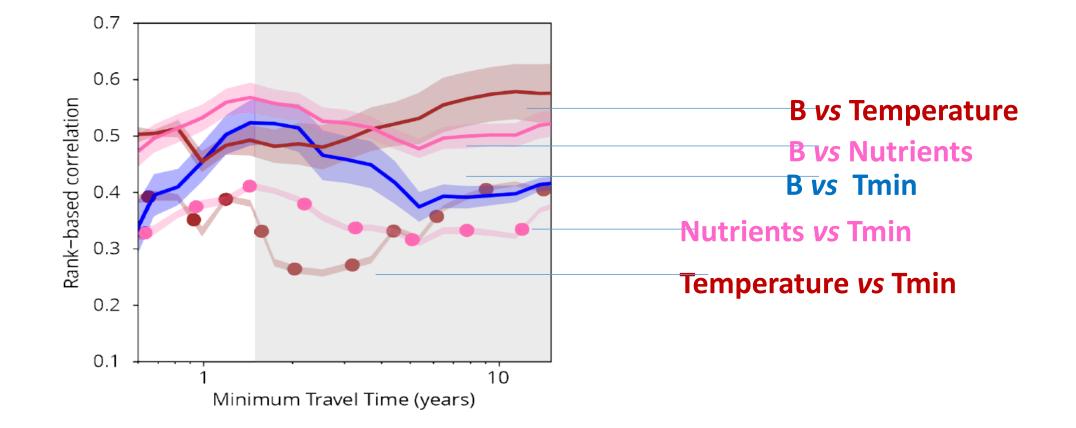


Travel Time (years)

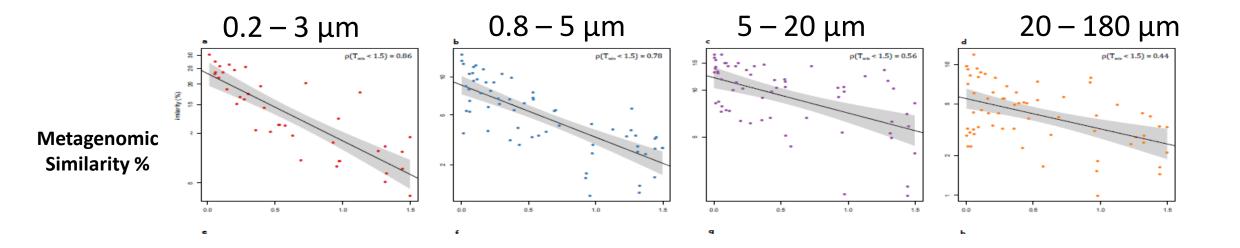
Metagenomic connectivity for pairs of stations < 1,5 year s

Richter el al. elife 2022

Measurable Influences of environmental drivers along plankton travel time

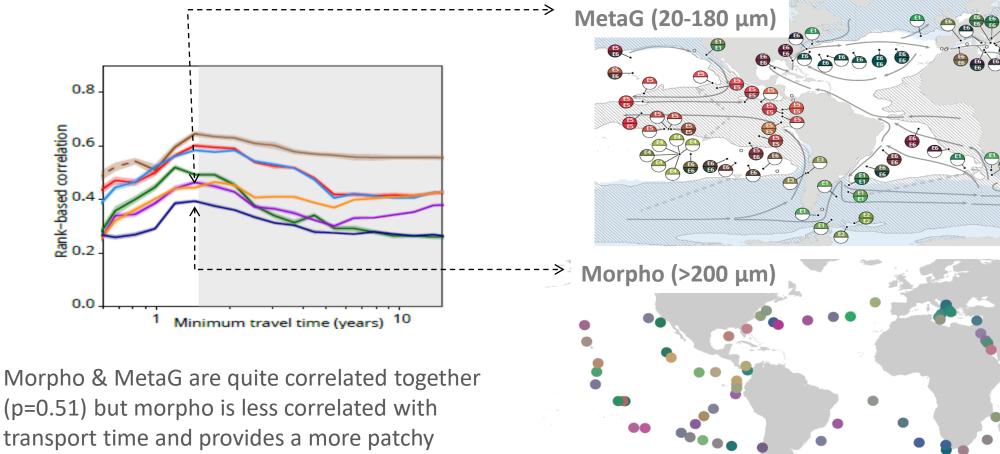


Influence of Transport Time : organism size dependent

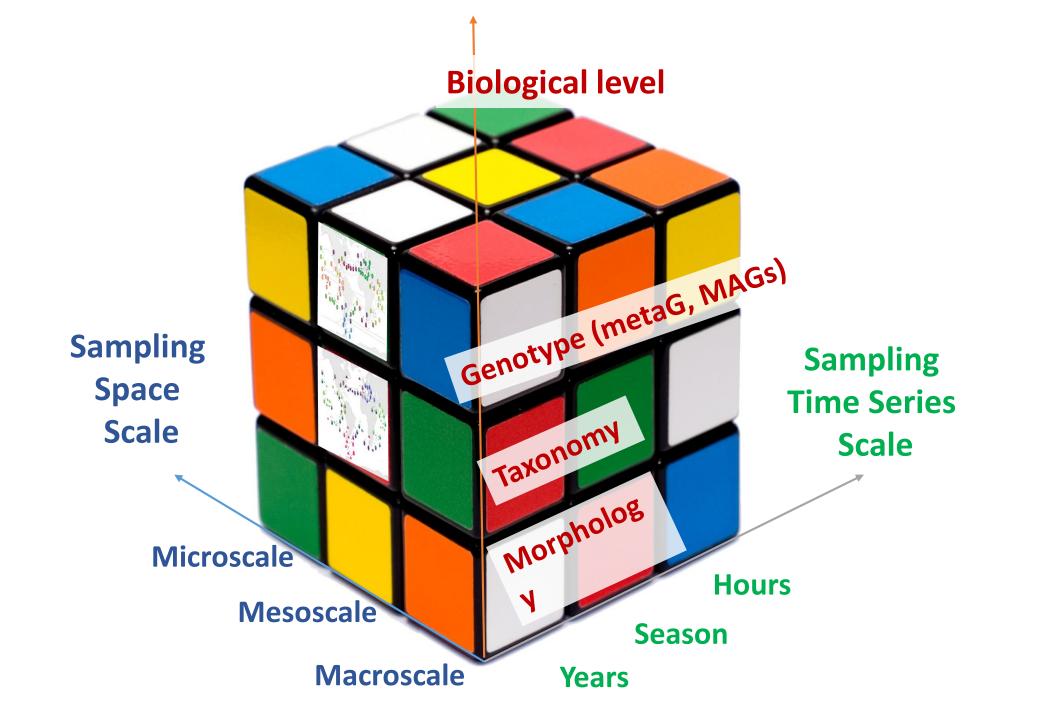


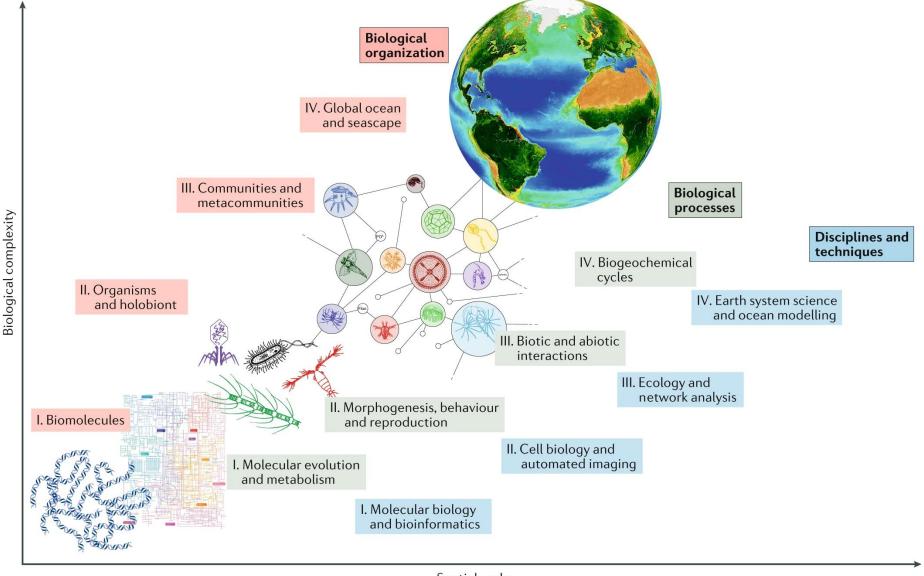
- The slope decreases with organism size
- => At genomic resolution, the turn over rate of populations along currents decreases with organism size

"Omics" resolution vs Morphology resolution



(p=0.51) but morpho is less correlated with transport time and provides a more patchy biogeography.



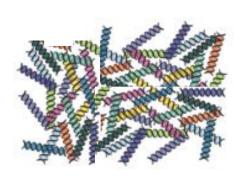


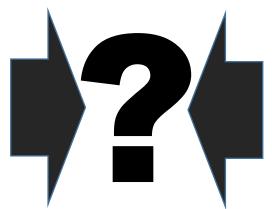
Spatial scale From nanometres to 40,000 km

Sunagawa, S., Acinas, S.G., Bork, P. et al. Tara Oceans: towards global ocean ecosystems biology. Nat Rev Microbiol 18, 428–445 (2020).

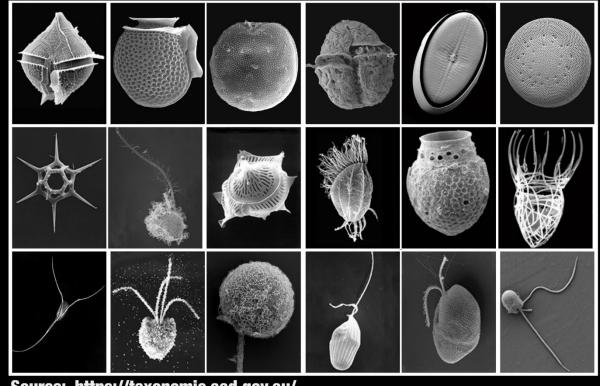
From kmers to genomes : Community level to genome level, and *vice versa*

A lot of fragments of DNA



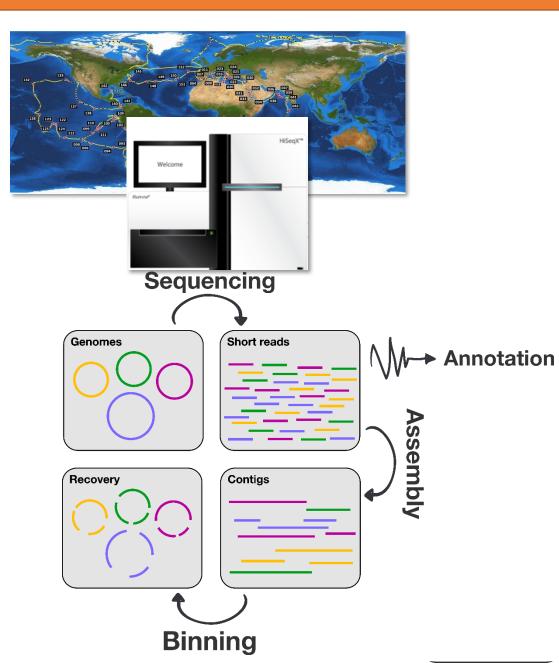


A lot of beautiful images but not that many genomes...



Source: https://taxonomic.aad.gov.au/

Reconstructing Genomes : Metagenome-based Assembled Genomes

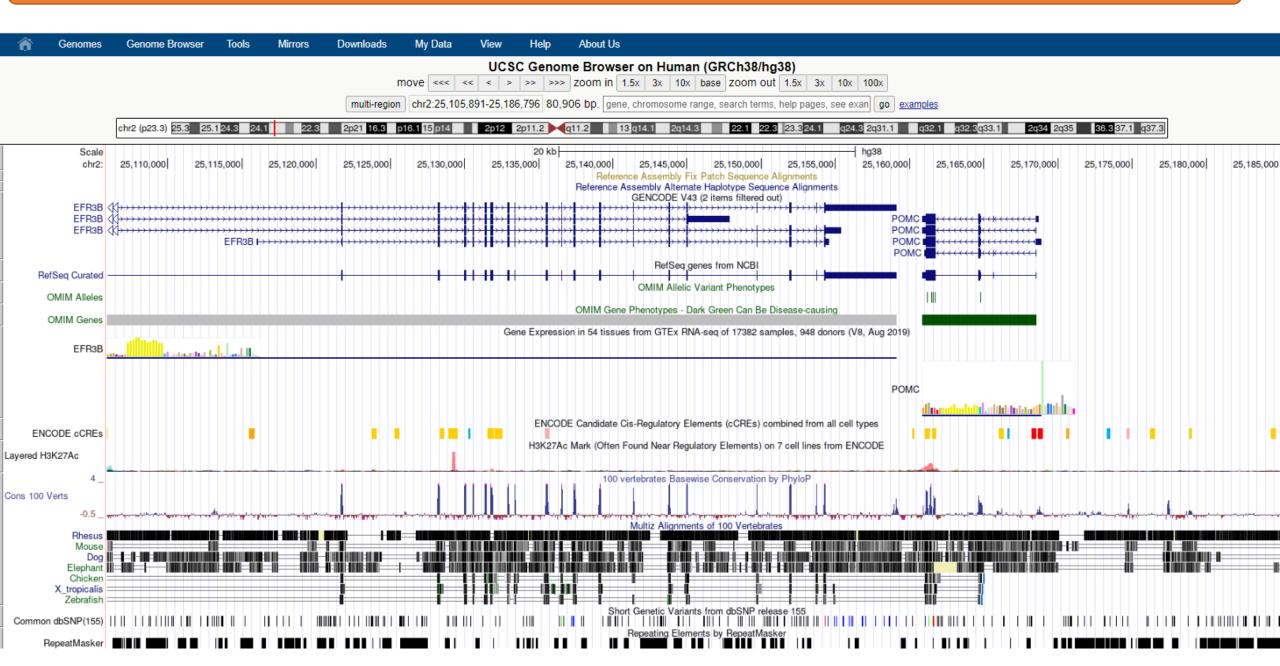


- Nearly 1,000 metagenomes
- 280 billion metagenomic reads
- 11 large co-assemblies by region
- 2 million contigs > 2.5 kbp

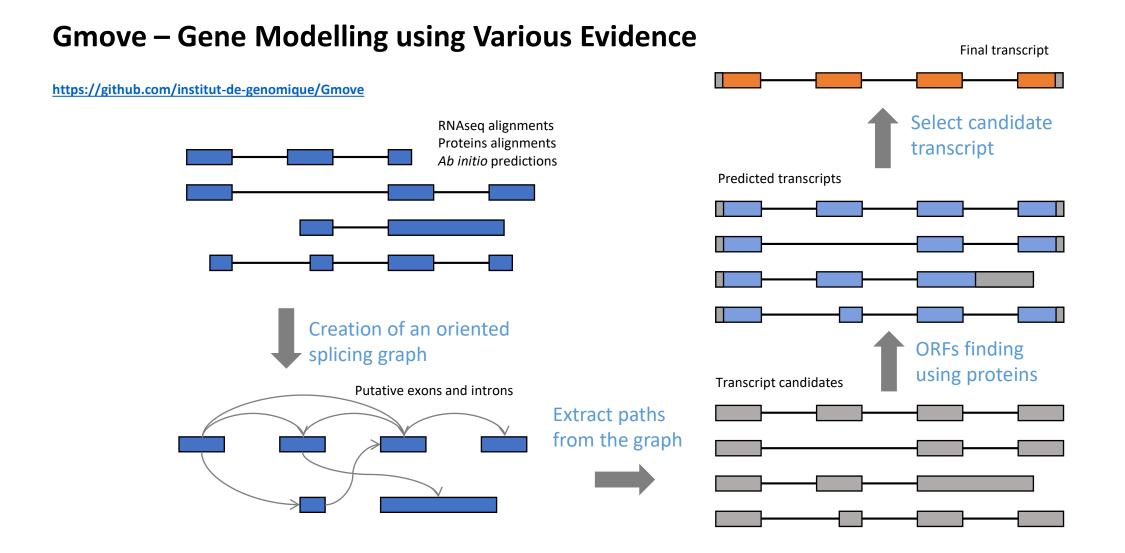
For each co-assembly:

- Automatic binning (2,550 bins in total)
- Extensive manual curation

Reconstructing Exon – Intron Gene Structure

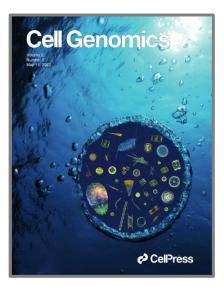


Combining data to predict genes structures

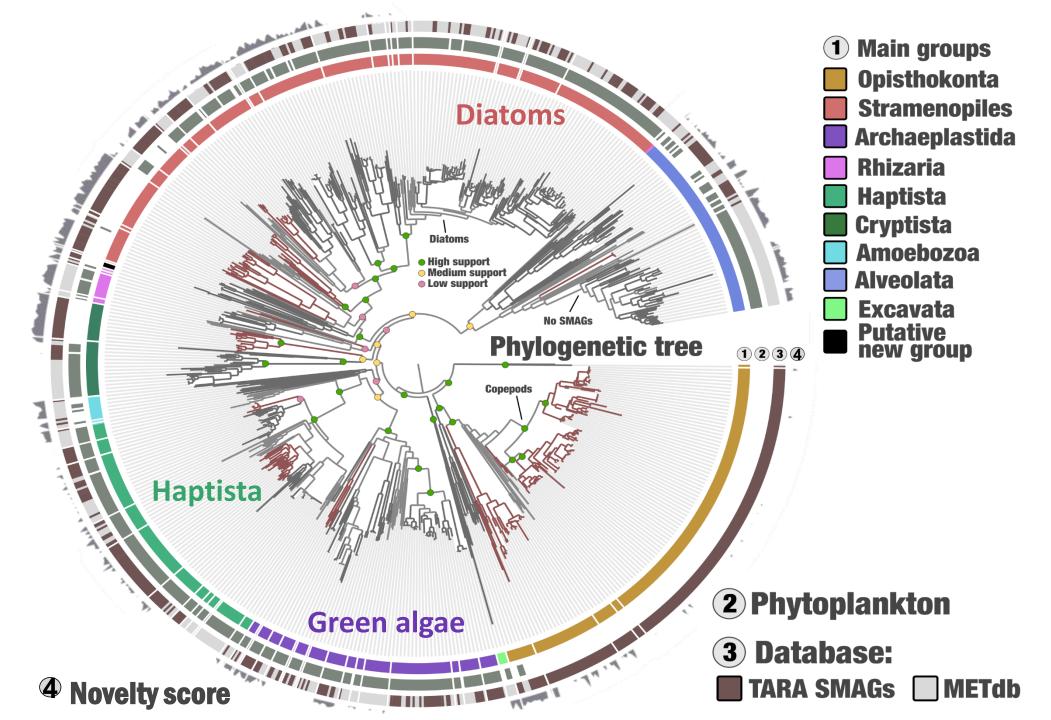


~ 700 eukaryotic genomes

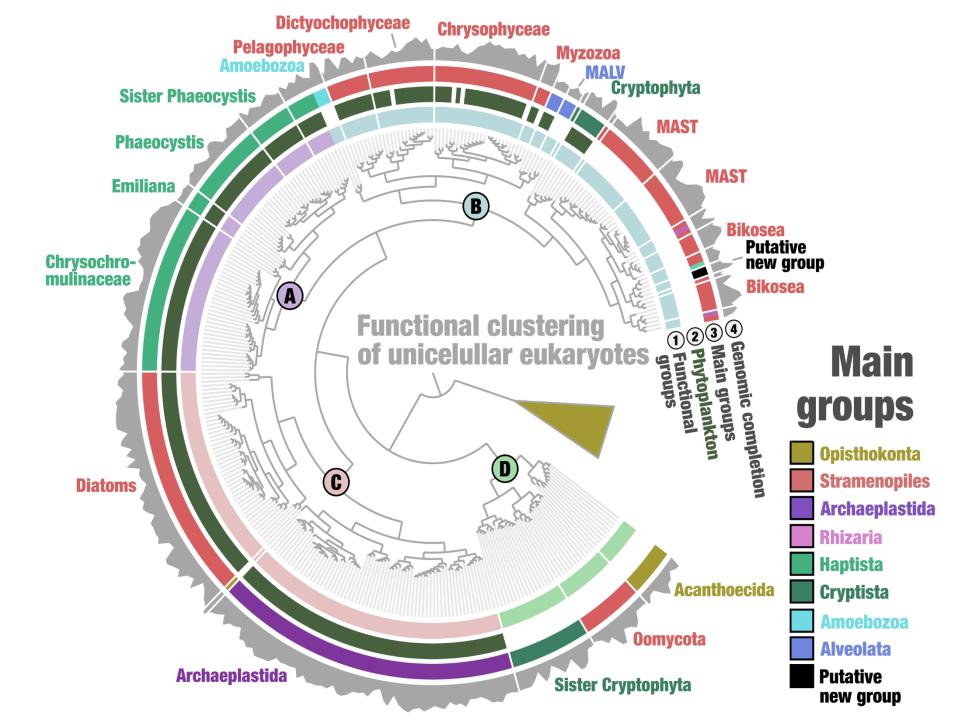
- Average completion ~40% (up to 93.7%)
- Average genomic length 35.4 Mbp (up to 1.32 Gbp)



Delmont et al. Cell Genomics 2022

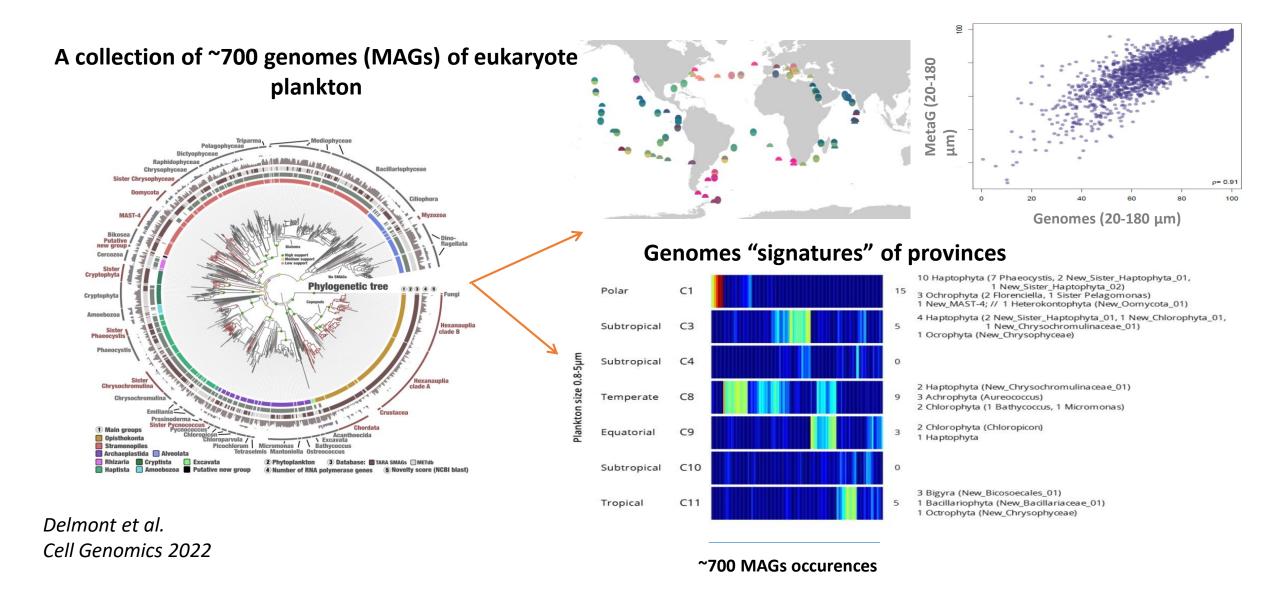


Delmont et al. Cell Genomics 2022



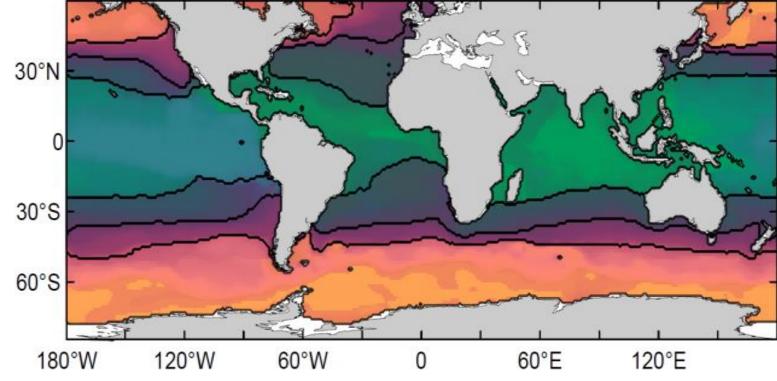
Delmont et al. Cell Genomics 2022

From kmers to genomes : Community level to genome level, and vice versa

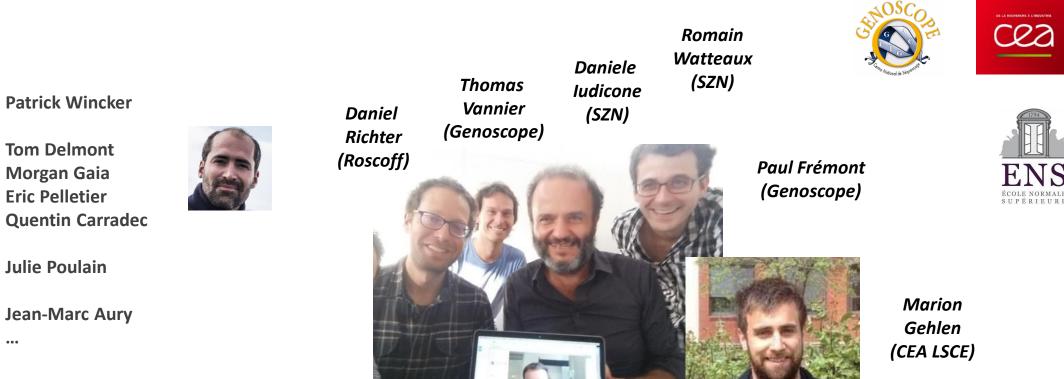


Going Further with Modeling approach (Paul's Talk Teasing)

- Can we extrapolate to :
 - Whole Oceans Biogeography & Merge Plankton size fractions ?
 - End of Century Modification in Biogeography & Biogeochemical Functions
 - => ML / Niche Modeling + Climate Models + Carbon Export Models



Fremont et al. Nature Clim. Ch. 2022



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Fabien Lombard (CNRS Villefranche)

Mathieu Vrac (CEA LSCE)

Colomban; **De Vargas** (Roscoff))













EMBL

Ínría_



All Tara Oceans Coordinators and actors