



Microbial Oceanomics using High-Throughput DNA Sequencing

Ramiro Logares

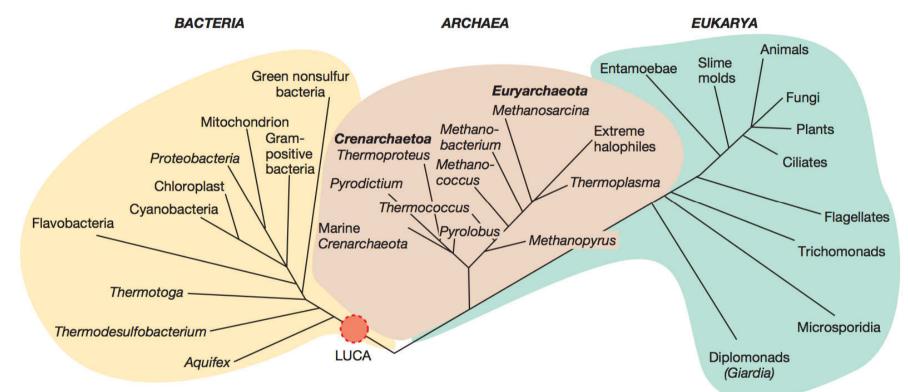
Institute of Marine Sciences, CSIC, Barcelona

9th RES Users'Conference – 23 September 2015

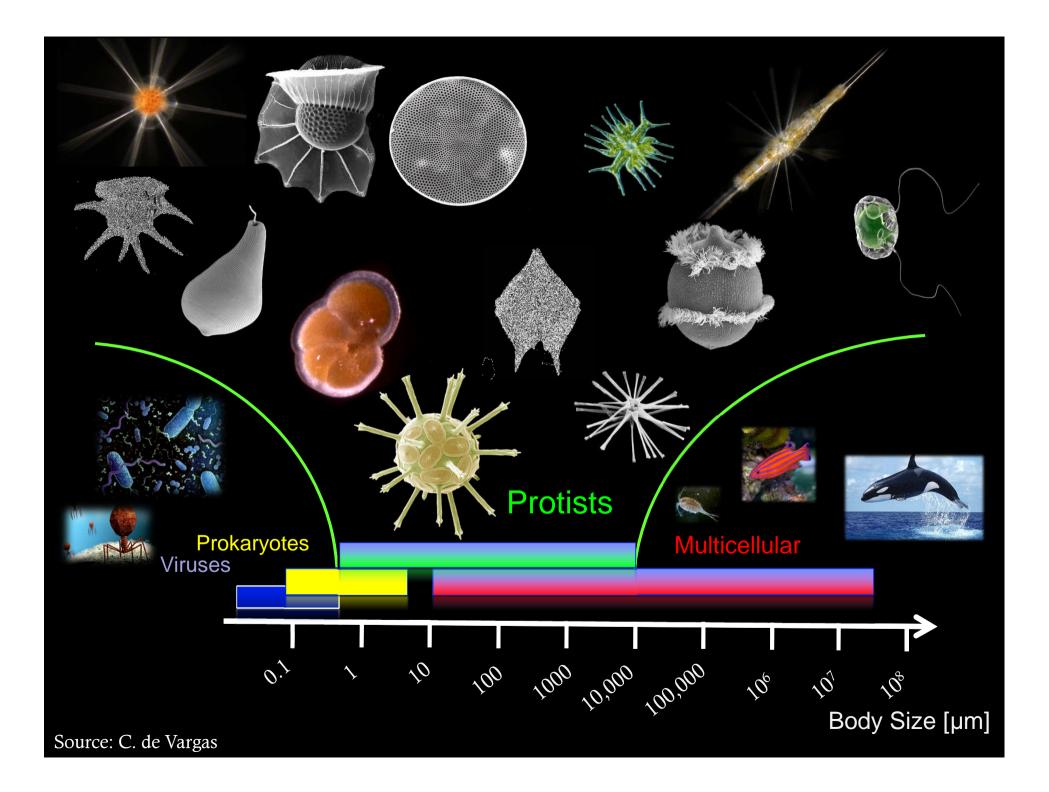
Importance of microbes in the sunlit ocean

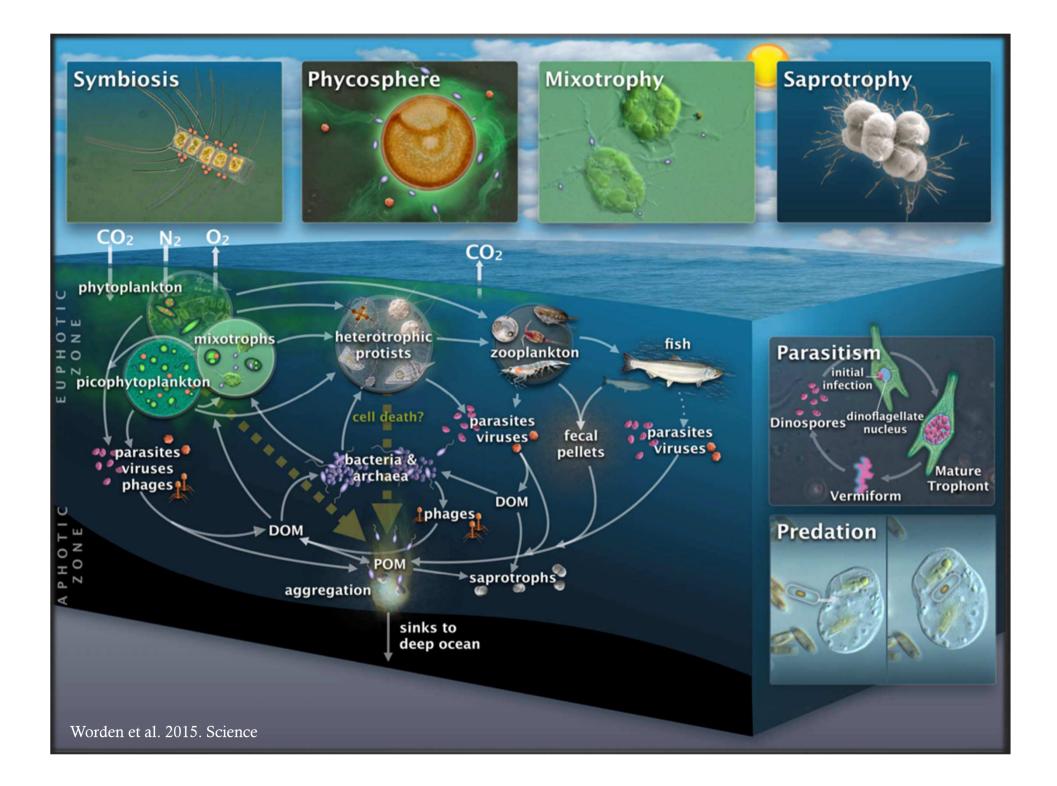
- Phytoplankton: 50% primary production of the Earth (Field et. Al 1998)
- Microplankton crucial for the marine food chain
- Biogeochemical cycling
- Large phylogenetic and metabolic diversity

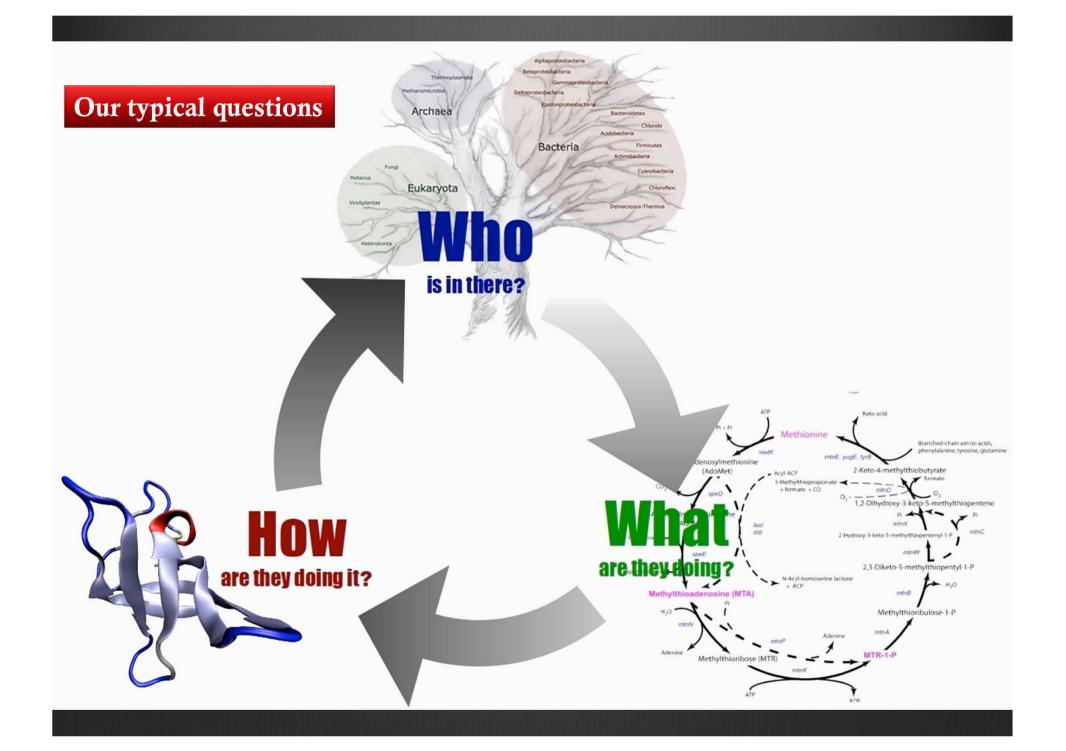
Microbial phylogenetic diversity



Brock "Biology of Microorganisms"



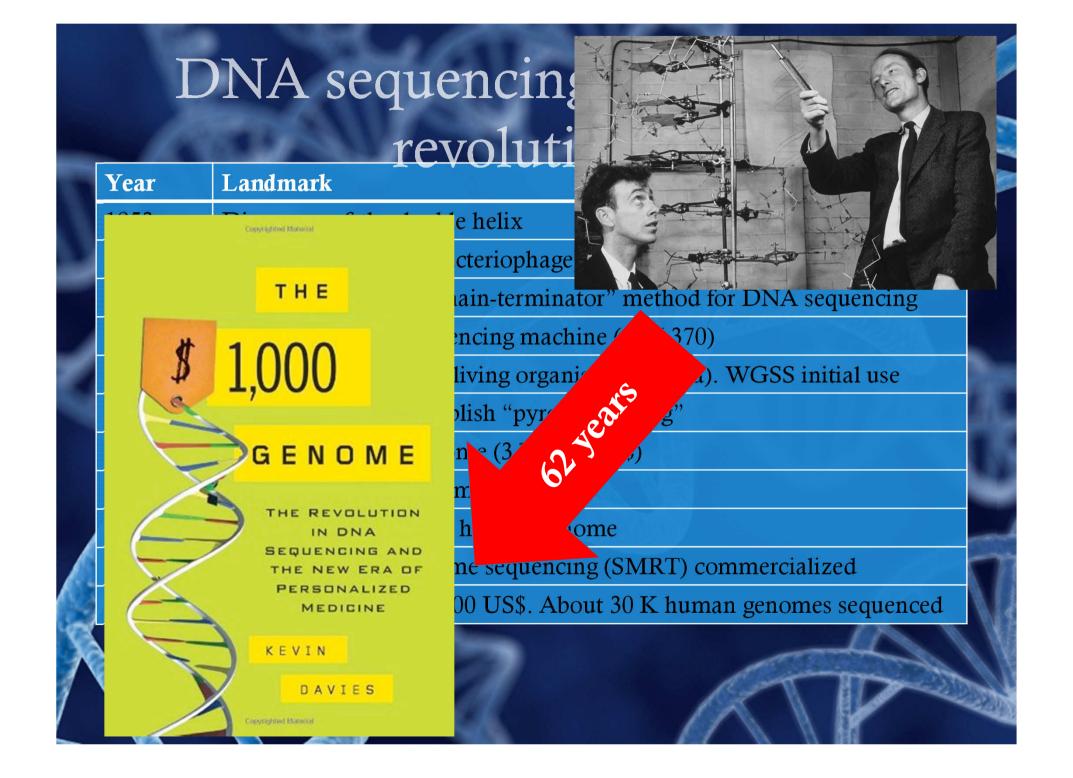




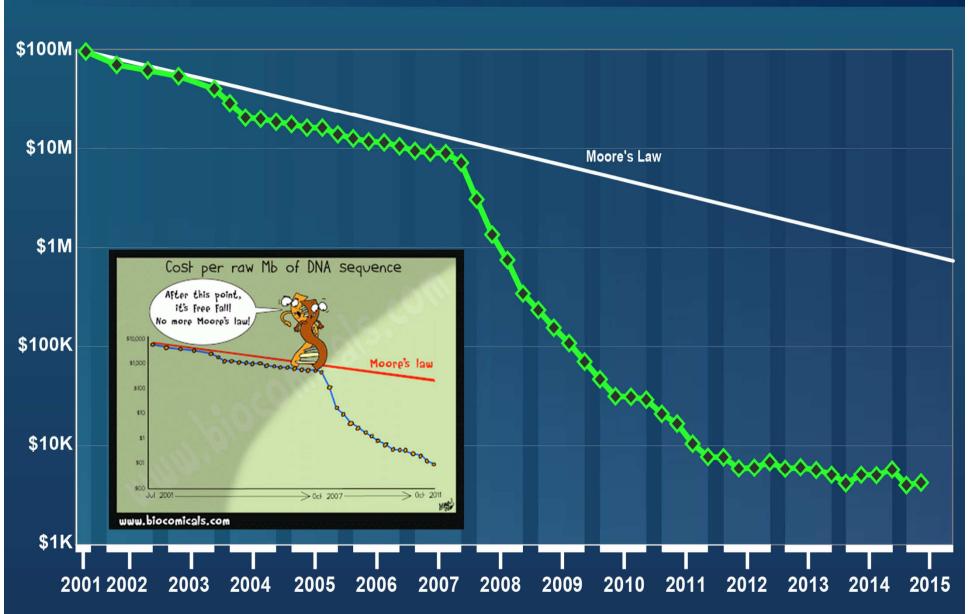
High throughput DNA sequencing technologies: a new lens for viewing the microbial world

DNA sequencing: an accelerating revolution

| ļ | Year | Landmark |
|---|------|---|
| | 1953 | Discovery of the double helix |
| | 1977 | First DNA genome (bacteriophage) |
| ł | 1977 | F. Sanger publishes "chain-terminator" method for DNA sequencing |
| 1 | 1987 | First commercial sequencing machine (ABI 370) |
| l | 1995 | First genome of a free-living organism (bacteria). WGSS initial use |
| | 1996 | Nygren & Ronaghi publish "pyrosequencing" |
| | 2001 | First draft human genome (3 billion US\$) |
| 1 | 2004 | 454 pyrosequencing commercialized |
| | 2009 | Illumina 50 K US\$ per human genome |
| 1 | 2010 | Single molecule real time sequencing (SMRT) commercialized |
| | 2011 | Human genome for 8000 US\$. About 30 K human genomes sequenced |
| | 2015 | 1,000 US\$ - Human genome (Illumina X10; 18,000 per year-machine) |



Cost per Genome



Sequencing platforms evolution



1st Generation

2nd Generation

3rd Generation

Still used for smaller projects or when high quality is needed

Quality reference

Widely used in most sequencing projects

Not widely used yet, some devices still not in the market



SMRT seq









Next Generation Genomics: World Map of High-throughput Sequencers



http://omicsmaps.com/hts/centres/imppc/

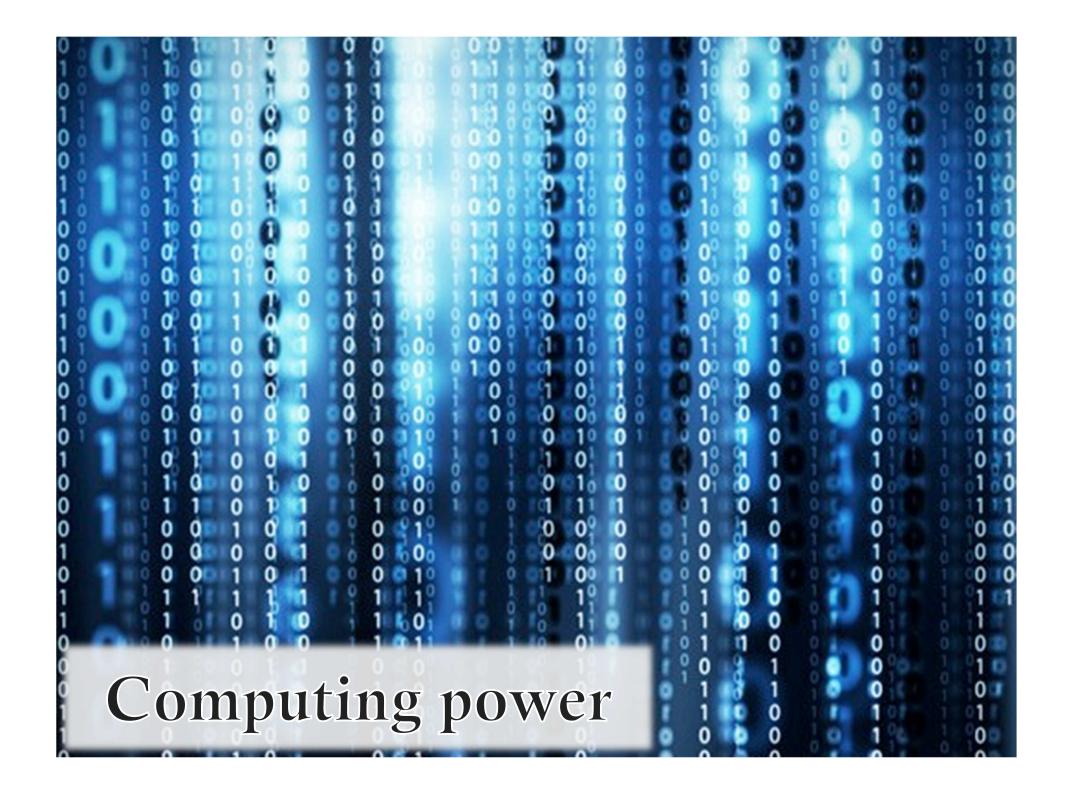
Amount of HTS DNA data produced now in the world:

 $\mathfrak{B} \approx 7,389$ functional HTS machines

 $\approx 35 \times 10^{15}$ bases / year == 35 PETAbases

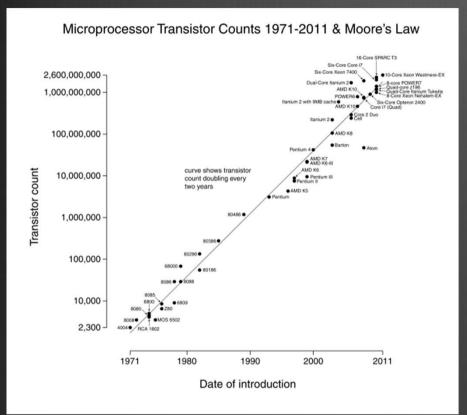
250,000 human genomes per year

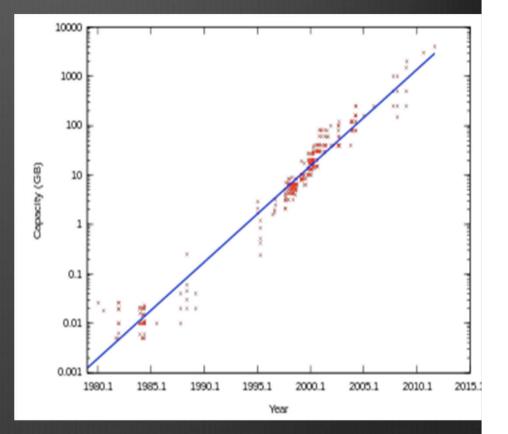




Moore's Law

Kryder's Law





Computer power tend to double every two years

Storage capacity doubles annually



TECHONOMY | 1/12/2012 @ 1:07PM | 4,108 views

DNA Sequencing Is Now Improving Faster Than Moore's Law!

Adrienne Burke, Contributor

+ Comment now

A "worldwide genomics revolution" is upon us.

The genomics industry marked a new milestone on Tuesday. As Forbes' Matthew Herper reported in three separate posts and nearly 100 related <u>Tweets</u>, the two leading manufacturers of DNA sequencing instruments announced almost simultaneously at an investors' conference that they would introduce new machines this



Image by World Economic Forum via Flickr

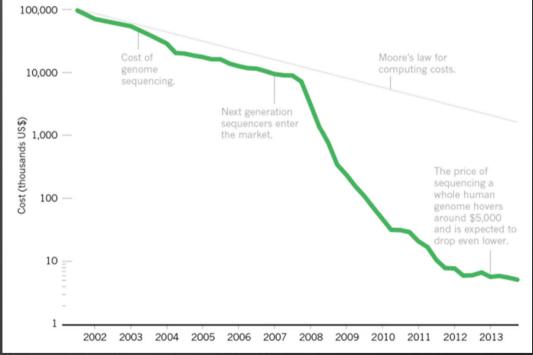
year capable of sequencing an entire human genome in a single day. Life Technologies said its forthcoming Ion Proton machine, which processes DNA on a semiconductor chip, will do it for a cost of \$1,000 per genome.

These advances are not just big news for biotech and medicine, but exciting for all Techonomists. They're proof that the pace of advances in genome sequencing technology has exceeded Moore's Law. The speed of genome sequencing has far better than doubled every two years since 2003, when the

Forbes Magazine, 2012

Falling fast

In the first few years after the end of the Human Genome Project, the cost of genome sequencing roughly followed Moore's law, which predicts exponential declines in computing costs. After 2007, sequencing costs dropped precipitously.



Worldwide sequencing capacity is growing at about 2-3 times per year

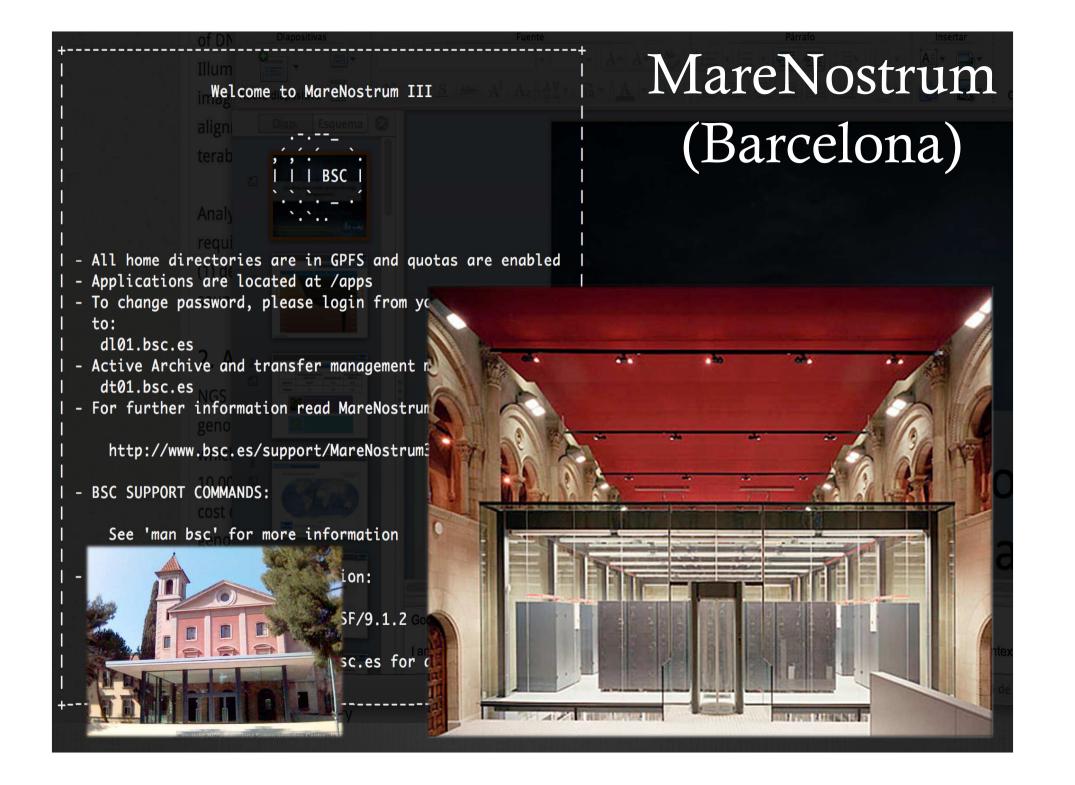
- Only one HiSeq2500 produces about 3 TeraBytes of data per month
- Data processing costs should be considered
 - Electricity
 - Costs for data-admin, and reparation
- Amortization (value decrease) of equipment (3yrs CSIC)
- Data storage cost is not trivial
- What to do with used data? And backups? (maybe cheaper re-sequencing than storing?)

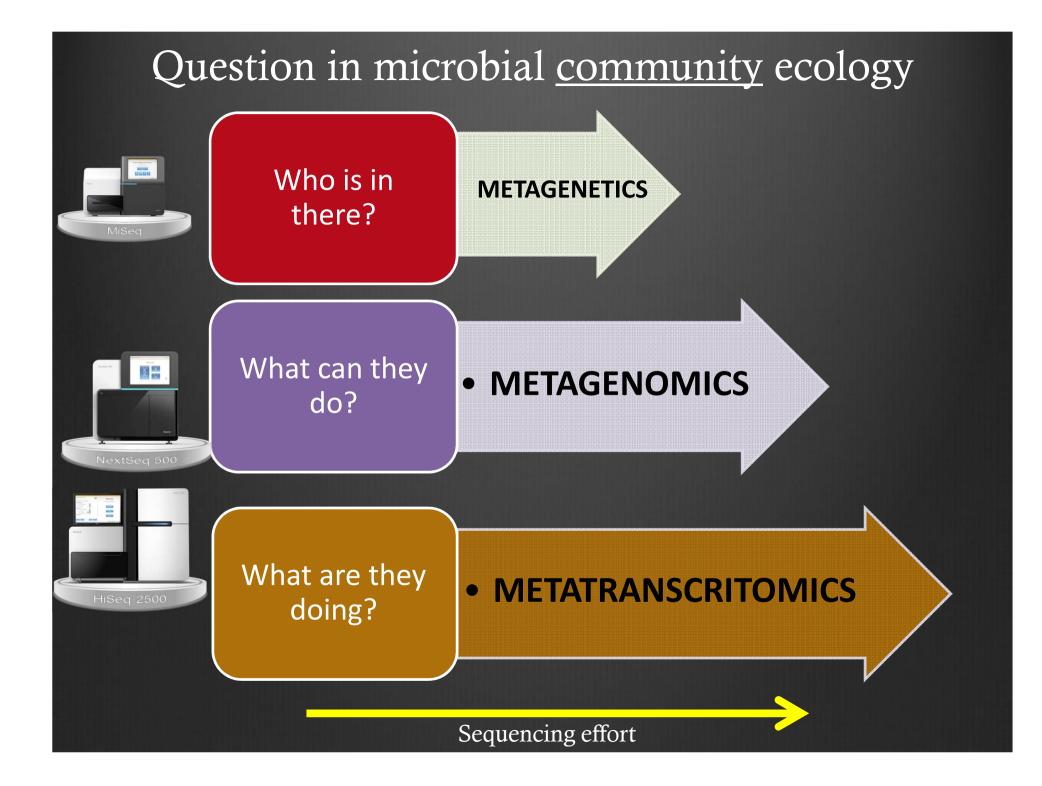
Data processing and computation

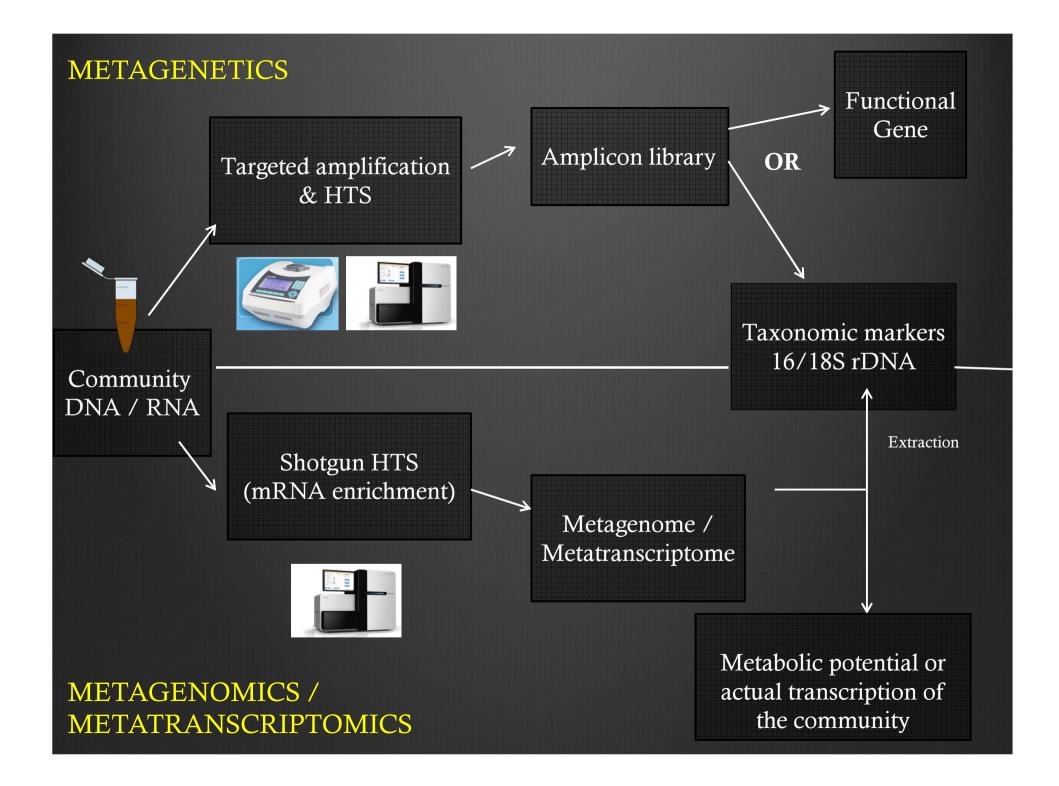


Minimum needed computer power

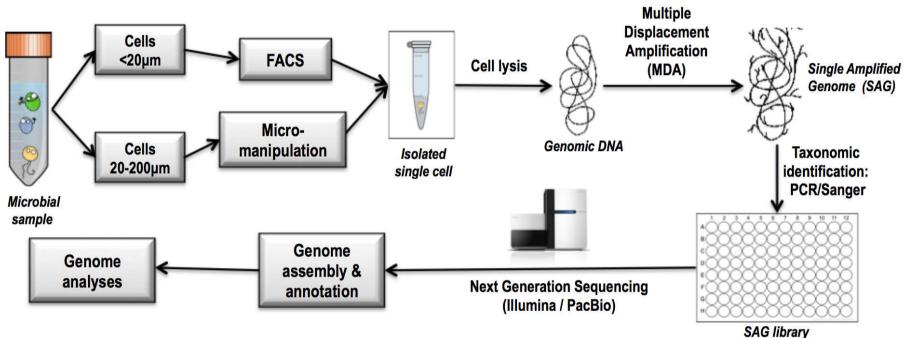
| I st Generation | 2nd Gene | ration | J rd Generation |
|----------------------------|-----------|----------------|----------------------------|
| | | | |
| SANGER | 454 Roche | Illumina HiSeq | PacBio |
| Cores= 1-2 | 16-32 | >64 (128) | >16 (32) |
| Mem= 1-4 GB | 32-64 | >64 (128) | >32 (64) |
| Disk= 0.2Tb | >2 Tb | > 10 Tb | >1Tb |
| | | Ex | |







Alternatively: Single Cell Genomics





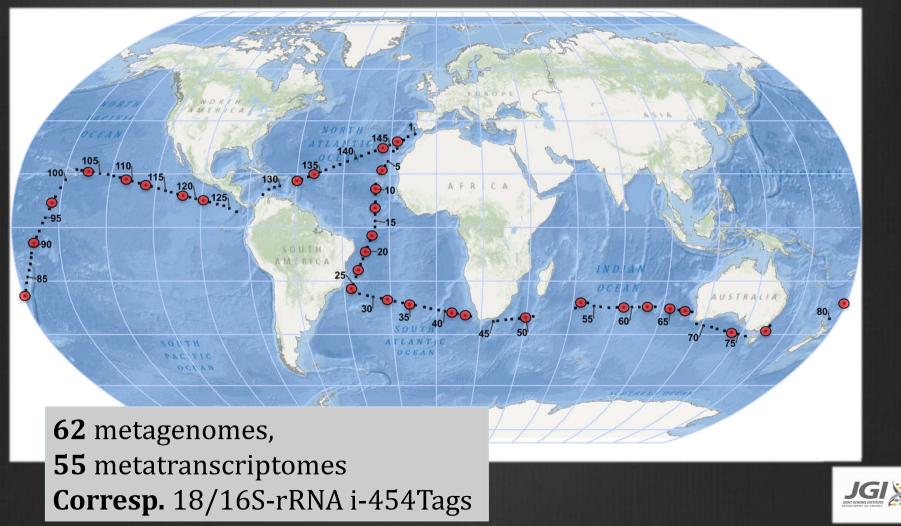
Some results using MareNostrum

Malaspina 2010 expedition

1) Deep-MalaspinOmics (24,000m samples)



Sample characteristics: 120 L filtered through two fractions: ✓ 0.2-0.8 μm (free-living bacteria and archaea) ✓ 0.8-20 μm (protists and attached bacteria)



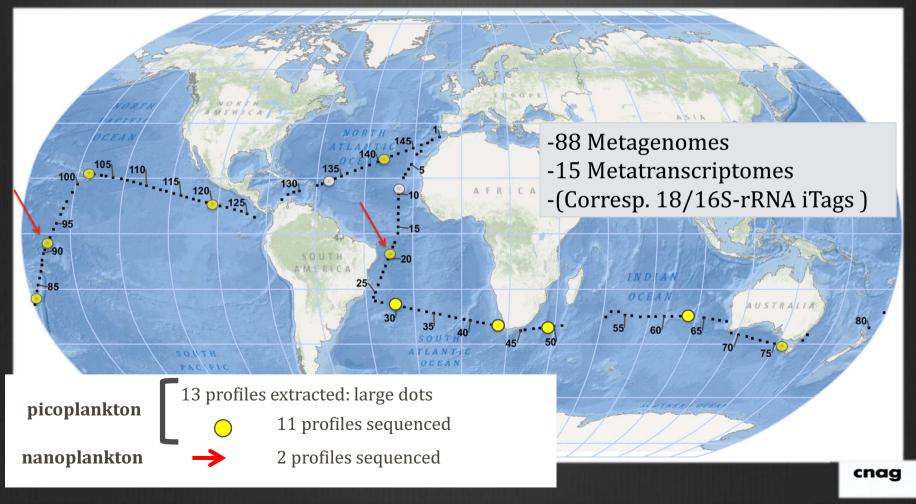
2) Malaspin*Omics* (0 to 4,000m samples)

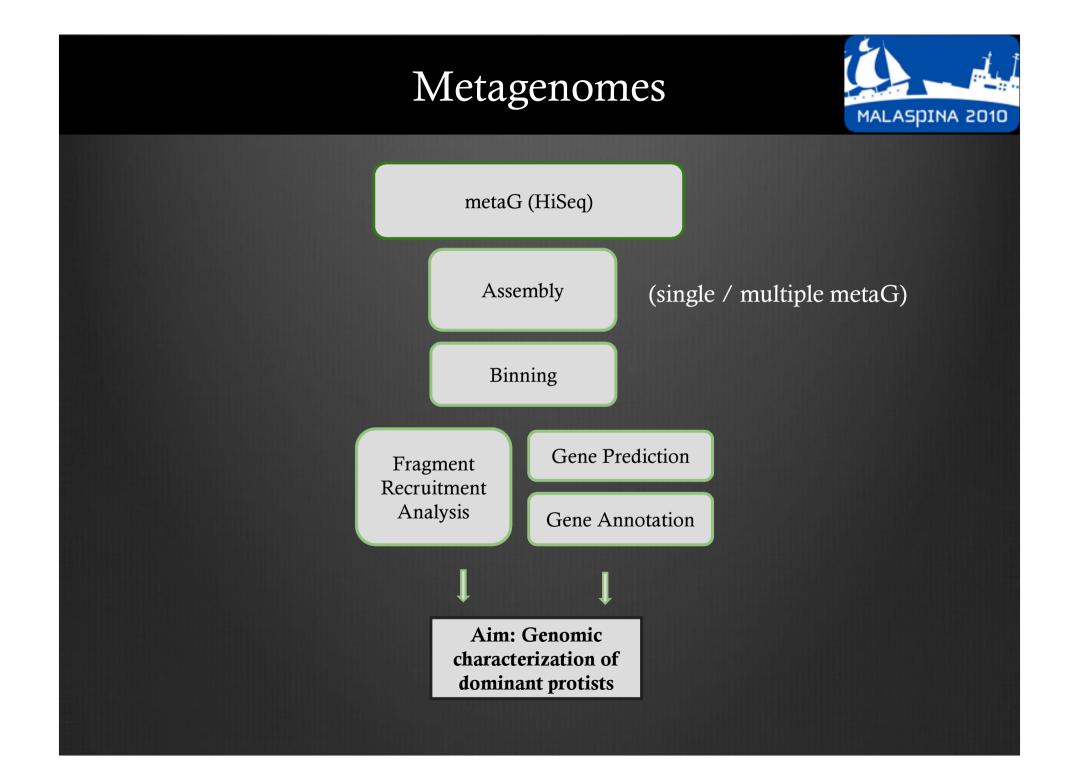
Sample characteristics:

✓ 0.2-3 µm (picoplankton)
✓ 3-20 µm (nanoplankton and attached bacteria)

MALASPINA 2010

Vertical profiles: 7 depths from surface to 4000m







4,000m samples



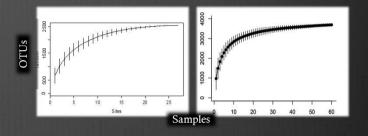
Assembly per sample \rightarrow eukaryotic few contigs

...but, we knew the same OTUs were found in several samples

Co-Assembling all samples together (58 metagenomes) should generate longer contigs

Co-assembly of all samples generated longer contigs

| | One Sample | Sum of All Samples | Assembly All Samples |
|---------------------------|------------|--------------------|----------------------|
| Contigs > 2Kb | 1,055 | 102,705 | 152,175 |
| Mean Coverage (> 2Kb) | 24.87 | - | 139.8 |
| Contigs > 10Kb | 21 | 5,823 | 23,086 |
| Largest Contig (bp) | 40,779 | 207,037 | 925,604 |
| % assembled reads (> 2Kb) | ~5% | - | ~40% |
| Largest Scaffold (bp) | 40,779 | | 1,275,015 |



MareNostrum Supercomputer 2,048 processors with Ray assembler



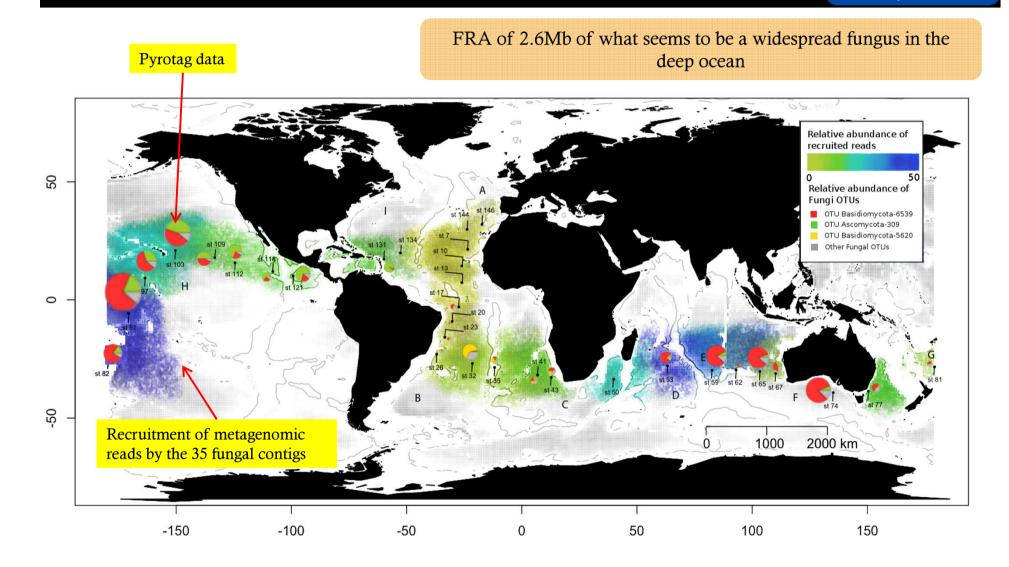
Metagenomes

Co-assembly of 1,500-4,000m samples:

- 58 | 4,000m (5Gb each)
- 29 | 1,500-4,000m (20-40 Gb each)
- Ray assembly with 2,048 threads @MN (18hs)

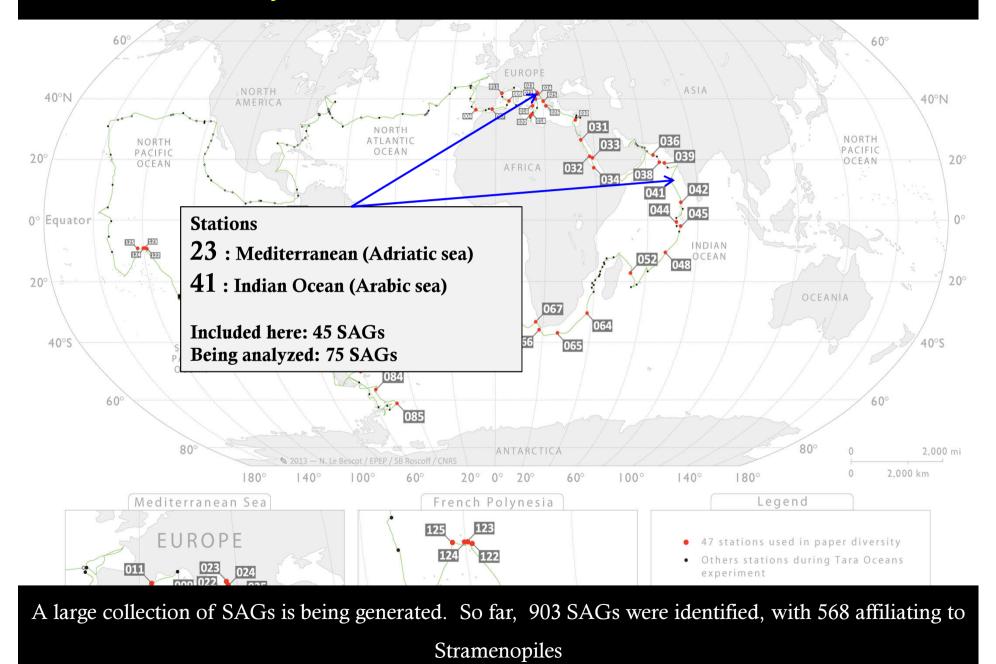
| | 4,000m | 1,500 - 4,000m |
|---------------------|---------------|----------------|
| # Contigs > 2kb | 152,175 | 339,898 |
| Largest contig (bp) | 925,604 | 1,119,237 |

Metagenomes: deep ocean fungi 4,000m samples





Analyzed Pico-euk SAGs from TARA



Co-assembly stats

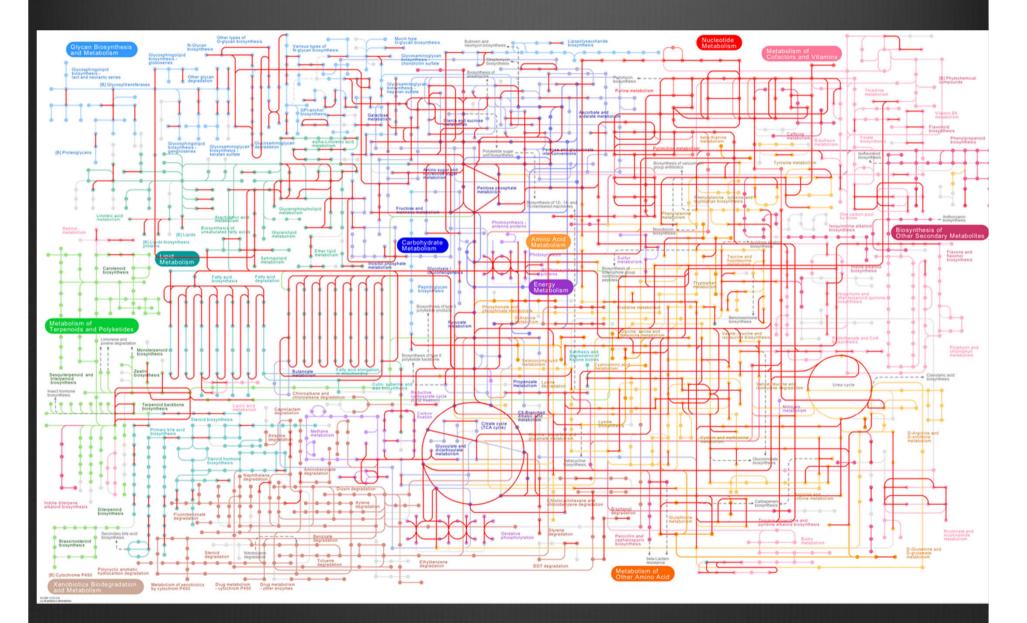
| | % Genome recovery (CEGMA) | Assembly size (Mb; contigs >1,000bp) | Contigs (> 1,000bp) | Max. contig | N50 (>1,000bp) |
|---|---------------------------------|--|------------------------|-----------------|----------------|
| Co-Assembly 14 SAGs MAST-4 clade A (SPAdes) | <u>89.1</u> | 47.5 | 14,564 | 57,905 | 4,563 |
| Co-Assembly 14 SAGs MAST-4 clade A (MegaHit) | 80.6 | 42.5 | 15,158 | 51,080 | 3,475 |
| MAST-4 clade A single SAG assembly (mean SD) | 20.6 10.2 | 9.1 4.5 | 1,694 763 | 72,570 20,347 | 11,041 3,121 |
| Co-Assembly 9 SAGs MAST-4 clade E (SPAdes) | <u>68.5</u> | 32.3 | 5,677 | 104,912 | 9,991 |
| MAST-4 clade E single SAG assembly (mean SD) | 14.3 5.5 | 6.2 2.4 | 1,098 350 | 63,915 18,608 | 10,567 1,920 |

Continuing analyses with the co-assembly

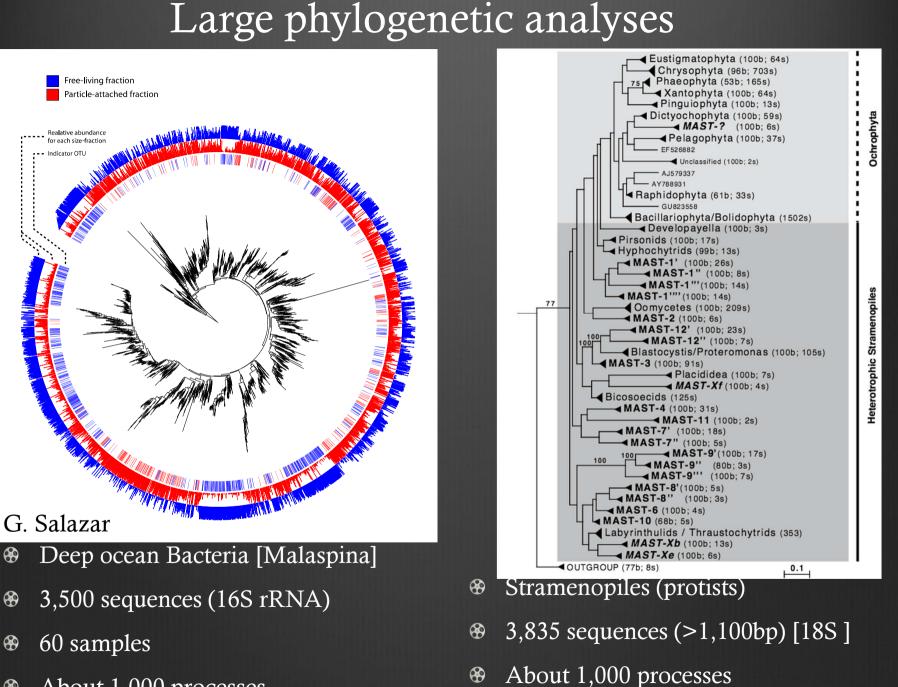
1) Gene prediction [Augustus]

2) Annotation (KEGG, KOG, Pfam, eggNOG, OMRGC, MMETSP)

General metabolic pathways



In Red: mapped MAST4 pathways/functions according to KO



About 1,000 processes

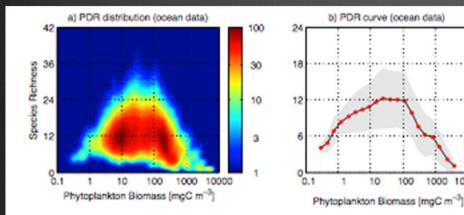
Simulations

Primary production vs. richness

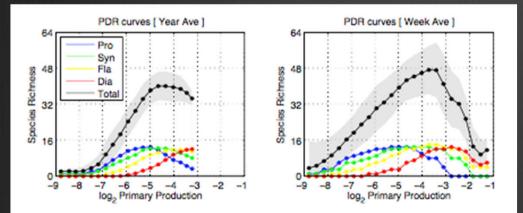
100

1000 10000

Real data



Simulation



MIT ecological selection model

- Global marine ecosystem model
- Several plankton and nutrient types
 - 64 phytoplankton species (small, large)
 - 2 zooplankton generic (small, large)
 - 4 nutrients (N,P,Si,Fe)
- Four phytoplankton functional groups with trade-offs
 - slow growth niche specialists (Prochlorococcus, Synechococcus)
 - fast growth niche opportunists (flagellates, diatoms)
- Self-assembly of the phytoplankton community
 - ecological selection by resource competition
 - survival of the most adapted to the environment

Vallina et al., 2014. Nat. Comm

Summary of results with RES support since 2011

2 in revision

Contributing mostly

- Metagenomics
- Genomics
- Phylogenetics
- Modelling

Conclusions: Microbial ecology

- Massive amounts of DNA data need powerful computers as well as programs that can deal with them
- Solution Future developments require further integration with highperformance computers and quantitative methods
- Analysis of large datasets will likely unveil patterns of genomic functioning as well as interactions between marine microbes







MASSIMO PERNICE



GUILLEM SALAZAR



SILVIA G. ACINAS



RAMON MASSANA



JULIA PERERA BEL



JOSE A. MARTIN CANO



PABLO SANCHEZ



EP GASOL

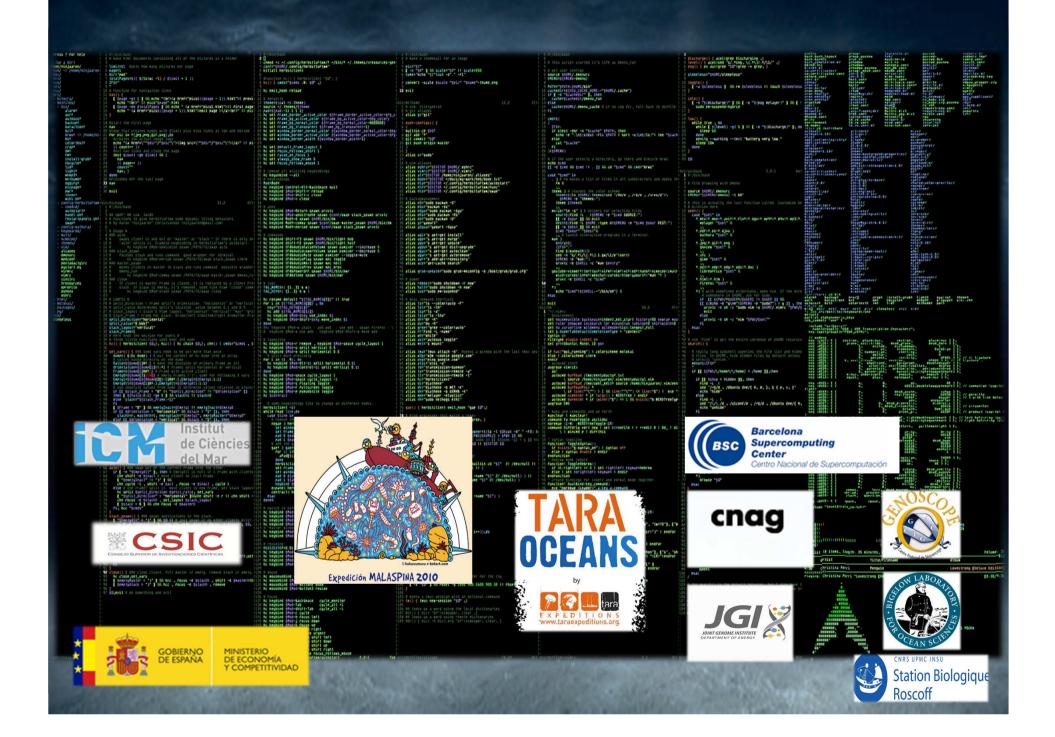
MARTA SEBASTIAN

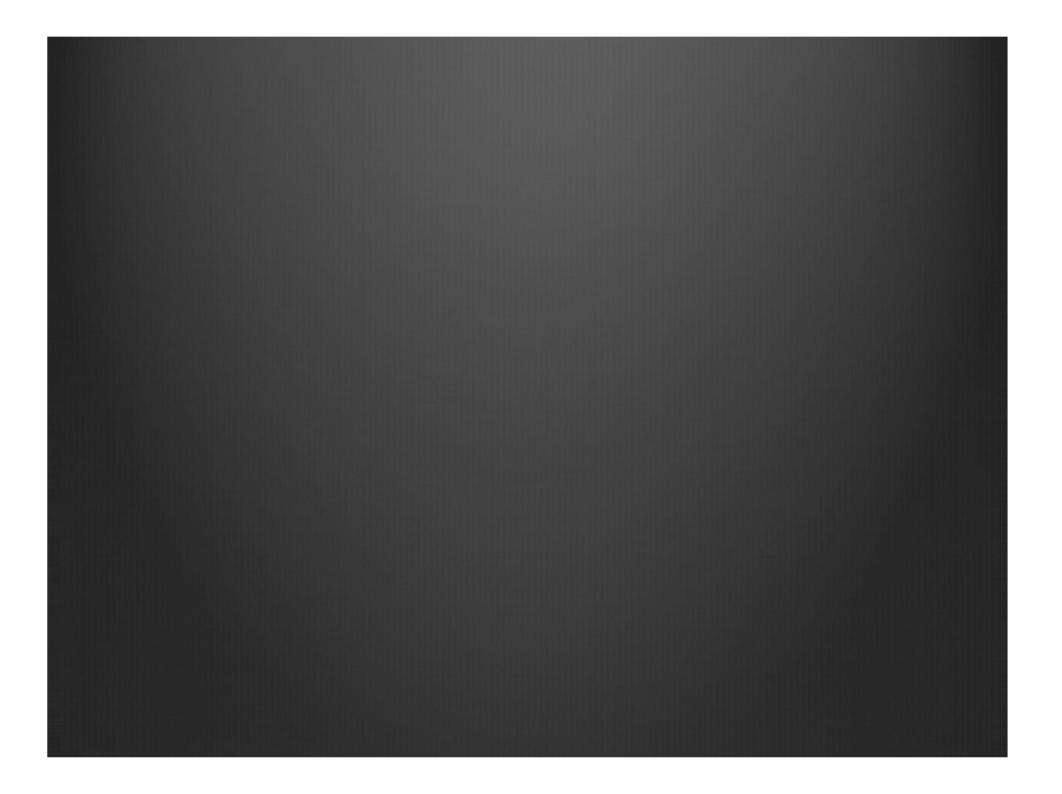


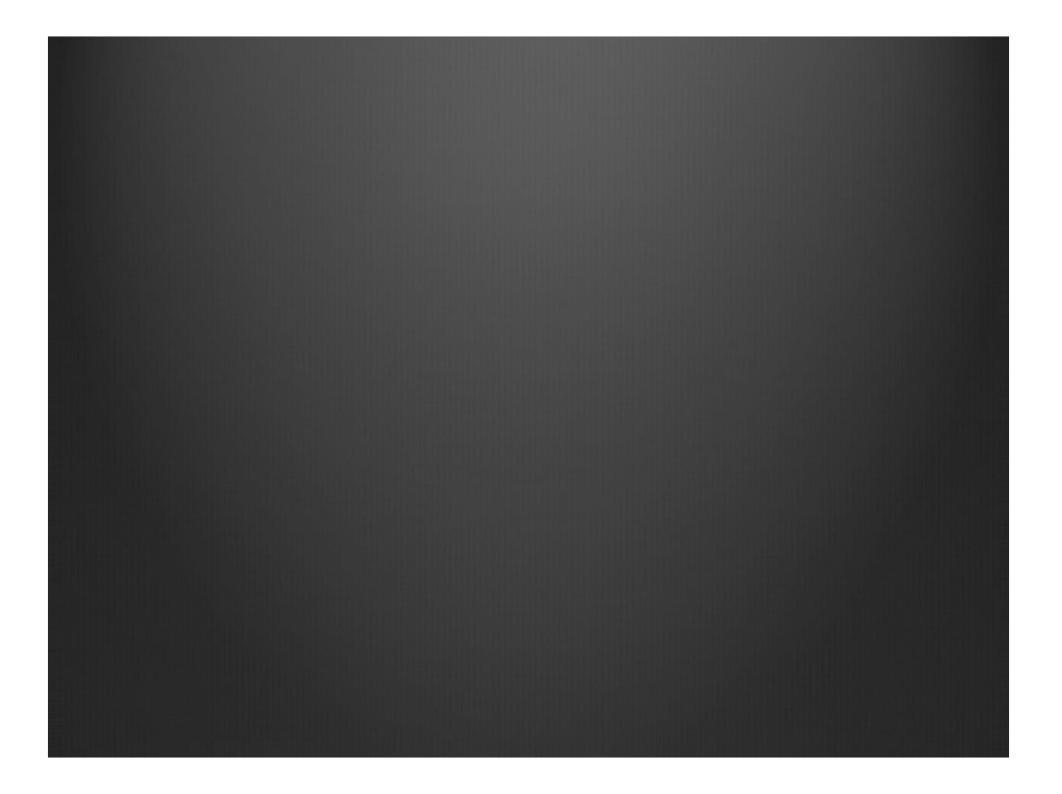
CARLES PEDRÓS-ALIÓ



CATERINA RODRÍGUEZ









- Single Molecule Real Time (SMRT) sequencing
- Average read length: 4,200-8,500 bp (longest read 30Kbp)
 - P4/C2: shorter reads, higher accuracy
 - ✤ P5/C3: longer reads, lower accuracy
 - 200-300 Mbp from each SMRT cell for 15-20kb insert size libraries
 - ✤ 100-150 Mbp for >20Kbp libraries
 - No multiplexing in genomics libraries (multiplexing in amplicons)

 - About 350 € per SMRT cell
 - Signal: colors







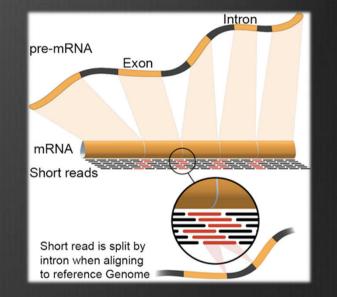
What if you are not interested in the whole community but in one species?

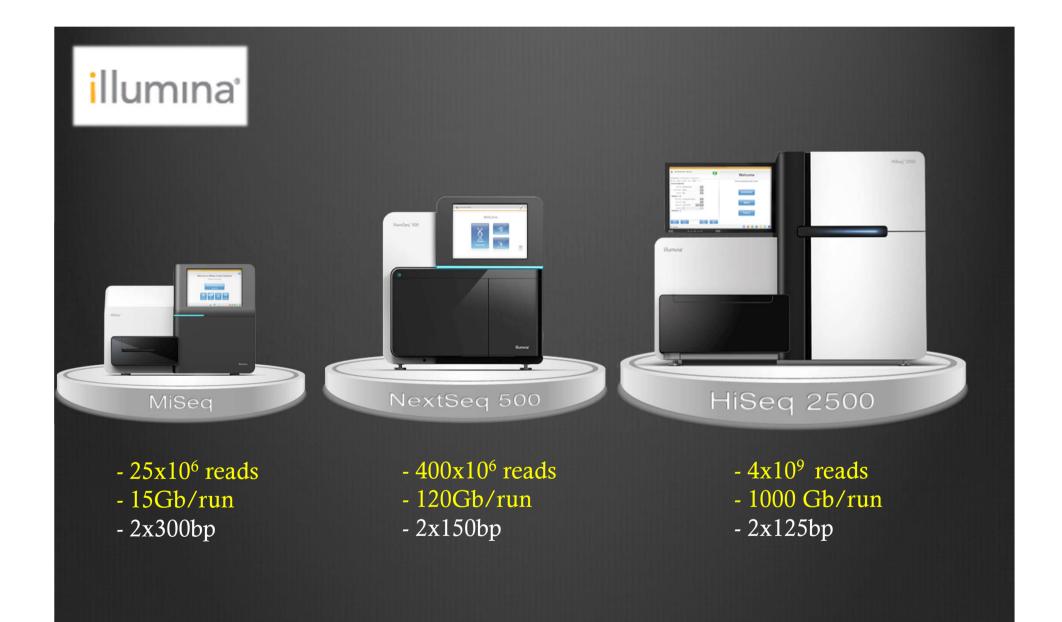
If you have a clonal culture:

Genomics and/or RNA-Seq

If you don't have a clonal culture:

Single-Cell genomics





Signal: colors 1Gb=1x10⁹ bases



+1 3

74 🖪 🗎

España, la falta de expertos en genética relativiza el logro

Por unos 750 euros, se puede comprar un buen ordenador portátil, viajar cinco días a las playas de Miami, y, desde hace unos días, secuenciar un genoma

hac tarr dor MinION USB stick gene sequencer finally comes to market

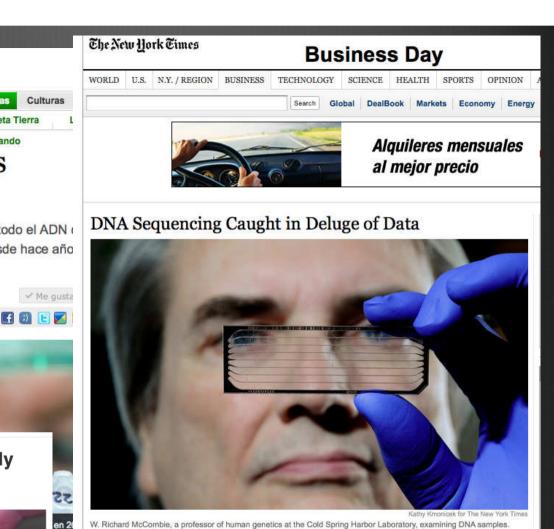
By John Hewitt on September 19, 2014 at 2:10 pm 22 Comments

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By ANDREW POLLACK Published: November 30, 2011

PHARMA & HEALTHCARE 9/24/2014 @ 6:00AM 3,497 views

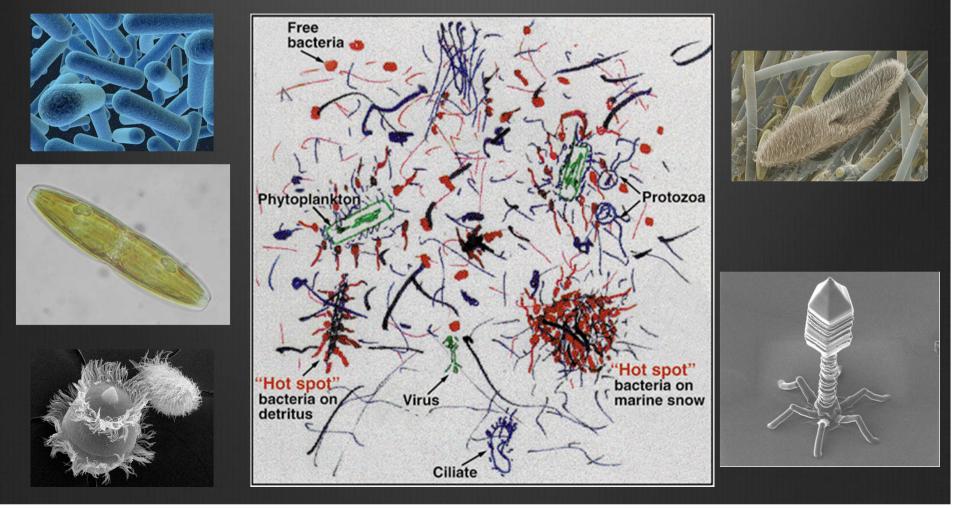
The (Unmet) Potential Value of Cancer Genome Testing

+ Comment Now + Follow Comments

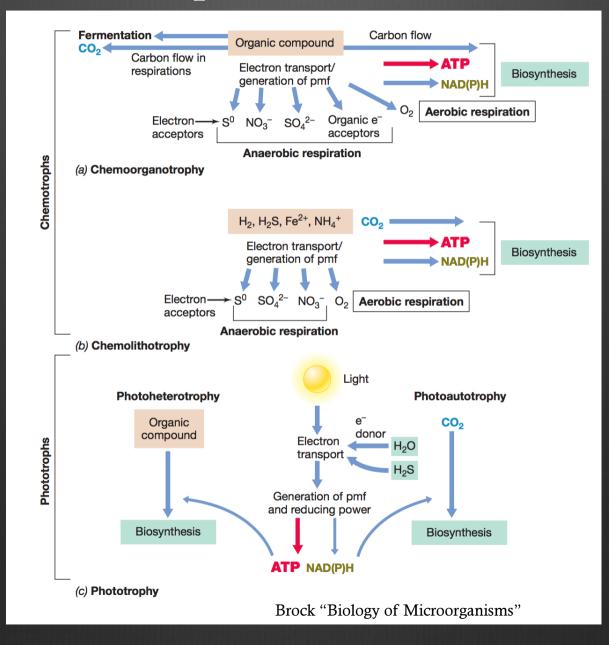
As the cost of DNA sequencing plummets, the possibility of testing all cancer patients' tumor genomes is becoming a reality. For just \$1000 or so, a doctor might submit most any malignant specimen for a complete genetic work-up. The sample might be a core

Oceanic microbial community

 Includes all species occurring at a particular site and their abundances



Multiple metabolisms



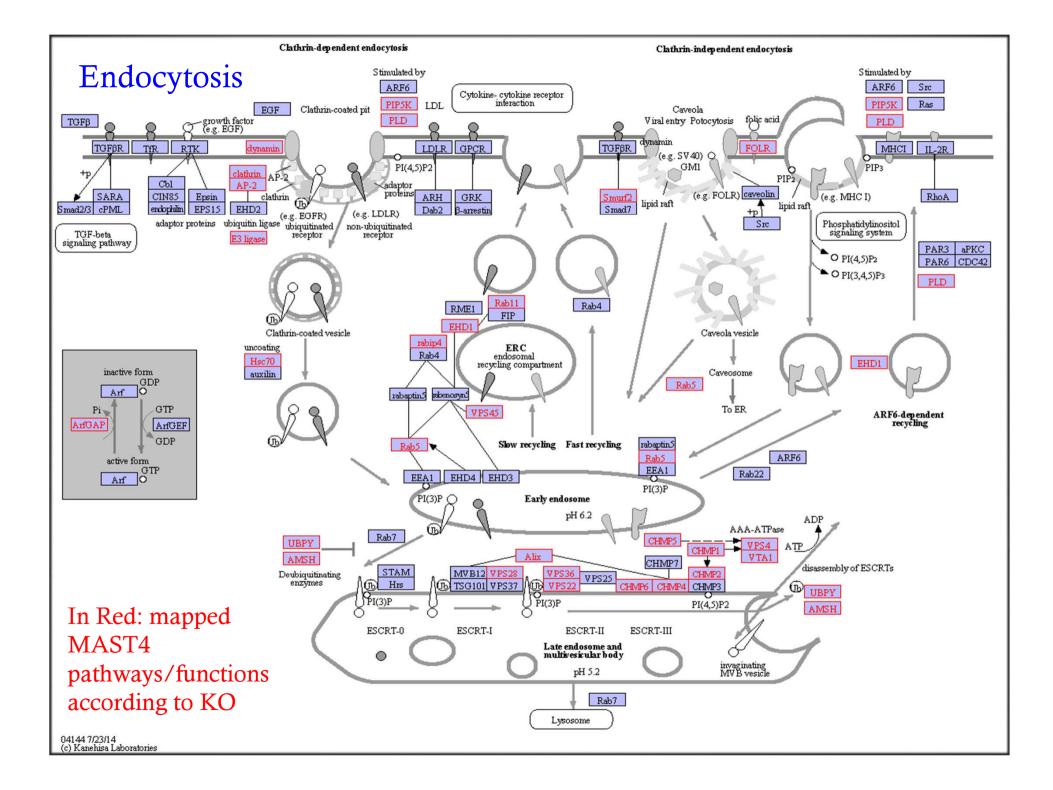
Comparison of platforms

| | Run time | Mreads/run | Read length | Mb/run | €/Mb |
|--------------------|-----------|------------|--------------|------------|--------|
| Sanger (3730xl) | 2h | 0.000096 | 400-900 | 0.06 | 1500 |
| 454 FLX Titanium | 10h | 1 | 400 | 400 | 15 |
| 454 FLX+ | 18-20 h | 1 | 700 | 900 | 9 |
| Ion Torrent | 2h | 80 | 400 | 32,000 | 1 |
| PacBio | 0.5-2h | 0.005 | 4-8 K | 300 (SMRT) | 0.33-1 |
| Illumina MiSeq | 55h | 25 | 2x300 | 15,000 | 0.1 |
| Illumina GAIIx | 14 days | 320 | 2x150 | 96,000 | 0.12 |
| | | | | | |
| Illumina HiSeq2500 | 1-11 days | 4000 | 2x125 | 1,000,000 | 0.05 |

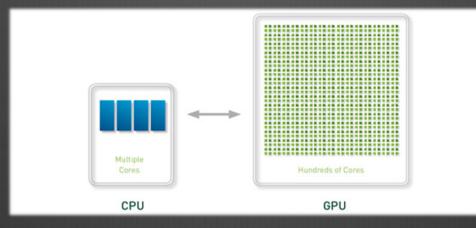
1Gb=1x10⁹ bases

Glenn 2011 updated in http://www.molecularecologist.com/next-gen-table-2-2014/





GPU (graphics processing unit) computing



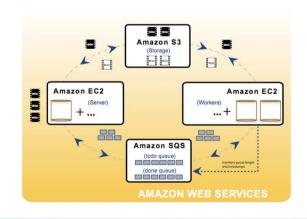
Serial part of an application runs on a CPU and the computationally-intensive part runs on a GPU



GPU Pipeline for HTS sequencing Centro de Investigación Príncipe Felipe http://docs.bioinfo.cipf.es/projects/ngs-gpu-pipeline/wiki

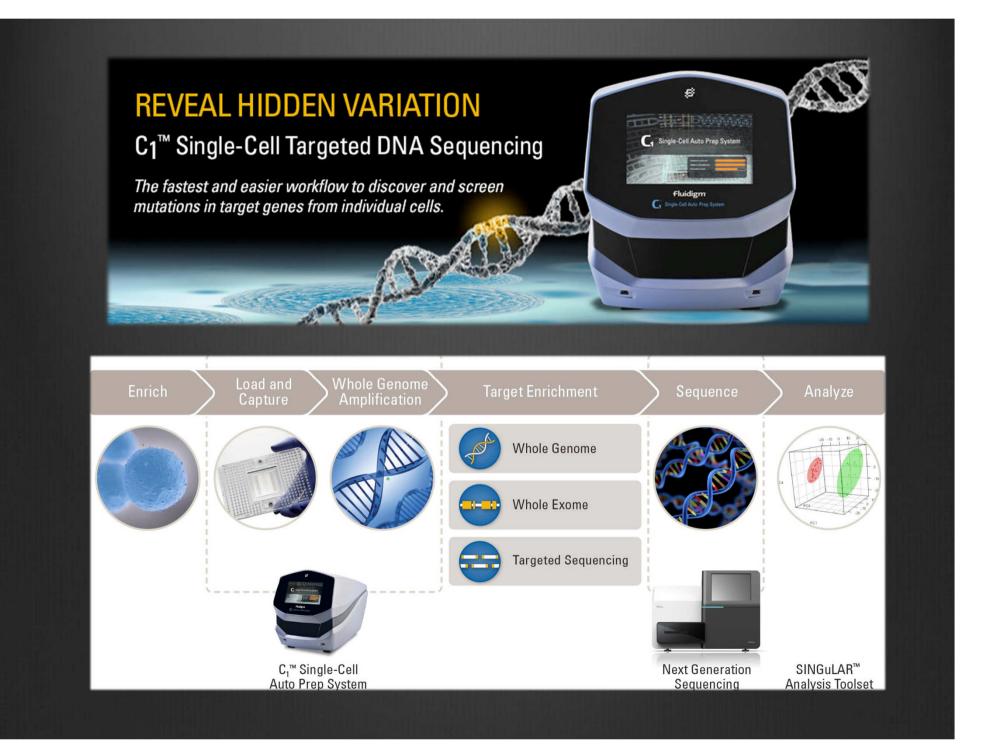
Cloud Computing...

- Purchase needed computer power
- Scalable (few to thousands of processors)
- No maintenance costs





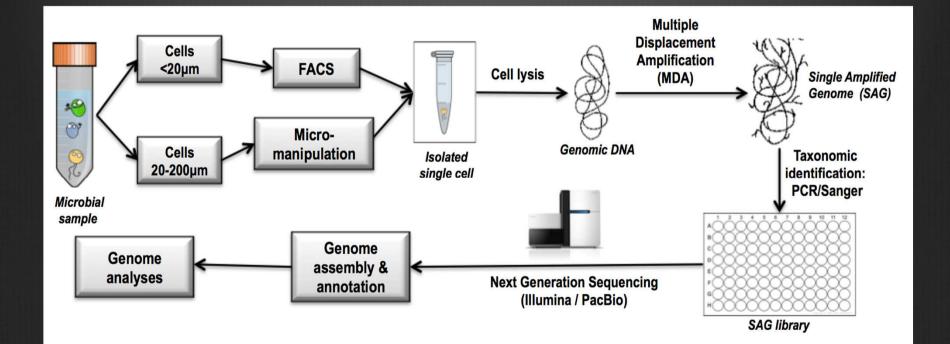
GALAXY (g2.bx.psu.edu)
 N3phele: HTS analyses at the EC2 with QIIME



Molecular biology + computers + stats

- **The next 20 years of genome research**
- M. Schatz (2015)
- http://biorxiv.org/content/early/2015/06/02/020289

General SAG construction strategy





Bigelow Laboratory for Ocean Sciences