



## NGS sequence application, a few examples

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Dr Francois Sabot & Christine Tranchant-Dubreuil

8th of October, 2018

IRD - UMR DIADE

## Analyses

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# Assembly

## 1. Fragment DNA and sequence



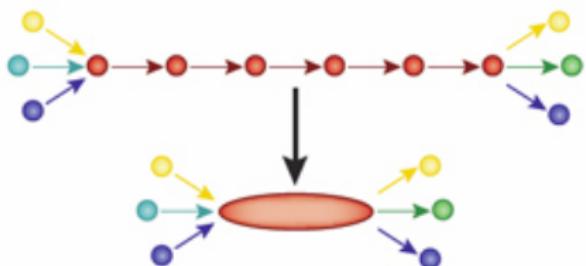
## 2. Find overlaps between reads

...AGCCTAGACCTACA**GGATGCGCGACACGT**  
**GGATGCGCGACACGT**CGCATATCCGGT...

From Baker, 2012

# Assembly

## 3. Assemble overlaps into contigs



## 4. Assemble contigs into scaffolds



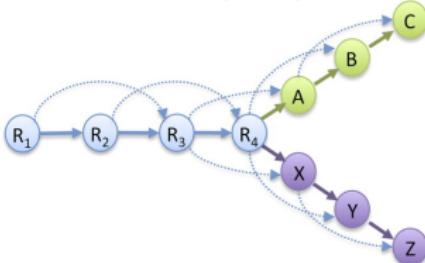
From Baker, 2012

# Assembly

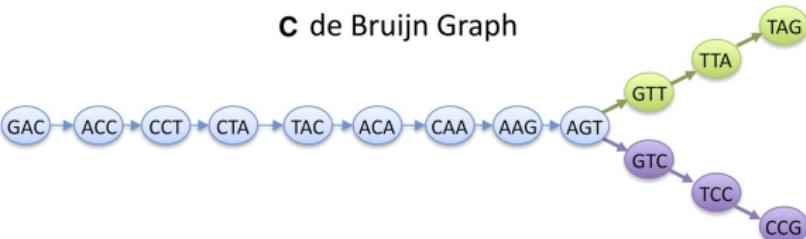
## A Read Layout

R <sub>1</sub> :	GACCTACA
R <sub>2</sub> :	ACCTACAA
R <sub>3</sub> :	CCTACAAAG
R <sub>4</sub> :	CTACAAAGT
A:	TACAAGTT
B:	ACAAGTTA
C:	CAAGTTAG
X:	TACAAGTC
Y:	ACAAGTCC
Z:	CAAGTCCG

## B Overlap Graph

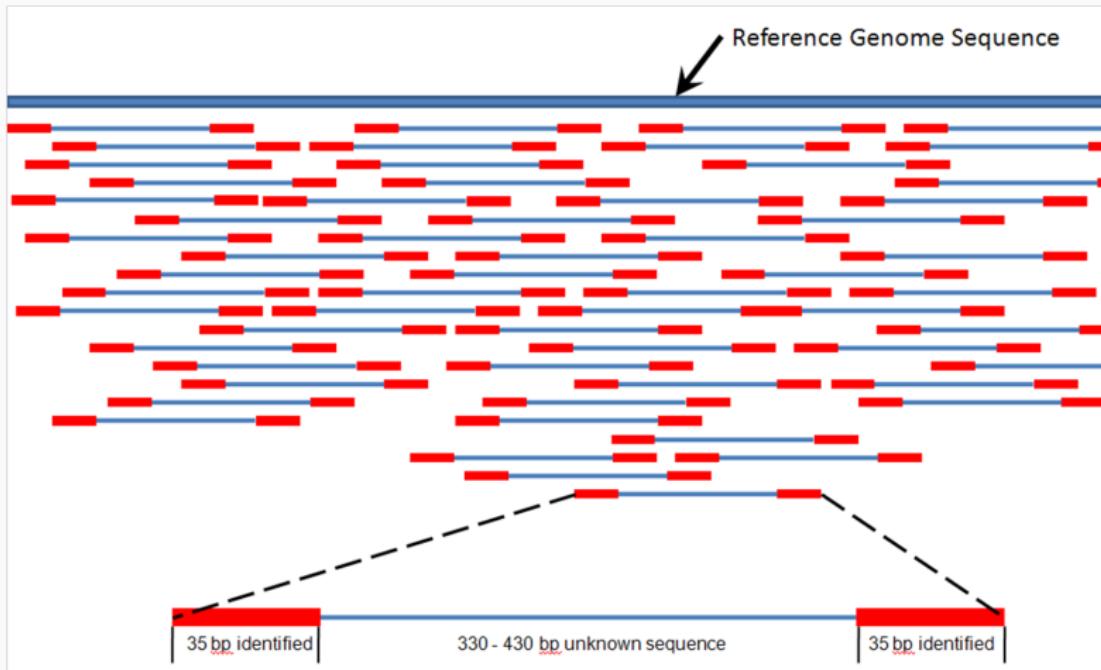


## C de Bruijn Graph



From Schatz, 2010

# Mapping

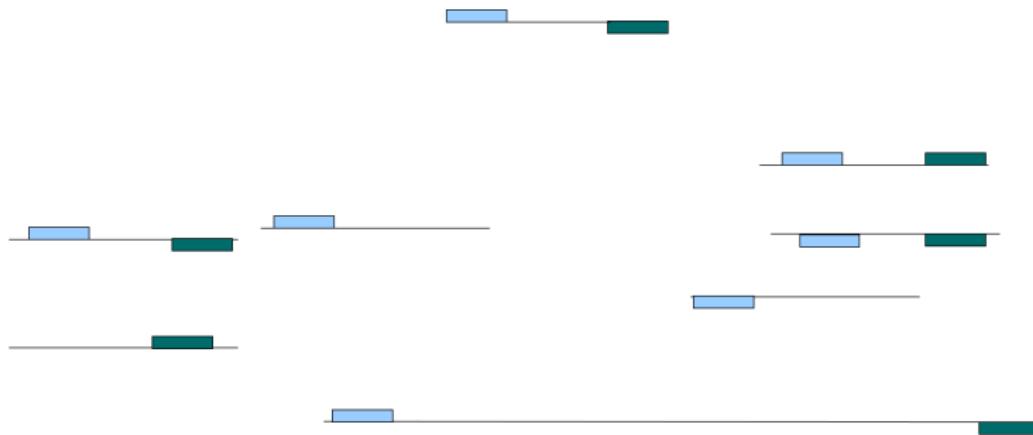


From Wikipedia

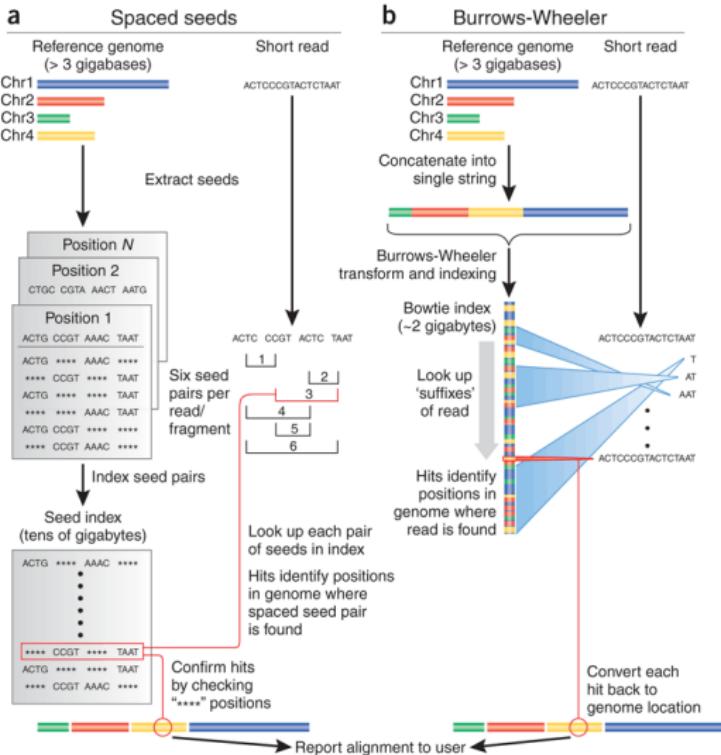
## Generally with **Pair-End** data



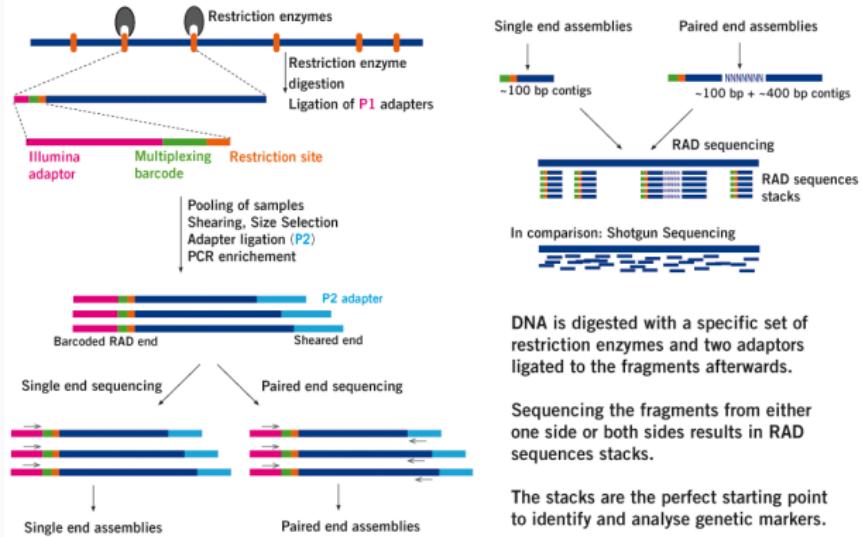
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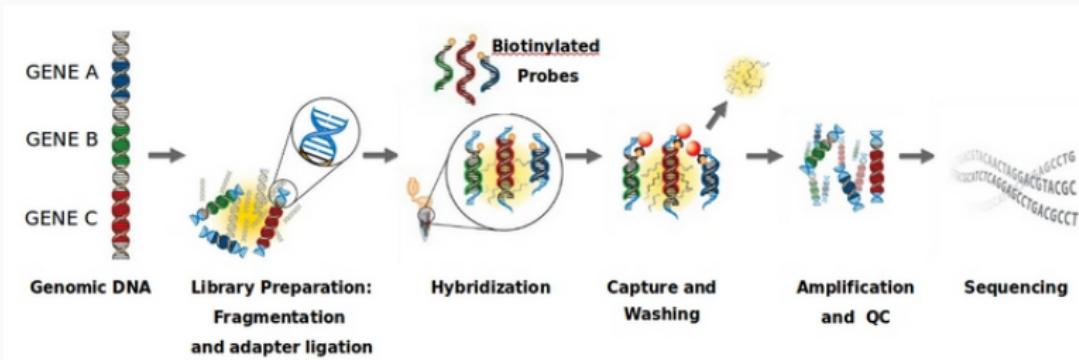
# Mapping



# RAD-Seq



From Eurofins



From CGFB, Bordeaux, France

- Mainly in RNA sequencing, but also in CNV (Copy Number Variation)

# Quantification

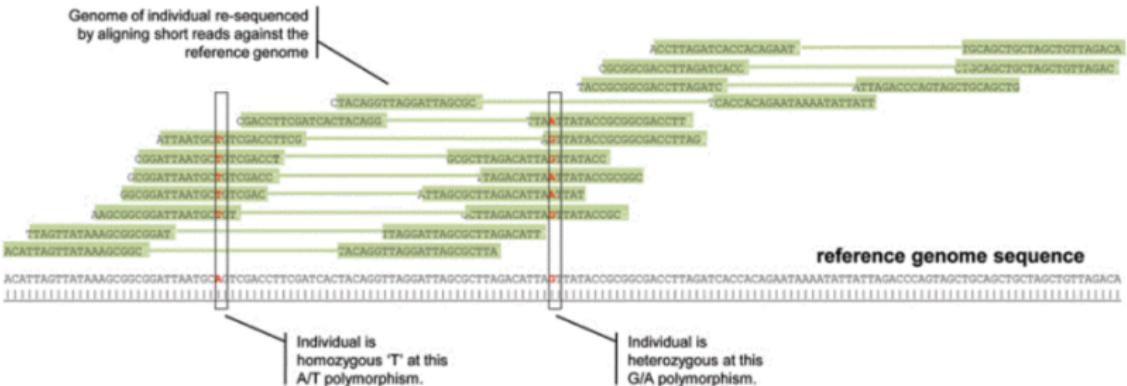
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- Counting the number of reads/bases at each position
- More precise than ChIP
- Need to be reproduced
- Lots of Statistical models and Controls behind

# SNP and InDel Detection



# SNP and InDel Detection

**Types of variants**

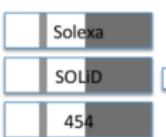
SNPs	Insertions
Alignment ACGT ATGT	Alignment AC-GT ACTGT
VCF representation POS REF ALT 2 C T	VCF representation POS REF ALT 2 C CT

Deletions	Complex events
Alignment ACGT A--T	Alignment ACGT A-TT
VCF representation POS REF ALT 1 ACG A	VCF representation POS REF ALT 1 ACG AT

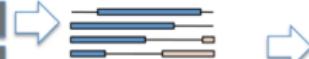
**Large structural variants**

VCF representation  
POS REF ALT INFO  
100 T <DEL> SVTYPE=DEL;END=300

## From unmapped reads to true genetic variation in next-generation sequencing data

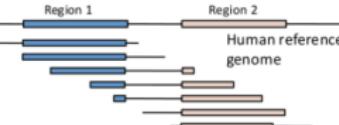


Raw short reads



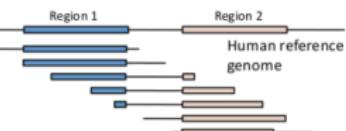
A single run of a sequencer generates ~50M ~75bp short reads for analysis

Mapping and alignment



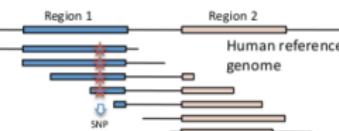
The origin of each read from the human genome sequence is found

Quality calibration and annotation



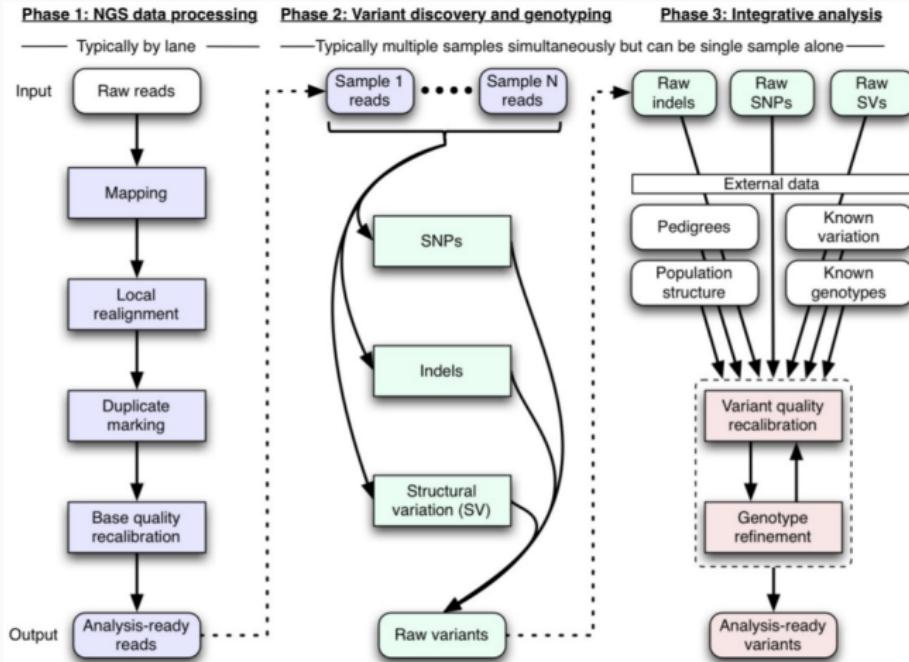
The quality of each read is calibrated and additional information annotated for downstream analyses

Identifying genetic variation

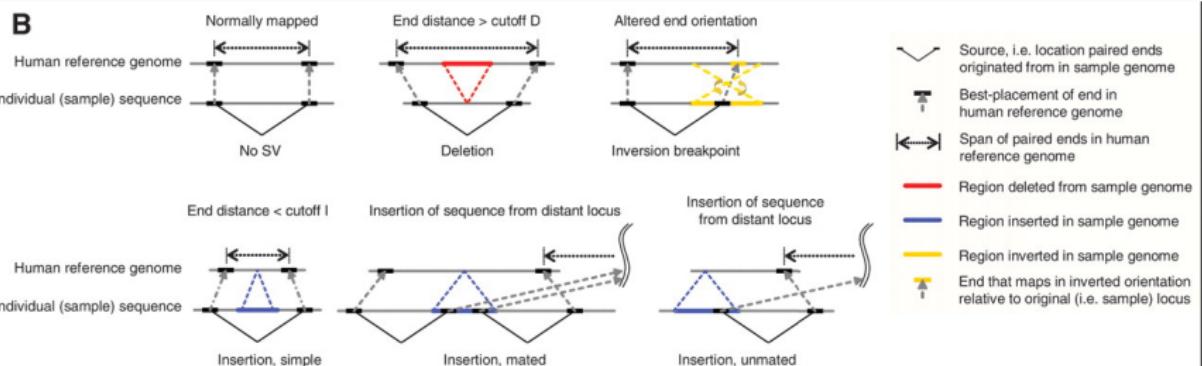


SNPs and indels from the reference are found where the reads collectively provide evidence of a variant

# And even more refined...

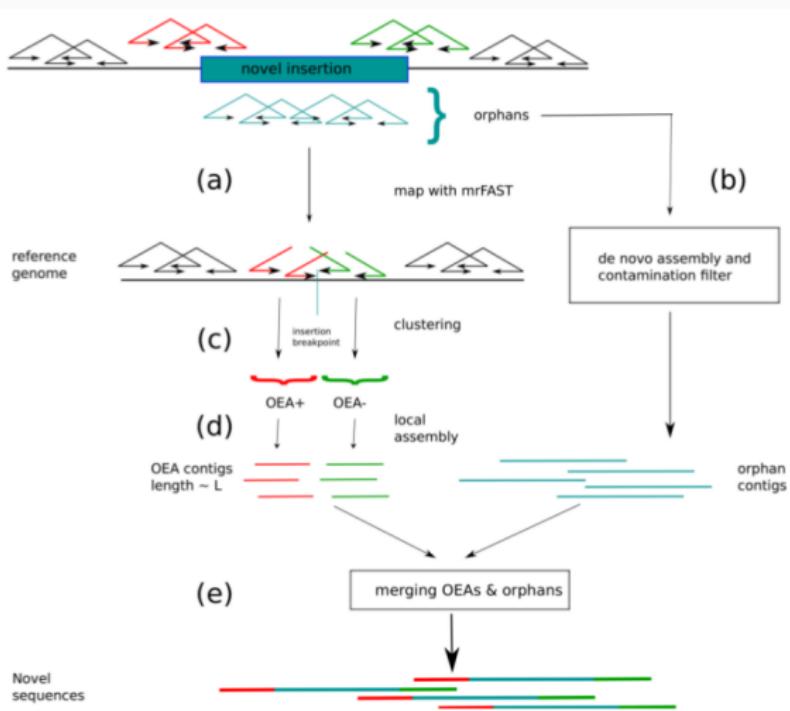


# Structural Variant Detection



From Korbel et al, 2007

# Structural variation, an approach



# Common File for all Variations, the VCF

## Example

VCF header									
<pre>##fileformat=VCFv4.0 ##fileDate=20100707 ##source=VCFtools ##reference=NCBI36 ##INFO=&lt;ID=AA,Number=1,Type=String,Description="Ancestral Allele"&gt; ##INFO=&lt;ID=H2,Number=0,Type=Flag,Description="HapMap2 membership"&gt; ##FORMAT=&lt;ID=GT,Number=1,Type=String,Description="Genotype"&gt; ##FORMAT=&lt;ID=GQ,Number=1,Type=Integer,Description="Genotype Quality (phred score)"&gt; ##FORMAT=&lt;ID=GL,Number=3,Type=Float,Description="Likelihoods for RR,RA,AA genotypes (R=ref,A=alt)"&gt; ##FORMAT=&lt;ID=DP,Number=1,Type=Integer,Description="Read Depth"&gt; ##ALT=&lt;ID=DEL,Description="Deletion"&gt; ##INFO=&lt;ID=SVTYPE,Number=1,Type=String,Description="Type of structural variant"&gt; ##INFO=&lt;ID=END,Number=1,Type=Integer,Description="End position of the variant"&gt;</pre>									
						Mandatory header lines			
						Optional header lines (meta-data about the annotations in the VCF body)			
Body									
<pre>#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT SAMPLE1 SAMPLE2</pre>									
1	1	.	ACG	A,AT	.	PASS	.	GT:DP 1/2:13 0/0:29	
1	2	rs1	C	T,CT	.	PASS	H2;AA=T	GT:GQ 0 1:100 2/2:70	
1	5		A	G	.	PASS	.	GT:GQ 1 0:77 1/1:95	
1	100		T	<DEL>	.	PASS	SVTYPE=DEL;END=300	GT:GQ:DP 1/1:12:3 0/0:20	
<p>Deletion      SNP      Large SV      Insertion      Other event</p> <p>Phased data (G and C above are on the same chromosome)</p> <p>Alternate alleles (GT&gt;0 is an index to the ALT column)</p>									

VCF = Variant Call Format From 1000 Genomes Project

# Which Technology for Which application ?

Application	GS FLX++	GS Junior	HiSeq 2500	MiSeq	PacBio RS
<b>Genome Sequencing</b>					
De novo sequencing of bacterial & fungal genomes	✓✓✓		✓	✓✓	✓
De novo sequencing of higher eukaryotic genomes	✓✓		✓✓✓		✓
De novo sequencing of BACs, viruses & plasmids	✓✓✓	✓✓✓			✓
Resequencing of genomes			✓✓✓	✓✓	
<b>Transcriptome Sequencing</b>					
De novo Transcriptome sequencing	✓✓✓		✓✓	✓✓	
Expression profiling			✓✓✓		
Small RNA sequencing			✓✓✓	✓✓	
ChIP sequencing			✓✓✓	✓✓	
<b>Resequencing &amp; Amplicons</b>					
Ultra deep amplicon sequencing	✓✓✓	✓✓✓	✓	✓	
Resequencing by Sequence Capture	✓✓	✓	✓✓✓		

From Eurofins

# Which Technology for Which application ?

From my own personal Experience:

**Assembly** : Nanopore, PacBio, Illumina (MySeq + HiSeq,  
various libraries)

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**Quantification** : Illumina

- Amount of original samples

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- Size of sequenced unit

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- Error rate

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- Volume of Outputted data

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All linked to technical constraints

- Cleaning data level

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- Mapping Conditions

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- Variation Calling level

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All linked to the Specificity/Sensitivity Informatics Paradox

# Limits, Biological

- Availability of Sample

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- Choice of Sample

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- Choice of Sample
- Amount of Sample

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- Amount of Sample
- Purity of Sample

- Availability of Sample
- Choice of Sample
- Amount of Sample
- Purity of Sample
- Size of sample (for Assembly/Mapping essentially)

## Applications

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- Gene discovery/GWAs

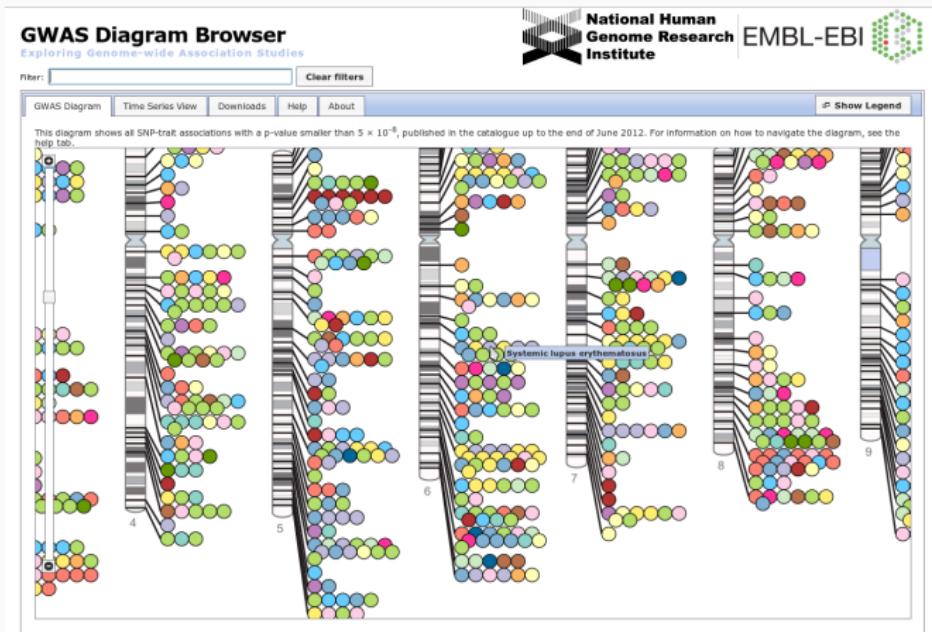
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- Species Definition

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- Species Definition
- Subspecies/specific subgroup definition

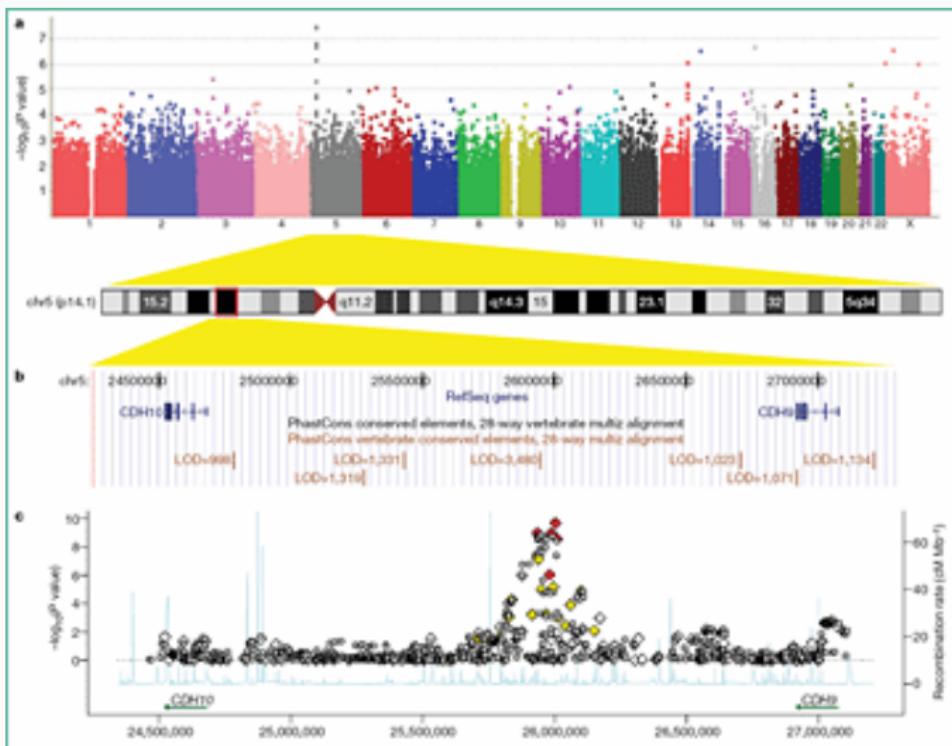
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- Global genotyping (for breeding in agriculture e.g.)

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- Genomic Ecology (Transposable elements, etc...)

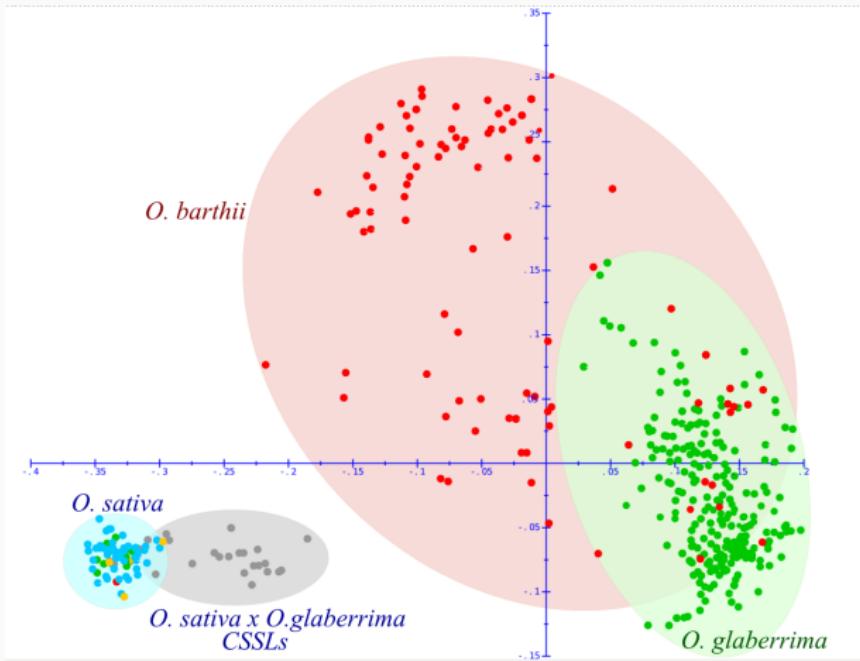
# Example in GWAs & Population Genomics



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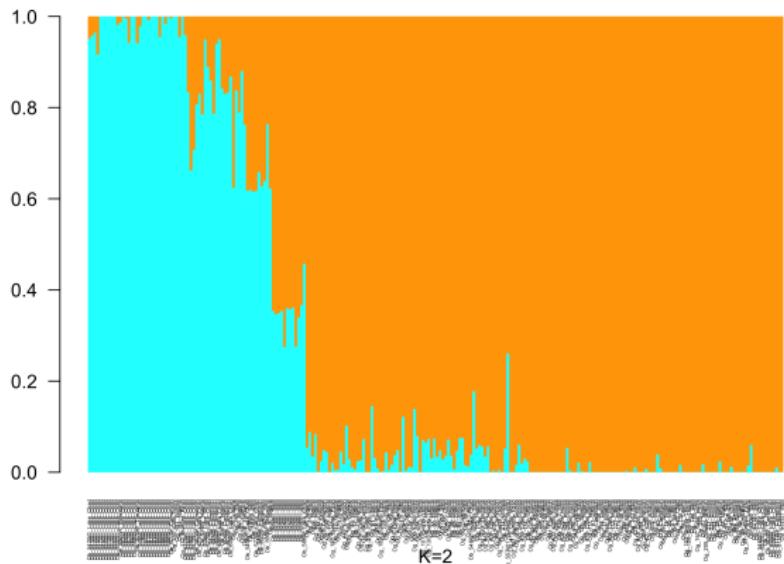
# Example in Global Genotyping & Population Genomics



From Orjuela et al, 2014

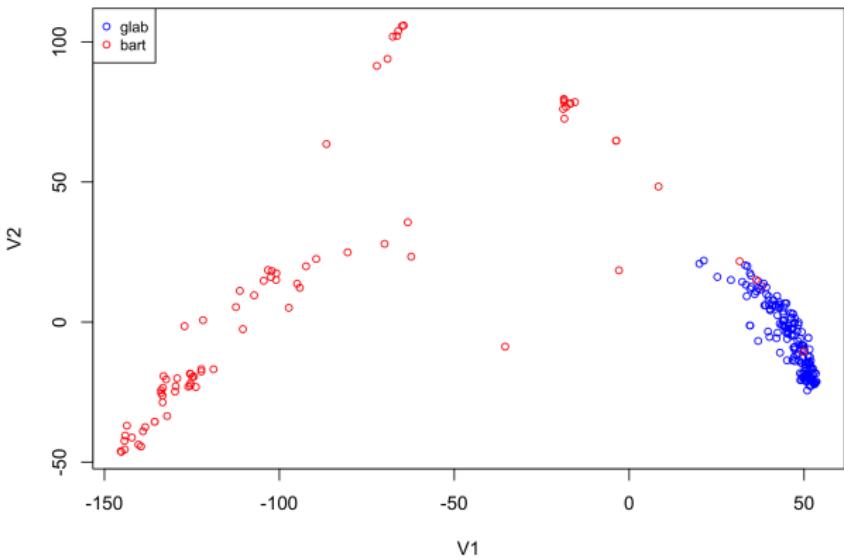
# Example in Global Genotyping & Population Genomics

163 *O. glaberrima* et 83 *O. barthii*



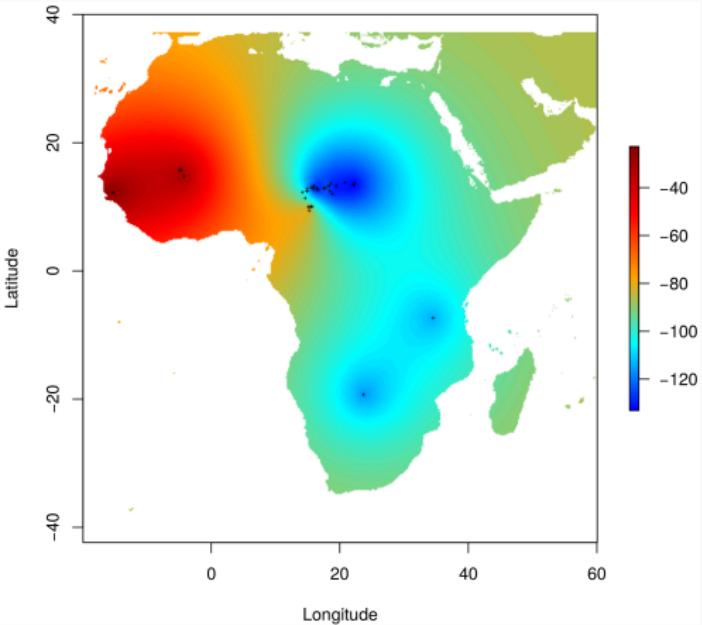
From Cubry et al, 2018

# Example in Global Genotyping & Population Genomics

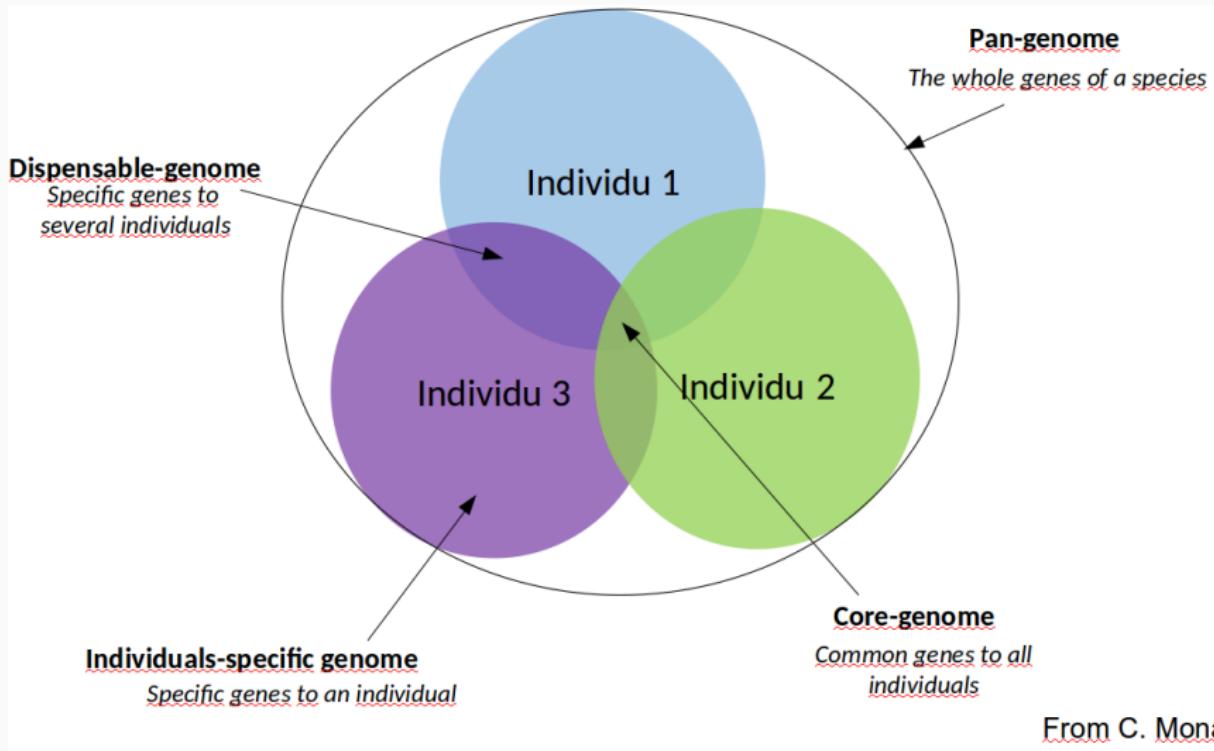


From Cubry et al, 2018

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From Cubry et al, 2018



From C. Monat

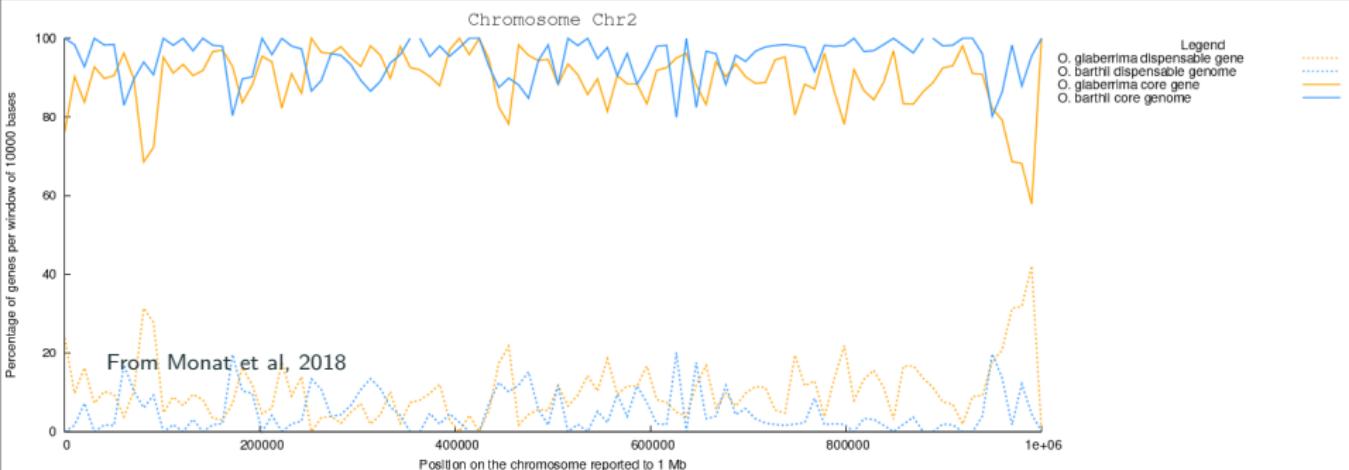
*O. glaberrima*  
current

86,44 %

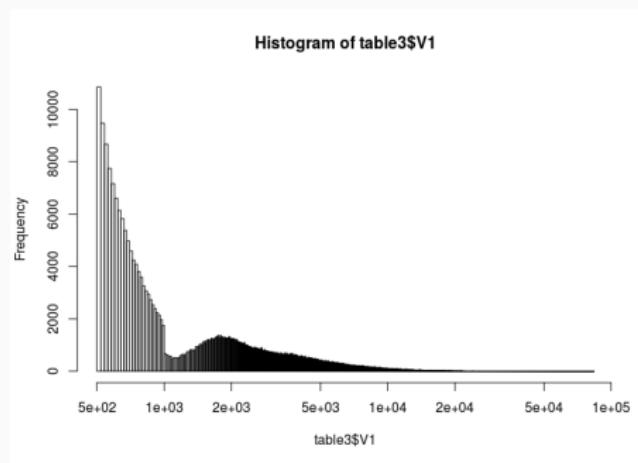
*O. barthii*  
current

98,15 %

From Monat et al, 2016



# Some really recent results...



**Table 3**  
Micro-Collinearity Statistics for CG14 vs. TOG5681

	Valid Scaffolds	Not Valid Scaffolds	Not Referenced Scaffolds
Number of sequences	48223	16672	93
Minimal size	200	201	202
Maximal size	86103	90835	3041
Mean size	4110	6087	447
Median size	1942	2592	320
Number of functionally annotated gene model	10685	2147	2
Number of GO	23634	4817	4

Sizes are given in bp.

From Monat et al, 2017

- Level of expression in different conditions or in different individuals

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- Variation in sequences

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- Detection of putative coding/active sequence

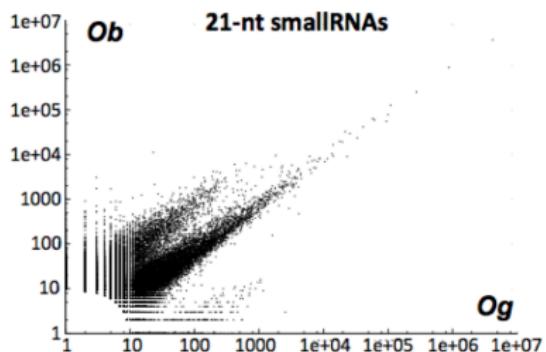
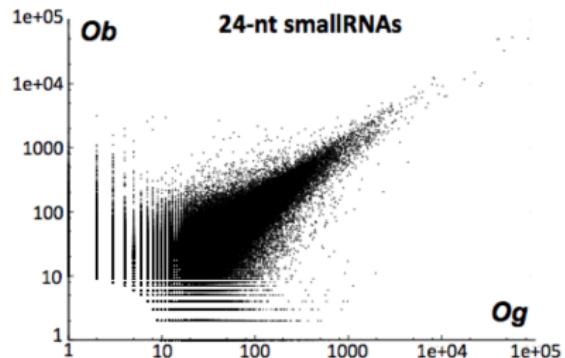
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- Variation in sequences
- Variation in specific forms

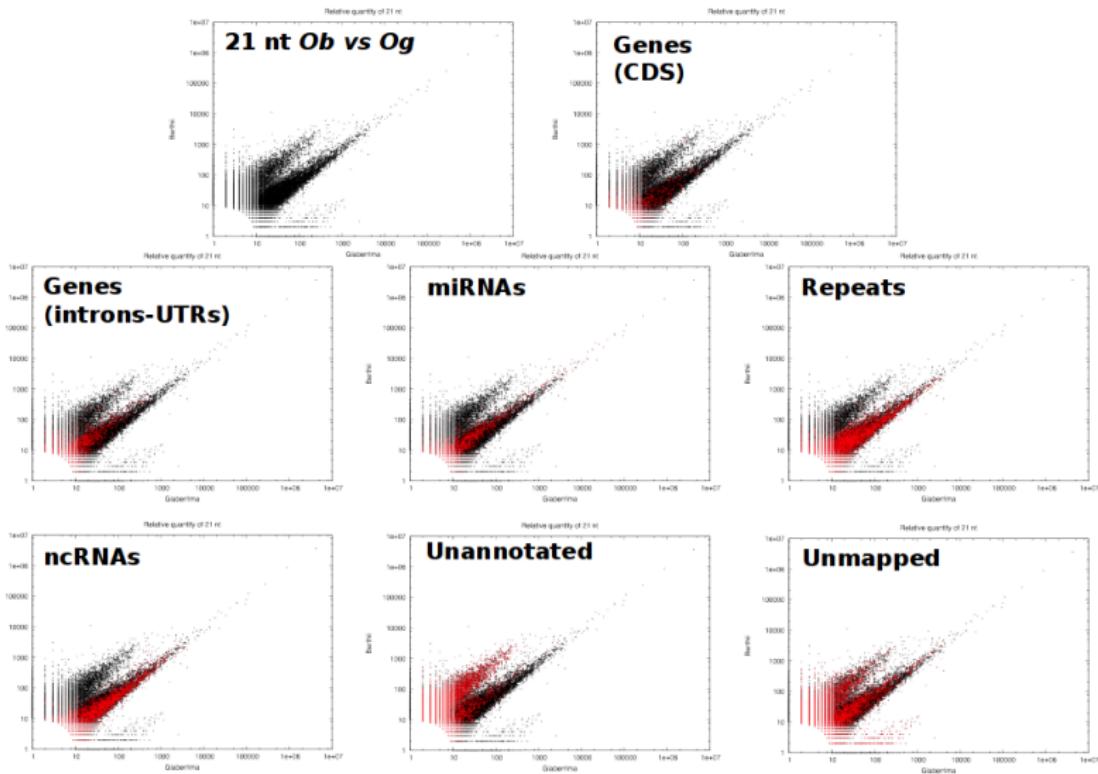
- Level of expression in different conditions or in different individuals
- Variation in sequences
- Variation in specific forms
- Detection of new forms

# Example in smallRNA Transcriptomics

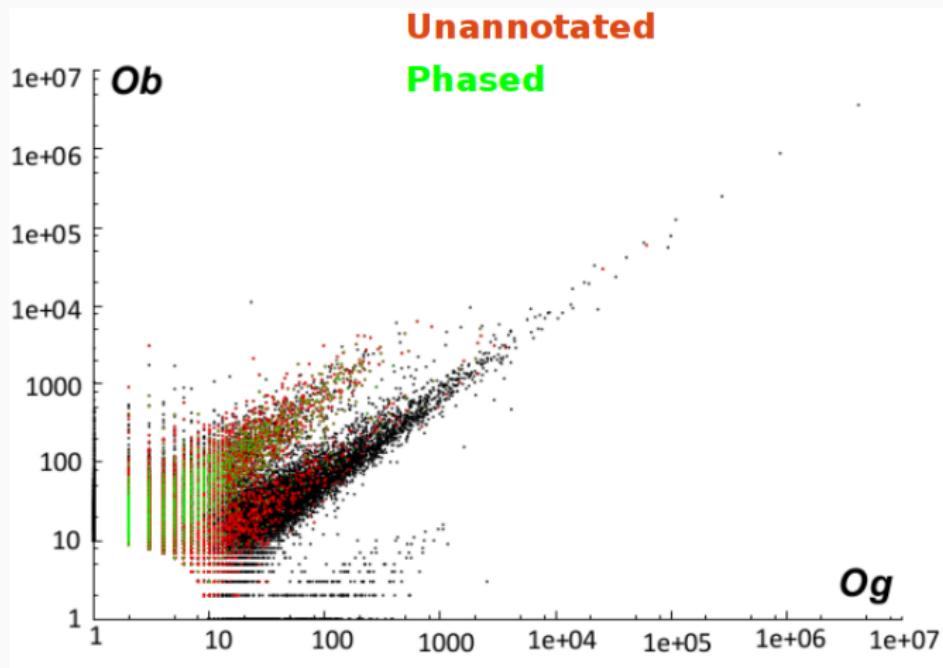


From Ta et al, 2015

# Example in smallRNA Transcriptomics



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From Ta et al, 2015

- Pre-diagnostic (Genetic illness, putative resistance)

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- Risk Assessment
- Epidemiological Studies

## THE METAGENOMICS PROCESS



Extract all DNA from  
microbial community in  
sampled environment



### DETERMINE WHAT THE GENES ARE

#### (Sequence-based metagenomics)

- Identify genes and metabolic pathways
- Compare to other communities
- and more...

### DETERMINE WHAT THE GENES DO

#### (Function-based metagenomics)

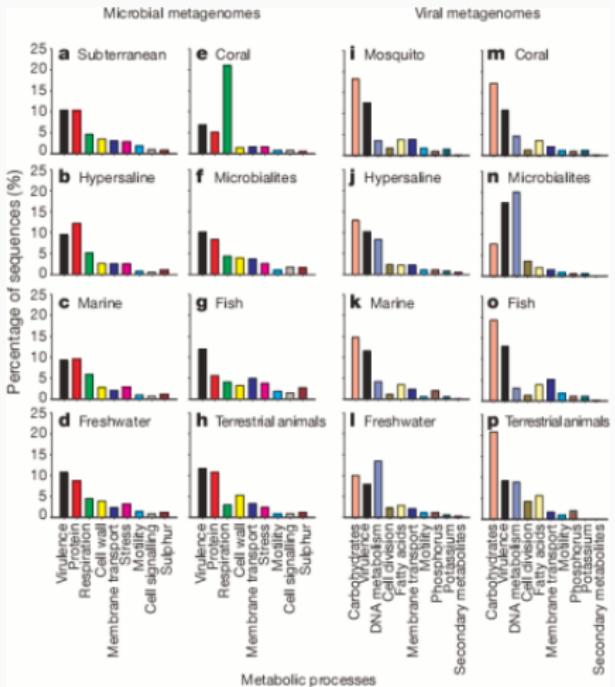
- Screen to identify functions of interest, such as vitamin or antibiotic production
- Find the genes that code for functions of interest
- and more...

# Large Metagenomic assays



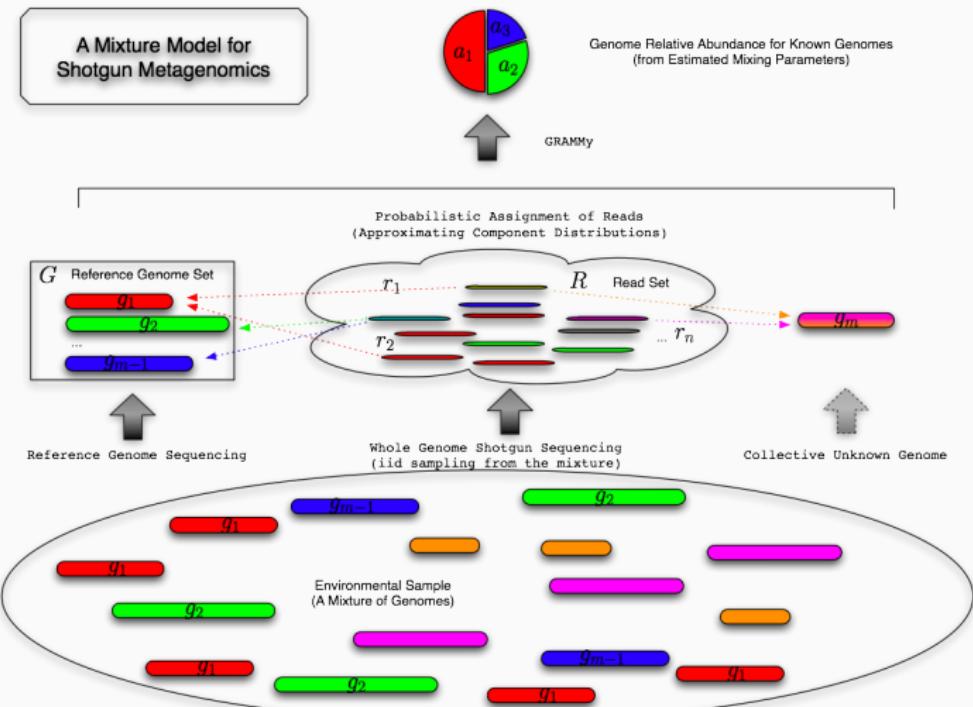
From Tara Ocean website

# Functional Metagenomics



From Dinsdale et al, 2008

# Barcoding



# Large Projects

**1000 Genomes**  
A Deep Catalog of Human Genetic Variation

Home About Data Analysis Participants Contact

LATEST ANNOUNCEMENTS

WEDNESDAY FEBRUARY 16, 2011

**February 2011 Data Up**  
**Full Project Indel Release**

Indels calls from [Dindel](#). These calls genome project. This release is ba

Data access links: [EBI](#) / [NCBI](#)

**1001 Genomes**  
A Catalog of *Arabidopsis thaliana* Genetic Variation

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Welcome to the 1001 Genomes Project

Links

Database & Species lists News Events Publications Participants For G10K Organizers (restricted)

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**GENOME 10K**  
Unveiling animal diversity

**Genome 10K Project**

To understand how complex animal life evolved through changes in DNA and use this knowledge to become better stewards of the planet.

April 2009—The Genome 10K project aims to assemble a genomic zoo—a collection of DNA sequences representing the genomes of 10,000 vertebrate species, approximately one for every vertebrate genus. The trajectory of cost reduction in DNA sequencing suggests that this project will be feasible within a few

Join us

Become a G10K affiliate

Genome assembly

# Possibilities in the next 5-10 years (From a presentation in 2013)



- Real-time Transcriptomics

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- Real-time Transcriptomics
- Single-Cell Genomics -> DONE in 2014
- Single-Cells Transcriptomics (and smallRNA) -> DONE in 2015
- Personal Genomics medicine (ethical problems...) -> Available
- And any new ideas you will have...

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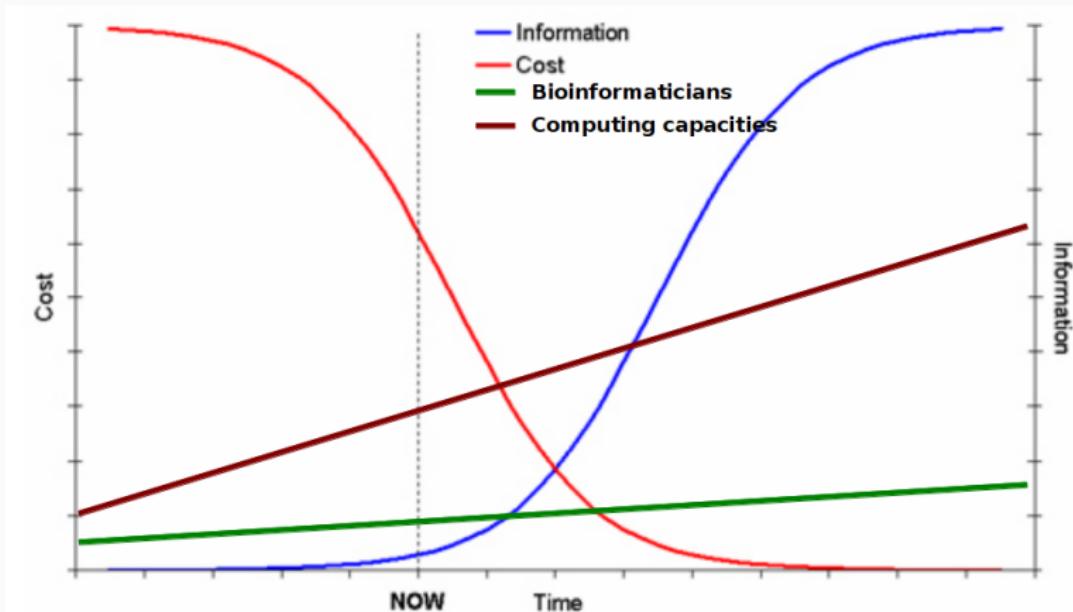
- NGS technologies change the way of abording Biology

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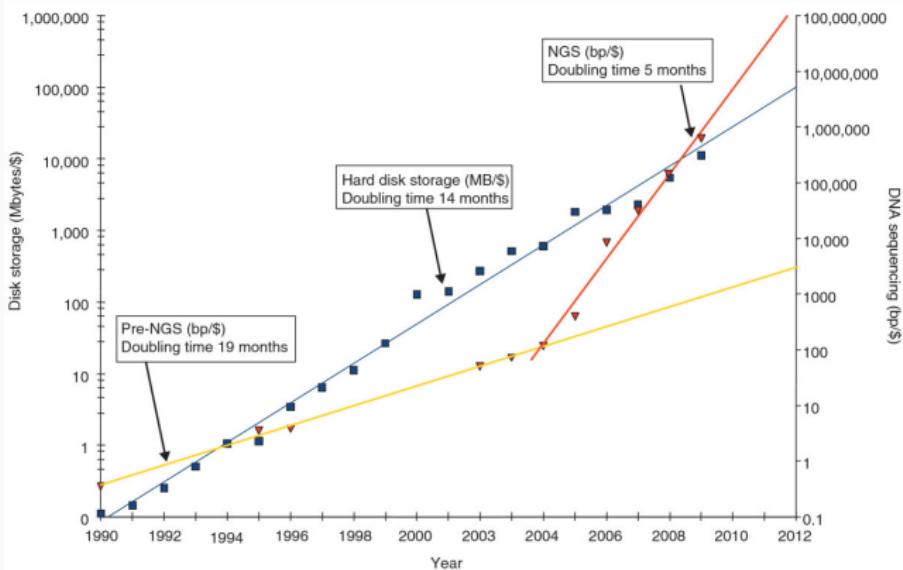
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- A lot of Possibilities, a lot of limits

- NGS technologies change the way of abording Biology
- A lot of Possibilities, a lot of limits
- The main limit is no more Sequence, but Sample acquisition and Data treatment

# Keep in mind!

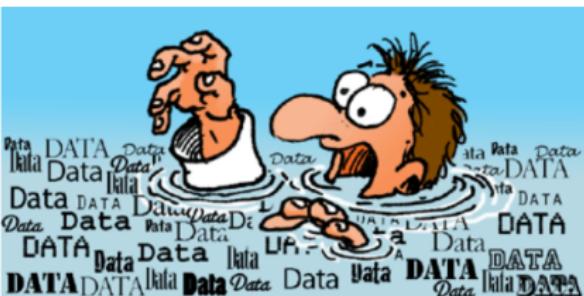


# ...From Data Rarity to Data Deluge



From L. Stein, 2010

# Be Careful to data drowning!



Thanks for your attention

