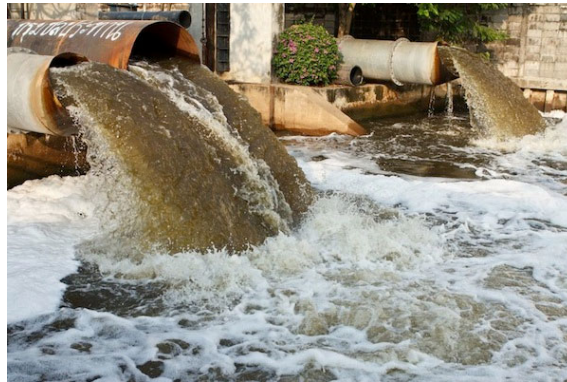


An introduction to metagenomics

Microbes are everywhere



<http://npic.orst.edu/envir/soil.html>



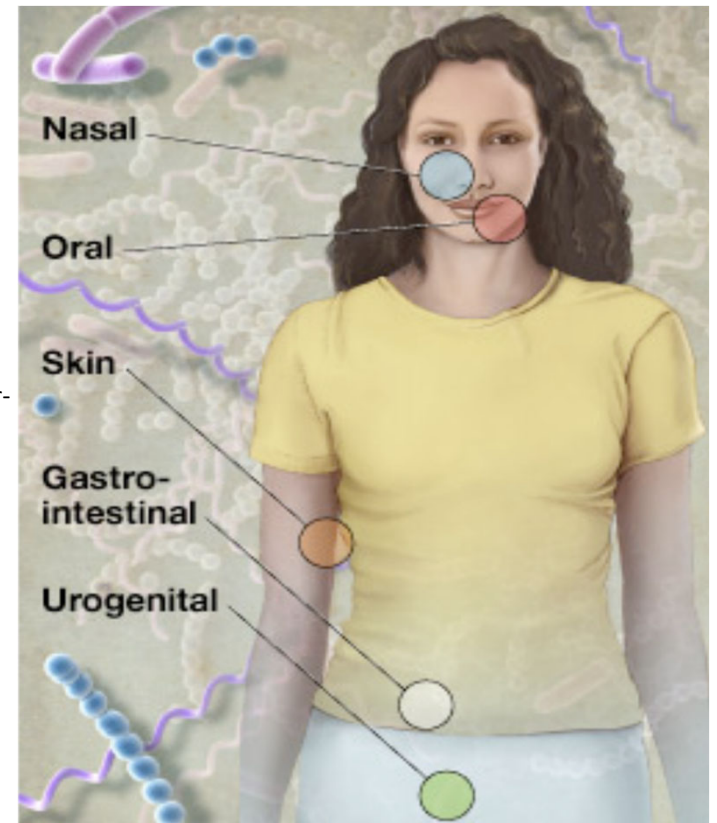
<http://www.rimpro-india.com/articles1/waste-water-treatment-technologies-and-techniques.html>



<http://ocean.nationalgeographic.com/ocean/>



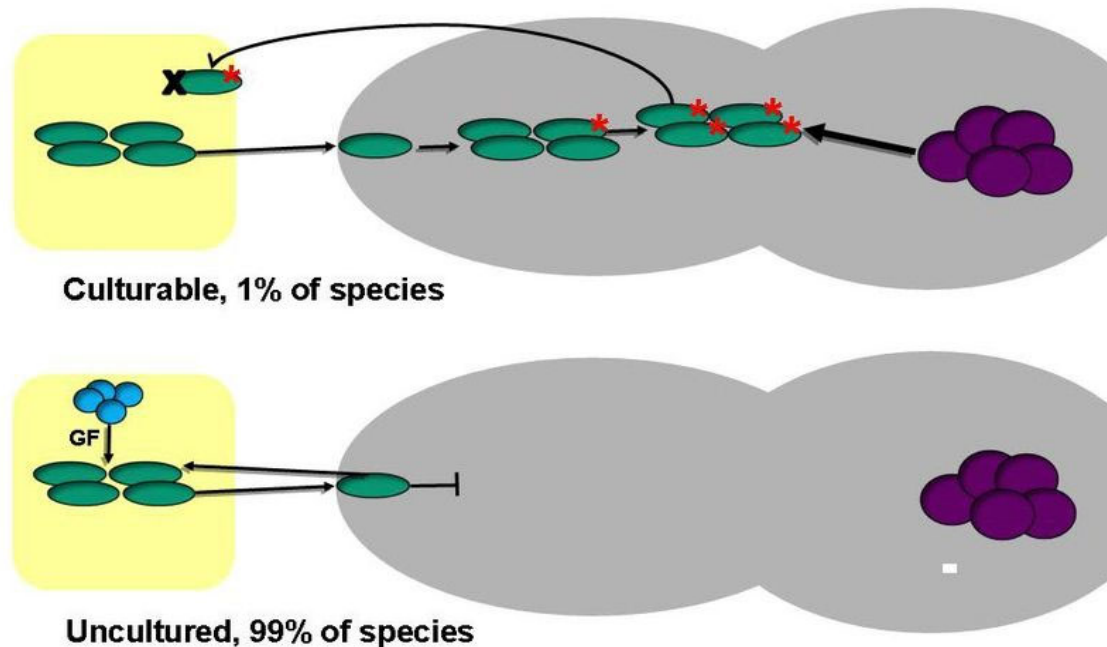
<https://en.wikipedia.org/wiki/Desert>



<https://www.bcm.edu/departments/molecular-virology-and-microbiology/research/the-human-microbiome-project>

Culture-independent methods are indispensable

more than 99% of organism genomes in the environment are **uncultured**



<http://schaechter.asmblog.org/schaechter/2010/07/the-uncultured-bacteria.html>

Metagenomics, the key to the microbial world

Metagenomics is the study of genetic material recovered directly from environmental samples.

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...

<http://blog.openhelix.eu/?p=431>

Early history of metagenomics

Norman R Pace propose the idea of cloning DNA directly from environmental samples as early as 1985.

Pace published their study that isolated and cloned bulk DNA from an environmental sample in 1991.

Edward DeLong laid the groundwork for environmental phylogenies based on signature 16S sequences in 1996

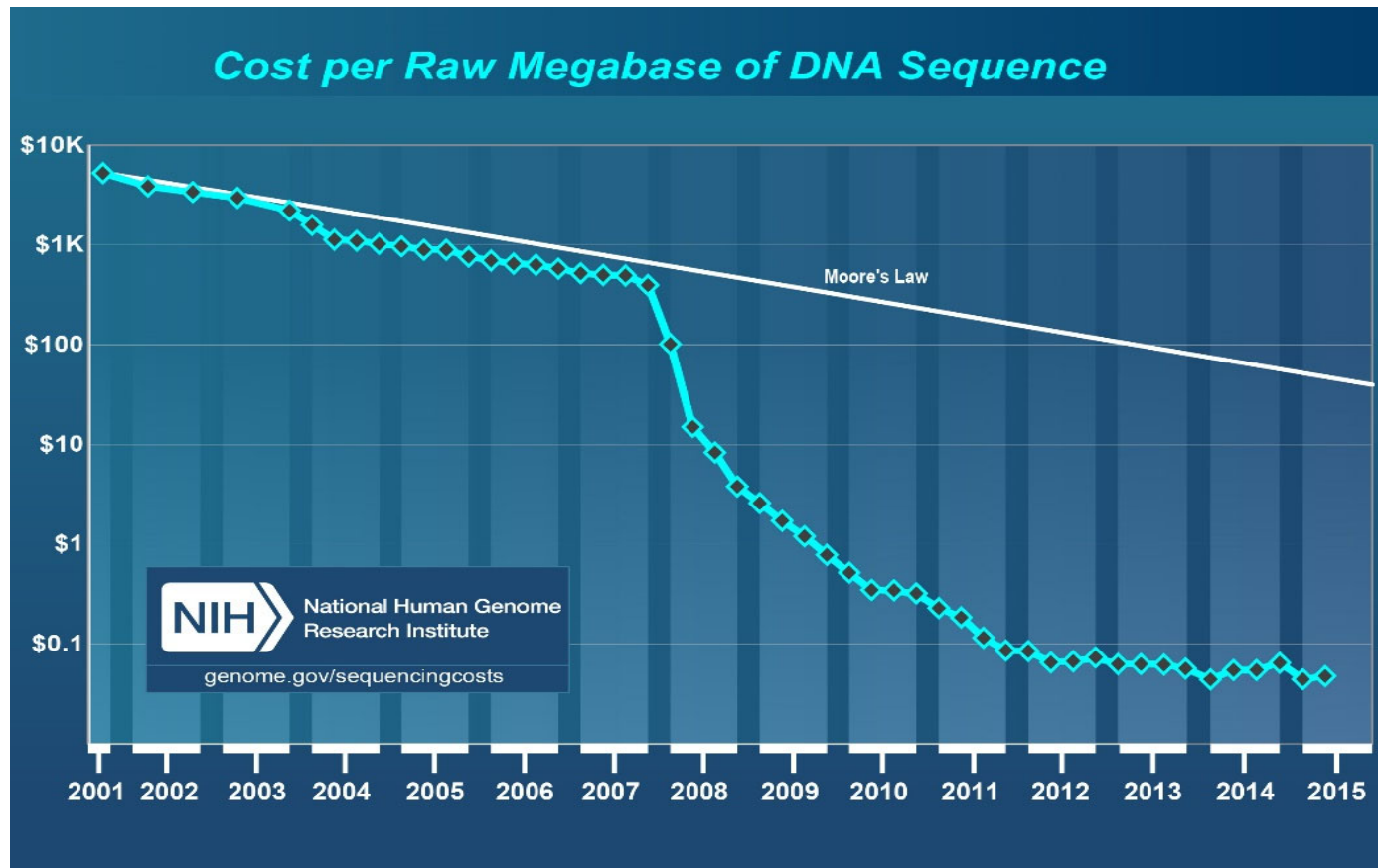
In 2002, Mya Breitbart and colleagues used environmental shotgun sequencing to show that 200 liters of seawater contains over 5000 different viruses.

In 2004, Gene Tyson et al acid mine drainage system; Craig Venter et al Global Ocean Sampling Expedition.

In 2005 Stephan C. Schuster et al studied mammoth DNA by 454 sequencer.



Sequencing cost reduced rapidly



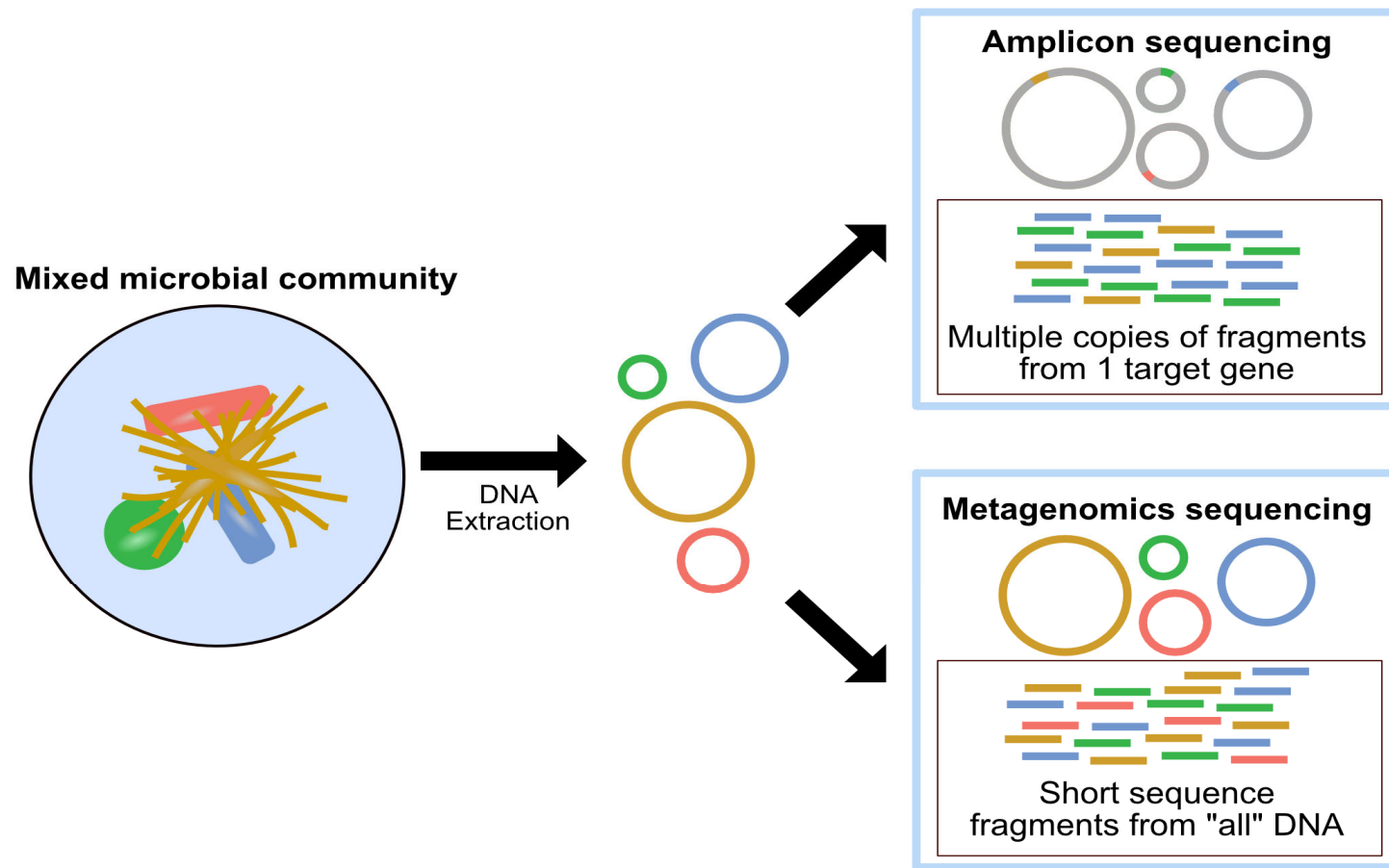
Images from <http://www.genome.gov/sequencingcosts/>

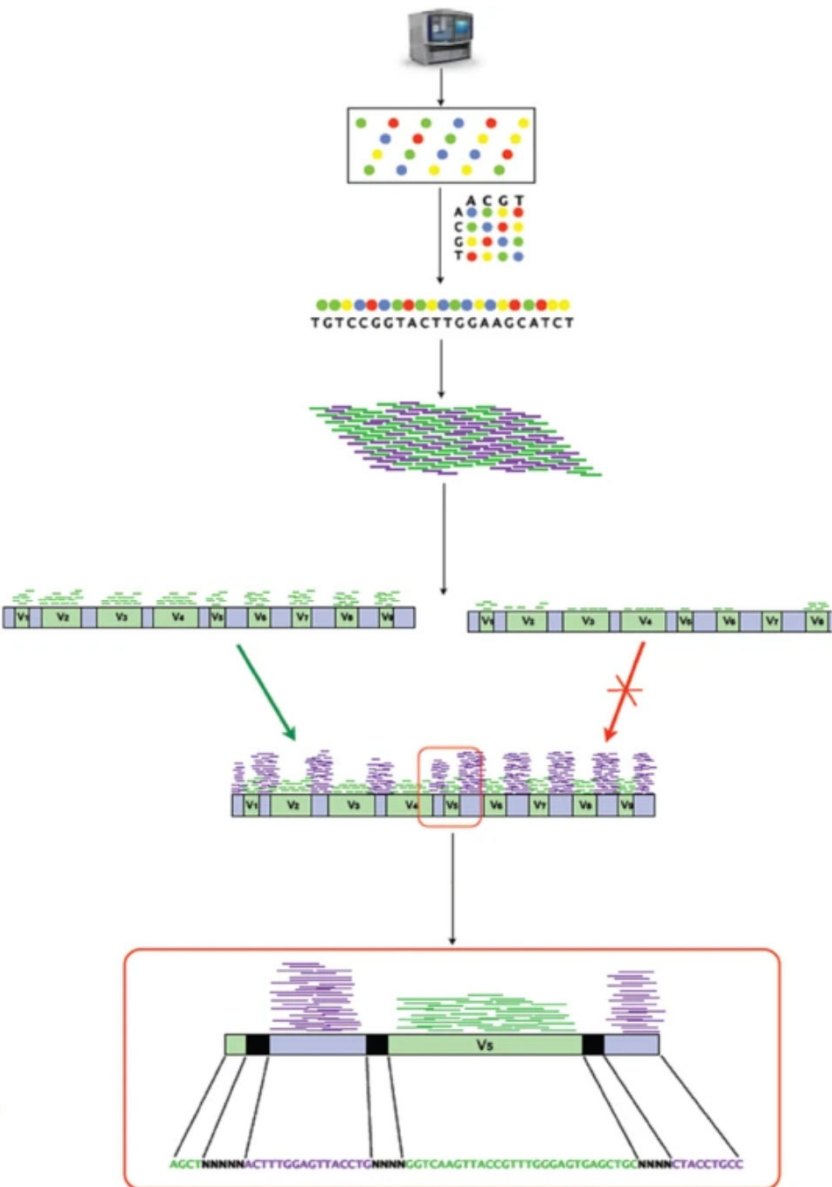
The almost there future

A human genome sequence

- 2000
€ 1,000,000,000 *in ~10 years*
- 2008
€ 50 - 100,000 *in ~4 months*
- 2010
€ 5 - 10,000 *in ~2 weeks*
- ...2015
€ 1,000 *in ~1 day*
- ...2020?
€ 10 *in ~1 hour to minutes*

Two types of sequencing





(a) Amplicon Sequencing

(b)

(c) Which marker genes?

(d) Variable regions

(e)

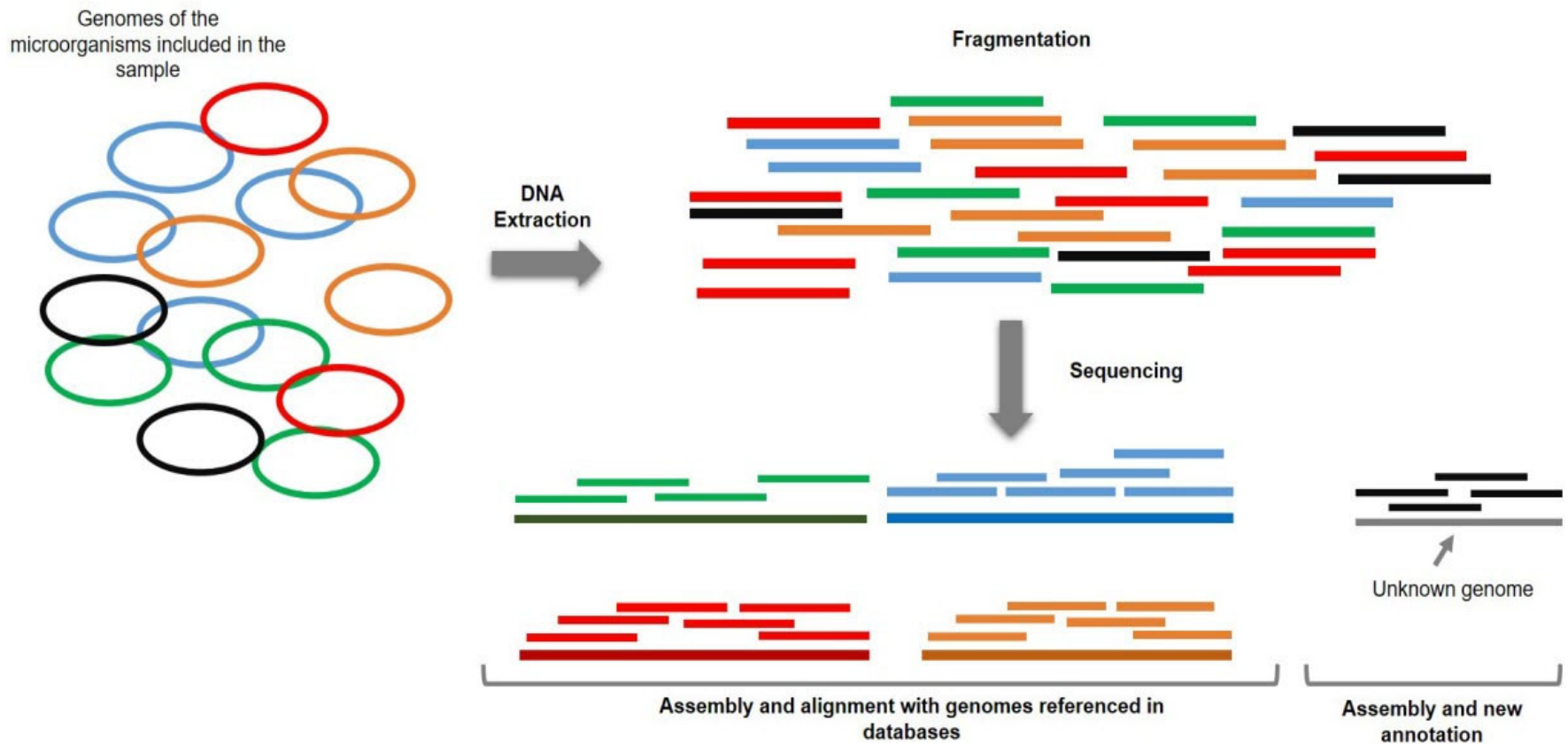
Neither universal nor unique

Three groups of widely used marker genes

	Mean			SD			Minimum			Median			Maximum		
	C	W	A	C	W	A	C	W	A	C	W	A	C	W	A
fetchMG	99.0	98.6	99.0	0.5	2.7	0.8	97.7	83.9	95.9	99.3	99.3	99.3	99.6	99.7	99.6
BLAST	96.1	88.9	94.2	3.0	7.0	4.5	87.1	58.9	76.7	96.7	90.9	96.0	99.7	96.8	99.0
universal	100	99.8	99.9	3.4	9.4	5.3	99.7	93.4	99.0	100	99.9	100	100	100	100
unique	98.5	97.3	98.2	1.0	3.9	2.2	95.4	82.6	89.0	99.0	99.1	99.1	99.5	99.6	99.5

The subcolumns with the name C, W, and A refer to the corresponding percentage for the USCGs from Creevey et al., Wu et al., and Alneberg et al., respectively.

Metagenomic shotgun sequencing



Shotgun reads in metagenomics

Sanger sequencing ~1000bp long reads, 99.99%

454 pyrosequencing typically produces ~400 bp reads

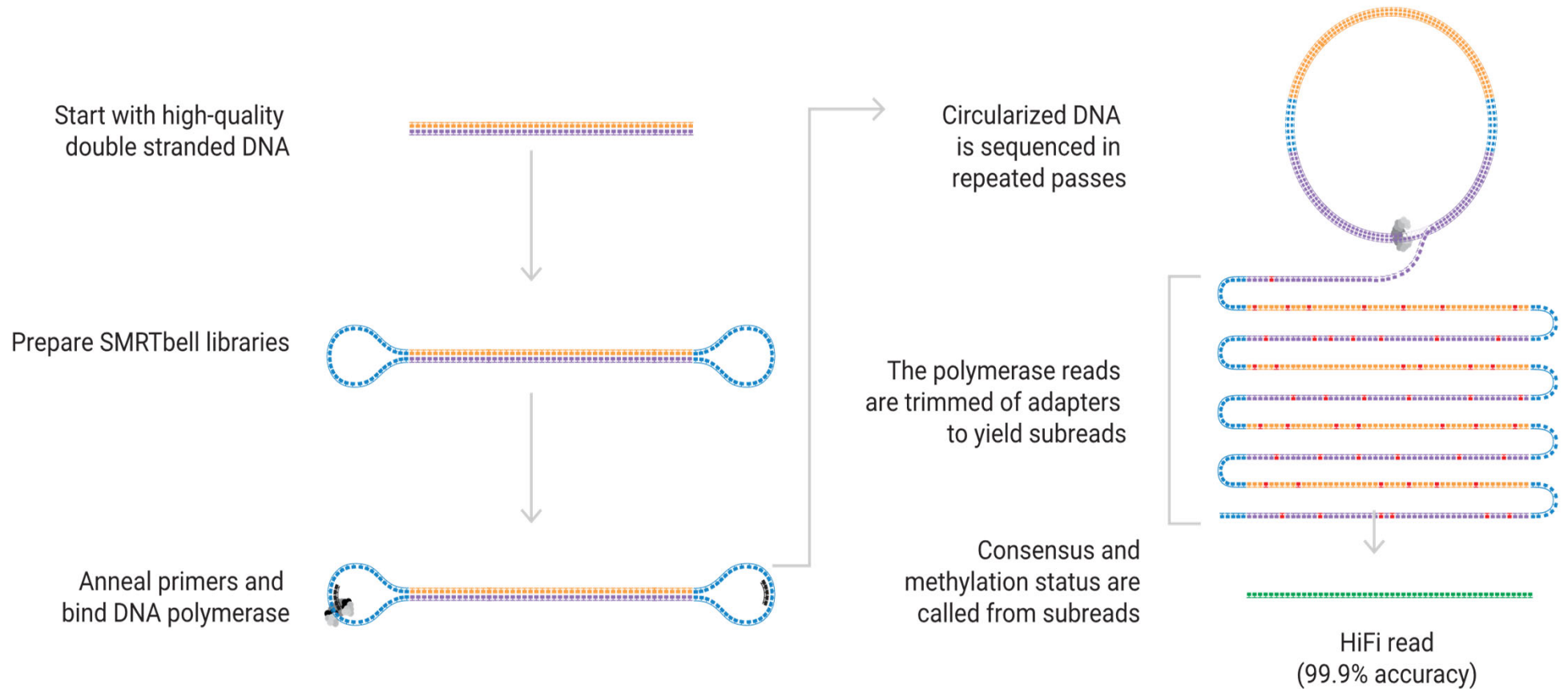
Illumina MiSeq produces 200-600bp reads (depending on whether paired end options are used),

Ion Torrent PGM System ~400 bp reads

Pacific bioscience 10 to 25kb, 99.9% accuracy


Oxford Nanopore minION flow cell R10.4, 10-25kb, 99% accuracy

HiFi reads from PacBio



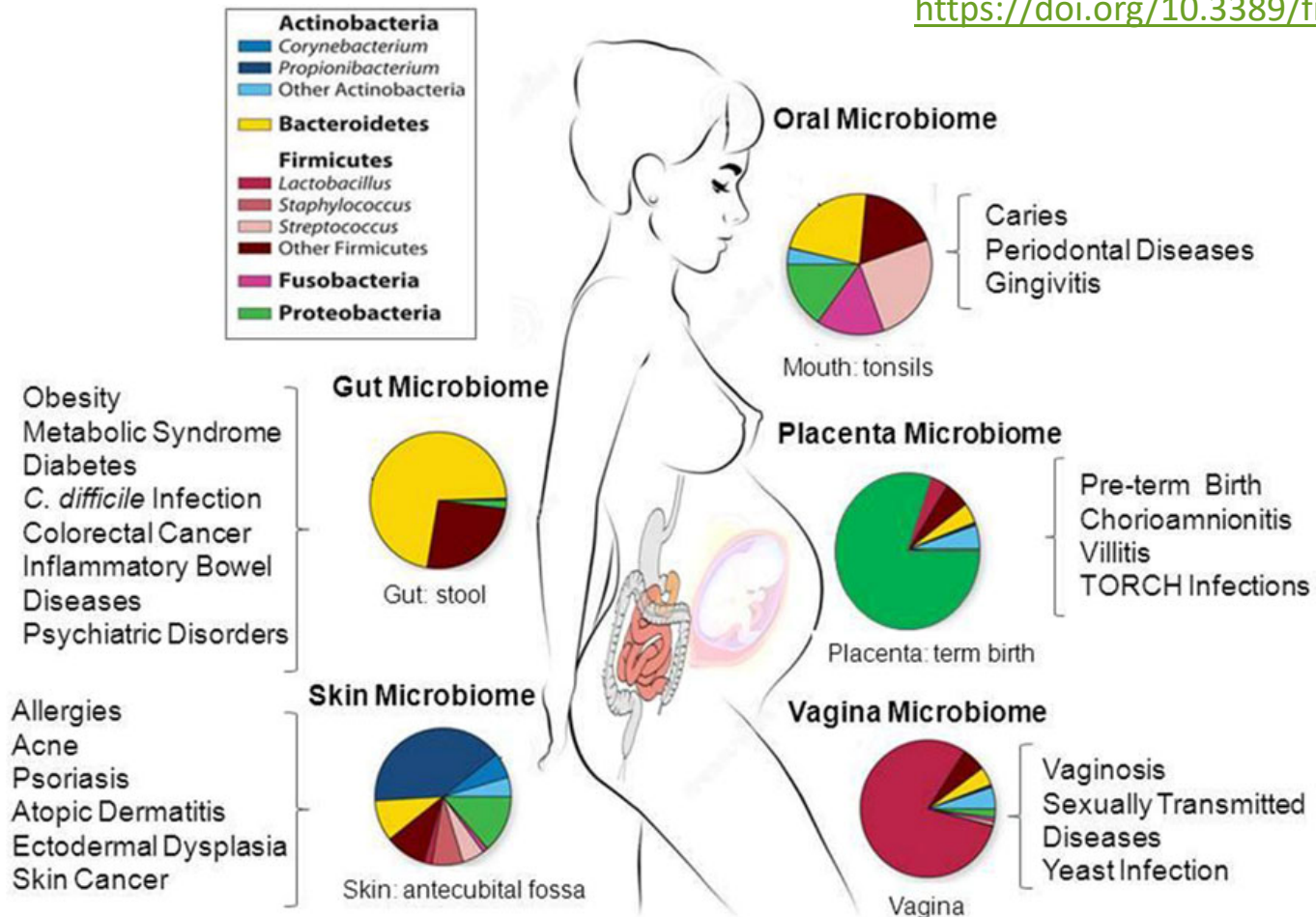
A simple math

How many species can we sequence in a instrumental run of Illumina sequencing (800 Gb)?

1. Each species is 1Mb long, coverage 1X
 2. Each species is 3Mb long, coverage 100X
- 
- A solid green horizontal bar spanning the width of the slide, located at the bottom.

The Human Microbiome

<https://doi.org/10.3389/fmicb.2015.01050>



What does the data look like

Mixed
genomes



Mixed
reads



What computational questions can be asked with the data

Mixed
genomes



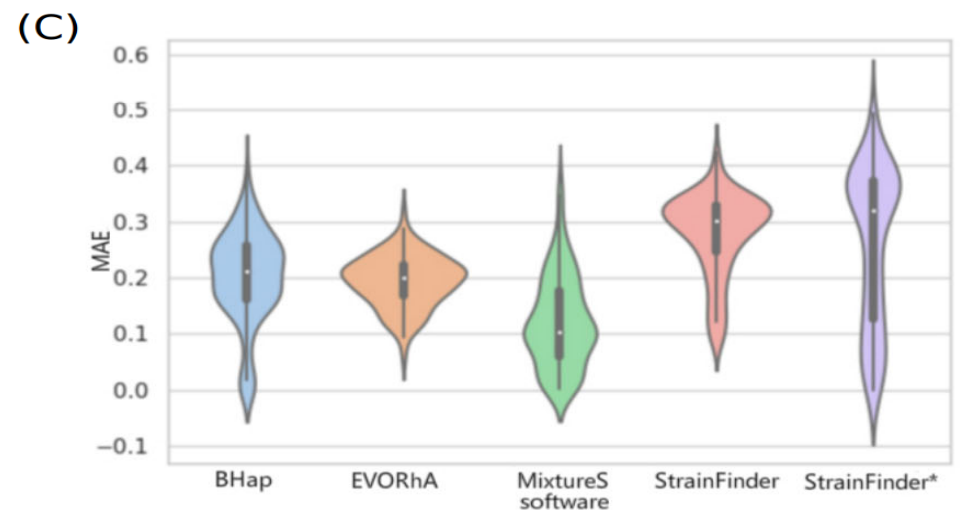
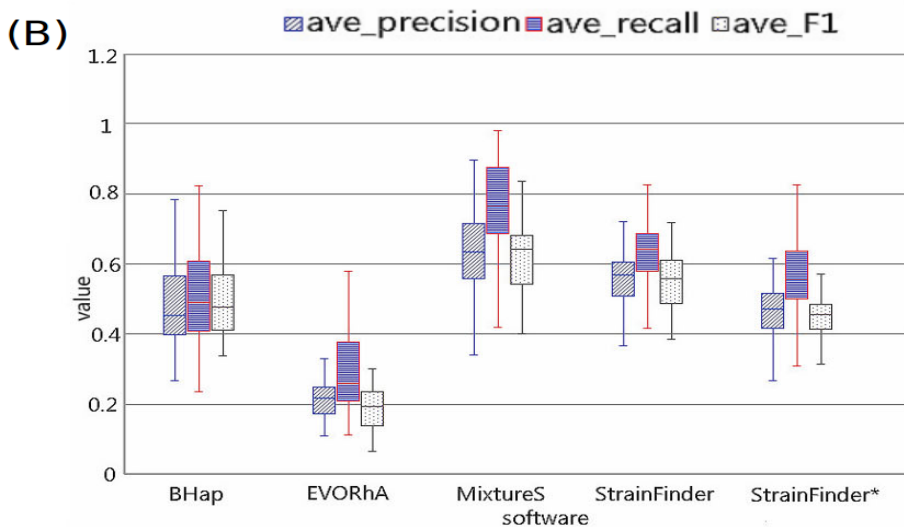
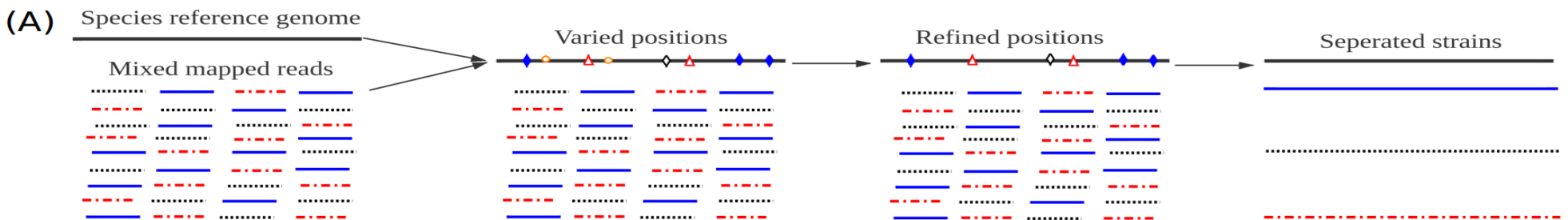
Mixed
reads



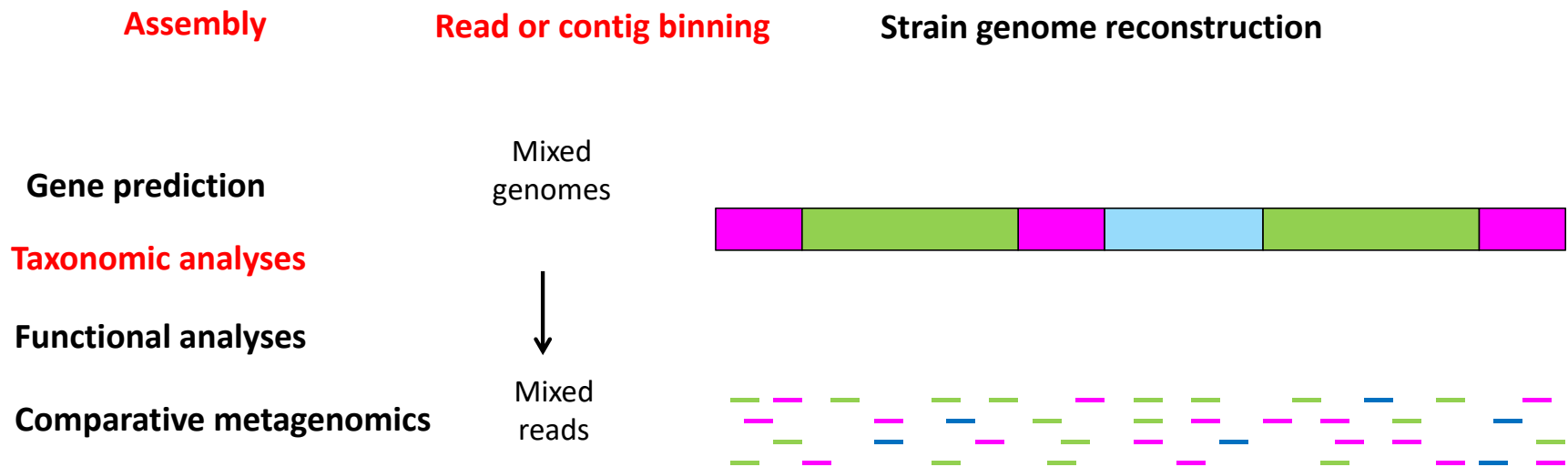
What computational questions can be asked with the data



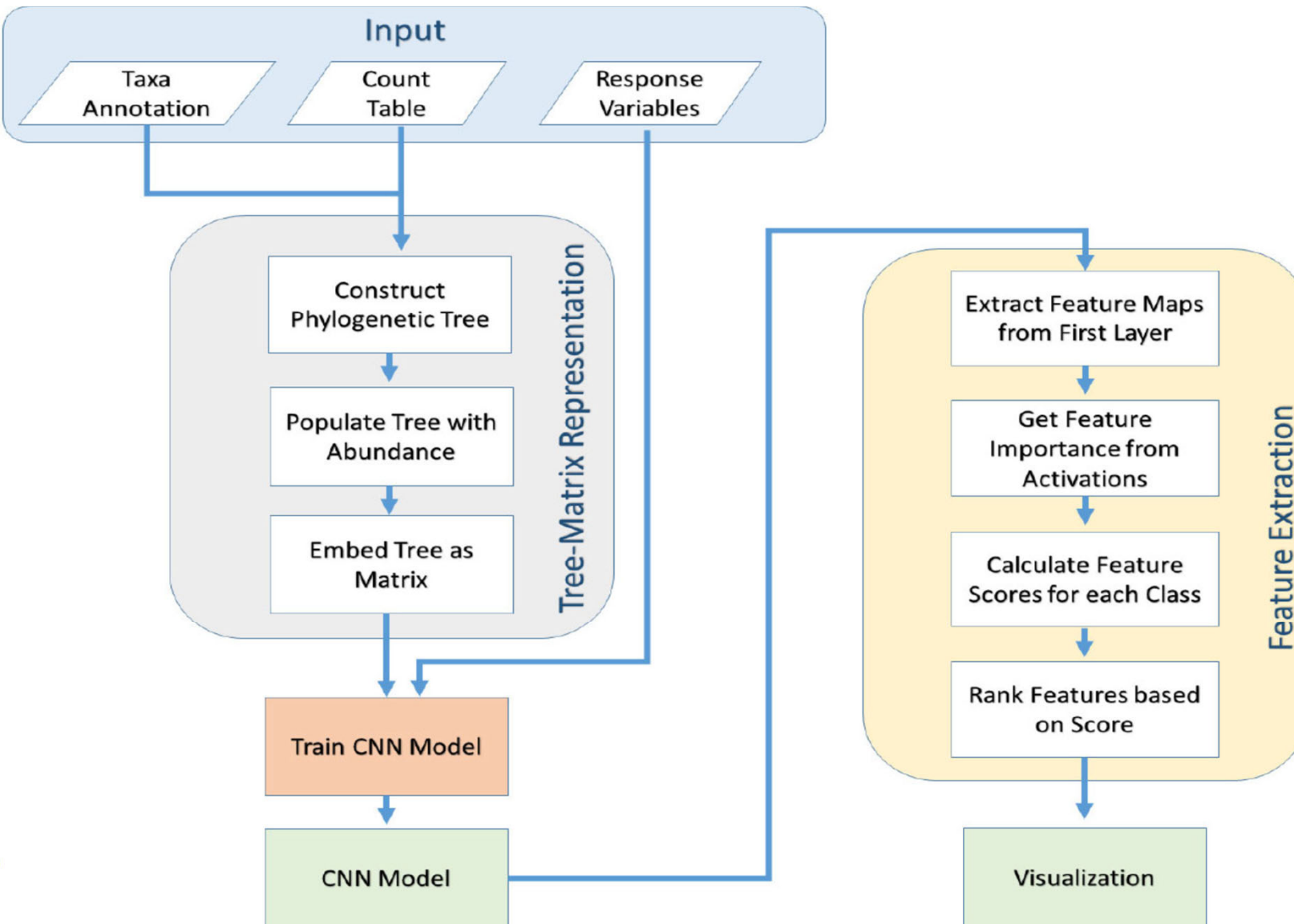
Strain genome reconstruction



Other questions can be asked ?



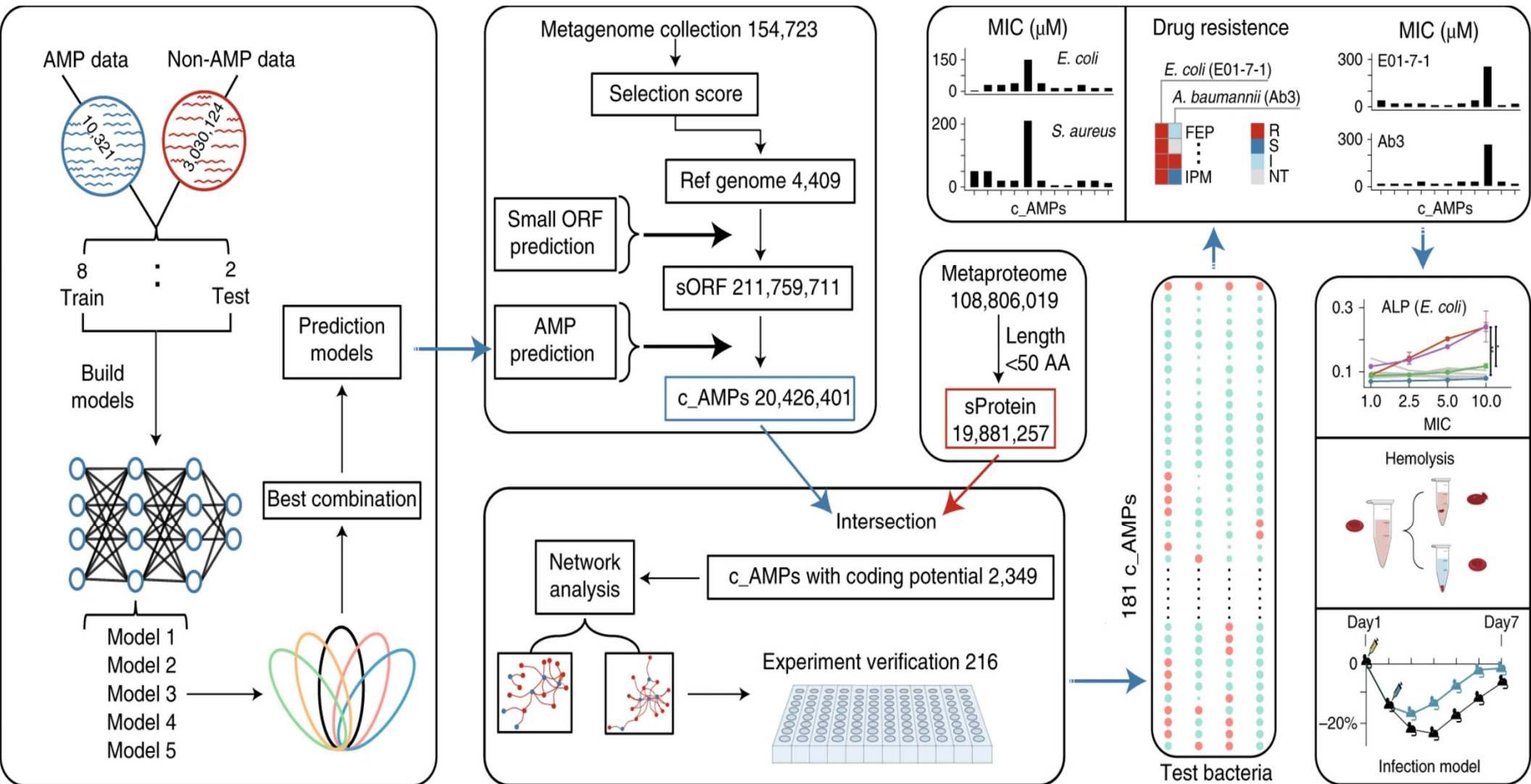
Disease status prediction



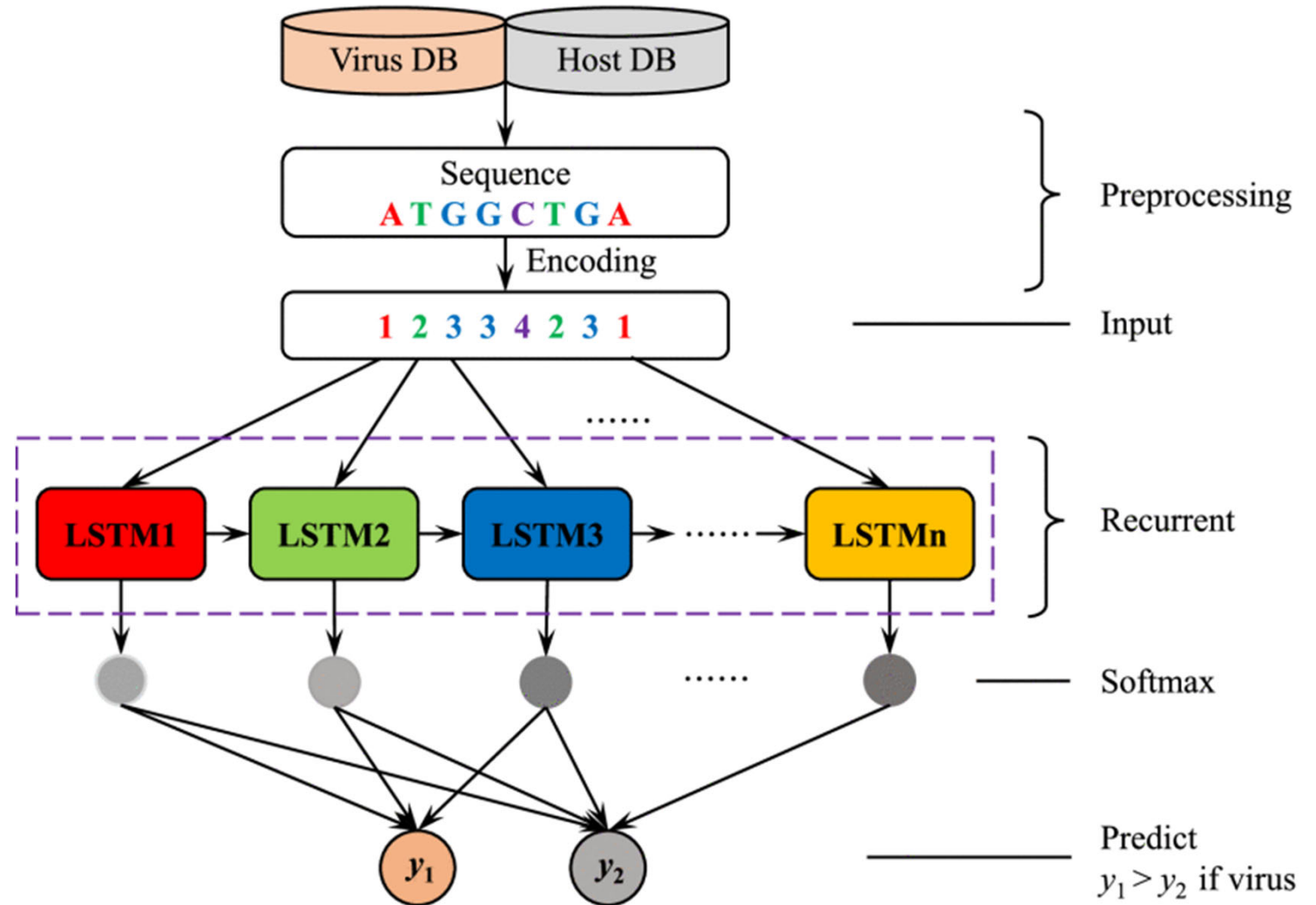
<https://www.biorxiv.org/content/10.1101/257931v1.full>

antimicrobial peptides

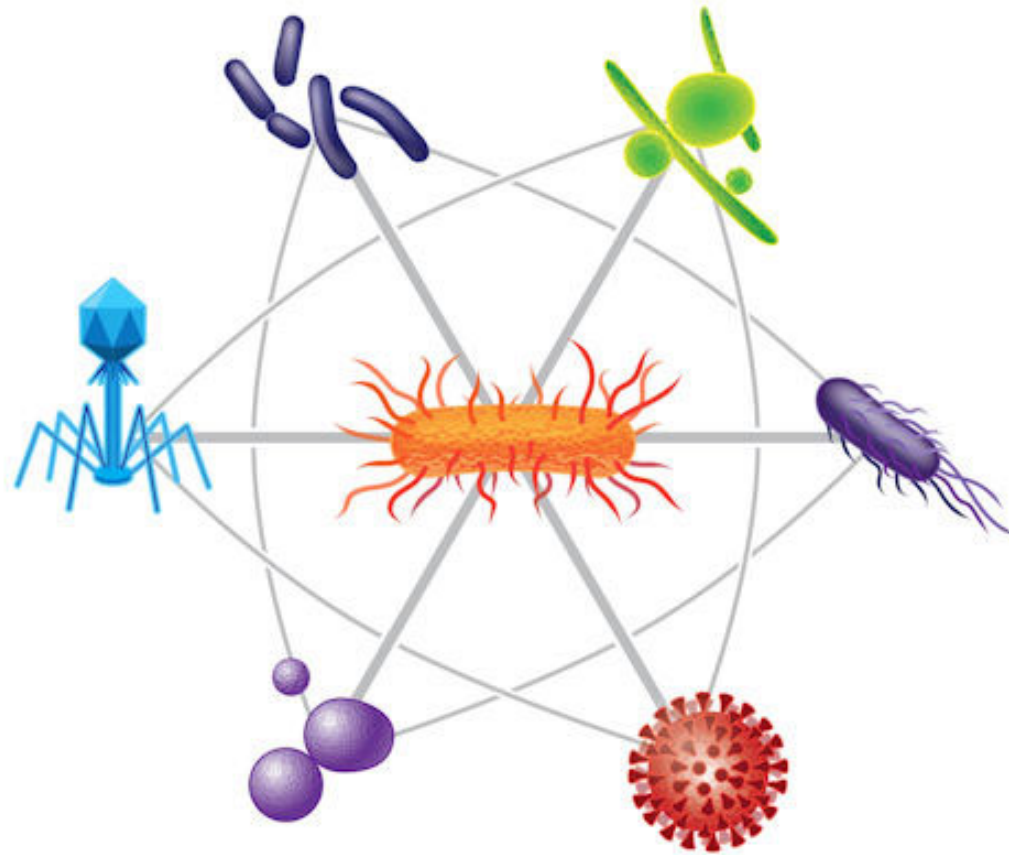
<https://www.nature.com/articles/s41587-022-01226-0>



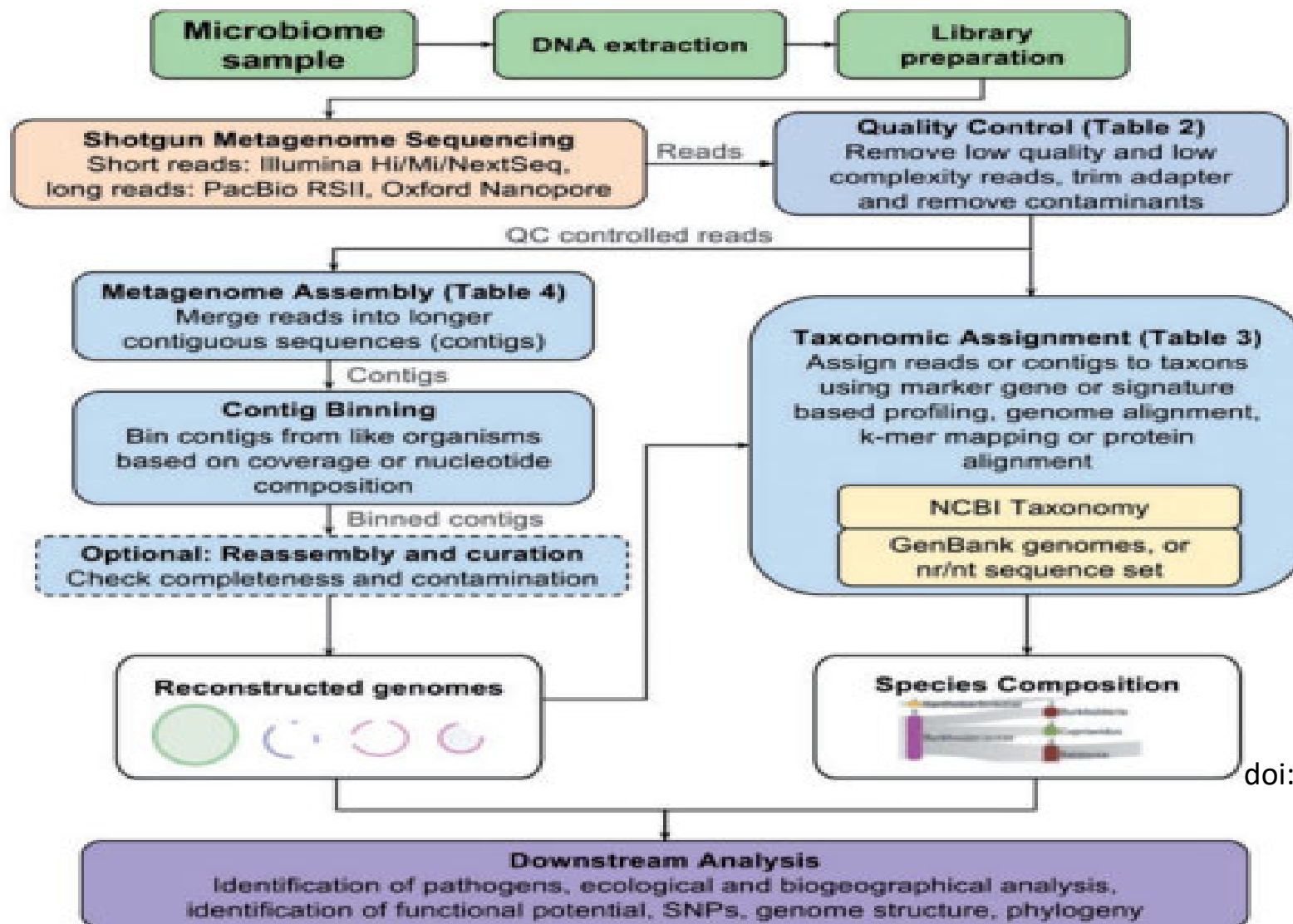
virus



Microbial Interaction

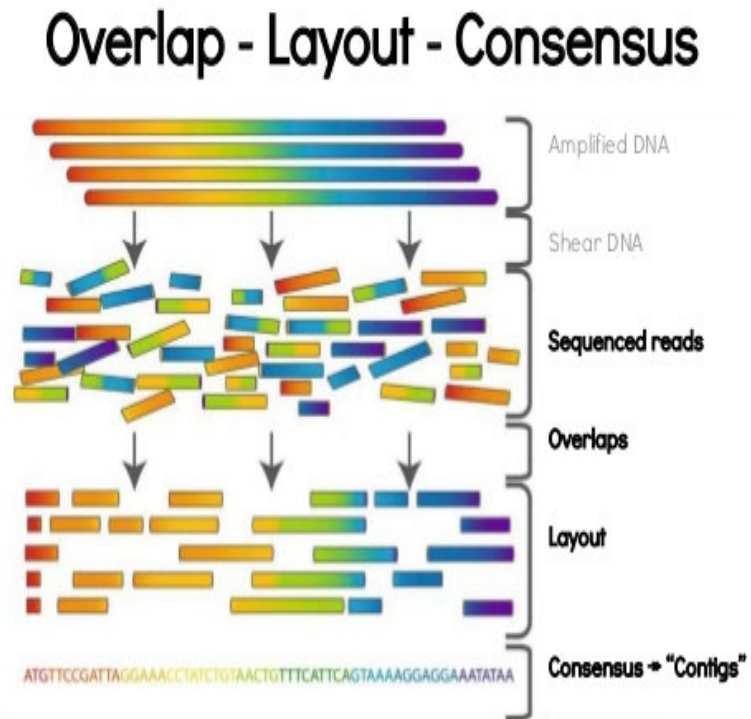


<https://asm.org/magazine/2023/spring/friends-foes-microbial-interactions-infections>

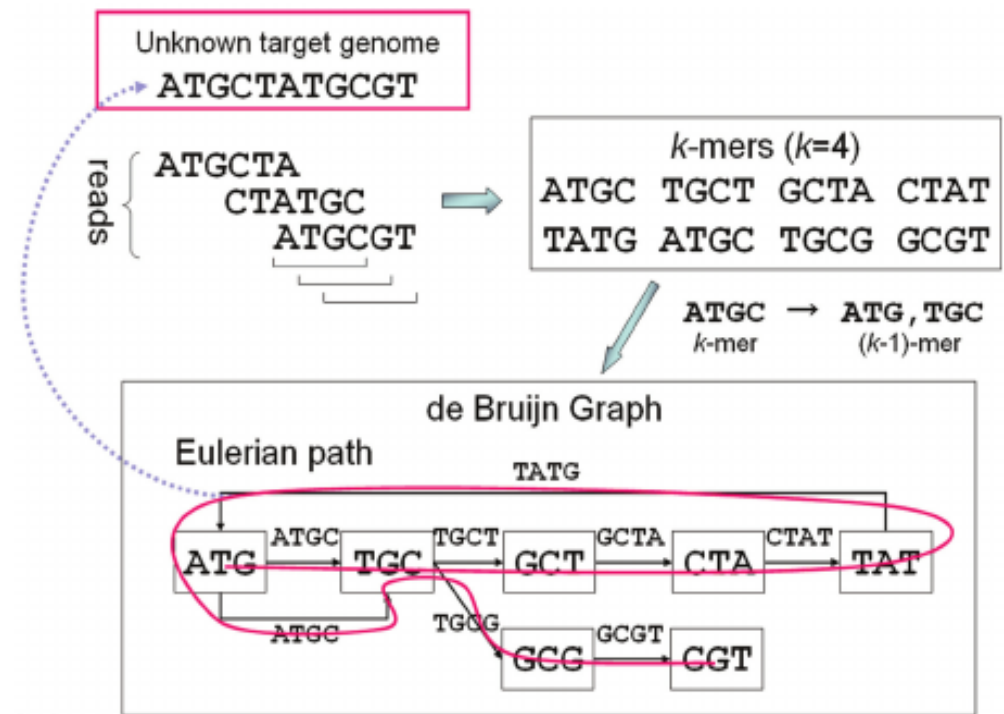


doi: 10.1093/bib/bbx120

Genome Assembly



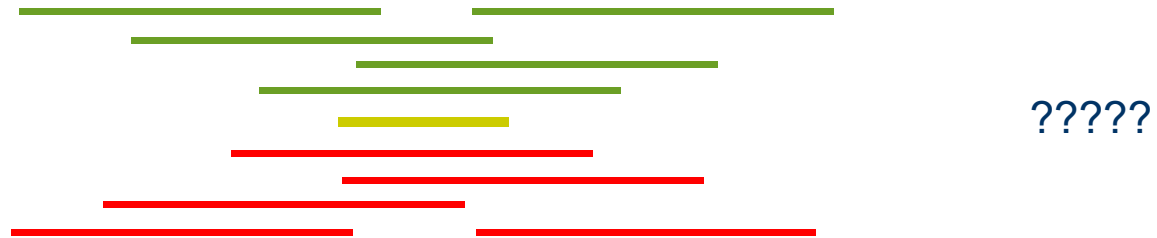
de Bruijn Graph



<http://www.chromnet.net/Genome%20De%20Novo%20Sequencing%20and%20Assembly.aspx>

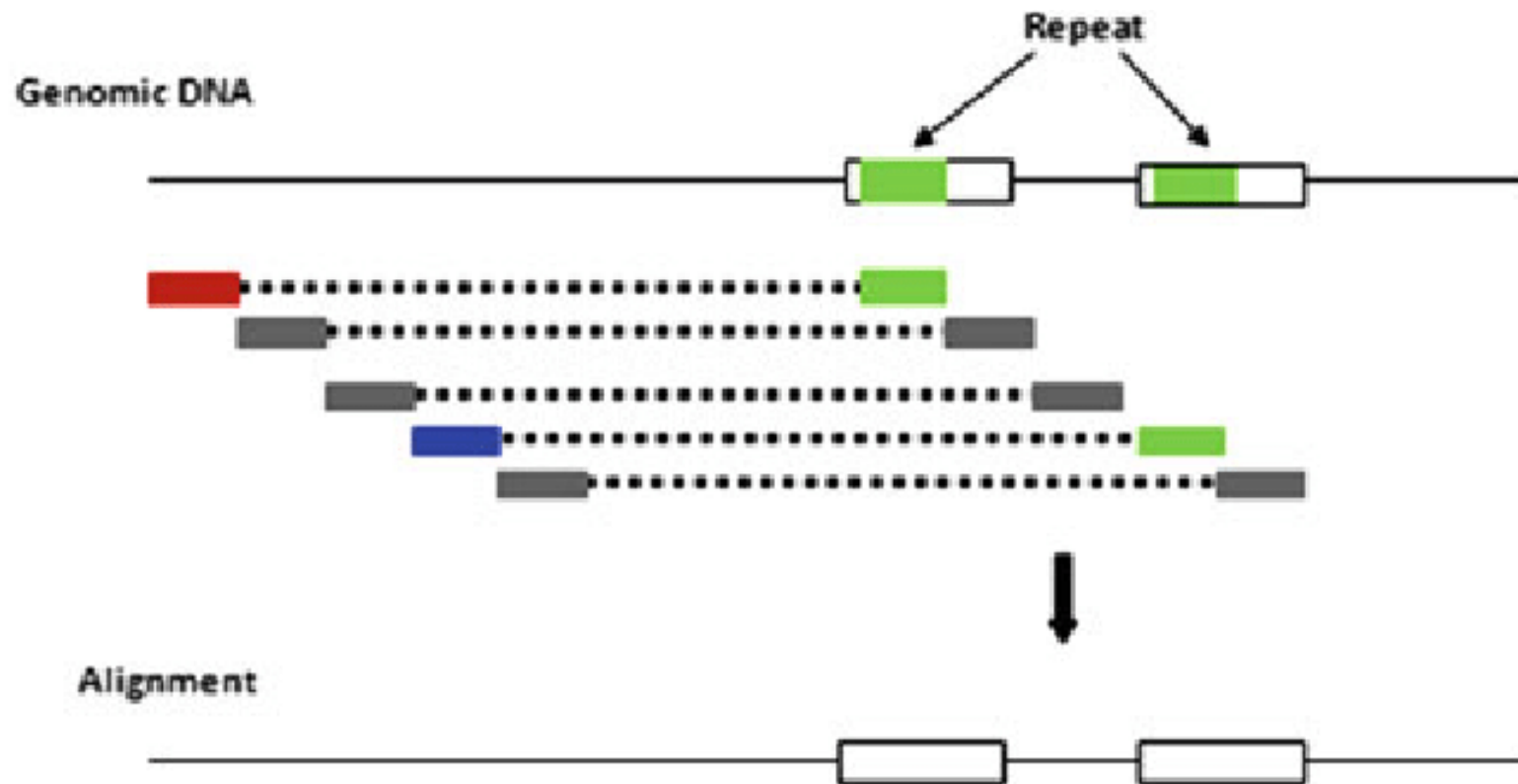
https://www.researchgate.net/figure/Illustration-of-de-Bruijn-graph-based-assembly_fig1_229437536

Troublesome Repeats



- Insert non-maximal reads whenever unambiguous

Paired-end reads and Repeats



Challenges in Metagenomic Assembly

Mixed reads

Horizontal gene transfer

Highly conserved genes

Uneven sequencing coverage

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Table 1

Overview and basic characteristics of currently used and freely available short read metagenome assemblers.

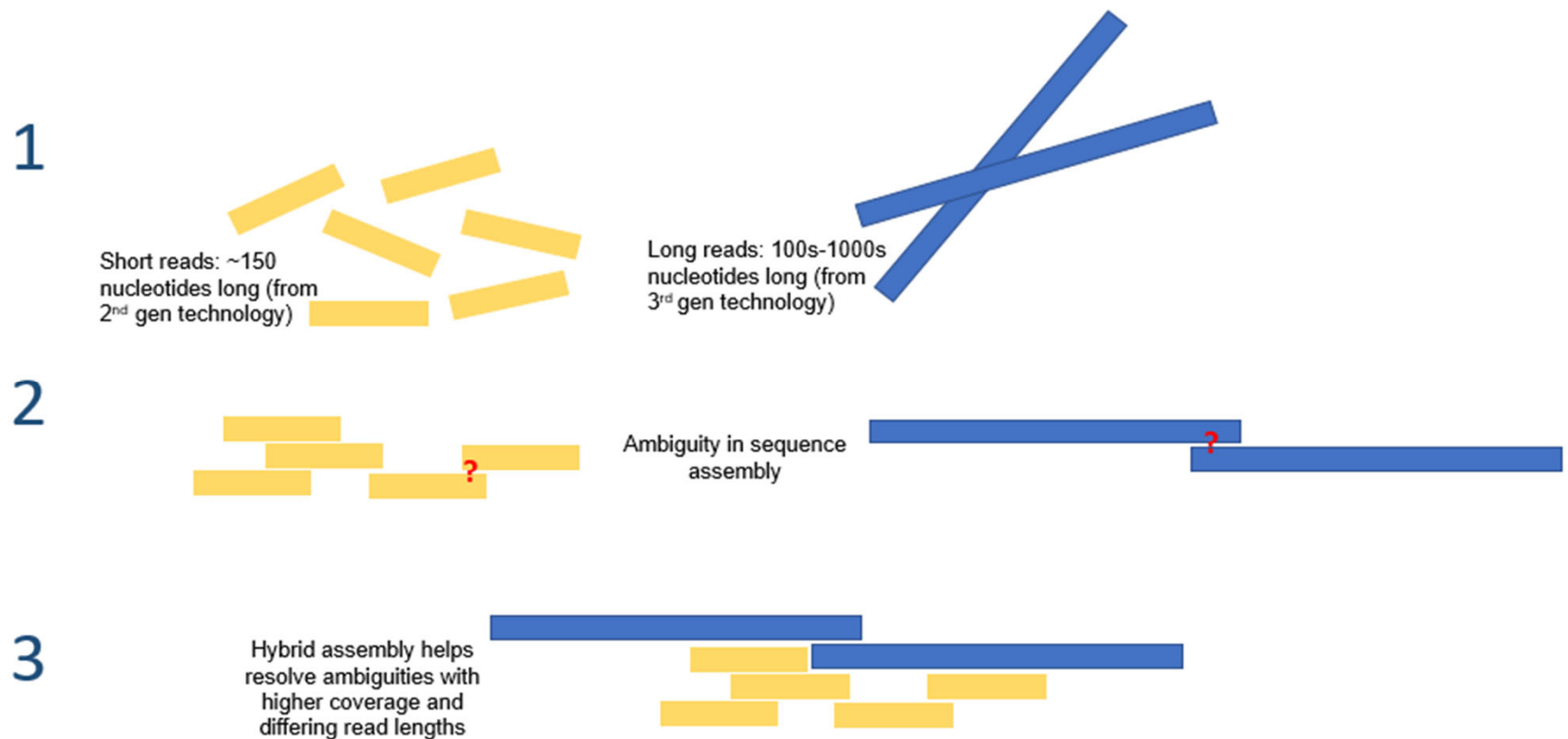
Included are basic characteristics influencing the user friendliness, such as the range of accepted input formats or extent of documentation. The number of total and recent citations indicates the past and present popularity as well as dissemination of the respective tool within the scientific community.

short read assembler	version	last release	Method	input seq format	read pair format	multiple libraries	extensive instructions available	Support	summary of user friendliness	Citations (total/2016)
IDBA-UD	1.1.2	2014	de Bruijn multiple K-mer	.fasta	interleaved only	yes	no	GitHub tickets, email	inflexible, incomplete documentation	481/189
MegaHit	1.0.3	2015	de Bruijn multiple K-mer	.fastq, .fastq.gz, .fasta, .fasta.gz, stdin	interleaved or separate	yes	yes	GitHub tickets, email	simple usage, flexible, well documented	59/39
MetaVelvet	1.2.01	2012	de Bruijn single K-mer	.fastq, .fastq.gz, .fasta, .fasta.gz, .sam, .bam, .stdin	interleaved or separate	yes	yes (mostly for velvet)	mailing list, email	flexible, well documented	187/72
MetaVelvet-SL	1.0	2015	de Bruijn single K-mer	.fastq, .fastq.gz, .fasta, .fasta.gz, .sam, .bam, .stdin	interleaved or separate	yes	no	email	Convolutated workflow, flexible	16/11
Ray Meta	2.3.1	2014	de Bruijn single K-mer	.fasta, .fasta.gz, .fastq, .fastq.gz	interleaved or separate	yes	yes	GitHub tickets, email	flexible, well documented	192/73
SOAPdenovo2	2.01	2015	de Bruijn single K-mer	.fastq, .fastq.gz, .fasta, .fasta.gz, .bam	interleaved or separate	yes	yes	GitHub tickets, email	well documented	938/334
Omega	1.0.2	2014	String graph prefix + suffix hashtable	.fastq, .fasta	interleaved only	no	yes	email	simple usage, well documented	21/13
metaSPAdes	3.8.0	2016	de Bruijn multiple K-mer	.fastq, .fastq.gz, .fasta, .fasta.gz, .bam	interleaved or separate	no	Yes	Sourcefourge/GitHub tickets, mailing list, email	flexible, well documented	5/5

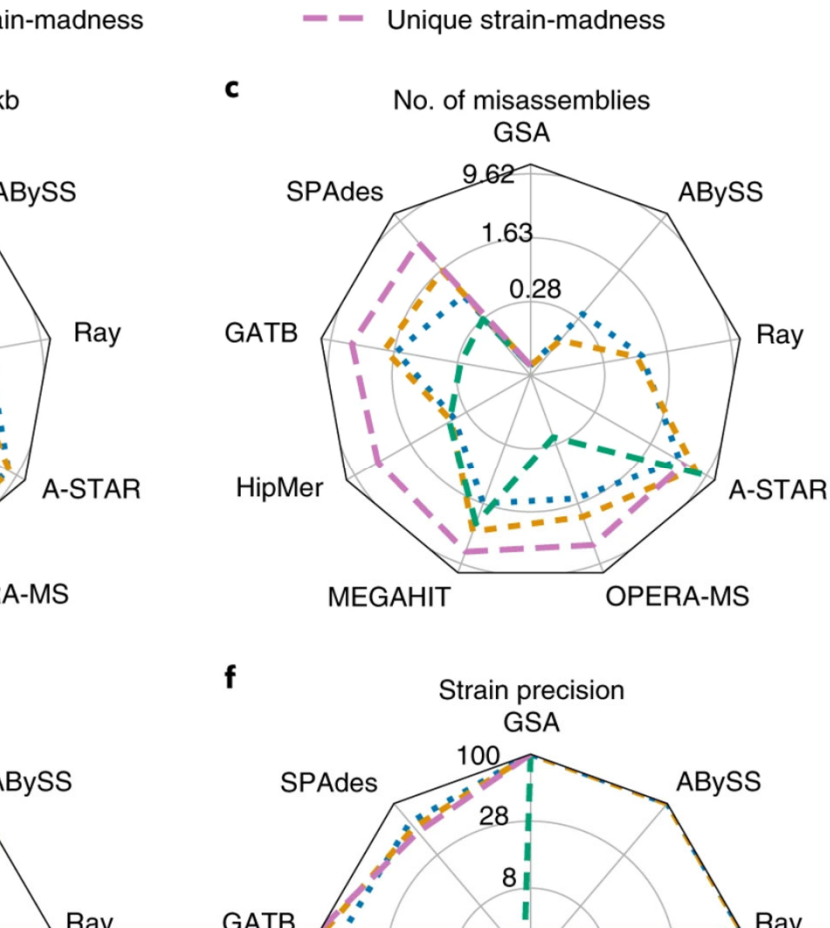
<https://www.ncbi.nlm.nih.gov/pmc/articles/PMC5242441/>

Assembly

Long and short read hybrid assembly



Current Status

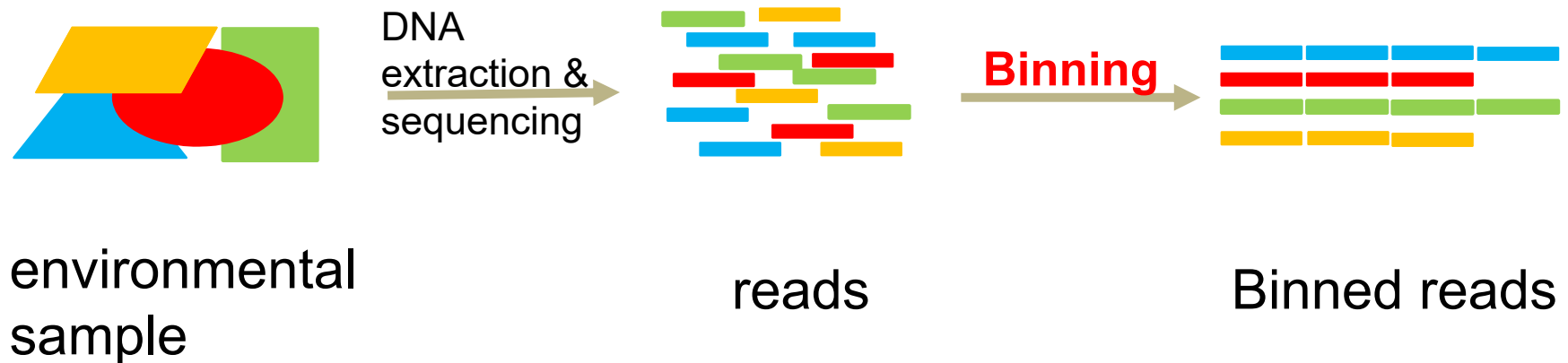


For difficult regions (repeats or highly conserved regions)

-STAR partially recovered 102 (78%) of 131 16S sequences. The hybrid assemblers GATB (mean completeness 60.1%) and OPERA-MS (mean 47.1%) recovered the most complete 16S sequences. Mean completeness for short-read assemblies ranged from 29.6% (HipMer) to 36.9% (MEGAHIT)

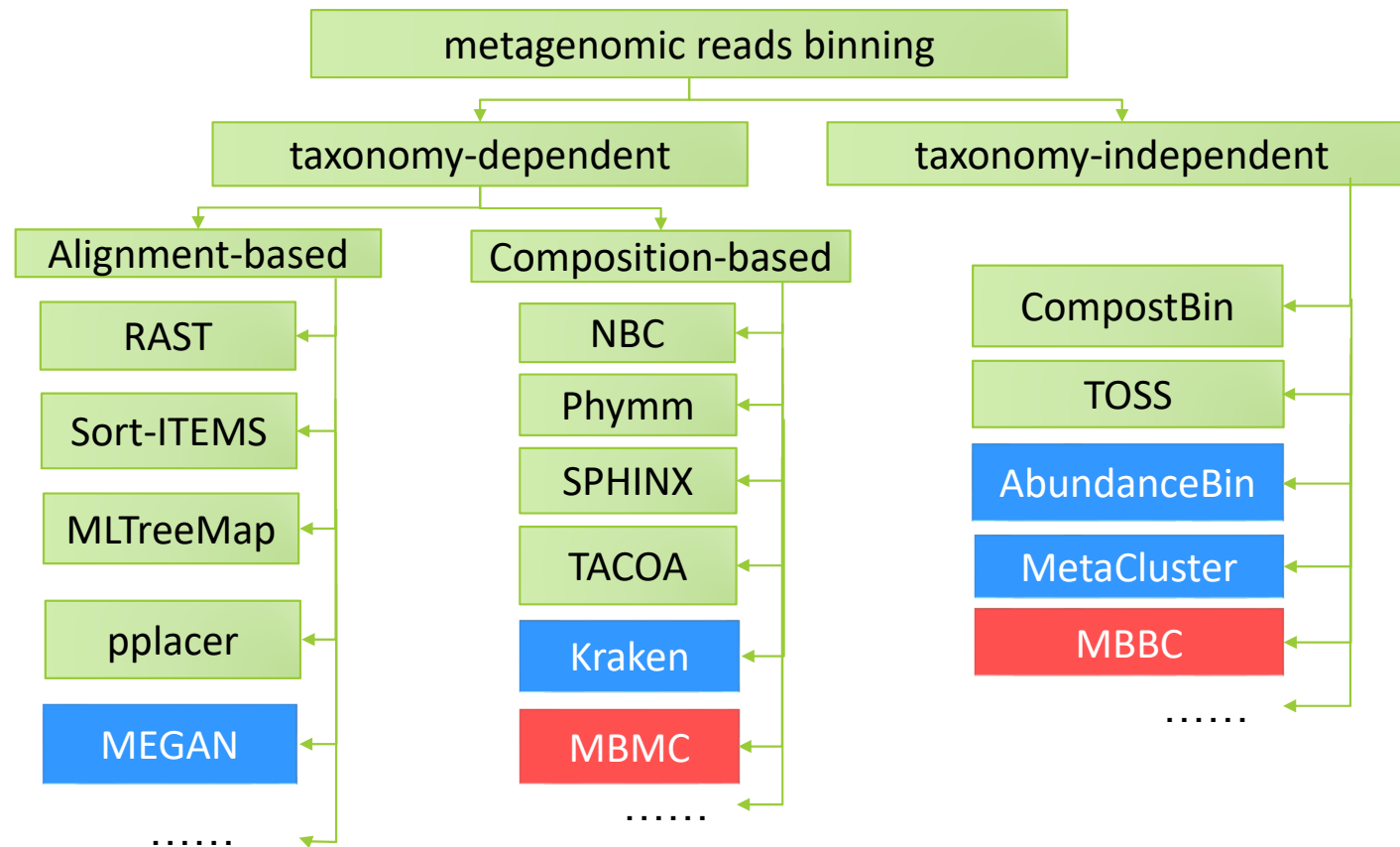
<https://www.nature.com/articles/s41592-022-01431-4/figures/1>

Binning



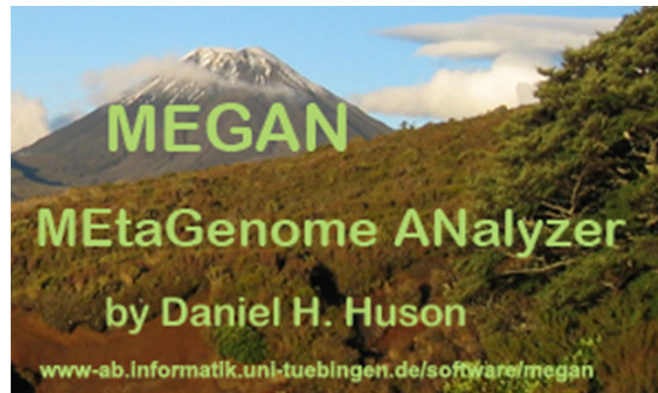
- Short reads length
- Sequencing errors
- high-throughput data

Current approaches



Taxonomy-dependent methods:

Alignment-based binning (**MEGAN**)



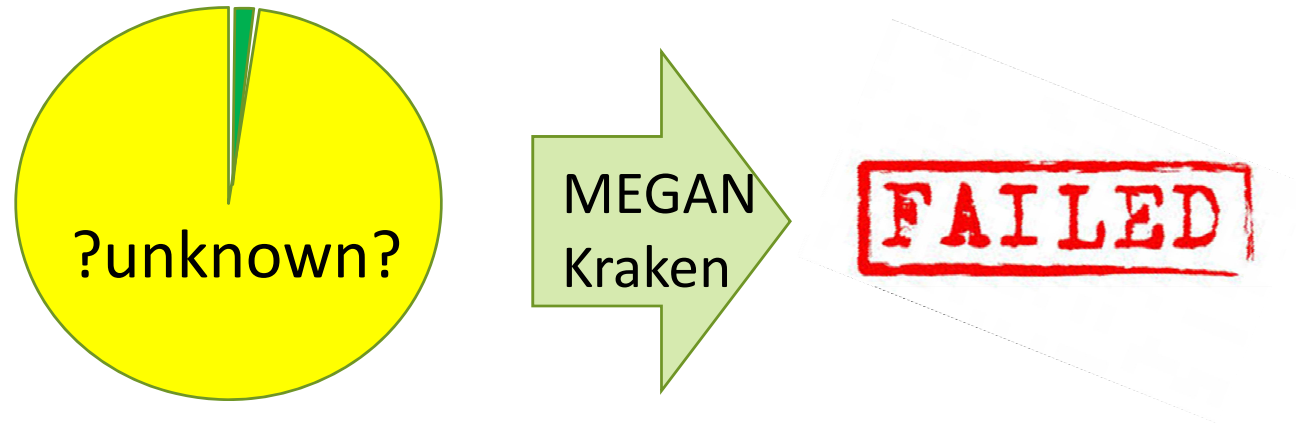
<http://ab.inf.uni-tuebingen.de/software/welcome.html/megan5>

Composition-based binning (**Kraken**)



<https://ccb.jhu.edu/software/kraken/>

Problems in taxonomy-dependent methods




MBMC works better for datasets that contain unknown species

Taxonomy-independent methods

AbundanceBin, MetaCluster

the difference of k-mer (short sequence with length k) frequencies of different microbes in the environmental samples

Observations:

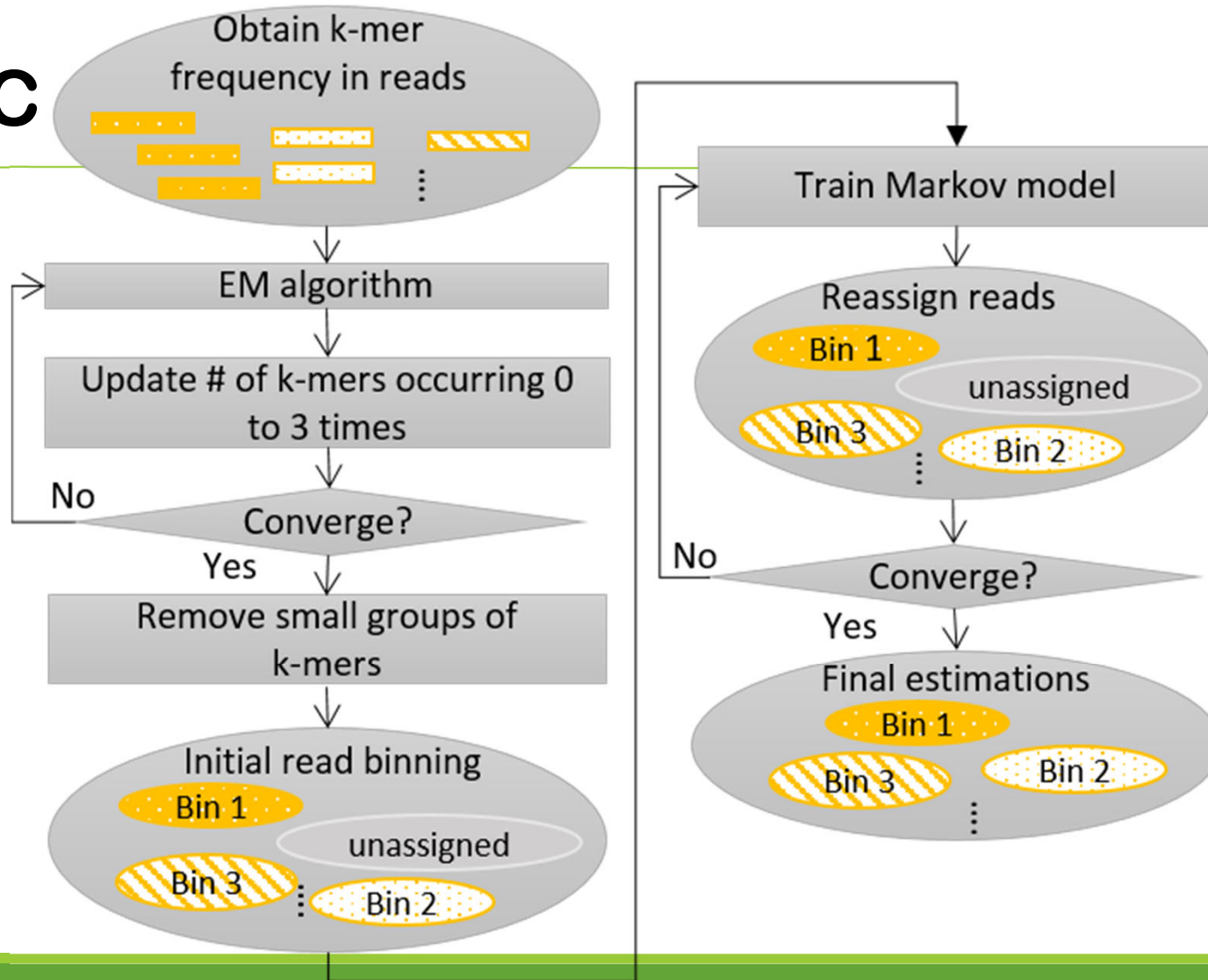
- k-mer frequency  genome's coverage
- long k-mers are usually unique in each genome
- k-mer frequency distribution from the same genome are similar

Problems in taxonomy-independent methods

➤ AbundanceBin/MetaCluster: only utilize the k-mers frequency.

→ MBBC: improve the method to utilize the k-mer frequency;
utilized Markov properties shared by a group of reads

MBBC



The procedure of read clustering in MBBC. The output on the right from each of the main steps on the left is connected with the corresponding steps.
 Wang et al. BMC Bioinformatics 2015 16:36 doi:10.1186/s12859-015-0473-8

Results1 MBBC reliably estimates the species number, genome sizes, relative species abundances, and k-mer coverage

A. Initial prediction of α , λ										
Initial Species	1	2	3	4	5	6	7	8	9	10
α	43.30%	22.97%	11.07%	20.84%	1.16%	0.51%	0.12%	0.03%	0.00%	0.00%
λ	3.88	11.14	16.57	23.61	38.71	51.62	74.22	105.37	158.79	329.53

↓

B. Prediction after updating #k-mers that occur 0 to 3 times										
α	31.59%	16.01%	25.79%	24.33%	1.38%	0.72%	0.14%	0.03%	0.01%	0.00%
λ	3.34	6.67	13.05	22.98	35.61	49.23	72.22	103.45	156.95	328.64

↓

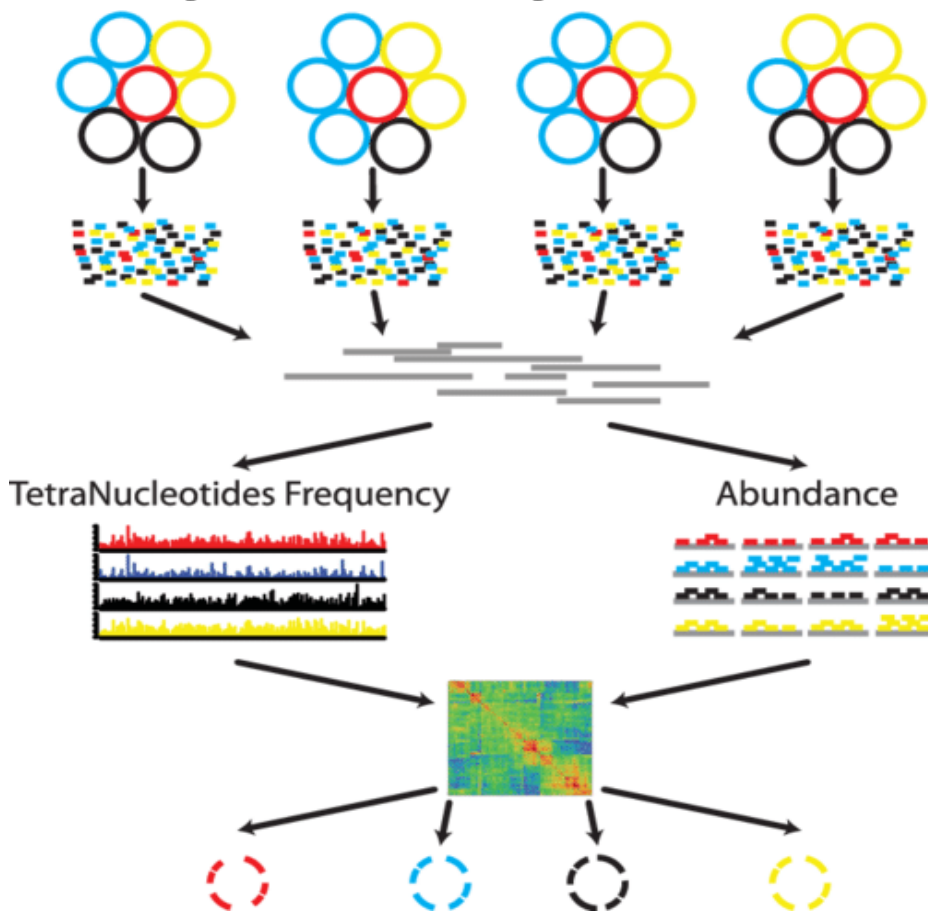
C. Prediction after removing small groups of k-mers										
Genome size	3009885	660737	1005524	948301	53786	27871	5352	1249	197	36
α	31.59%	16.01%	25.79%	24.33%	1.38%	0.72%	0.14%	0.03%	0.01%	0.00%
λ	3.34	6.67	13.05	22.98	35.61	49.23	72.22	103.45	156.95	328.64

↓

D. Prediction after iteratively binning read based on Markov chains: Predicted (real data)				
Predicted Species	1	2	3	4
Genome size	1498994 (1160554)	825923 (945296)	1138156 (1107344)	1212248 (1075140)
α	9.42% (6.98%)	10.35% (11.36%)	27.91% (29.95%)	52.33% (51.70%)
λ	3.34 (3.49)	6.67 (5.83)	13.05 (12.48)	22.98 (20.52)

MBBC: an efficient approach for metagenomic binning based on clustering
 BMC Bioinformatics. 2015;16(1):36.

Contig Binning



Preprocessing

- 1 Samples from multiple sites or times
- 2 Metagenome libraries
- 3 Initial de-novo assembly using the combined library

MetaBAT

- 4 Calculate TNF for each contig
- 5 Calculate Abundance per library for each contig
- 6 Calculate the pairwise distance matrix using pre-trained probabilistic models
- 7 Forming genome bins iteratively

Concoct

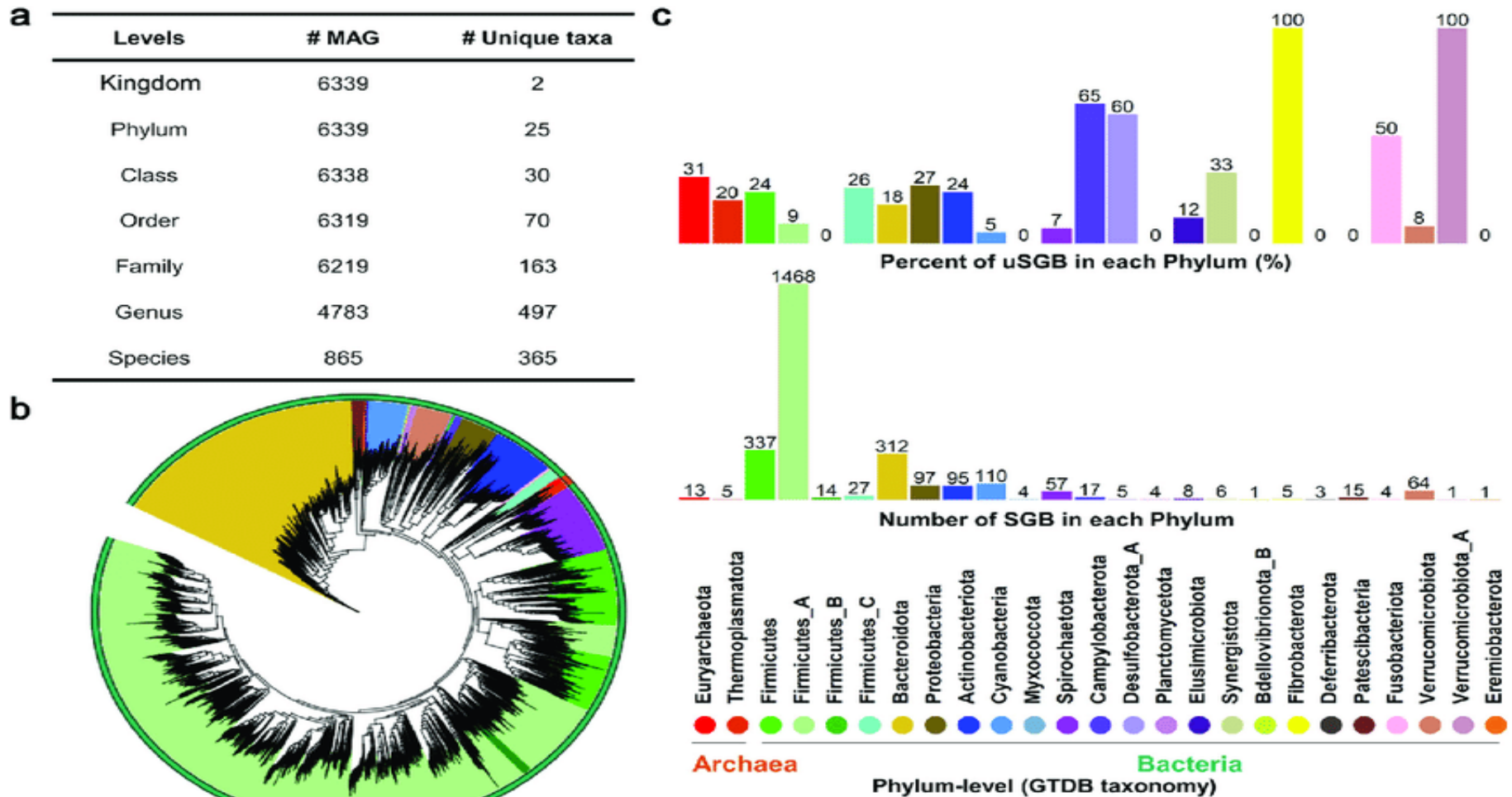
Tetramer frequency ($V = 136$ tetramers)

Normalized Coverage of a contig across M samples

Total coverage of a contig across M samples

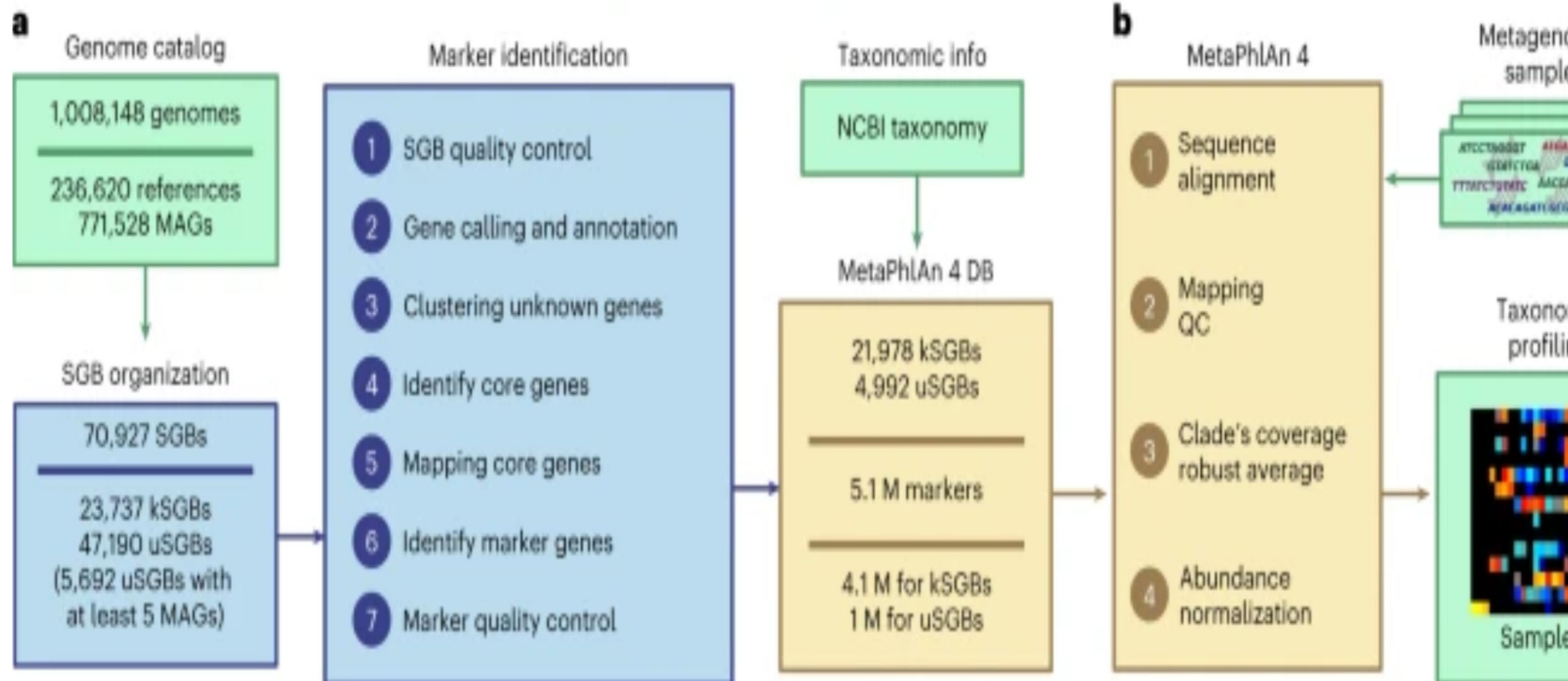
Dimension reduction by SVD on all N contigs' Log $M+1+V$ vectors to account for >90% variation of the vectors, then apply Gaussian mixture decomposition to obtain different MAGs.

Taxonomic Analysis



MetaPhlAn

● Input/output ● Genomic database ● MetaPhlAn 4



<https://www.nature.com/articles/s41587-023-01688-w>

Other metagenomic data

metatranscriptomics

metaproteomics

Single cell metagenomics

Hi-C