



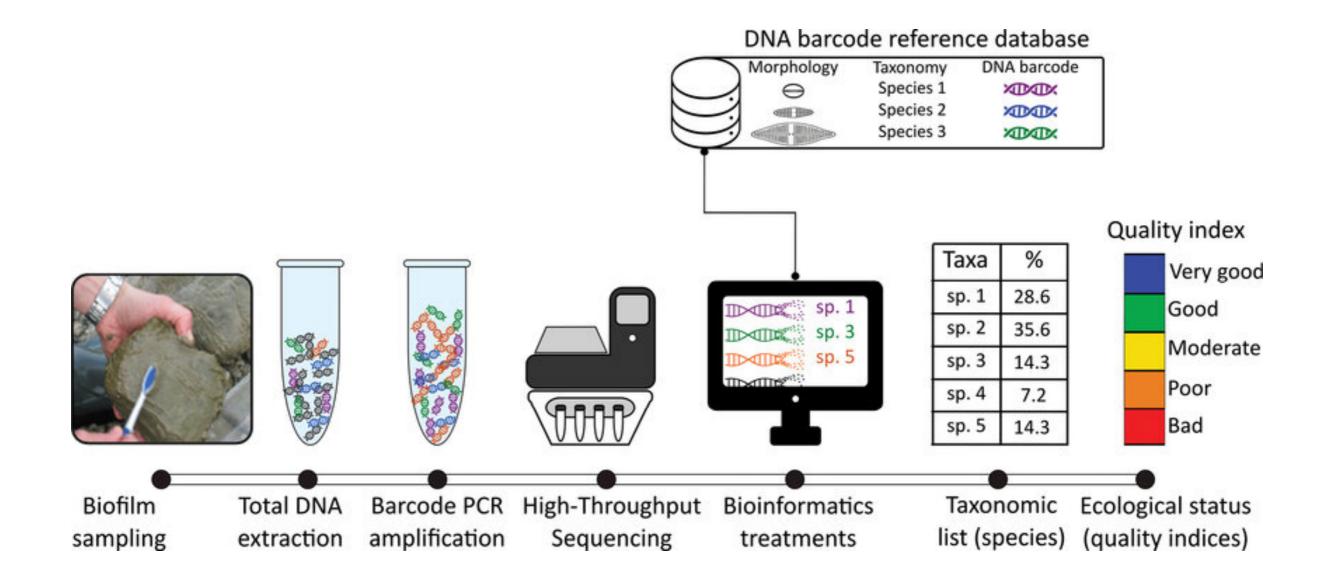
Bioinformatic analysis of metabarcoding data with DADA2

Theorical part

Clarisse Lemonnier

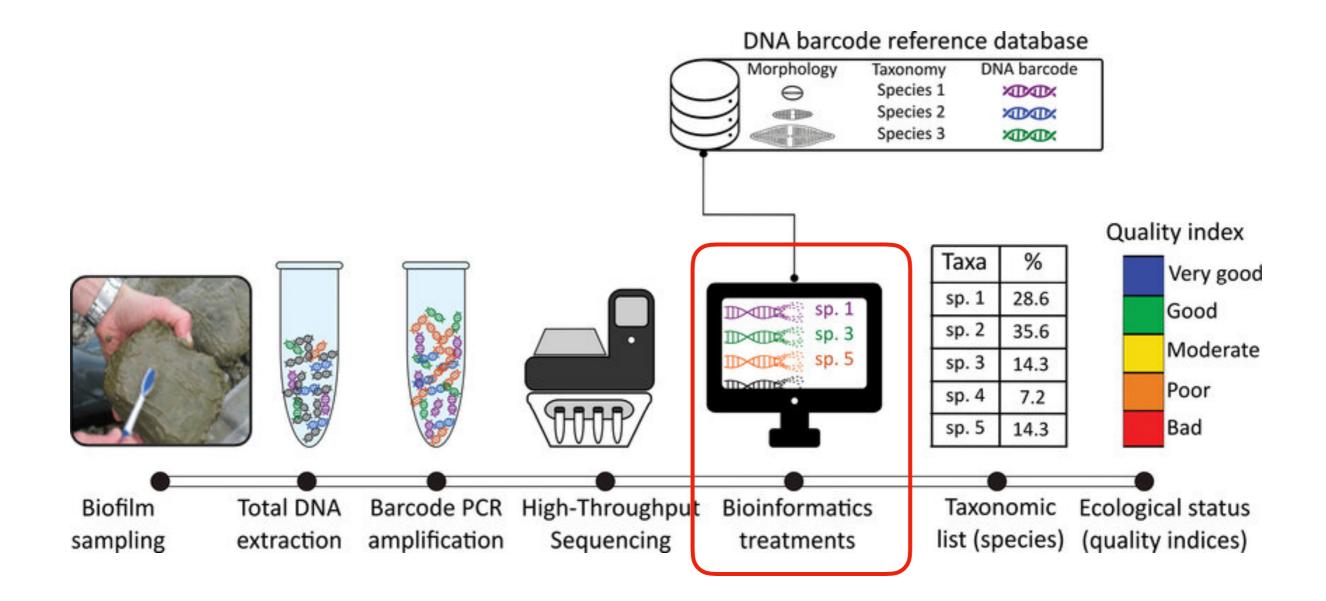


Importance of the bioinformatic steps





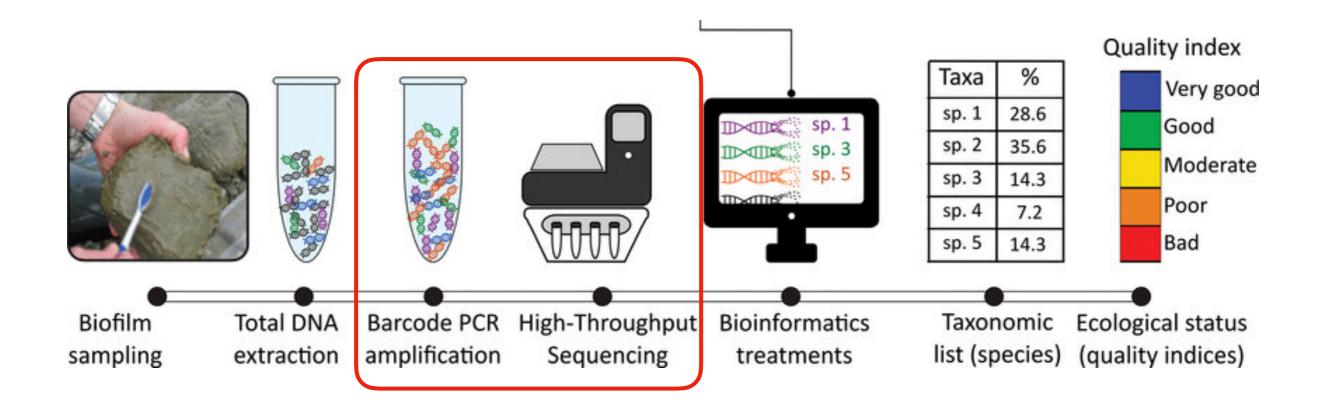
Importance of the bioinformatic steps



It is this step that you will produce an ASV/OTU table that will be at the beginning of taxonomic identification, ecological analysis and quality indices calculation



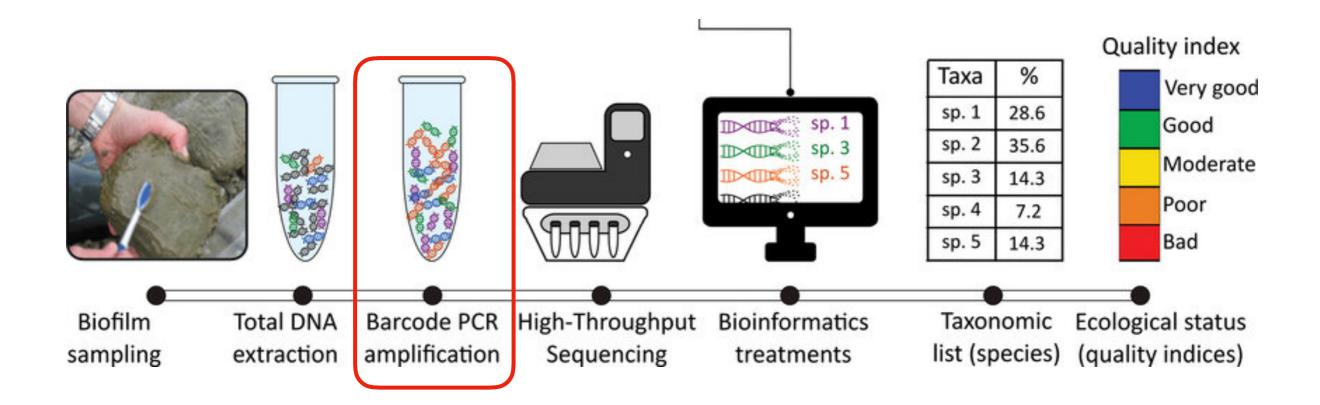
Some reminders on the metabarcoding steps :



It is important to know what's happening during these steps to understand all the bioinformatic pipeline



Some reminders on the metabarcoding steps :



1. PCR

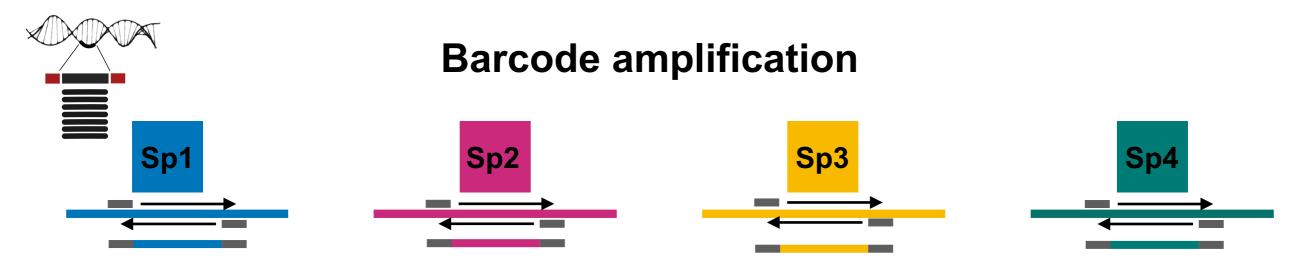


Let's do metabarcoding in a sample with 4 species



Reminders

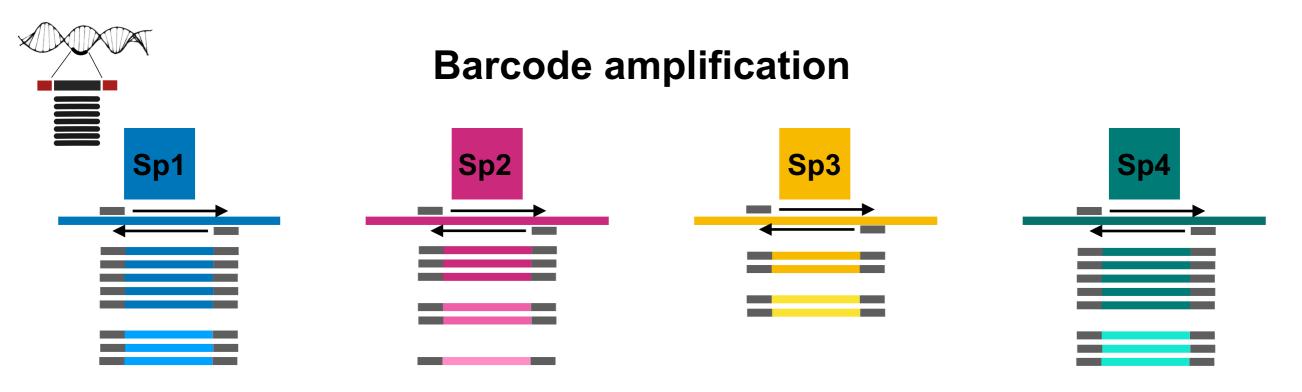




Primers are added

Reminders



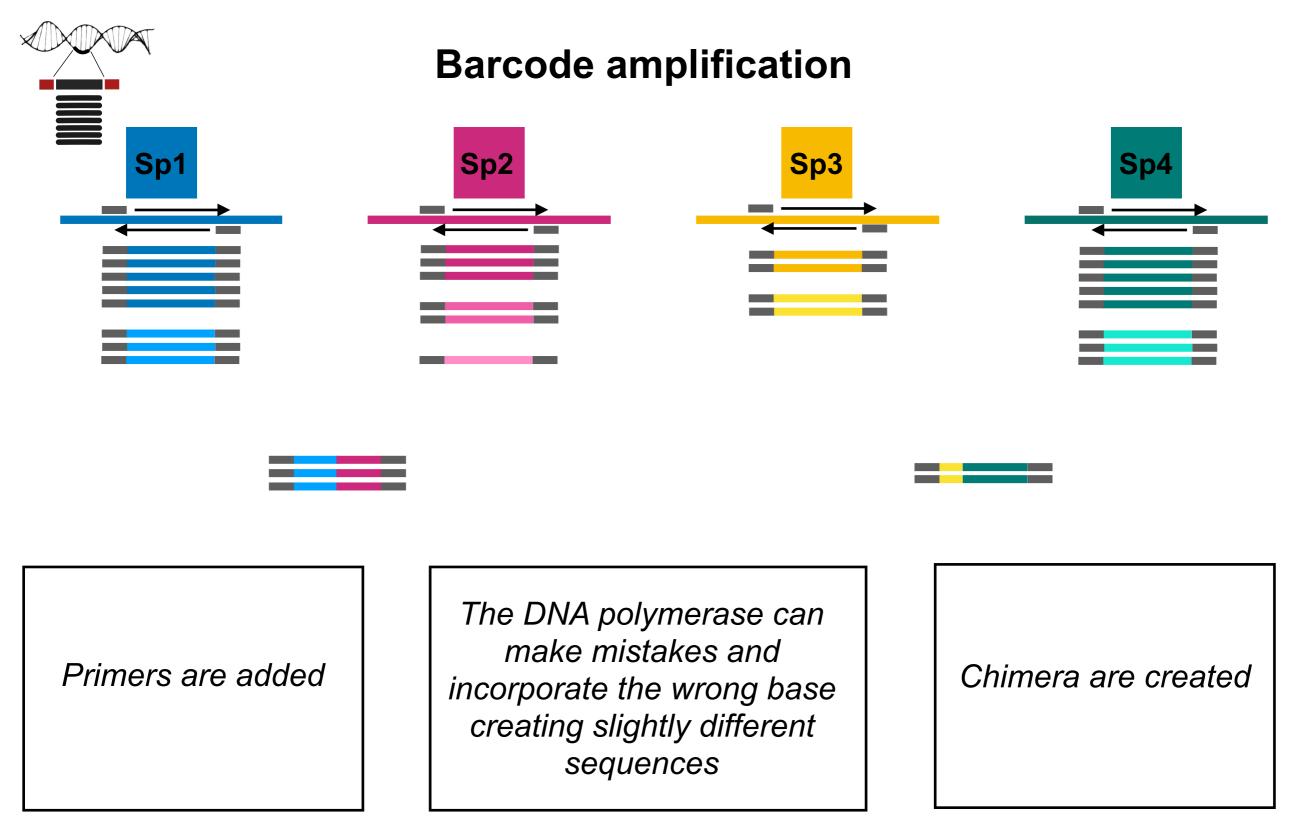


Primers are added

The DNA polymerase can make mistakes and incorporate the wrong base creating slightly different sequences

Reminders

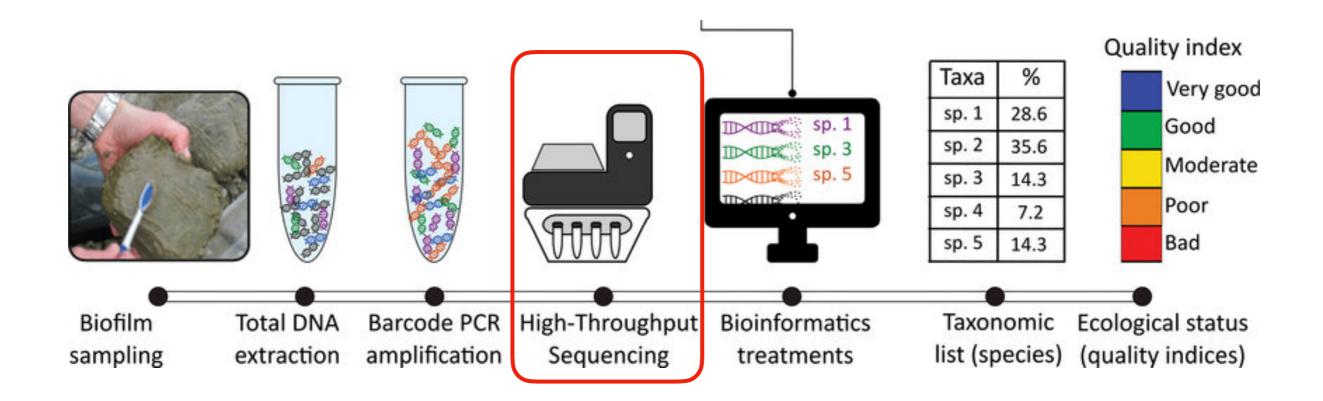




Vocabulary : at this step you have <u>amplicons</u> of the targeted barcode



Some reminders on the metabarcoding steps :

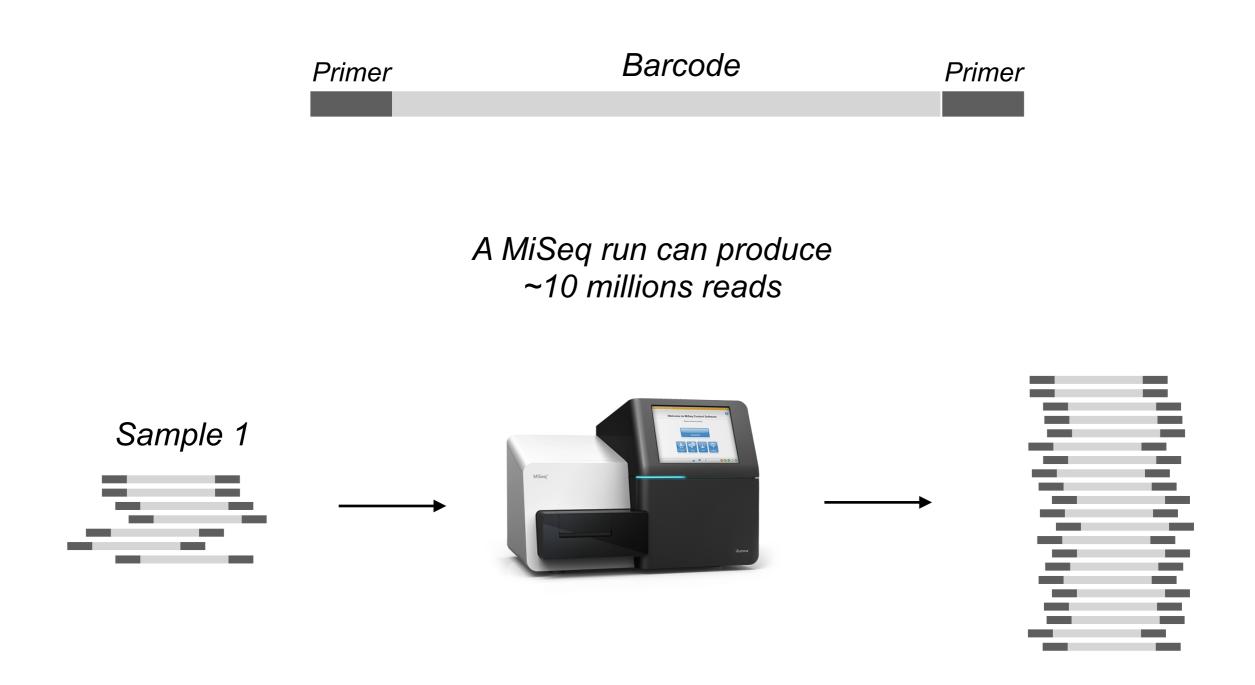


2. Sequencing

Note : I will be only speaking of Illumina MiSeq sequencing



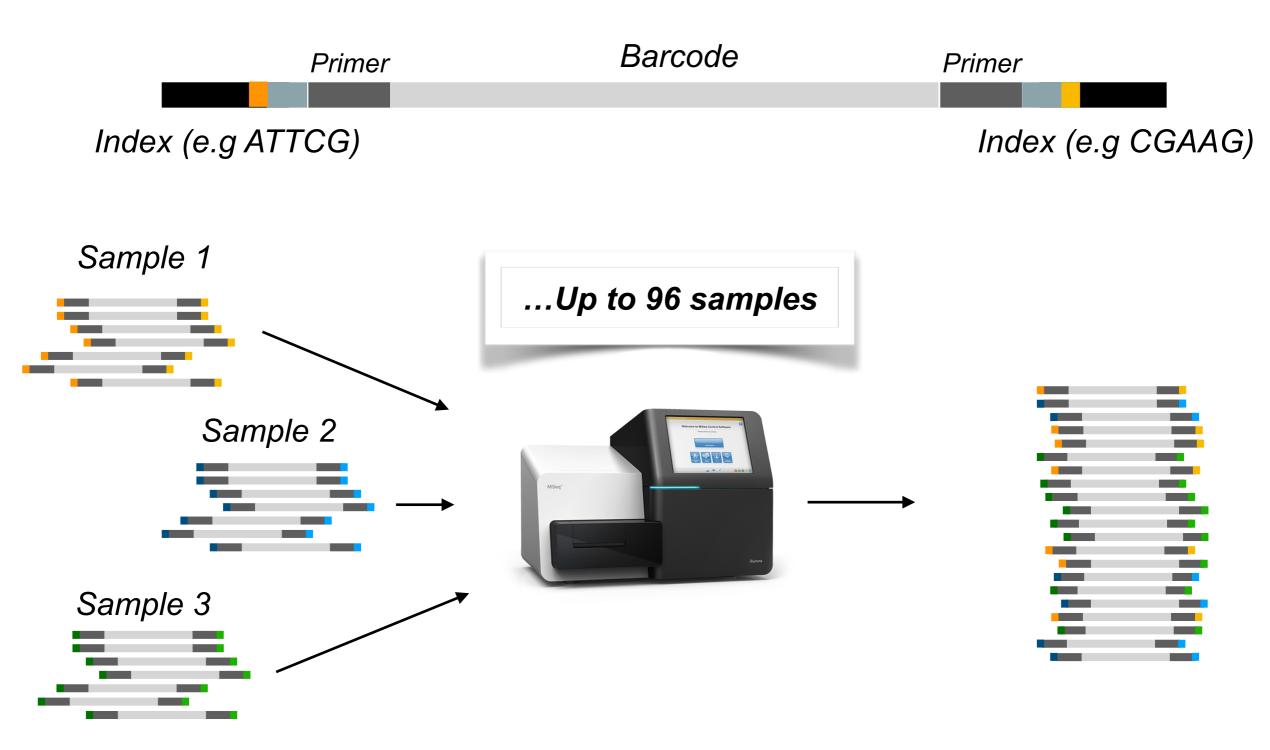
1. Multiplexing



... but 100 000 reads in a sample is often enough



1. Multiplexing



Demultiplexing is usually done by the sequencing platform



2. Paired-end sequencing

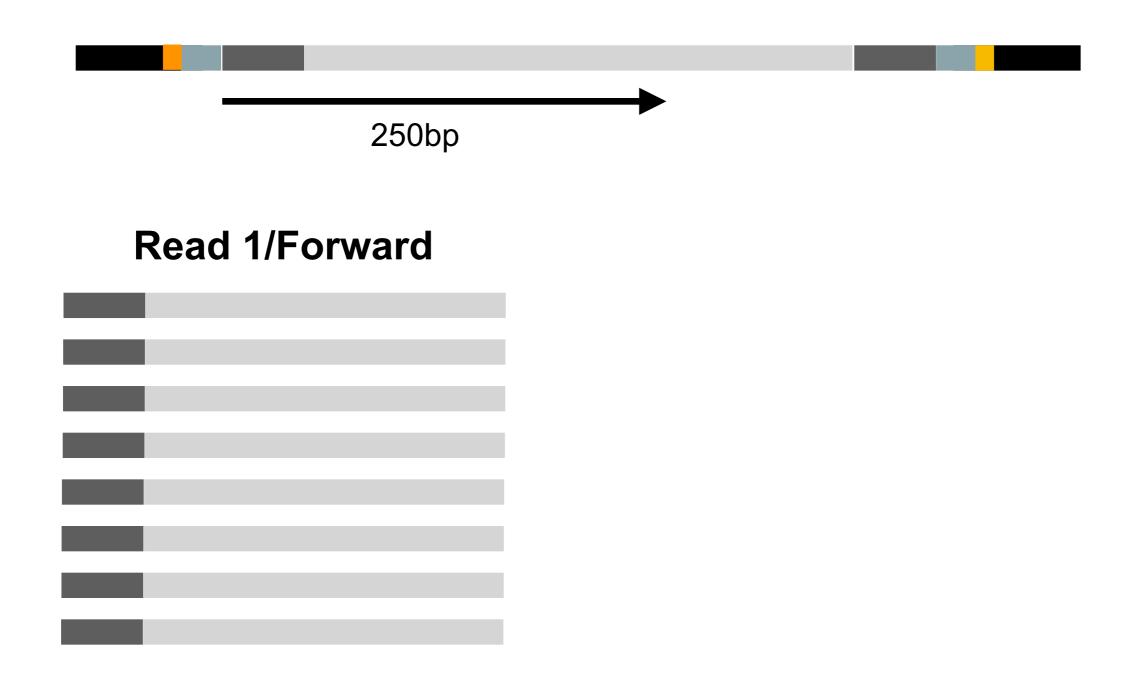


Illumina technology : 300bp max (often 250bp)

23S phytoplankton barcode + primers = ~400bp

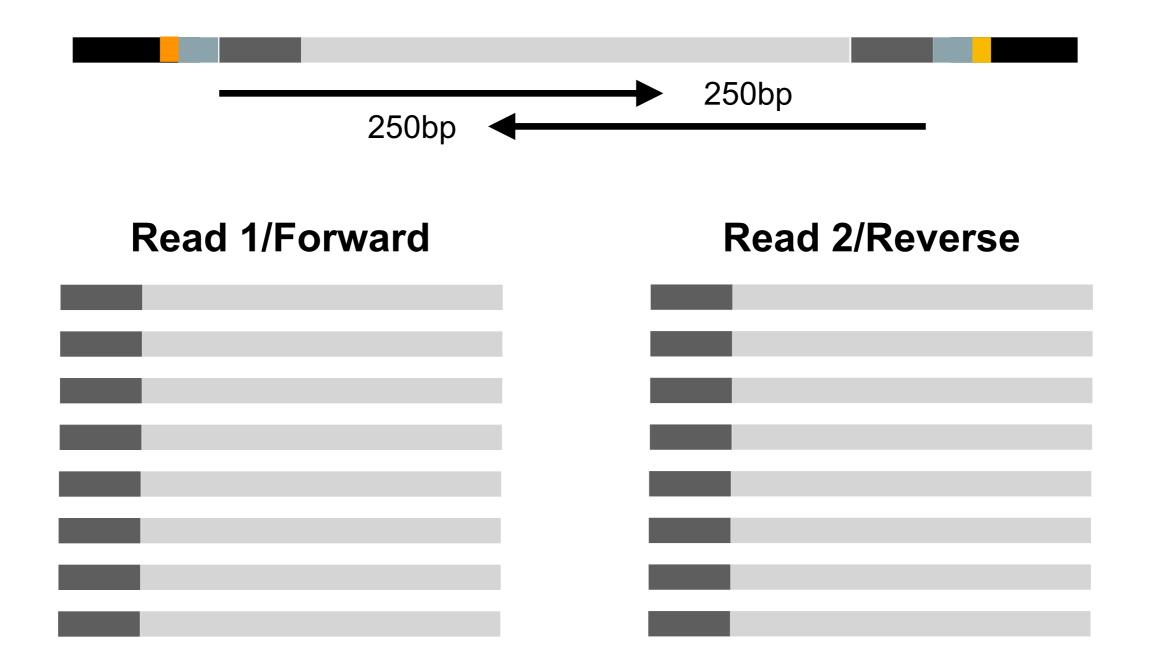


2. Paired-end sequencing





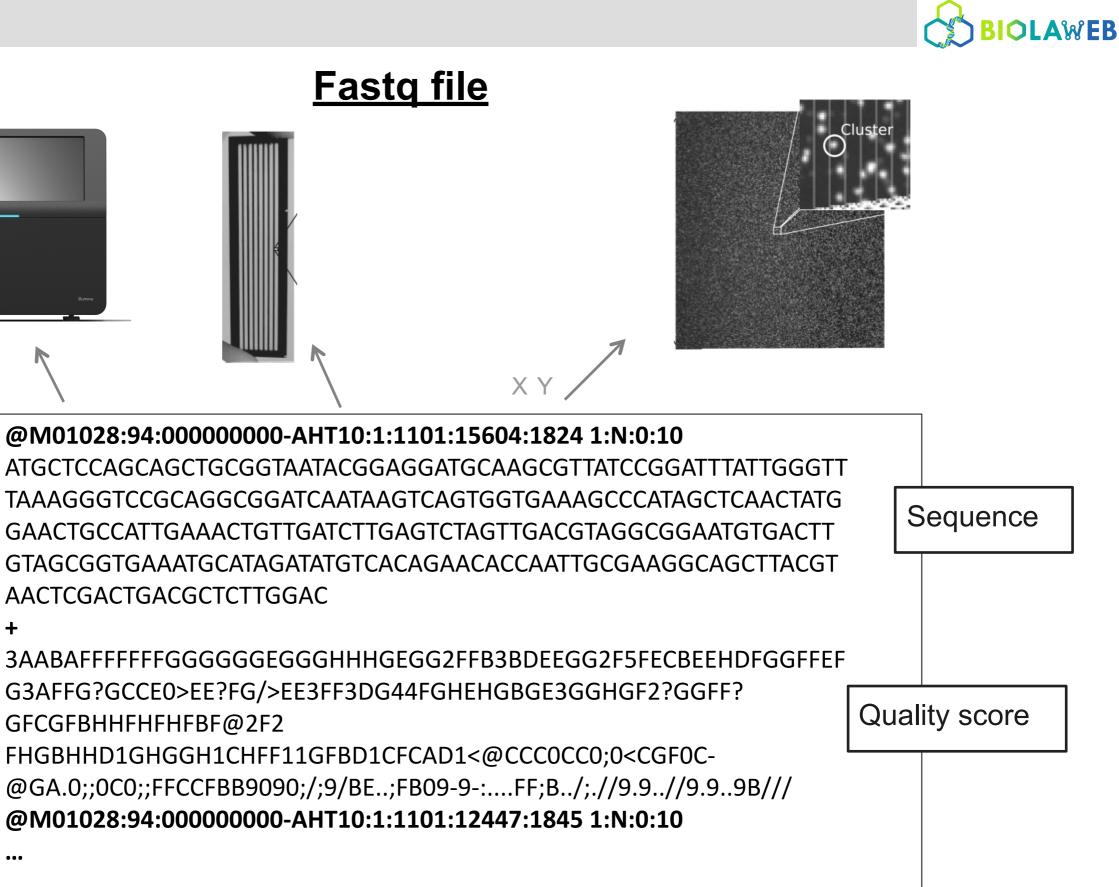
2. Paired-end sequencing



At the end of a run, you have 2 fastq files for each sample

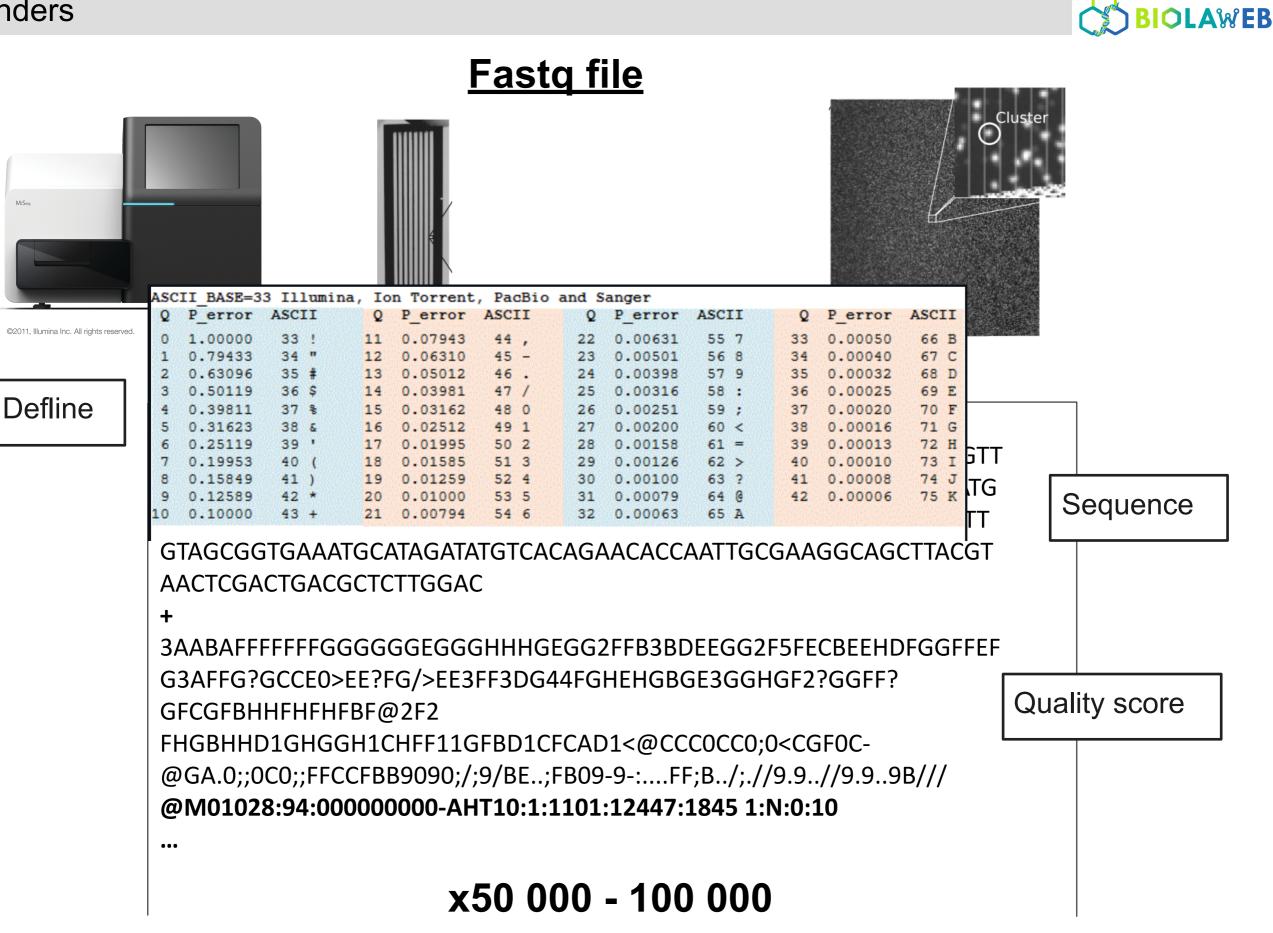
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Defline



x50 000 - 100 000

Reminders





...and now your dealing with this !

BIOLAWEB



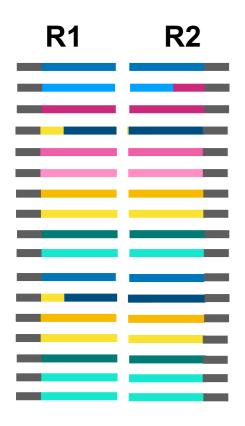
A lot of noise is added during the PCR and sequencing steps



BIOINFORMATIC PIPELINE - the theory

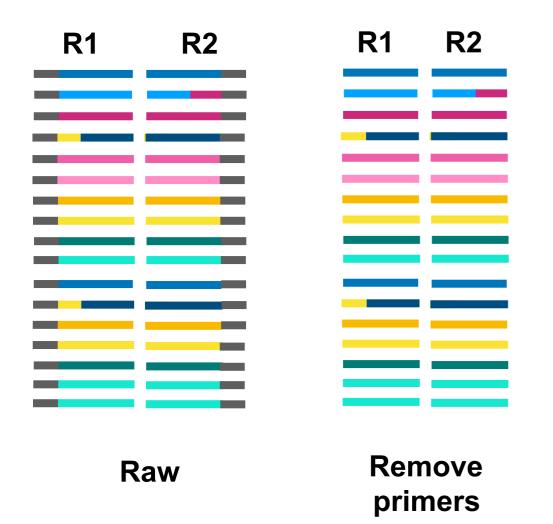






Raw









Raw

Remove primers Remove sequences of bad quality

Q	P_error	ASCII	Q	P_error	ASCII	Q	P_error	ASCII	Q	P_error	ASCII
0	1.00000	33 !	11	0.07943	44 ,	22	0.00631	55 7	33	0.00050	66 B
1	0.79433	34 "	12	0.06310	45 -	23	0.00501	56 8	34	0.00040	67 C
2	0.63096	35 #	13	0.05012	46 .	24	0.00398	57 9	35	0.00032	68 D
3	0.50119	36 \$	14	0.03981	47 /	25	0.00316	58 :	36	0.00025	69 E
4	0.39811	37 %	15	0.03162	48 0	26	0.00251	59 ;	37	0.00020	70 E
5	0.31623	38 &	16	0.02512	49 1	27	0.00200	60 <	38	0.00016	71 G
6	0.25119	39 '	17	0.01995	50 2	28	0.00158	61 =	39	0.00013	72 H
7	0.19953	40 (18	0.01585	51 3	29	0.00126	62 >	40	0.00010	73 I
8	0.15849	41)	19	0.01259	52 4	30	0.00100	63 ?	41	0.00008	74 J
9	0.12589	42 *	20	0.01000	53 5	31	0.00079	64 @	42	0.00006	75 K
0	0.10000	43 +	21	0.00794	54 6	32	0.00063	65 A			



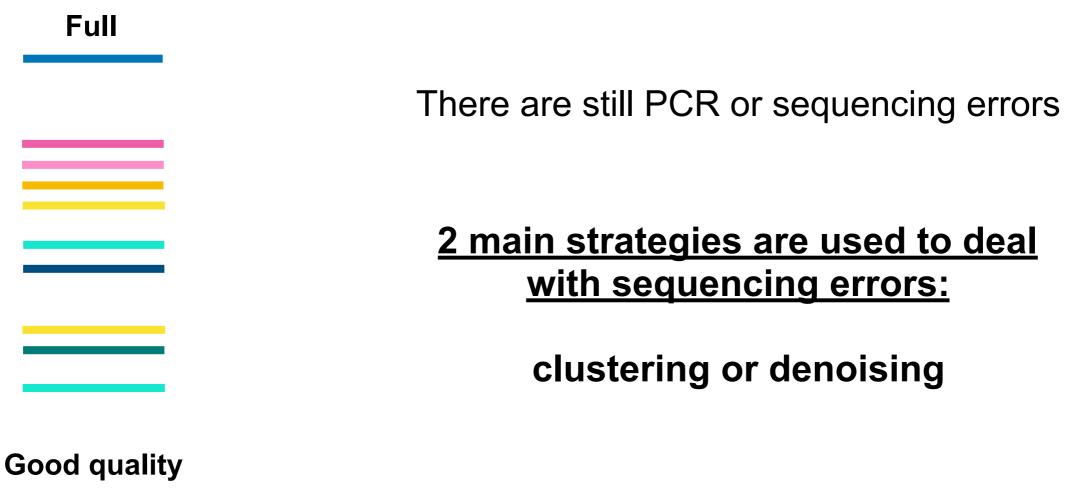
R1	R2	R1 R2	R1 R2	Full
R	aw	Remove primers	Remove sequences of bad quality	Merge



R1 R2	R1 R2	R1 R2	Full	Full
Raw	Remove primers	Remove sequences of bad quality	Merge	Remove chimera

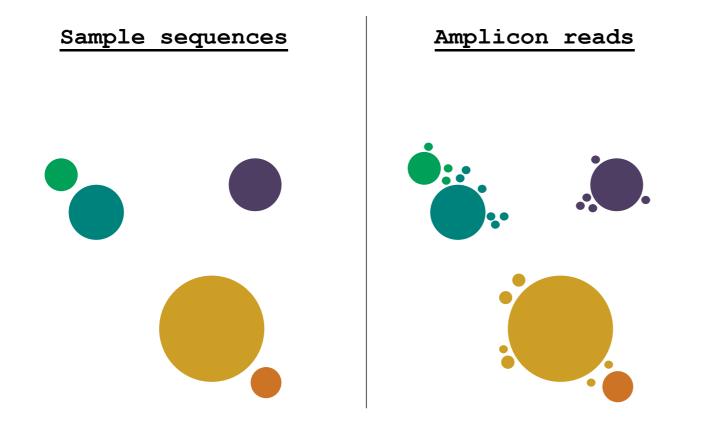


Second objective of bioinformatic pipeline : Make sense of these sequences in a biological/ ecological way

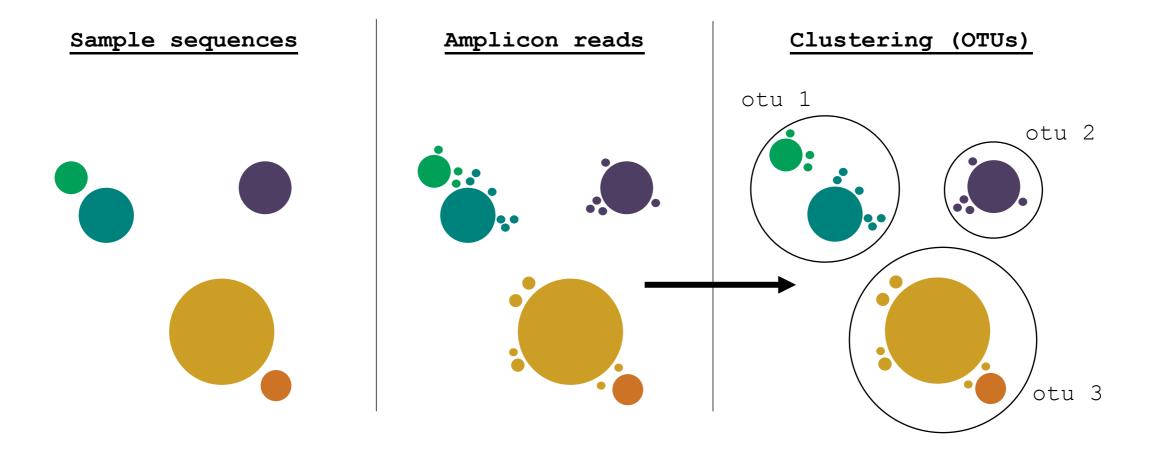


sequences









The first method used: cluster sequences based on their similarity, with a user-defined threshold (often 97%)

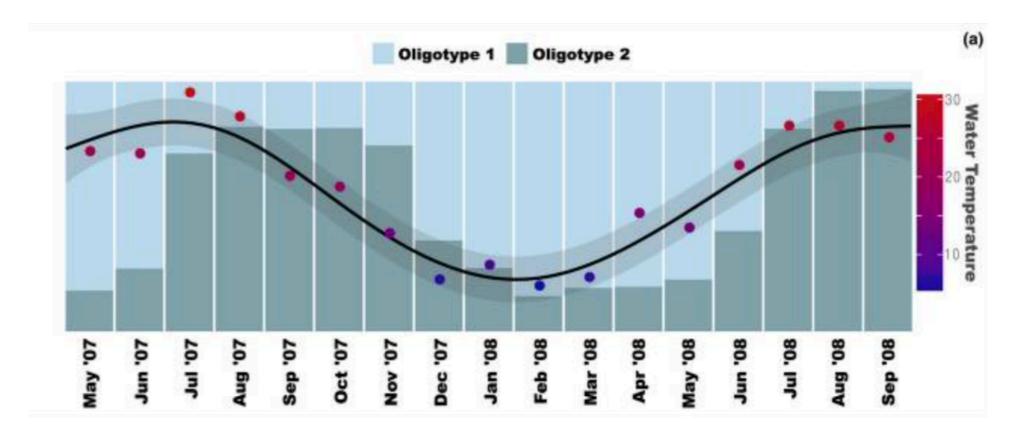
Scientists at the beginning wanted to be close to taxonomic concepts as well as getting rid of sequencing errors. This 97% threshold was thought to correspond to the species level.

Clustered sequences are then called Operational Taxonomic Units (or OTU)



However, defining an arbitrary threshold has some limits

Very similar sequences can have distinct ecological patterns

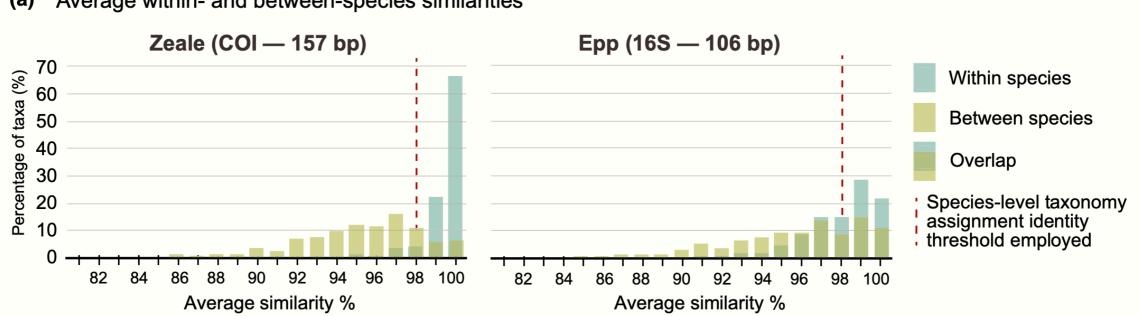


Here the two sequences of 459 nucleotides are identical at 99.57%



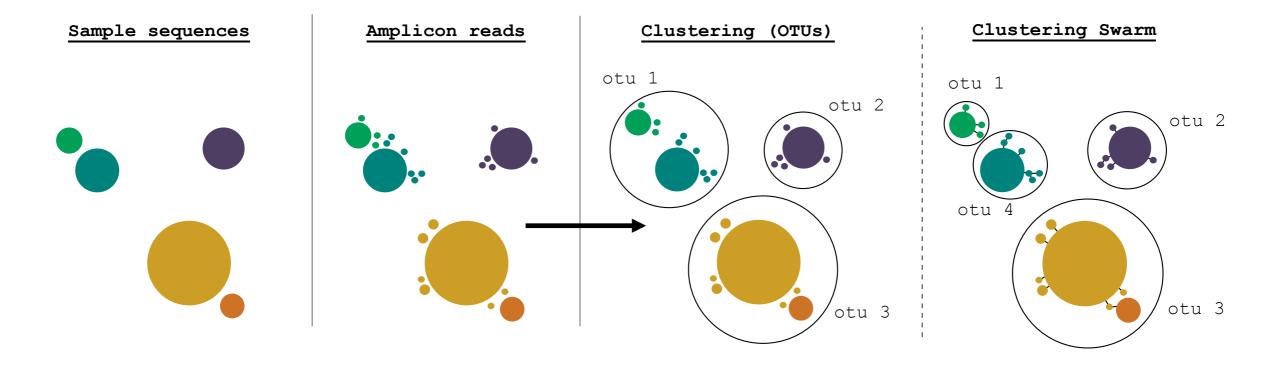
However, defining an arbitrary threshold has some limits

And the similarity threshold varies depending on the species

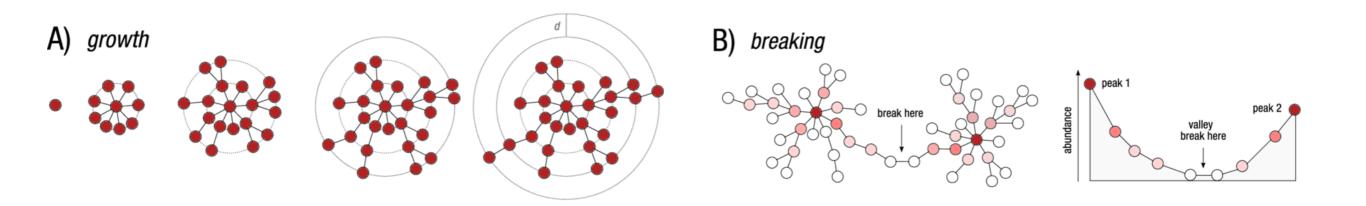


Average within- and between-species similarities (a)



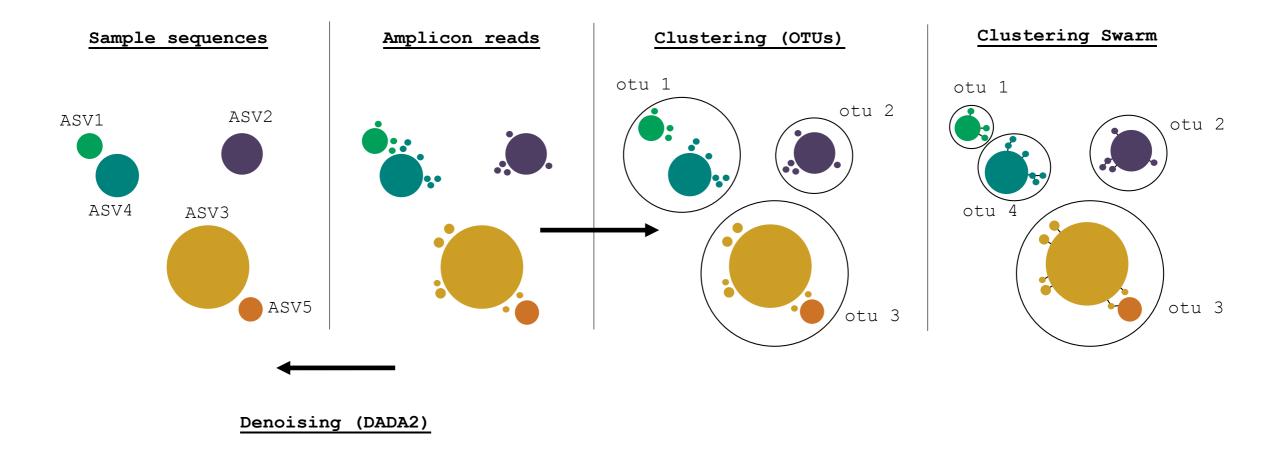


Exemple of a clustering method using a softthreshold : Swarm (Mahé et al., 2014)





2/ Denoising

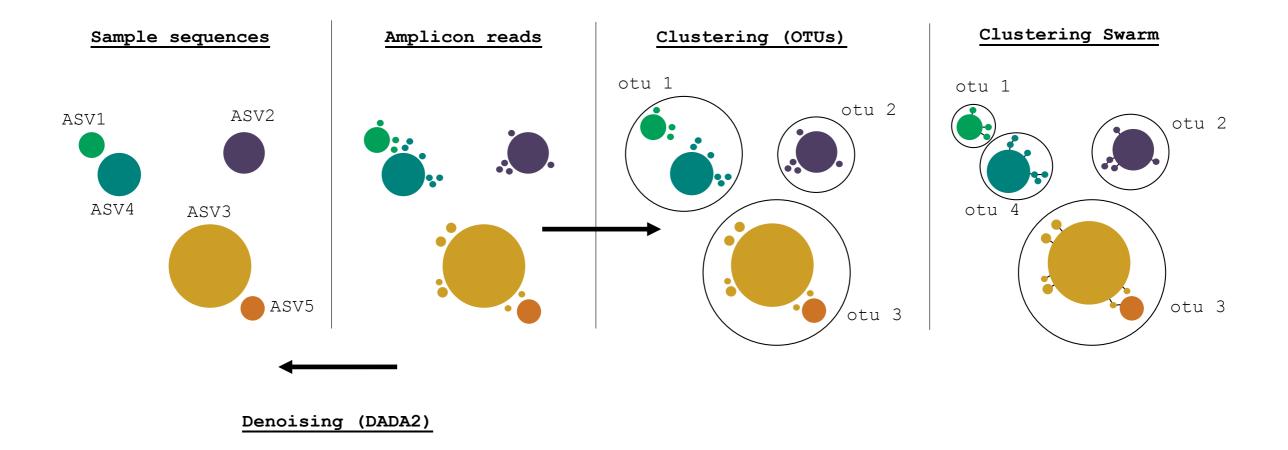


A more recent approach : **denoising** (e.g. DADA2, Deblur, UNOISE3)

Try to correct the sequencing errors directly to retrieve « true » sequences - or Amplicon Sequence Variants (ASV)



2/ Denoising



However, 2 different ASVs can represent the same species (different copies of the same gene)





Clustering or denoising ?

With ASVs you better discriminate ecological patterns but you can also be too efficient (the same genome can contain multiple ASVs as it can have multiple copies of the targeted genetic locus)

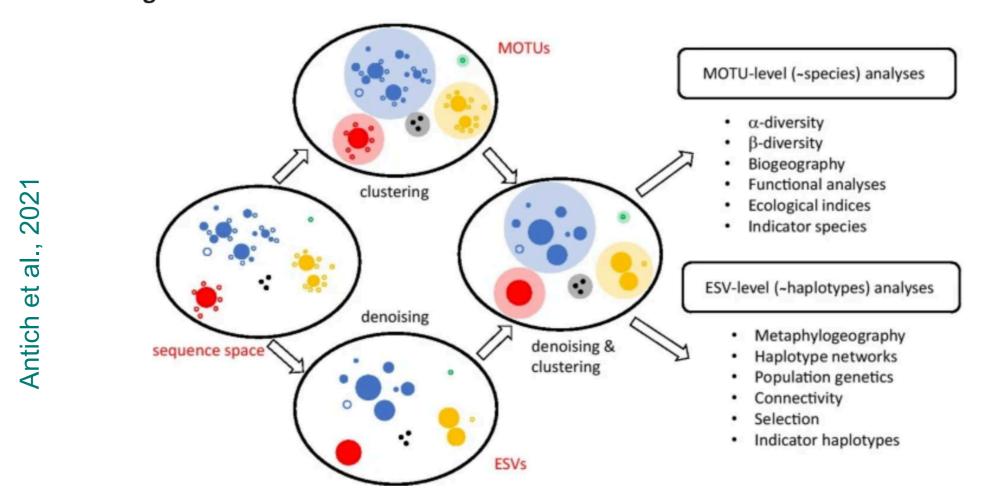
With OTUs you can reduce this variability but with a drawback of missing ecologically important variants.

+ reconstructing OTUs is dataset-dependent - if you analyse more samples, you need to re-run entirely the analysis which is not the case with DADA2 as you infer exact variants.



Clustering or denoising ? It will depend on the question

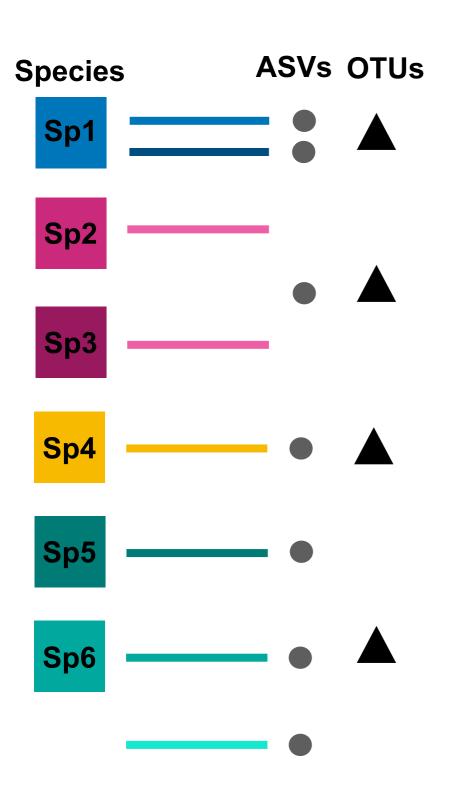
Fig.1



For microbial organisms (protists, phytoplankton, bacteria), denoising with DADA2 is now the most used approach.



OTUs/ASVs are not species



One sequence does not mean one species

Whatever approach you use you don't have species at the end, but operational taxonomic units (OTUs) or amplicon sequences variants (ASVs).



BIOINFORMATIC PIPELINE - the steps

each: function(e, t, n) (
0 = e.length,
af (n) (
17 (a) (for (; 0 > i; i++)
<pre>if (r = t.apply(e[i], n), r === 1) break } else</pre>
<pre>for (i in e) if (r = t.apply(e[i], n), r === [1] break</pre>
<pre>} else if (a) {</pre>
<pre>for (; 0 > i; i++) if (r = t.call(e[i], i, e[i]), r === !1) break</pre>
<pre>} else for (i in e)</pre>
<pre>if (r = t.call(e[i], i, e[i]), r === !1) break; return e</pre>
<pre>}, trim: b && !b.call("\ufeff\u00a0") ? function(e) {</pre>
<pre>return null == e ? "" : b.call(e)</pre>
<pre>} : function(e) { return null == e ? "" : (e + "").replace(C, "")</pre>
<pre>}, makeArrmy: function(e, t) {</pre>
<pre>var n = t []; return null != e && (M(Object(e)) ? x.merge(n, "string" == typeof e ? [e] : e) : h.call(n, e)), n</pre>
indrray: function(e, t, n) {
1f (t) { 1f (m) neturn m.call(t, e, n);
<pre>for (r = t.length, n = n ? 0 > n ? Math.max(0, r + n) : n : 0; r > n; n++) if (n in t 86 t[n] === e) return n</pre>

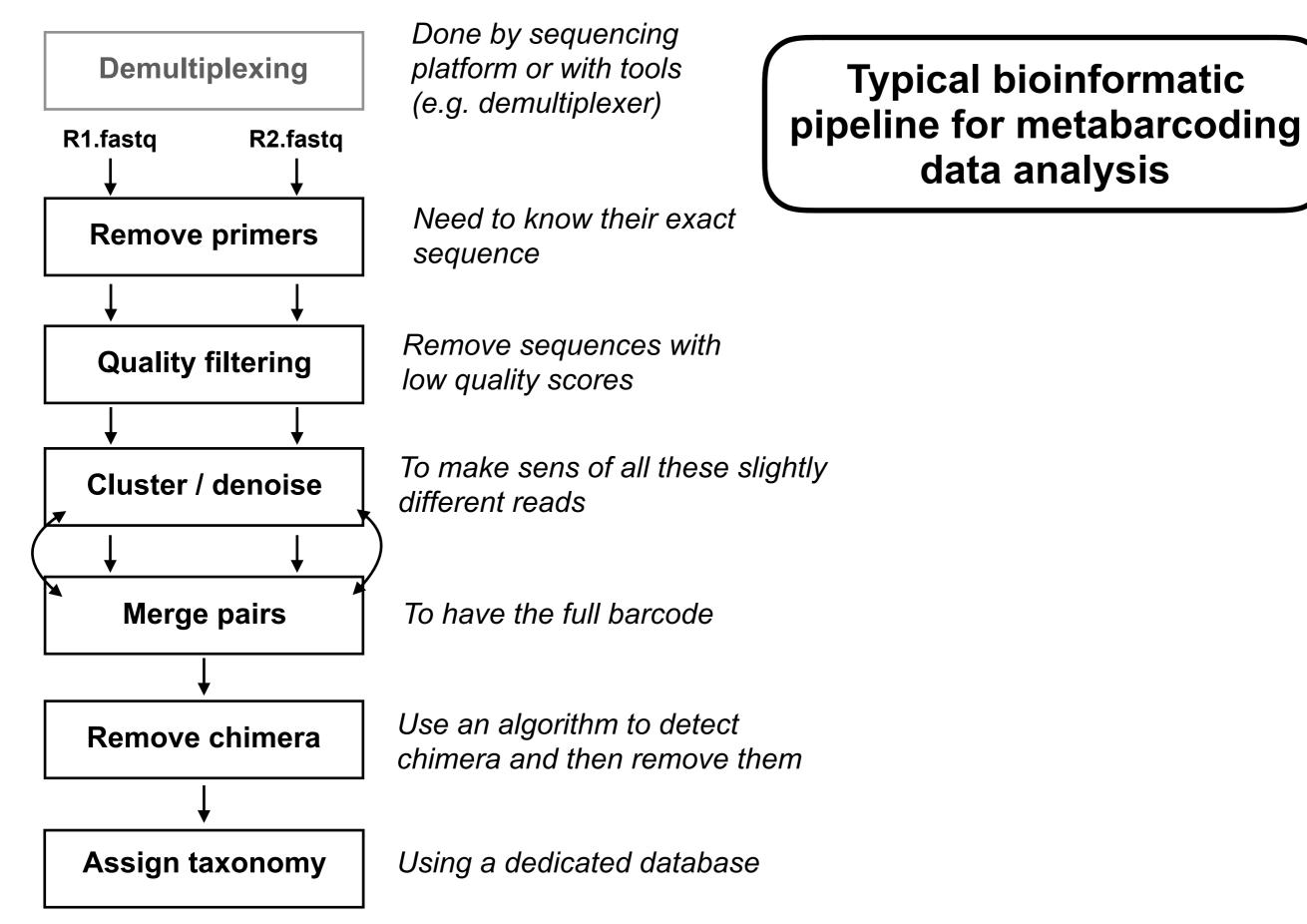
Bioinformatic pipeline





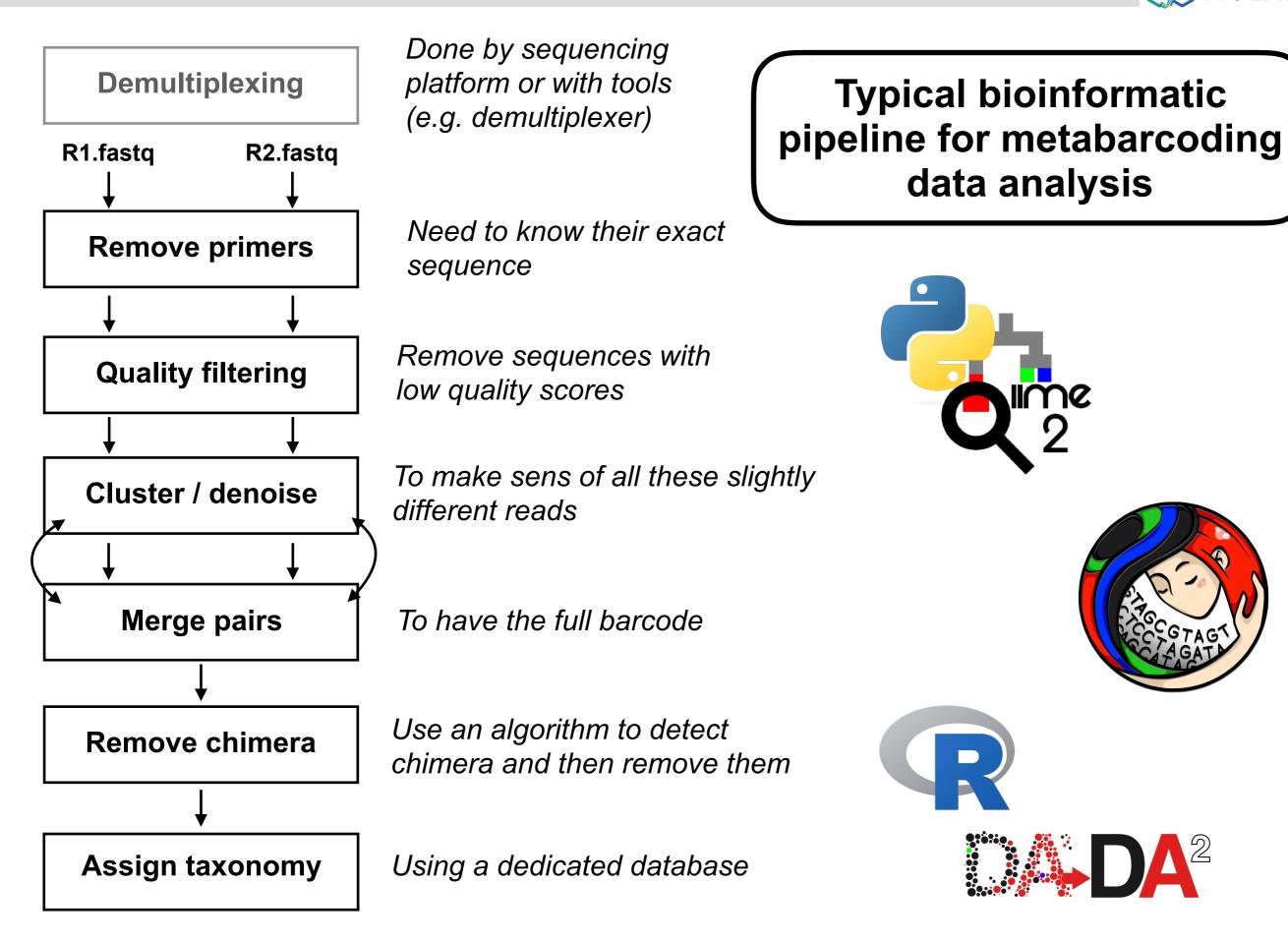
Bioinformatic pipeline





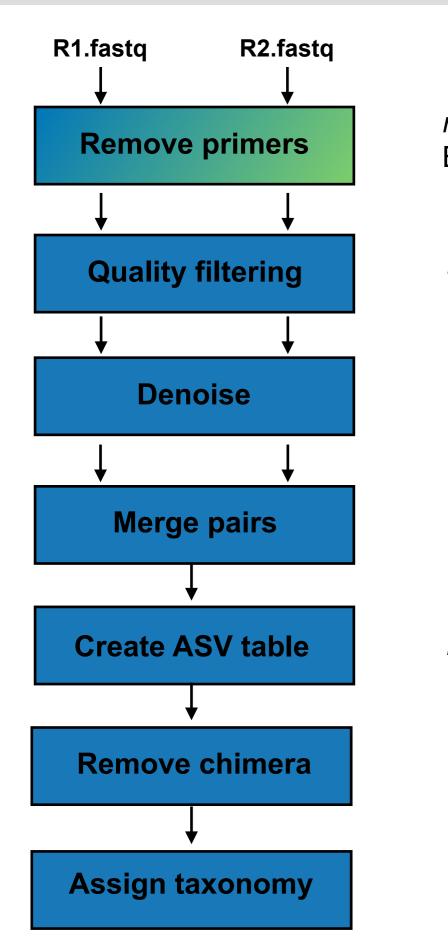
Bioinformatic pipeline





DADA2 pipeline





removePrimers() But cutadapt is better

filterAndTrim()

learnErrors() dada()

mergePairs()

makeSequenceTable()

removeBimeraDenovo()

assignTaxonomy()





marcelm/cutadapt



Cutadapt removes adapter sequences from sequencing reads



ANY QUESTIONS ?



Informations you need before running the pipeline :

- Primers sequences / or size
- Barcode expected length
- Dedicated reference library for your barcode.

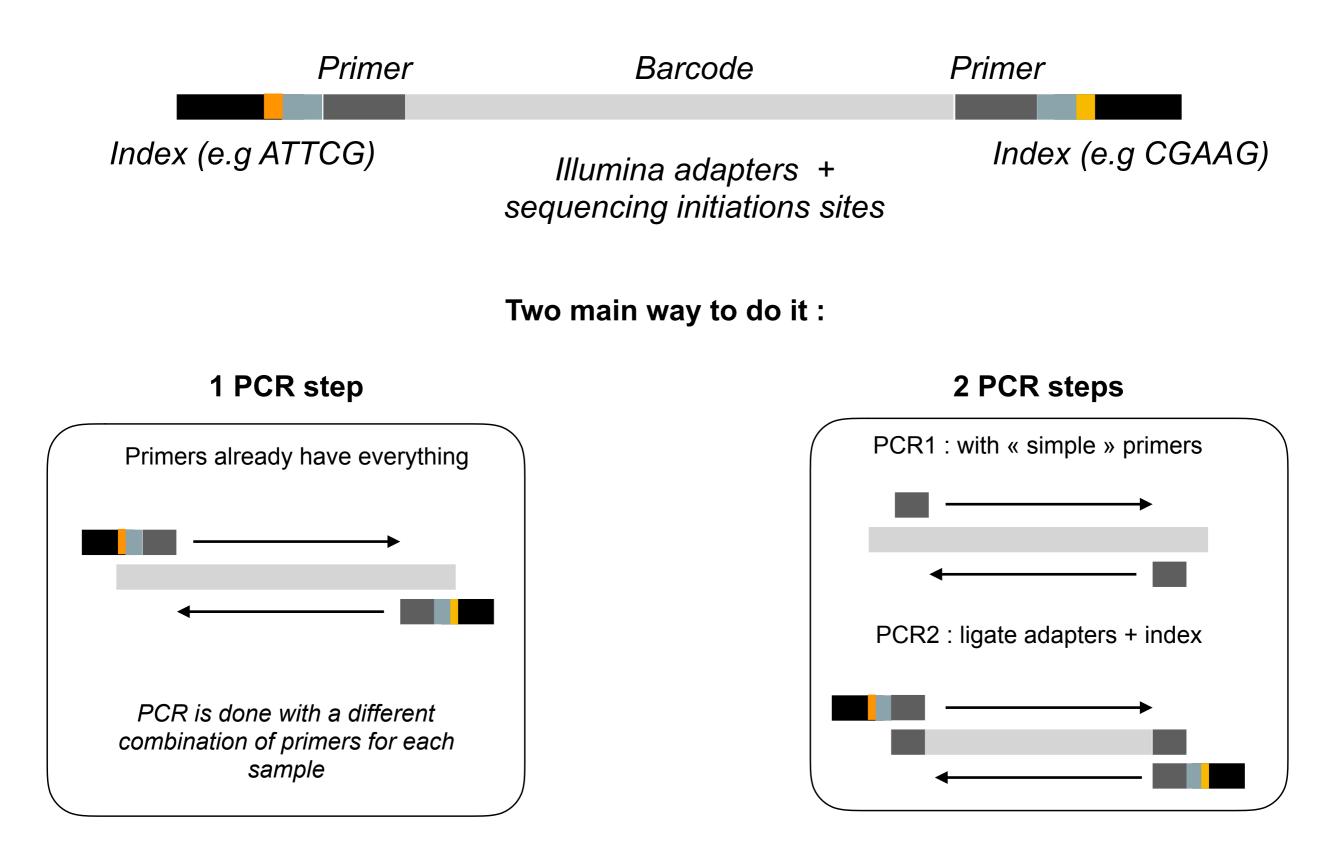
They will be essential for some command lines and also to check if all the steps went well.

Remember : it is one thing to run the pipeline, it is another one to be confident of what you have at the end !

Do not take any result as acquired. You need some checking point to be sure of what you have at the end.

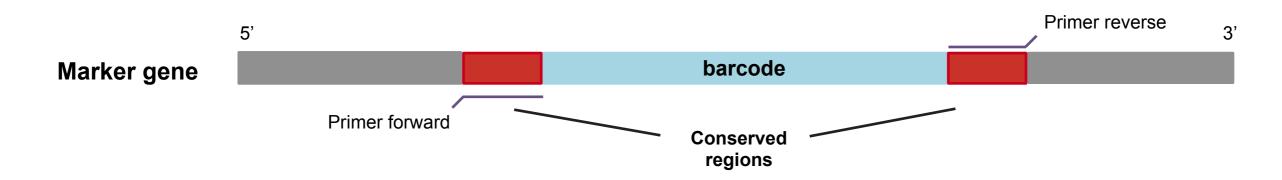


Preparation for Illumina sequencing



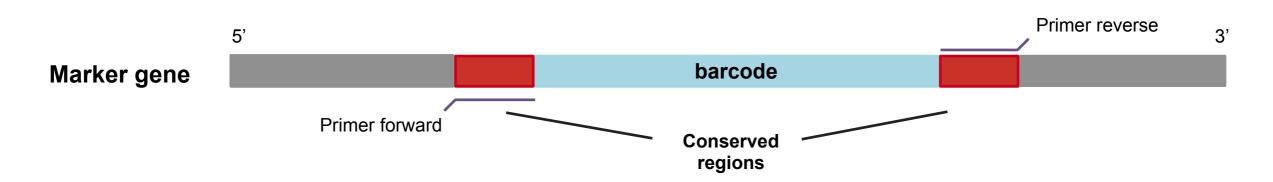


Primers : some reminders

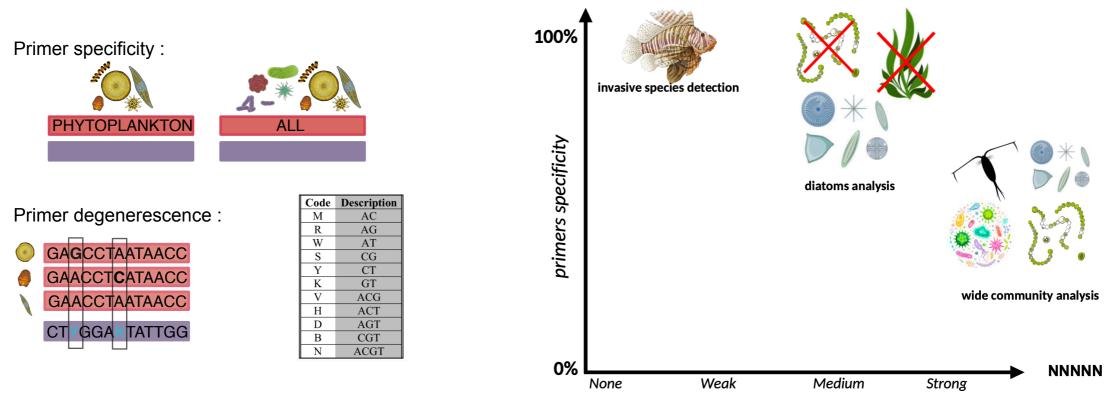




Primers : some reminders



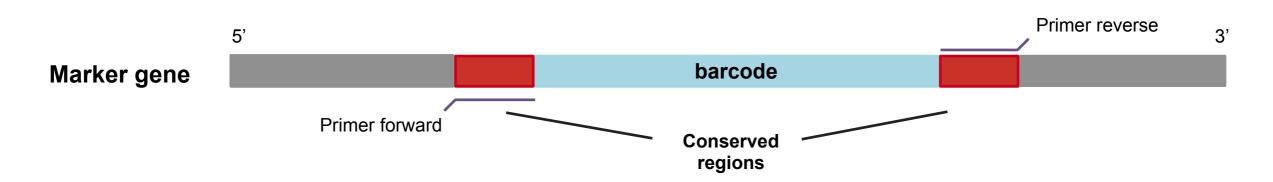
For wide community analysis, they are often degenerated - used in multiple versions



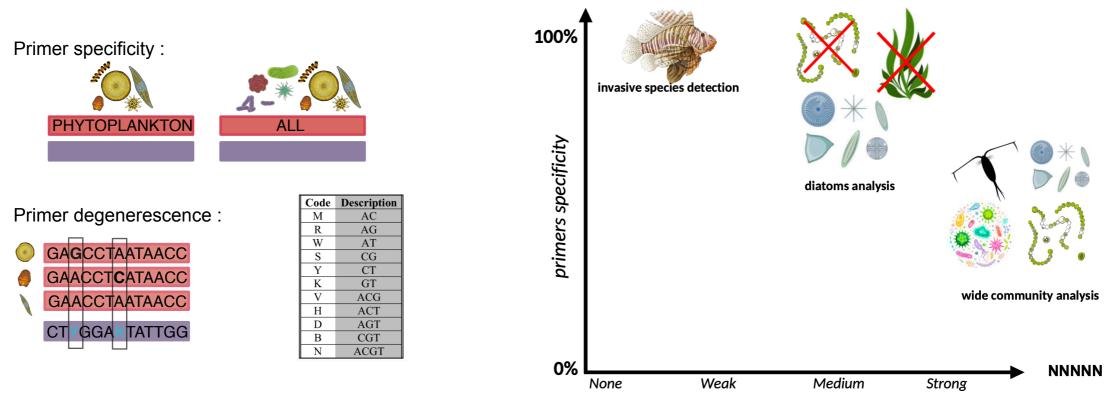
degenerescence level of the primers



Primers : some reminders



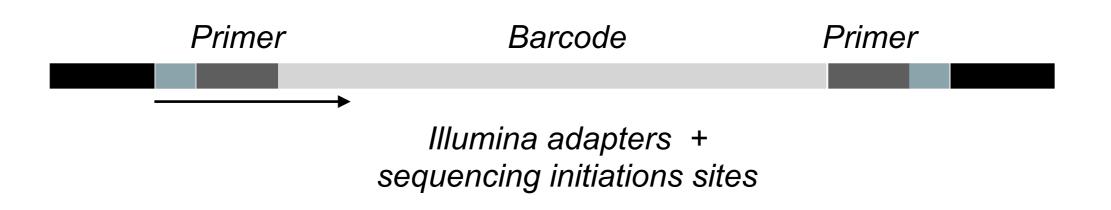
For wide community analysis, they are often degenerated - used in multiple versions



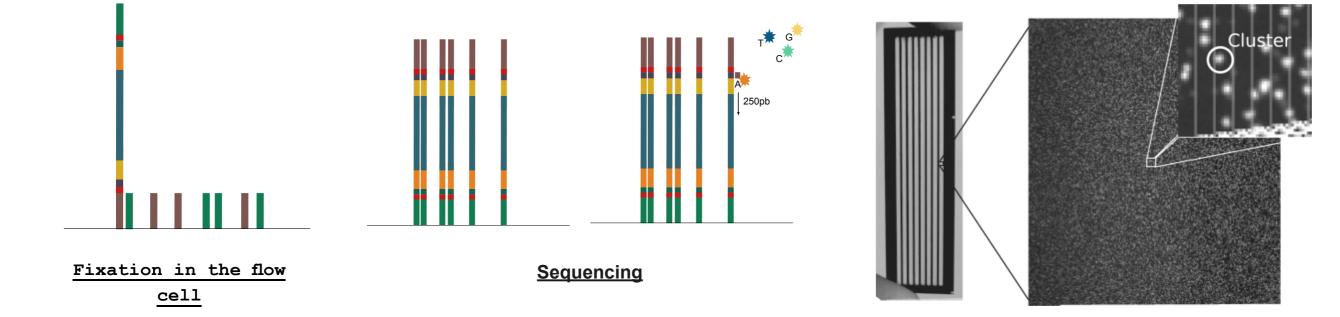
degenerescence level of the primers



Preparation for Illumina sequencing



1. Illumina adapters and sequencing initiation sites



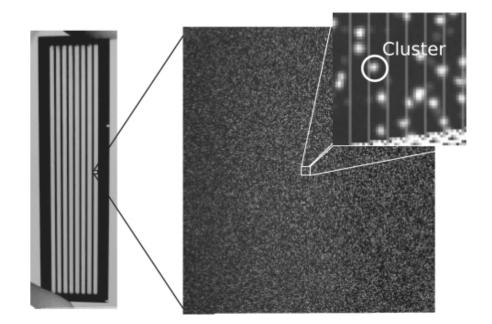


Preparation for Illumina sequencing



2. Index for multiplexing

(for MiSeq, up to 96 samples)



~ 10 000 000 reads (per Ilumina MiSeq run)