



CHARLES UNIVERSITY



# **Bioinformatics and Microbiome Analysis MB140P94**

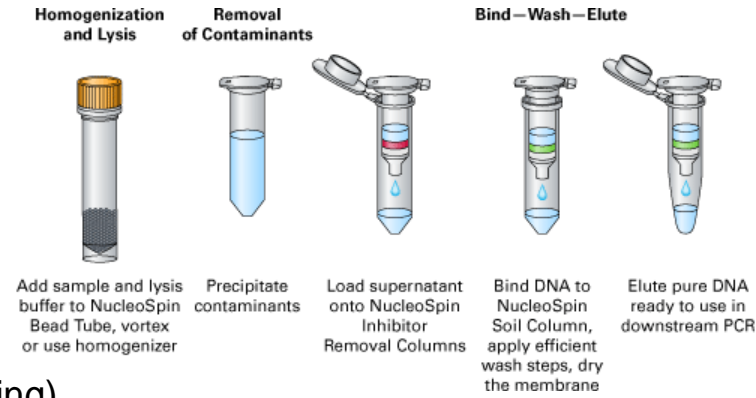
## **Amplicon data**

**Tomáš Větrovský, Iñaki Odriozola and Petr Baldrian**  
**Laboratory of Environmental Microbiology**  
**Institute of Microbiology of the CAS**

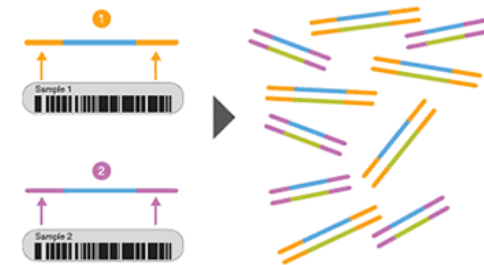
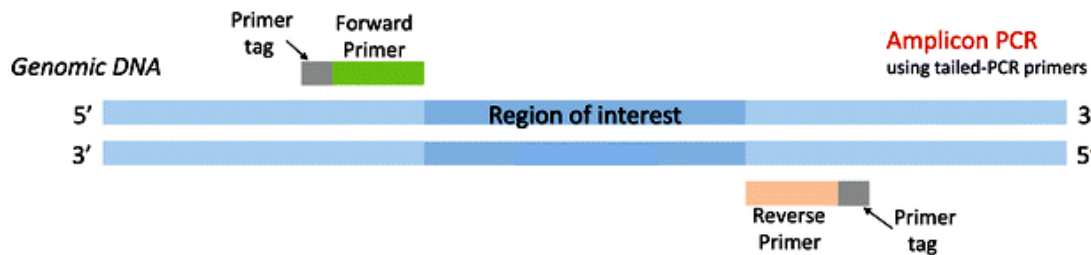


# Illumina amplicon sequence library preparation

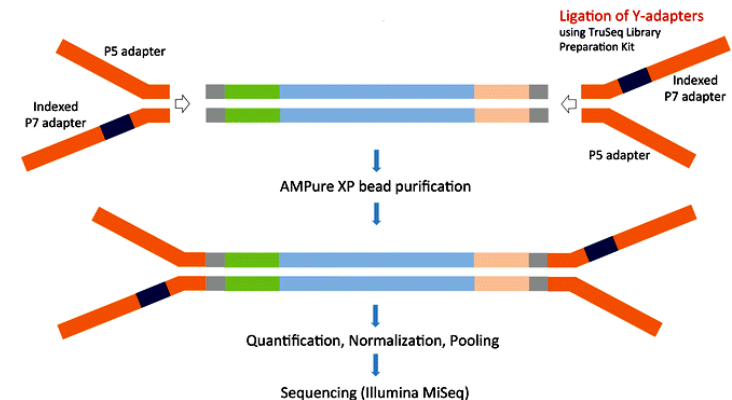
## 1. DNA isolation



## 2. PCR with barcoded primers (multiplexing)



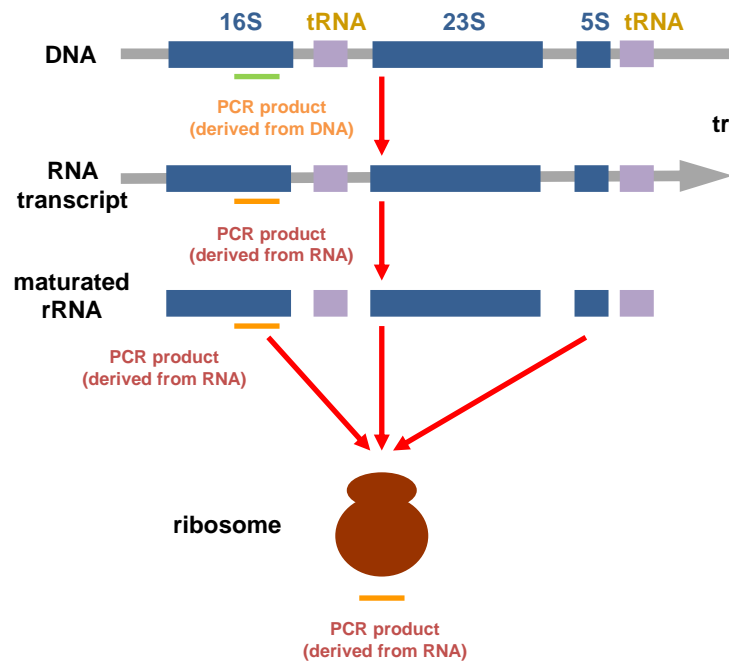
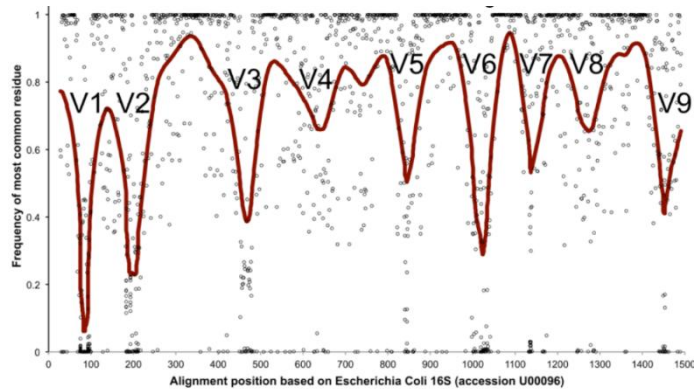
## 3. Ligation of sequencing adapters – to attach short oligonucleotides (60bp) to your DNA fragments, these oligonucleotides are used to attach to the sequencing flow cell and they are also used as barcode of library



## 4. Quantification of the library by qPCR – to quantify of the exact amount of ligated fragments

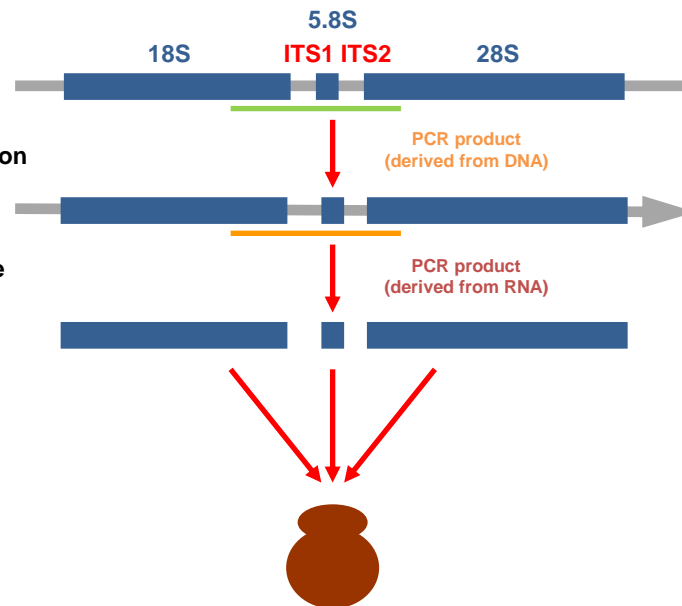
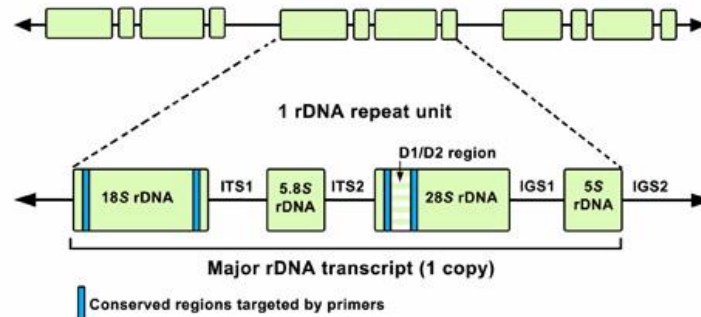
# Most used marker genes

## Bacterial 16S ribosomal RNA gene



1-15 copies of rDNA per genome

## Fungal internal transcribed spacer (ITS)



X0-X00? copies of ITS per genome

Total community

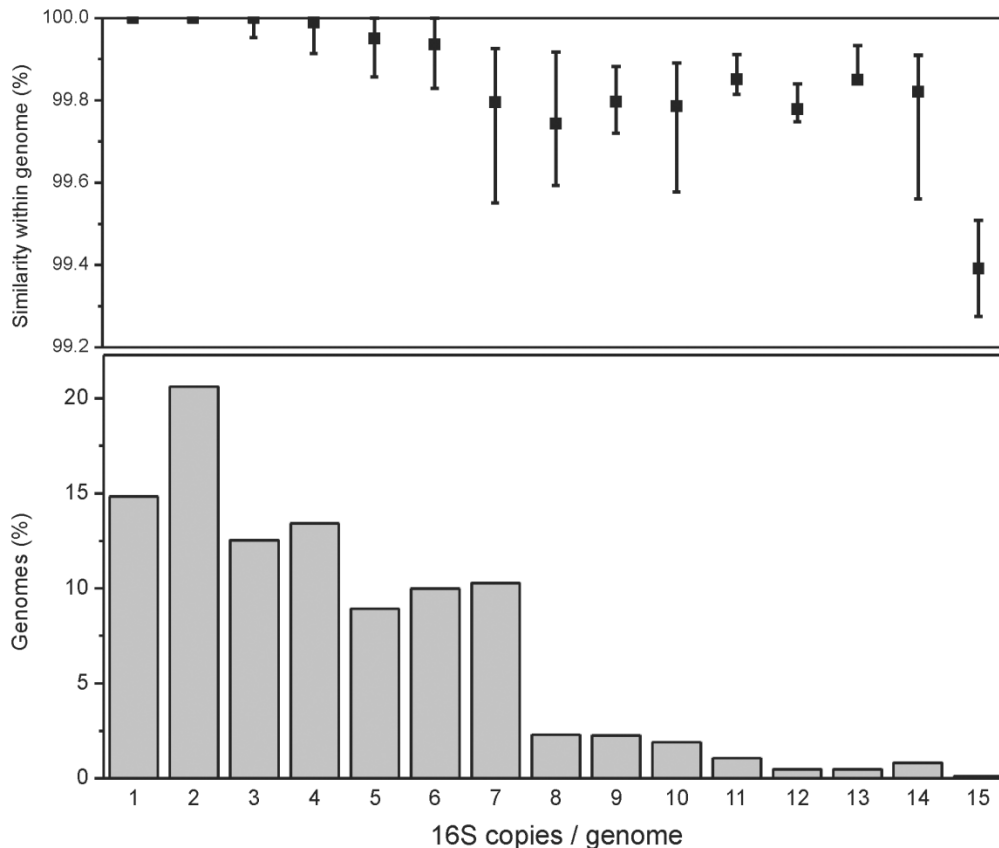
Active community

# 16S rDNA gene vs. alternative (low-copy) markers

**Pros:** highly populated reference databases

**Cons:** multicopy nature of bacterial 16S rDNA gene

- possibility of high intragenomic variability - diversity over estimation (number of OTUs)
- relative abundance estimation is skewed -> normalisation by 16S copy number of closest taxon



16S rRNA within-genome similarity and copy numbers in bacterial genomes.

Upper panel: the similarity of genomes with various copy numbers: the values indicated represent the first, the second and the third quartile.

Lower panel: distribution of 16S rRNA copy numbers per genome in 1,690 sequenced bacterial genomes.

# Sequencing platforms for amplicon sequencing (most used)



## 454 Pyrosequencing

Not supported anymore  
(most studies 2009-  
2012)

Errors in homopolymeric  
regions

long reads  
(up to 700 bp)



illumina®

## Illumina

The most used

Error rate less than  
1%

pair-end data  
(Illumina)



ion torrent  
by life technologies™

## IonTorrent

Very cheap  
sequencing

Lot of errors

Medium read  
size  
200-600 bp



PACIFIC  
BIOSCIENCES™

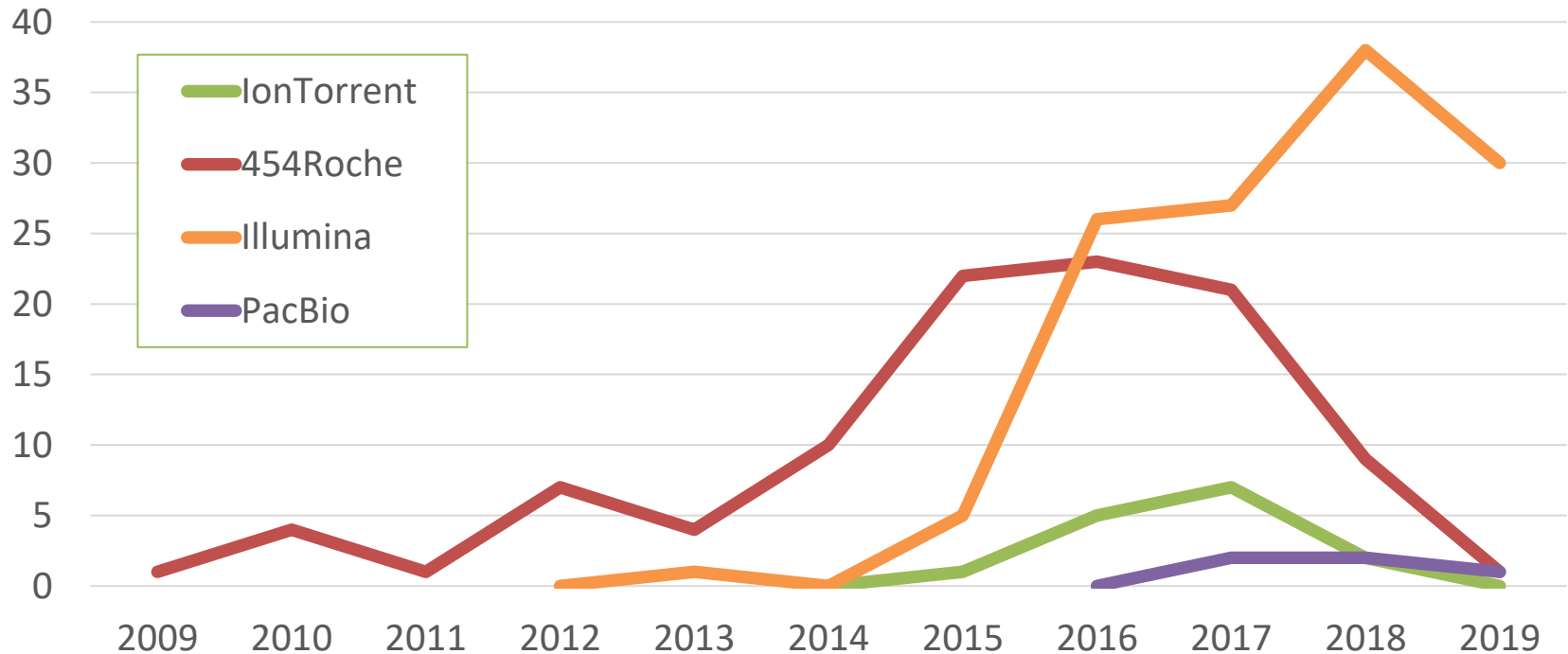
## PacBio

Still rare

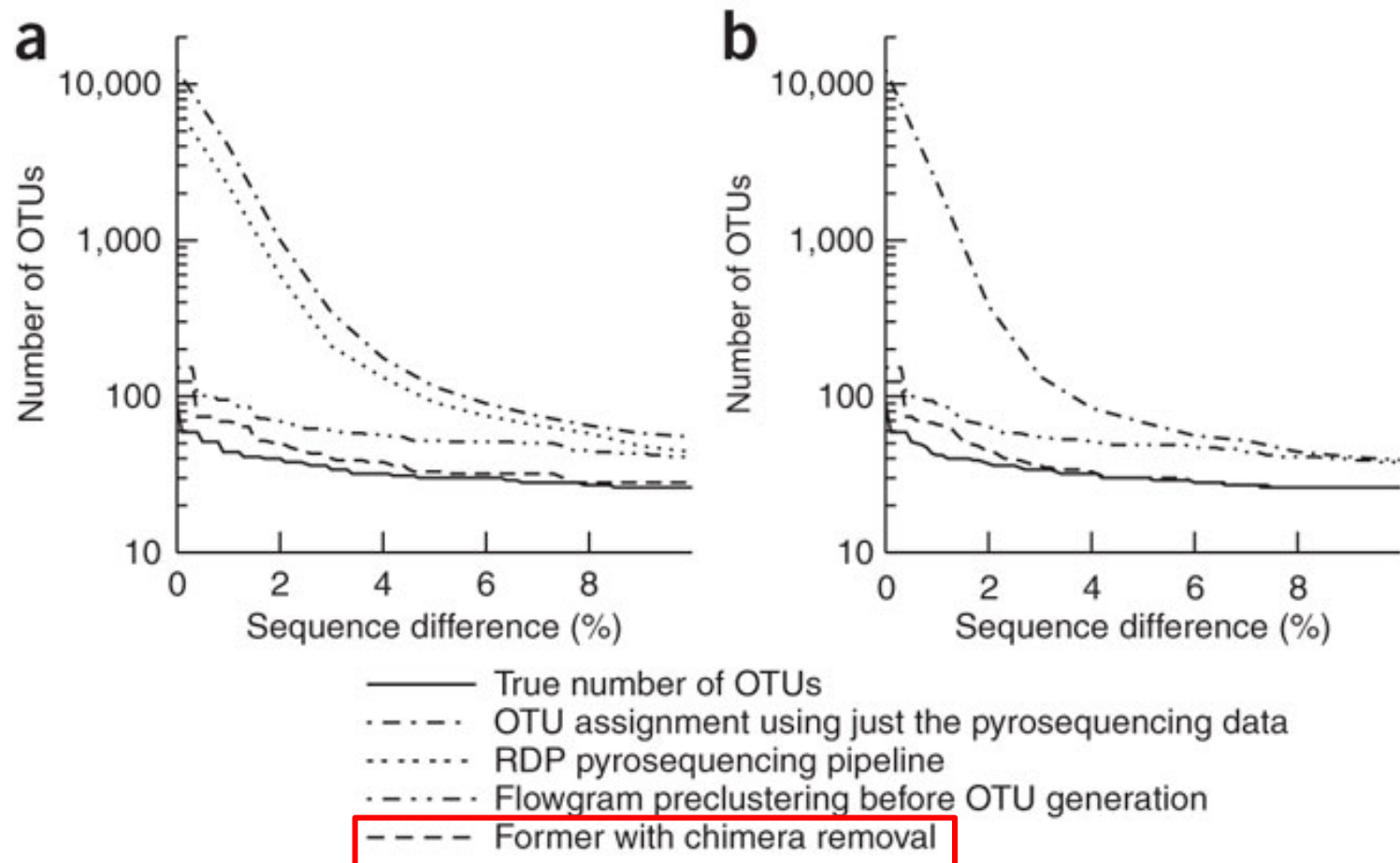
Repeated  
sequencing of the  
same region

Extra long reads  
(up to 10.000 bp)

## Sequencing platforms for amplicon sequencing (based on GlobalFungi data sources)

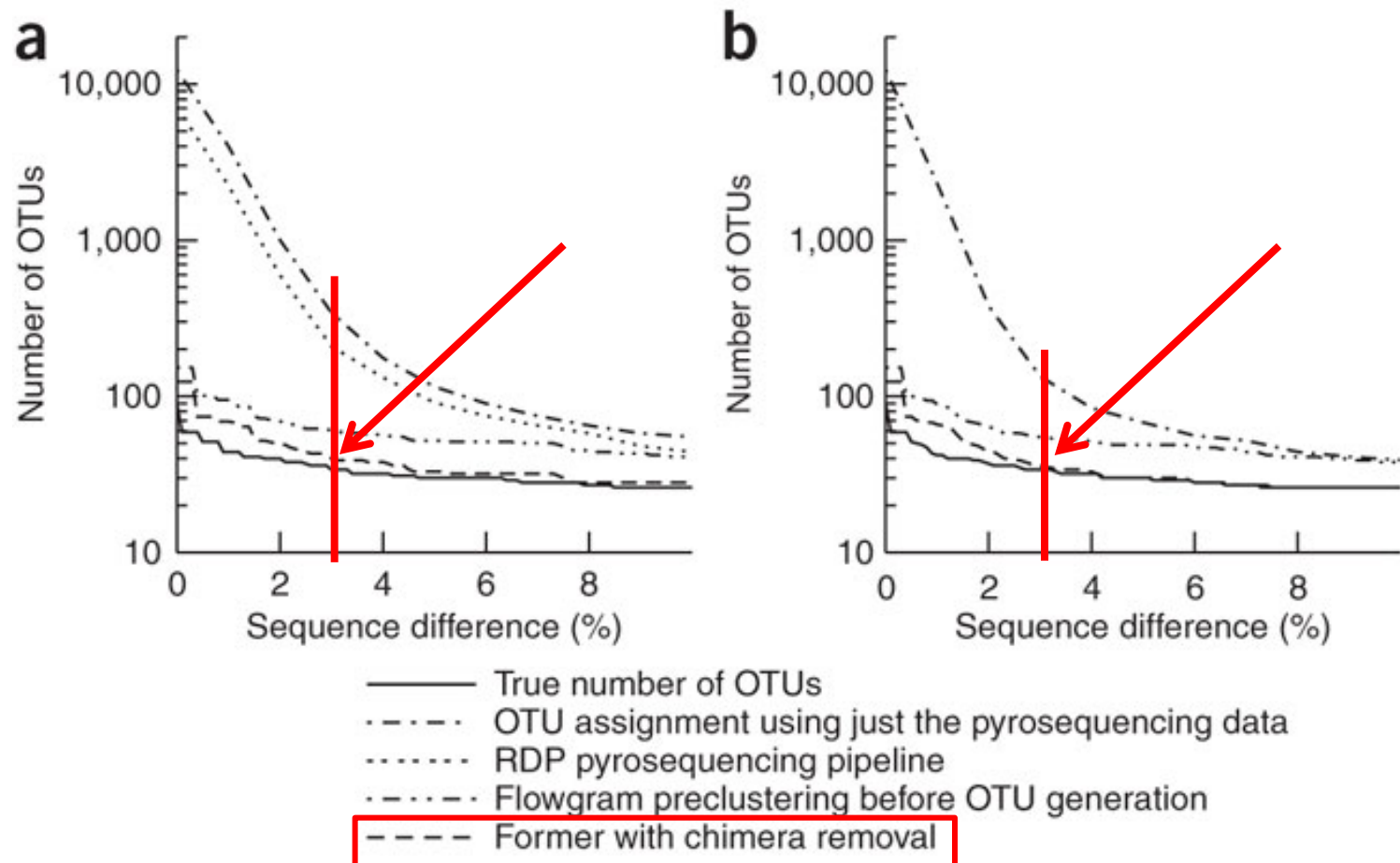


# Searching for true number of taxa (OTUs) in the data



Quince, Christopher, et al. "Accurate determination of microbial diversity from 454 pyrosequencing data." *Nature methods* 6.9 (2009): 639.

# Searching for true number of taxa (OTUs) in the data




Quince, Christopher, et al. "Accurate determination of microbial diversity from 454 pyrosequencing data." *Nature methods* 6.9 (2009): 639.



This is still a big concern to all microbial community analysis pipelines!



**Mick Watson**  
@BioMickWatson


Seguindo

PhD student put 21-OTU mockrobiota sample through bioinf SOPs:  
  
QIIME: over 1000 OTUs!  
Mothus: over 500 OTUs!  
  
Be afraid - be very afraid!

Traduzir do inglês

Retweets  
**76**

Curtidas  
**105**





07:00 - 16 de jun de 2017

21

76

105

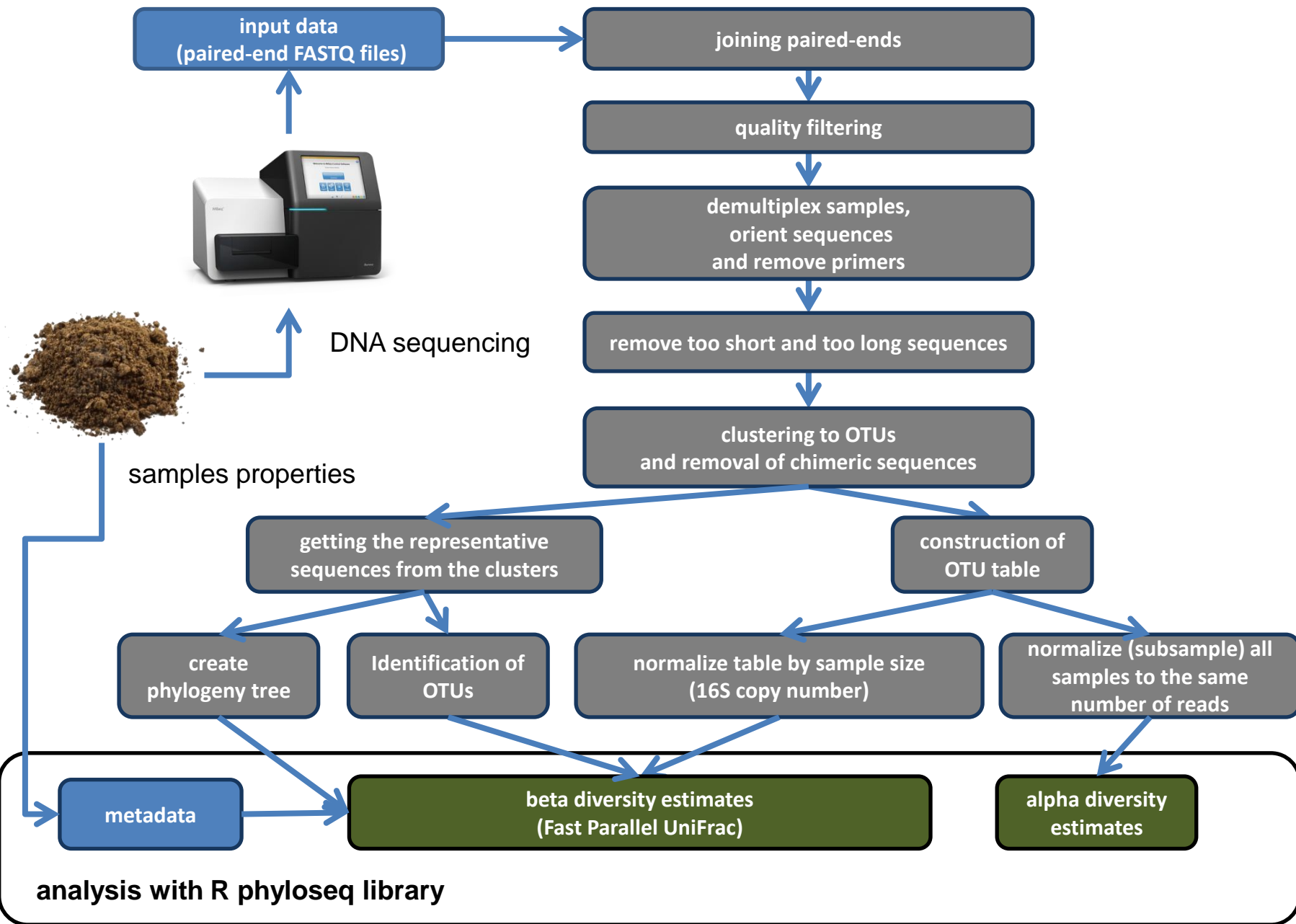
Tweete sua resposta

**Mick Watson** @BioMickWatson · 16 de jun

Em resposta a @BioMickWatson  
mostly low-abundance OTUs, but still.

Traduzir do inglês

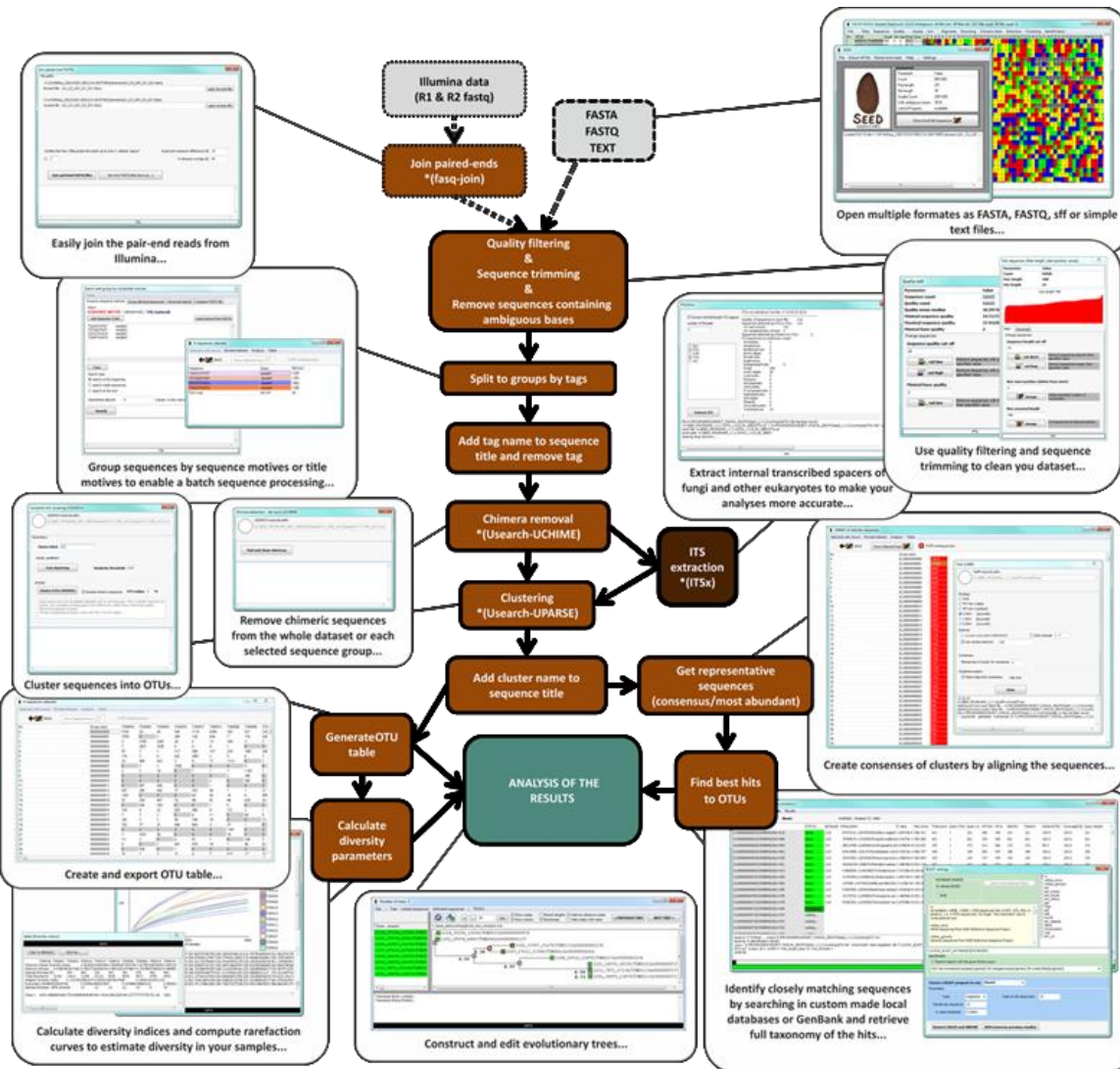
# Amplicons pipeline workflow



# GUI based alternative for Windows (<http://www.biomed.cas.cz/mbu/lbwrf/seed/>)

## SEED 2: a user-friendly platform for amplicon high-throughput sequencing data analyses

- editing of sequences and their titles
- sorting
- quality trimming
- pair-end joining
- grouping of sequences based on sequence motifs or sequence titles
- batch processing of sequence groups
- denoising
- chimera removal
- ITS extraction
- sequence alignments and clustering
- OTU table construction
- construction of consensus sequences
- creation of local databases for BLAST
- searching either local databases or the whole NCBI
- retrieval of taxonomical classification from the NCBI
- calculation of diversity parameters
- many more...



## SEED is alternative to

<https://qiime2.org/>

QIIME 2™ is a next-generation microbiome bioinformatics platform that is extensible, free, open source, and [community developed](#).

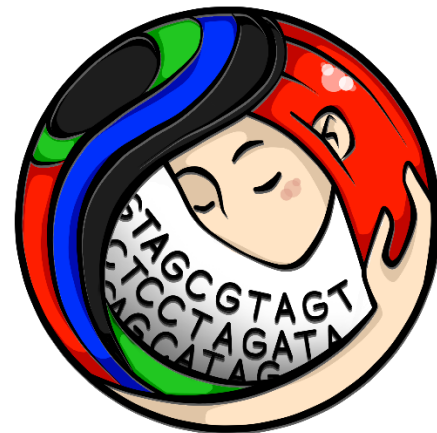


- command line, Unix (Linux) based

<https://mothur.org/>

mothur - one of the most widely used tools for analyzing 16S rRNA gene sequence data

- command line, multiplatform



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## BAC\_R1.fastq - first sequence

@M03794:8:000000000-AJCUU:1:2114:9990:17907 1:N:0:7

AACAGCCGGACTACTGGGGTTTCTAATCCTGTTTGCTCCCCACGCTTTCGTGCCTCAGTGTCAATGACCGTGTAGC  
AAGCTGCCTTCGCAATTGGTGTTCTATGTCATATCTAAGCATTTACCGCTACATGACATATTCGCTTACCTCCAC  
GATATTCAAGACTAATAGTATCAATGGCAGTTCCCAAGTTAAGCTCGGGGATTTCACCACGGACTTACTAGCCCACC  
TACGCACCCCTTAAACCCAGT

+

BCCCCFCCCCCGGGGGGGFGGHHHCHHHHHHHHHHHHHHGG2FGGGGGHGHGGHHHFFHHHHHHHHHHHHHGGH  
HHHHHHHHHHHHHHHHHGGGGHHHHHGHGHHHHHHHHHHFFHHHHHHHHHGHHHHHHGGGGHHHHHGHHHHHHHHHHGG  
GGHHHHHHHGCDFDHHGGHHHHHHHHHGG>GGGHHG/GHGHHHHHHFFHGHGHGHG?DGGGGGGGGGGGGGAC  
GGGGGGGGGGFGGFAAFF;@ADFFFFFFB/FFFFF;

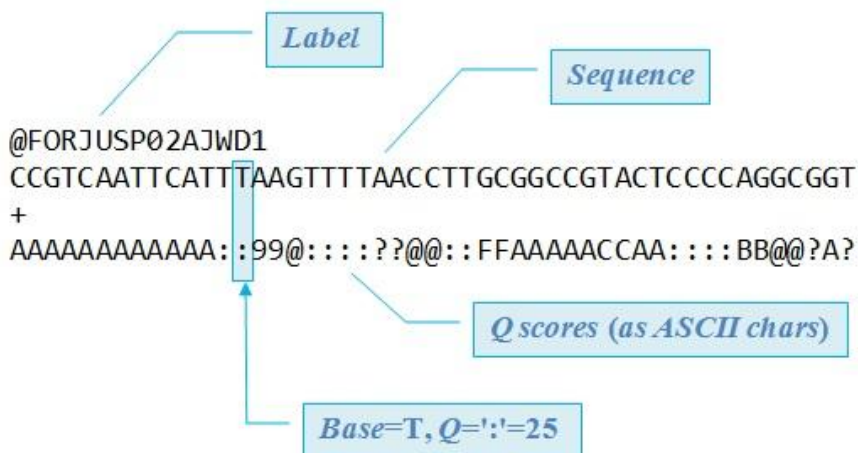
## BAC\_R2.fastq - first sequence

@M03794:8:000000000-AJCUU:1:2114:9990:17907 2:N:0:7

ACGAAGTGTGCCAGCAGCCGCGGTAATACGGAGGGTGCAAGCGTTATCCGGATTCACTGGGTTTAAAGGGTGC  
GTAGGTGGGCTAGTAAGTCCGTGGTGAATCCCCGAGCTTAACCTGGGAAGTGCATTGATACTATTAGTCTTGA  
ATATCGTGGAGGTAAGCGGAATATGTCATGTAGCGGTGAAATGCTTAGATATGACATAGAACACCAATTGCGAAG  
GCAGCTTGCTACACGGTCATTGACACTG

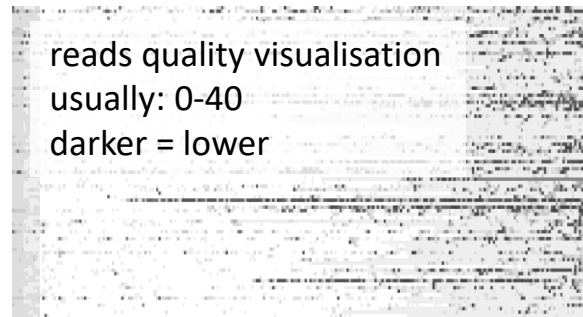
+

BBBBBBBFFBFDGFFFGGGCGGCGGGHHHGHDEDEDGGGFGGHHHGGGGGFDEE?EEGBGFHHHHGFFGGFFGGFE  
FGDFDGFEGGHHHGFEGFHFHEEFFFHHHHFHGGGFGFHHHFFHHHHHHBGHHFBCHFBGGGHHHHHGHHHHH  
HGFFHHHHGHGD<GGAGHHHGGG@CFHHFCGHH:CCG?AAAGFEFEFGGGBFFFFFFFGBBFGFFBDE?BBBB.@9  
-9..A.B/:AFFFEF.@;AAF//99FFF/

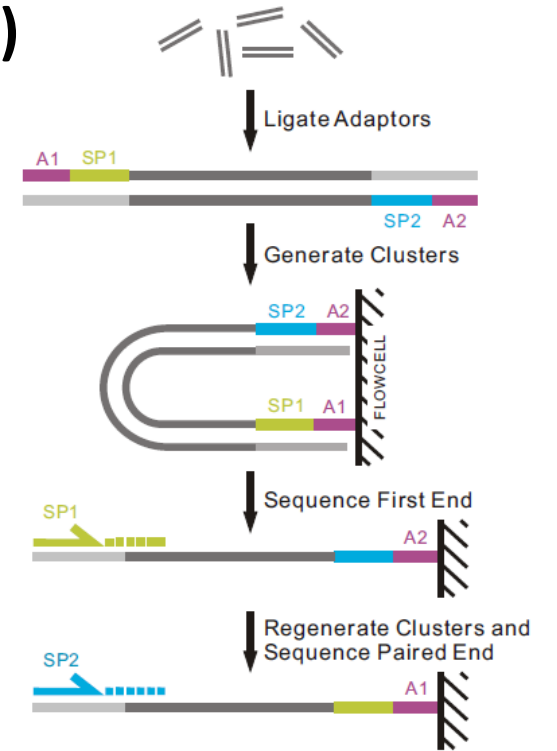
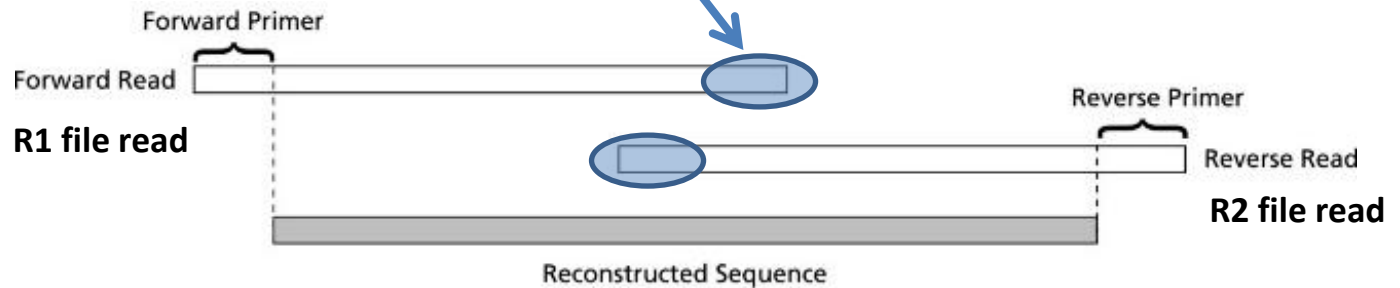


## RAW DATA

# Joining of pair-end data (Illumina)



reads quality is dropping at the ends



reconstructed sequence are based on bases with higher quality

PROGRAM: **FastqJoin**

set a minimal overlap length and overlap precision

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# Sequence quality and quality filtering

## FASTQ format



$$Q_{illumina} = -10 \times \log_{10} \left( \frac{P_e}{1 - P_e} \right),$$

where  $P_e$  is the probability of identifying a base incorrectly.  
For Sanger and other platforms, the formula is as follows [8]:

$$Q_{PHRED} = -10 \times \log_{10}(P_e).$$

$$Q_{illumina} = 10 \times \log_{10} \left( 10^{\left\{ \frac{Q_{PHRED}}{10} \right\}} + 1 \right)$$

Table 2. Phred quality scores are logarithmically linked to error probabilities ([http://en.wikipedia.org/wiki/Phred\\_quality\\_score](http://en.wikipedia.org/wiki/Phred_quality_score))

Phred quality score	Probability of incorrect base call	Base call accuracy
10	1 in 10	90%
20	1 in 100	99%
30	1 in 1000	99.90%
40	1 in 10 000	99.99%
50	1 in 100 000	99.999%
60	1 in 1 000 000	99.9999%

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# Multiple samples in one library (Multiplexing)

name	FWDprimer	REVprimer
SAMPLE001	515F_T103	806R_T007
SAMPLE002	515F_T002	806R_T052

**spacer** is not presented in native sequences, it is used to prevent overestimation of any taxa

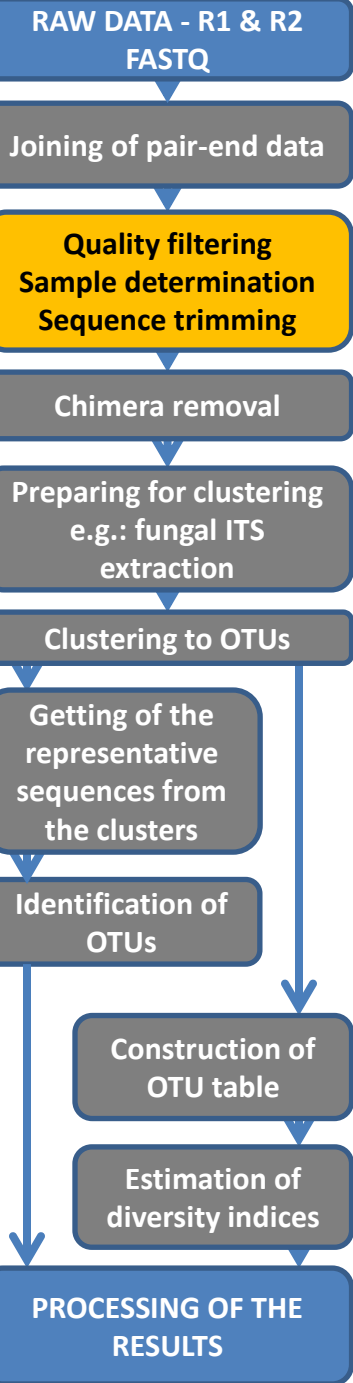
**TAG SPACER ORIGINAL PRIMER**

515F\_T103  
515F\_T002

806R\_T007  
806R\_T052

AATATACGTGTGCCAGCMGCCGCGGTAA  
ACGAAGTGTGCCAGCMGCCGCGGTAA

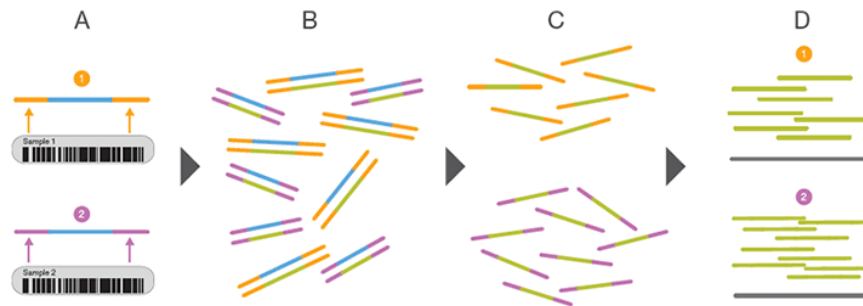
AGCCACCGGACTACHVGGGTWTCTAAT  
ATCCTCCCGGACTACHVGGGTWTCTAAT





# Sample determination (de-multiplexing)

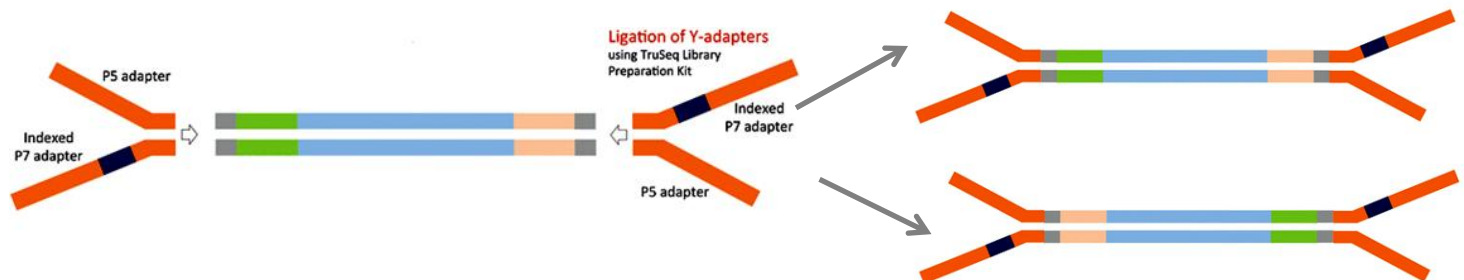
## Demultiplex samples



put sample names to  
sequence titles

```
>SAMPLE034|M03794:8:00000000  
00-AJCUU:1:2114:9990:17907  
CCTGTTTGCTCCCCACGCTTTC  
GTGCCTCAGTGTCAATGACCGT  
GTAGCAAGCTGCA...
```

## Orient sequences



cca 50 % of the reads are  
reverse complement oriented  
due to ligation of library adapters

## Remove primers

since primer sequences are not native to the sample, they need to be removed before clustering to OTUs

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RESULTS

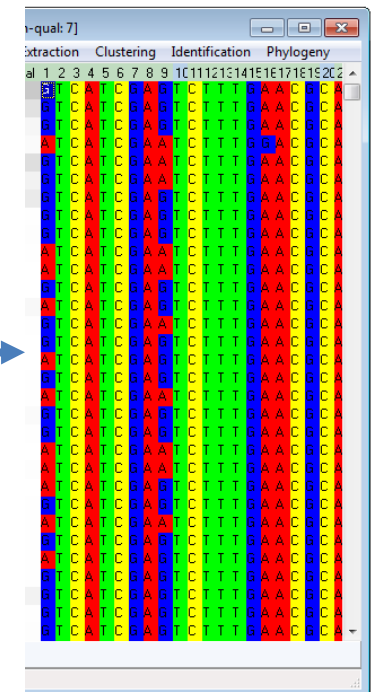
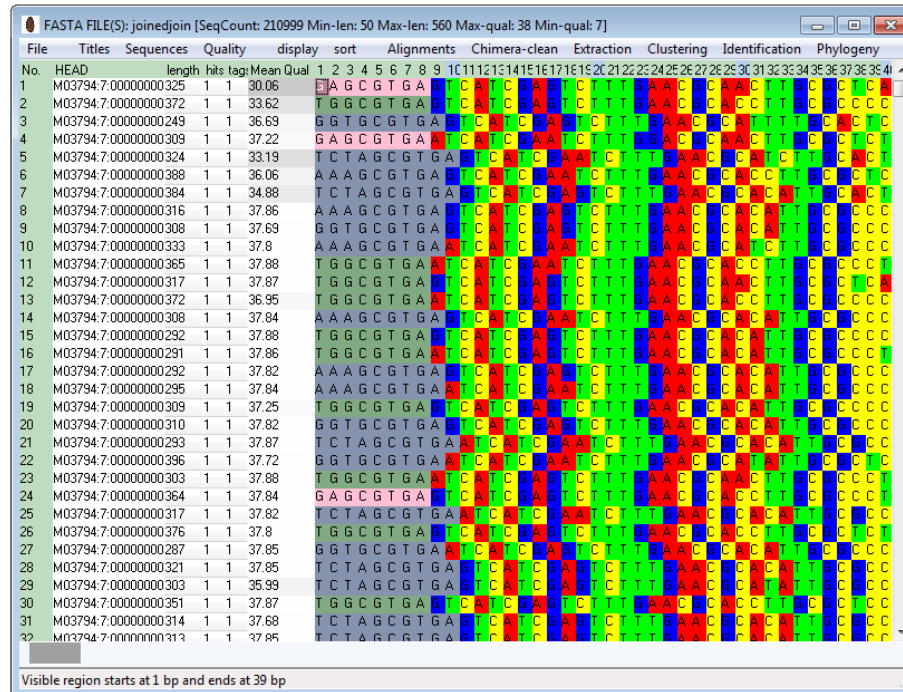
# Sample determination (de-multiplexing) and removing barcodes and primers

```
GAGCGTGA    gITS7_T02
TGGCGTGA    gITS7_T03
GGTGCGTGA   gITS7_T06
AAAGCGTGA   gITS7_T08
TCTAGCGTGA  gITS7_T10
```

search for the sample barcodes  
at the beginning of reads

Sequence	Query	RESULT
GAGCGTGA	gITS7_T02	14687
TGGCGTGA	gITS7_T03	22568
GGTGCGTGA	gITS7_T06	16835
AAAGCGTGA	gITS7_T08	19258
TCTAGCGTGA	gITS7_T10	20585

remove barcode after putting the name of sample to sequences header...

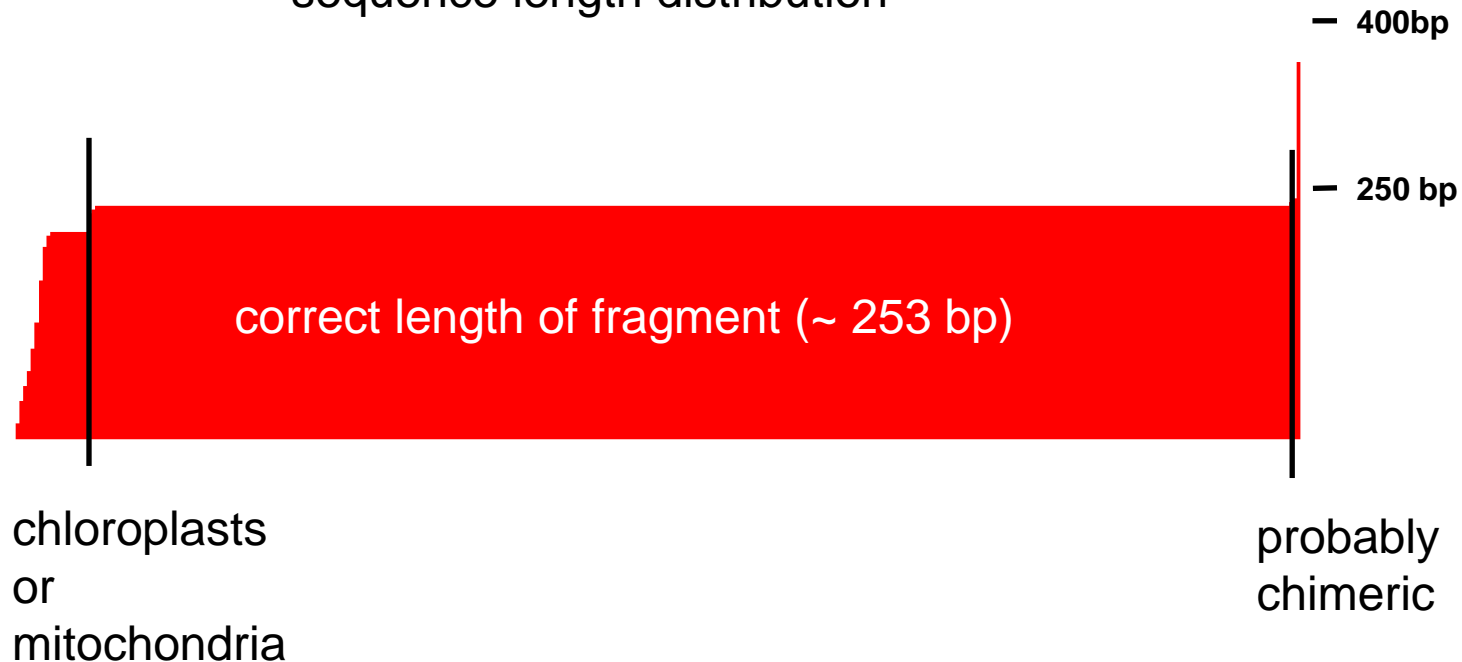


# Sequence trimming

removing sequences with aberrated length - depends on the marker gene

Too short – nonspecific PCR products/erroneous sequences  
Too long - nonspecific PCR products/chimeric sequences

sequence length distribution



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# Chimera removal

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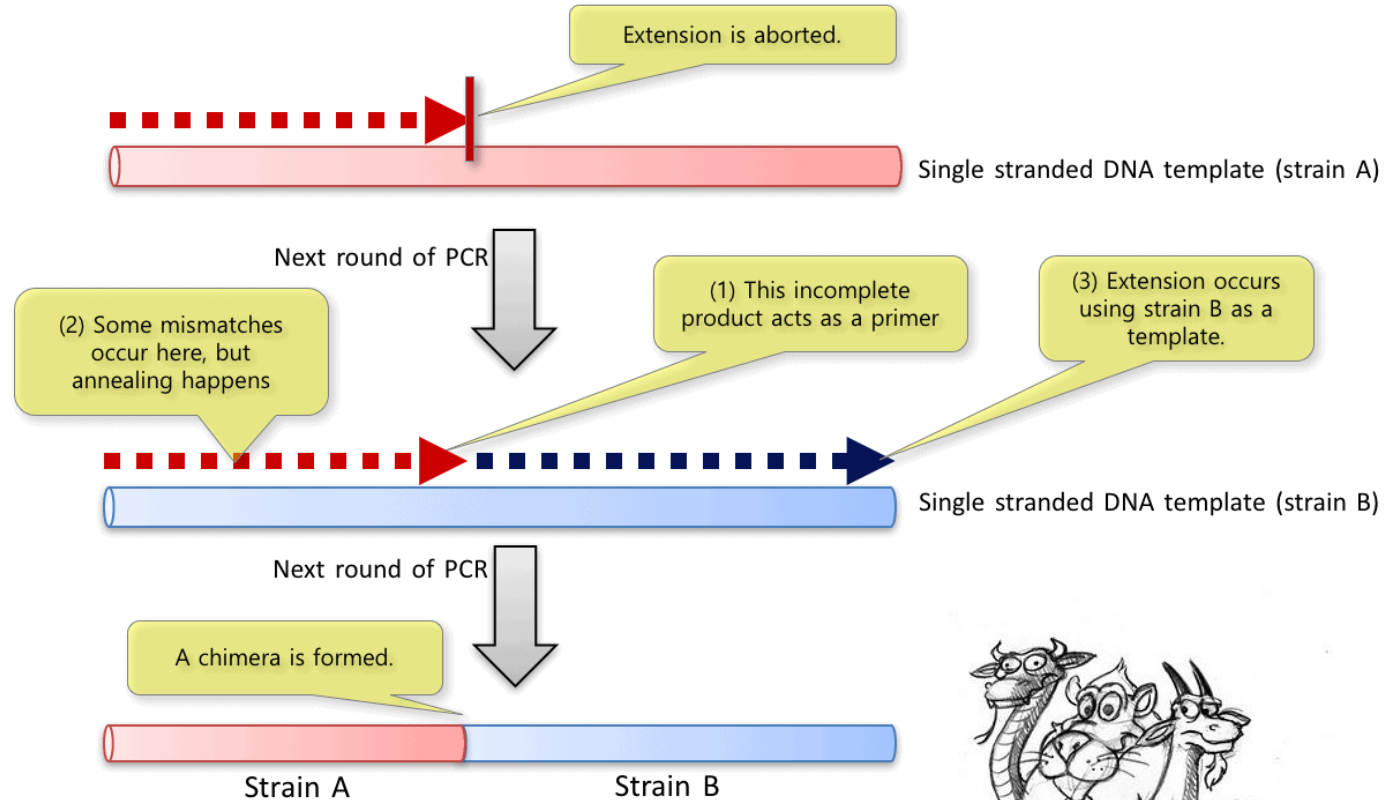
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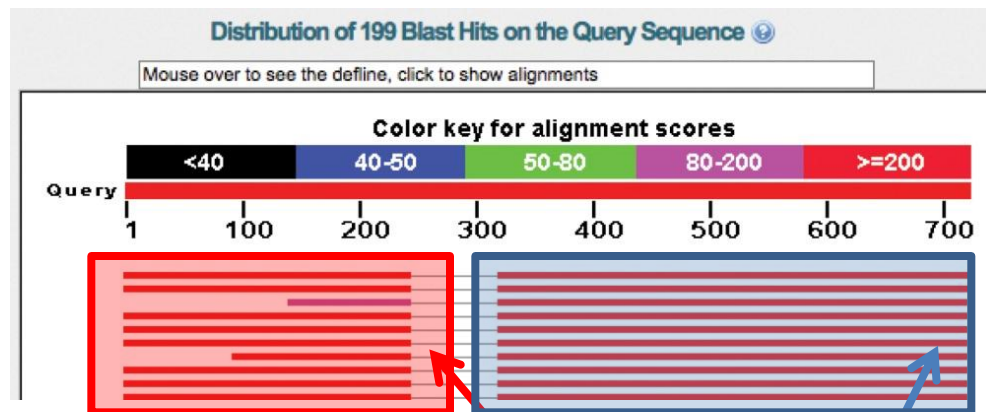
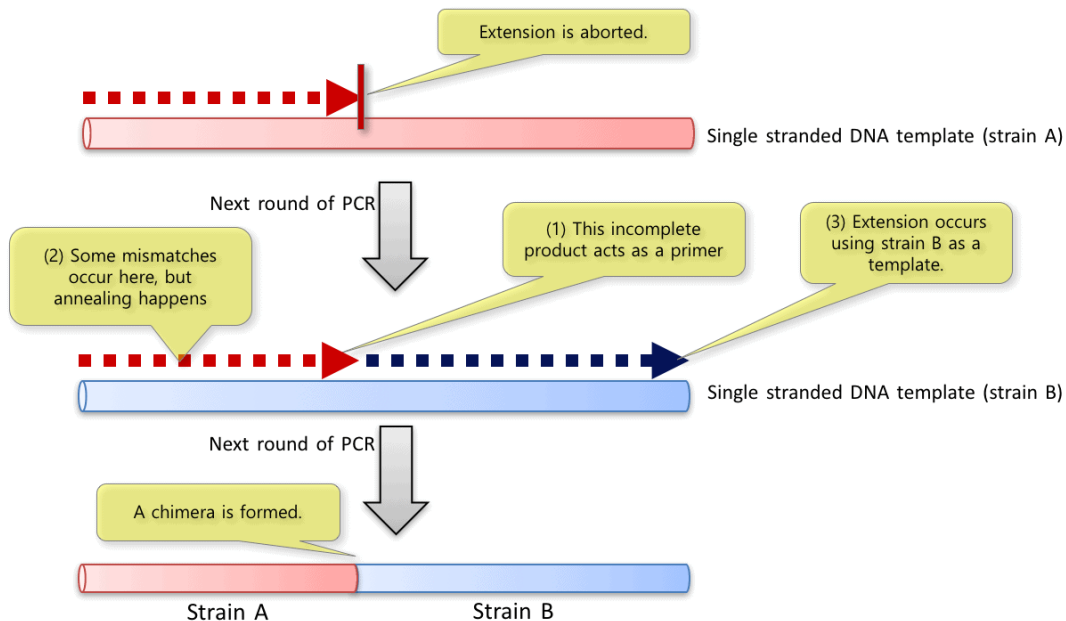
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# Chimera removal



Parental sequence 1  
Parental sequence 2  
Chimeric sequence

AAAAAAAAAAAAAAAAAAAAA  
TTTTTTTTTTTTTTTTTTTTT  
AAAAAAAAATATTTTTTTTTT

# Chimera removal

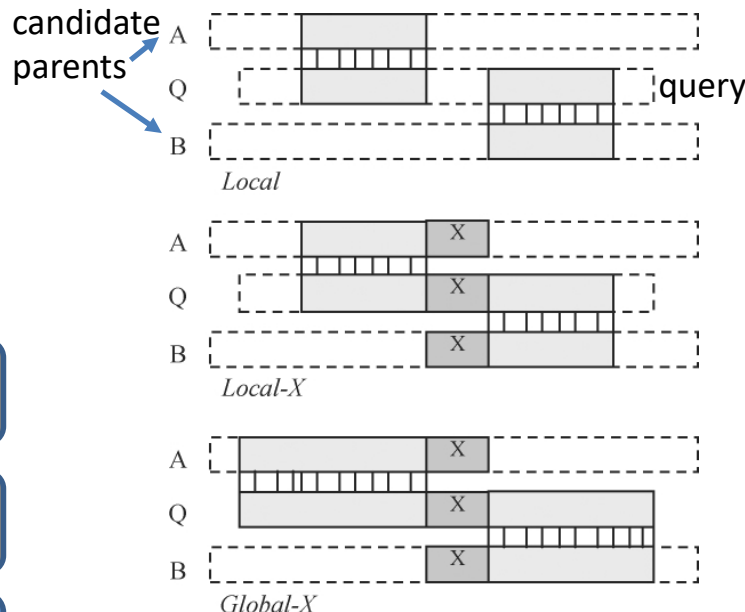
chimeric sequences are derived from parental sequences  
-> program search for parents.

## *de novo* approach:

- no reference database
- usually based on measuring of several sequence parts abundances – parents should be more abundant than its offspring (chimeras)

## reference based approach:

- pair-wise alignment with reference sequence database



Pair-wise alignments with all reference sequences

TGGGCGTAAAGCGCGCTAGG Read  
TGtagcggtgaaacGCGTAGa Refseq 1

TGGGCGTAAA GCGCGCGTAGG Read  
TGGGCGTAAA taGttacTgat Refseq 2

+3 +1  
TGGGCGTAAA GCGCGCGTAGG Read  
caagCatcAt GCGCGCGTcGG Refseq 3

Model with minimum score = nr. diffs + 3 x crossovers = 4.

TGGGCGTAAA GCGCGCGTAGG Read  
TGGGCGTAAA GCGCGCGTcGG Model m=2, d=1

PROGRAMS: UCHIME, UPARSE

# Chimera removal

chimeric sequences are derived from parental sequences  
-> program search for parents.

## *de novo* approach:

- no reference database
- usually based on measuring of several sequence parts abundances – parents should be more abundant than its offspring (chimeras)

## reference based approach:

- pair-wise alignment with reference sequence database

## Problems

- algorithms are not optimal
- computation cost could be high
- problems with highly similar sequences

## Problems

- no appropriate reference database for environmental samples
- variable quality of reference databases

PROGRAMS: **UCHIME**, **UPARSE**

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# Fungal ITS extraction



## ITSx

Improved software for detection and extraction of ITS1 and ITS2 from ribosomal ITS sequences of fungi and other eukaryotes for use in environmental sequencing

relies on **HMMER**

searching sequence databases for sequence homologs, and for making sequence alignments. It implements methods using probabilistic models called profile hidden Markov models (profile HMMs).

using the extracted variable ITS improves resolution when the OTUs are created

<http://microbiology.se/software/itsx/>

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# Clustering to OTUs

**OTU (Operational taxonomic unit)** group of similar sequences grouped based on some similarity threshold  
usually 97% similarity (16S, ITS) represents Species

## Heuristic

- comparison of each sequence with representative sequence („seed“)
- depends on sequence order

FAST

## Hierarchical

comparison of each sequence with each other (tree construction)

SLOW

## Model based

- probabilistic, iterative
- uses more information than the sequence identity

VERY SLOW

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# Clustering to OTUs

## programs

Tool	Distance calculation	Clustering algorithm	Reference
DOTUR (+ MUSCLE)	MSA <sub>denovo</sub>	Hierarchical	Schloss et al. 2005
<b>MOTHUR</b>	<b>MSA<sub>profile</sub></b>	<b>Hierarchical</b>	<b>Schloss et al. 2009</b>
ESPRIT	PSA	Hierarchical	Sun et al. 2009
SLP	PSA	Hierarchical	Huse et al. 2009
ESPRIT-TREE	PSA	Hierarchical	Cai et al. 2011
JMOTU	PSA	Hierarchical	Jones et al. 2011
CD-HIT	PSA	Heuristic	Li et al. 2006
<b>USEARCH/UPARSE</b>	<b>PSA</b>	<b>Heuristic</b>	<b>Edgar et al. 2010/2013</b>
GRAMCLUSTER	PSA	Heuristic	Russell et al. 2010
DNACLUSt	PSA	Heuristic	Ghodsi et al. 2011
<b>CRUNCHCLUST</b>	<b>PSA</b>	<b>Heuristic</b>	<b>Hartmann et al. 2012</b>
DYSC	PSA	Heuristic	Zheng et al. 2012
MS-CLUST	PSA	Heuristic	Chen et al. 2013
TBC	PSA	Heuristic	Lee et al. 2012
TSC	PSA	H&H combination	Jiang et al. 2012
<b>CROP</b>	<b>PSA</b>	<b>Model-based (BC)</b>	<b>Hao et al. 2011</b>
BEBAC	PSA	Model-based (BC)	Cheng et al. 2012
DBC454	composition	Model-based	Pagni 2013
DBC	PSA	Model-based	Preheim et al. 2013
M-PICK	graphical	Model-based	Wang et al. 2013

**PSA - Pairwise Sequence Alignment**

**MSA - Multiple Sequence Alignment**

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**Clustering to OTUs**

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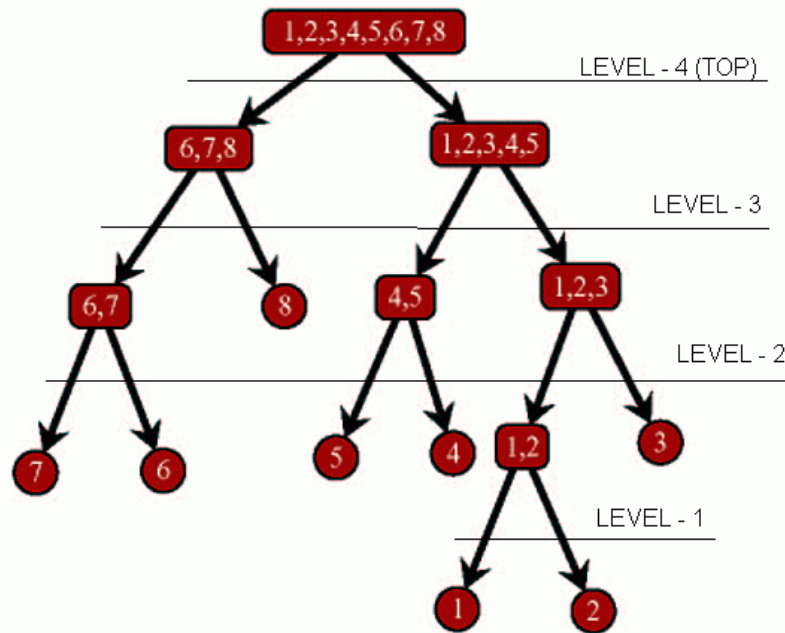
Identification of  
OTUs

Construction of  
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Estimation of  
diversity indices

PROCESSING OF THE  
RESULTS

## Clustering to OTUs (hierarchical)



all pairwise comparisons  
are performed and  
OTUs are delineated at  
fixed distance level

linking method is an important driver of the outcome:

- **single-linkage clustering** (SL) clusters may be merged together due to single sequences being close to each other, even though many of the sequences in each cluster may be very distant to each other
- **complete-linkage clustering** (CL) tends to find compact clusters of approximately equal diameters. With CL, all objects in a cluster are similar to each other
- **average-linkage clustering** (AL) can be seen as an intermediate between single and complete linkage clustering, resulting in more homogeneous clusters than those obtained by the single-linkage method

# Clustering to OTUs (heuristic - USEARCH)

RAW DATA - R1 & R2  
FASTQ

Joining of pair-end data

Quality filtering  
Sample determination  
Sequence trimming

Chimera removal

Preparing for clustering  
e.g.: fungal ITS  
extraction

**Clustering to OTUs**

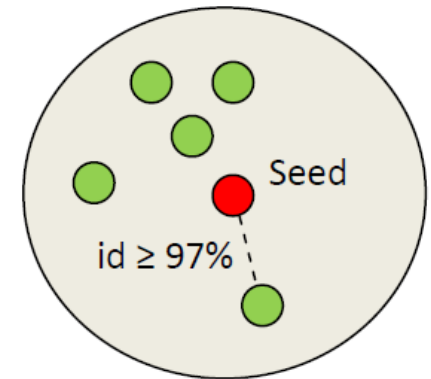
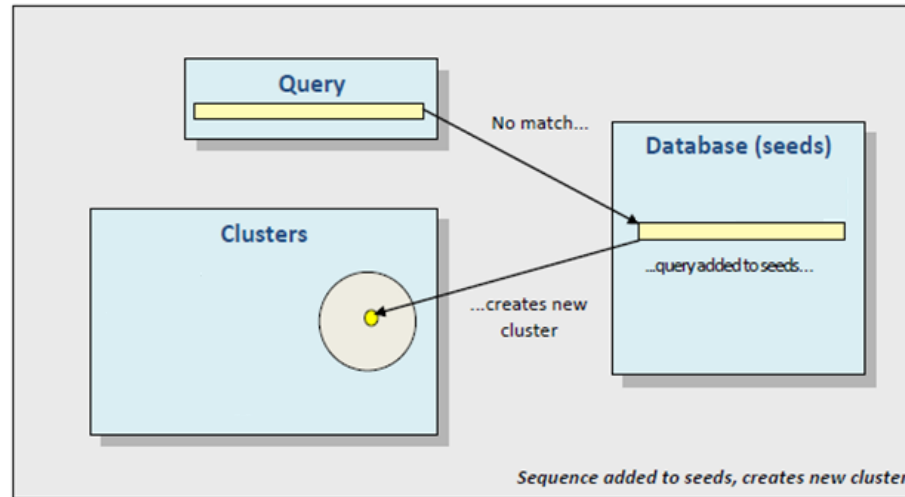
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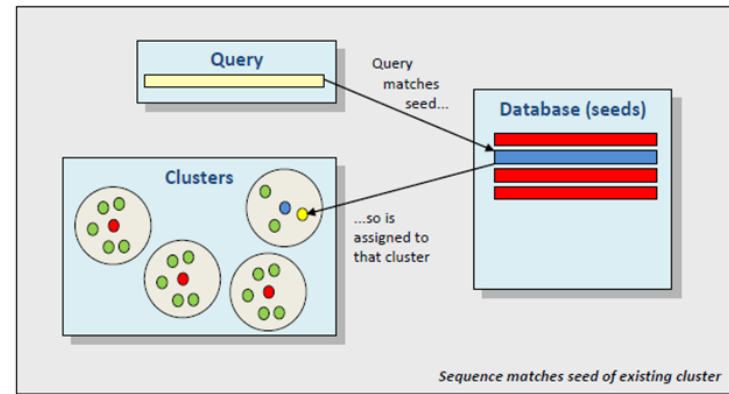
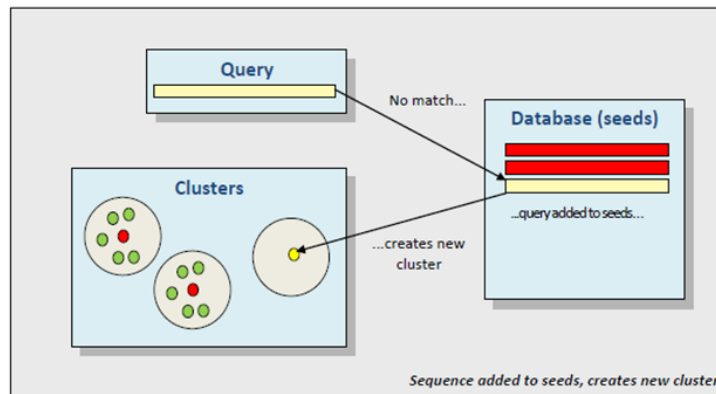
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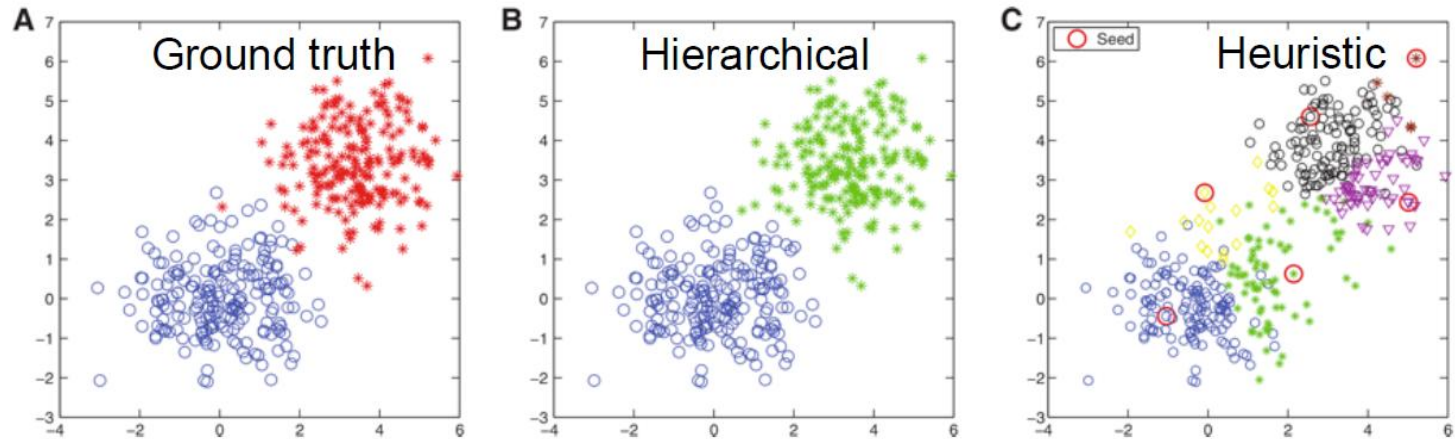
**PROCESSING OF THE  
RESULTS**



cluster definition



## Clustering to OTUs (hierarchical vs. heuristic)



### Hierarchical clustering

- is able to identify the real clusters (ideally)
- computationally expensive

X

### Heuristic clustering

- computationally cheap
- often generates artificial clusters (overestimated diversity)

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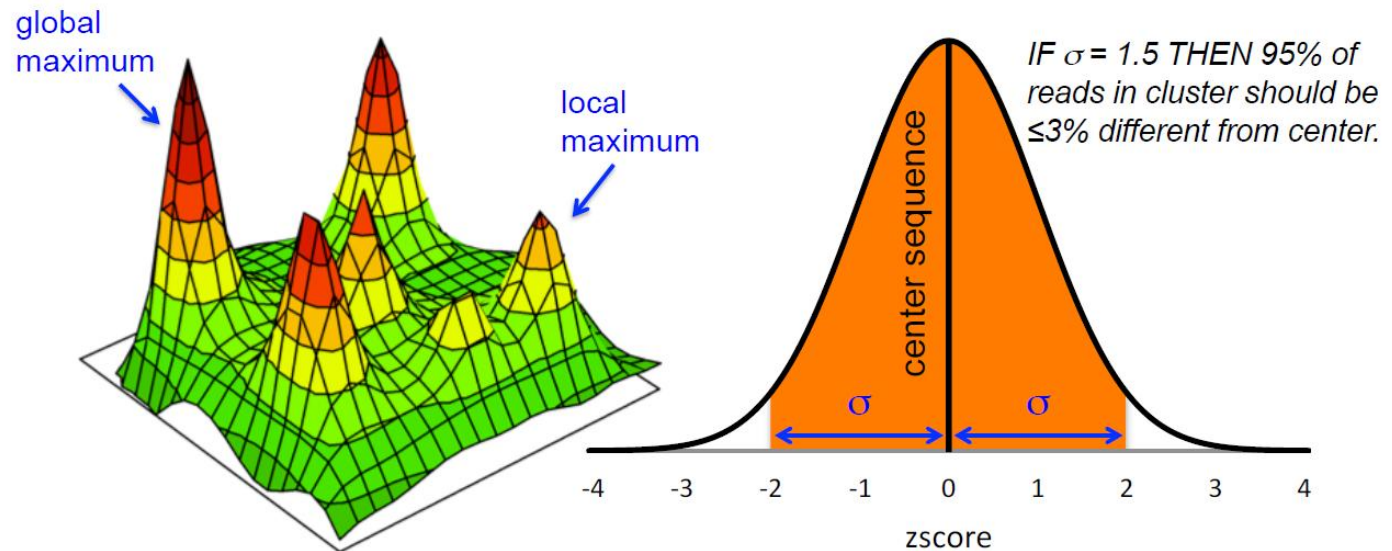
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# Clustering to OTUs (model based)

## CROP (FILTER – PSA – BAYESIAN)

Hao et al. 2011 (Bioinformatics): “If we consider the sequences as data points in a high-dimensional space [...], then the probability that a sequence belongs to a cluster becomes a function of the distance between the sequence and the center.”

**CROP** uses a mixture model to find subpopulations among all sequences under the assumption that they are independently drawn from a mixture of Gaussian distributions.



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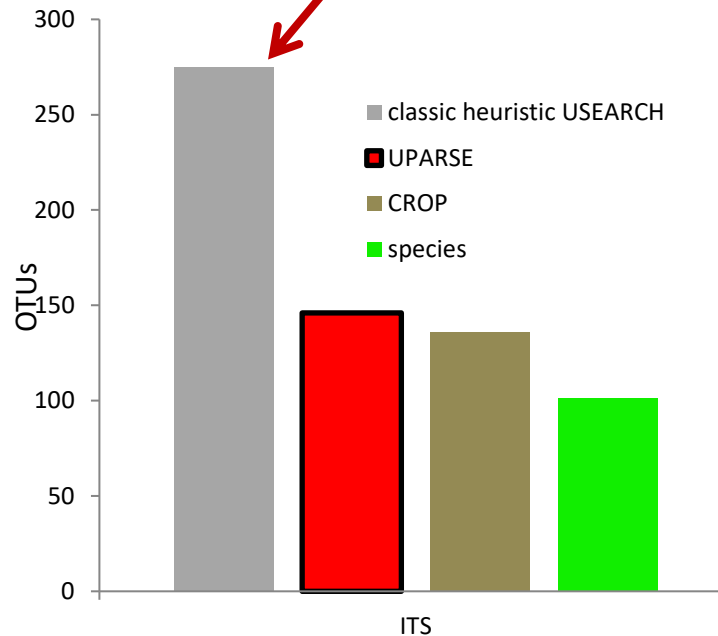
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RESULTS

# Clustering to OTUs (comparison)

Problems with overestimation of obtained OTUs...



mock fungal community  
about 100 fungal species

solution

**UPARSE (USEARCH)** – improved heuristic algorithm which is able to recognize chimeric sequences

<http://drive5.com/uparse/>

Edgar, R.C. (2013) UPARSE: Highly accurate OTU sequences from microbial amplicon reads, *Nature Methods* [[PubMed:23955772](https://pubmed.ncbi.nlm.nih.gov/23955772/), [dx.doi.org/10.1038/nmeth.2604](https://doi.org/10.1038/nmeth.2604)].

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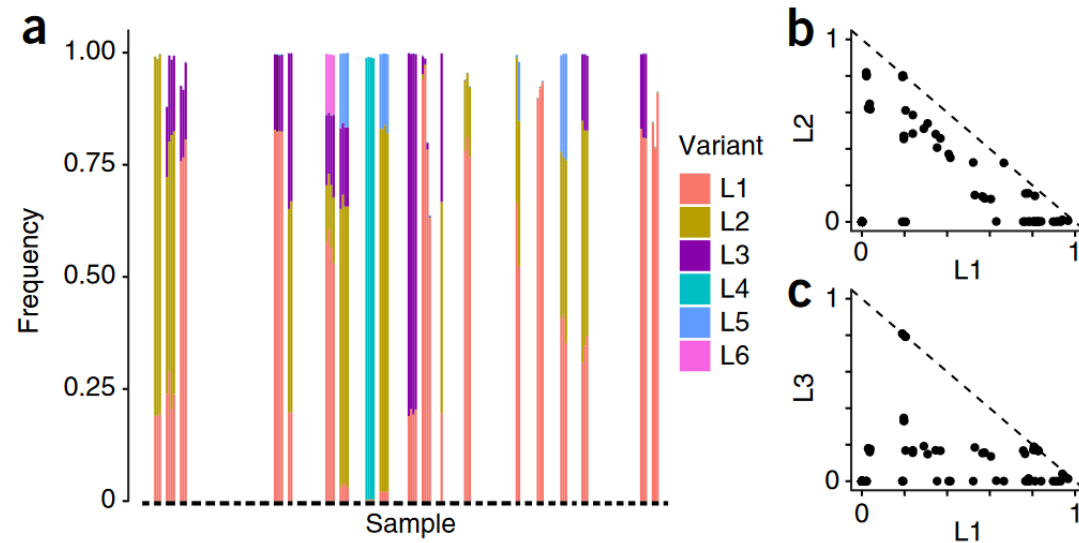
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RESULTS**



## Clustering-independent methods

Callahan, Benjamin J., et al. "**DADA2**: high-resolution sample inference from Illumina amplicon data." *Nature methods* (2016).



*L. crispatus* sequence variants in the human vaginal community during pregnancy. DADA2 identified six *L. crispatus* 16S rRNA sequence variants present in multiple samples and a significant fraction of all reads.

Amir, Amnon, et al. "**Deblur** Rapidly Resolves Single-Nucleotide Community Sequence Patterns." *mSystems* 2.2 (2017).

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representative  
sequences from  
clusters

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OTU table

Estimation of  
diversity indices

PROCESSING OF THE  
RESULTS



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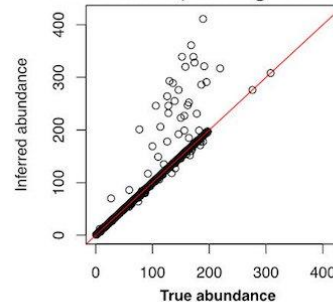
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PROCESSING OF THE  
RESULTS



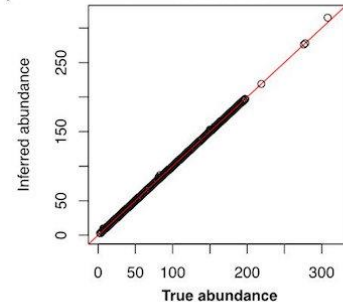
## Accuracy: Simulated data

3% OTUs (average linkage)



TP: 978  
FP: 272  
FN: 77  
cor: 0.935

DADA2



TP: 1042  
FP: 0  
FN: 13  
cor: 0.999

Data: Kopylova, et al. mSystems, 2016.

## Advantages

**Resolution:** DADA2 infers exact amplicon sequence variants (ASVs) from amplicon data, resolving biological differences of even 1 or 2 nucleotides.

**Accuracy:** DADA2 reports fewer false positive sequence variants than other methods report false OTUs.

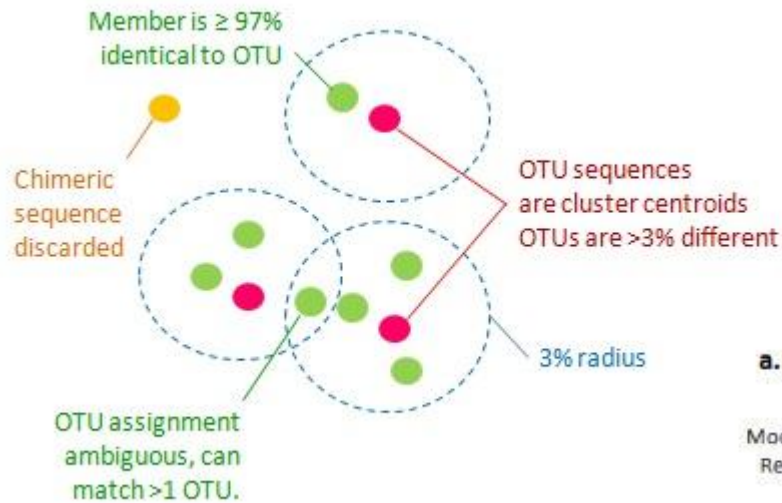
**Comparability:** The ASVs output by DADA2 can be directly compared between studies, without the need to reprocess the pooled data.

**Computational Scaling:** The compute time of DADA2 scales linearly sample number, and memory requirements are essentially flat.

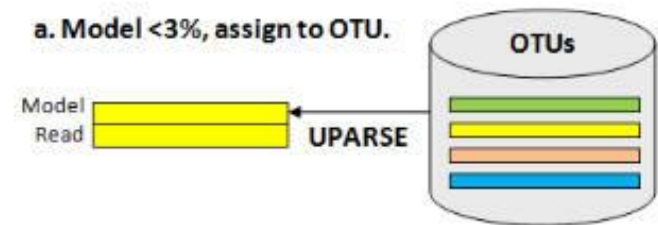
## Disadvantage

sequence variants are not representing the real sequences (they are estimated based on the errors modeling)

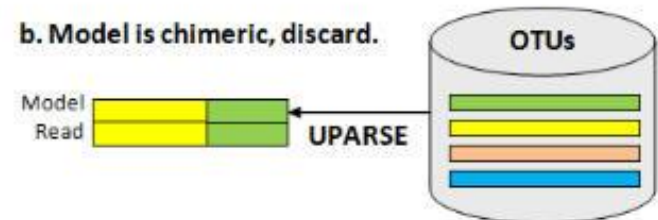
# UPARSE: Clustering and chimera removal in the same time



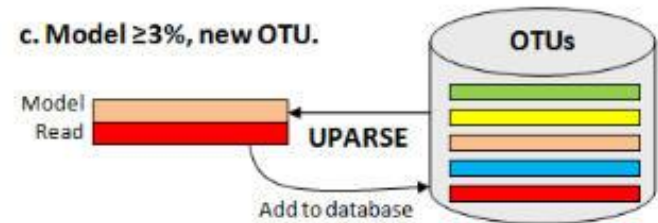
a. Model  $<3\%$ , assign to OTU.



b. Model is chimeric, discard.



c. Model  $\geq 3\%$ , new OTU.



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representative  
sequences from  
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**PROCESSING OF THE  
RESULTS**

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diversity indices

PROCESSING OF THE  
RESULTS

# Construction of OTU table

**OTU table** - matrix that gives the number of reads per sample per OTU

OTU_ID	SAMPLE_1	SAMPLE_2	SAMPLE_3	SAMPLE_4	SAMPLE_5	SAMPLE_6	SAMPLE_7	SAMPLE_8
CL00001	249	189	220	311	1	16	68	2
CL00002	201	19	169	438	1	8	12	0
CL00003	190	39	176	210	0	21	20	1
CL00004	183	36	195	177	1	16	16	0
CL00005	0	26	2	35	20	164	4	116
CL00006	0	0	0	0	1	0	0	0
CL00007	133	71	125	89	0	3	26	0
CL00008	106	42	96	158	0	10	14	0
CL00009	95	46	108	134	2	7	24	0
CL00010	0	0	0	0	0	0	0	3
CL00011	0	1	0	0	0	0	0	0

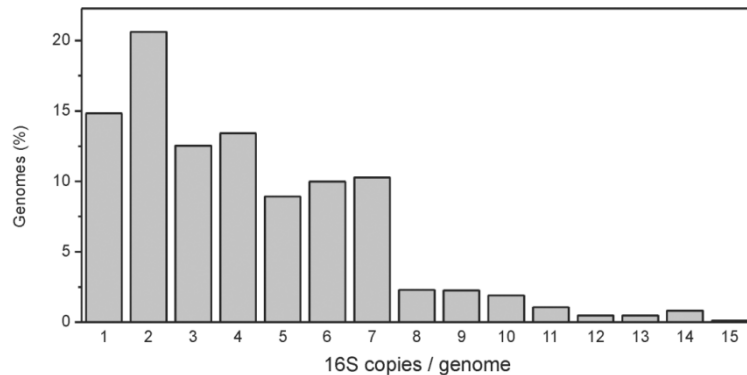
## OTU frequency does not correlate with species frequency

This means, for example, that the most abundant OTU does not have to be the most abundant species – especially because of multi-copy nature of target genes as 16S and ITS

## Singleton counts are especially suspect

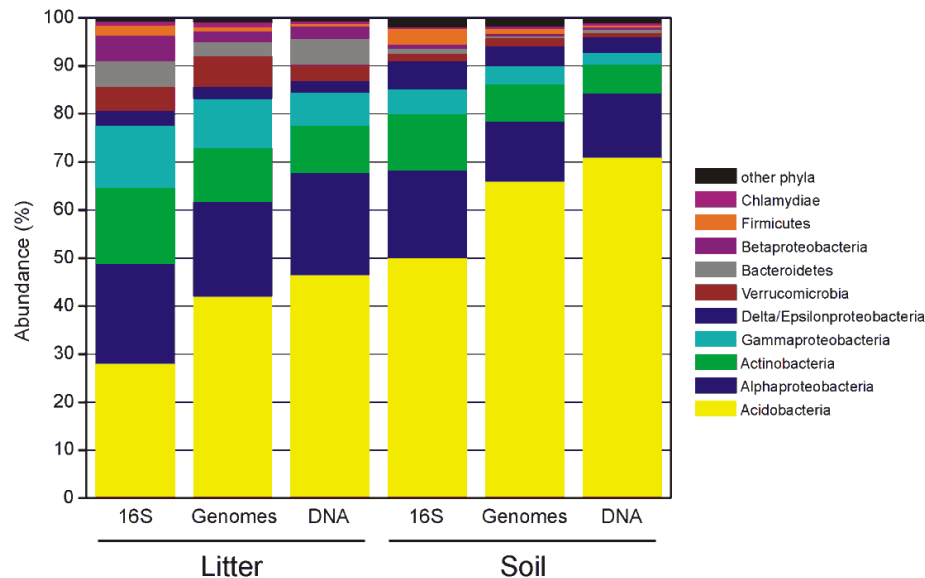
- many OTU table entries are often singletons (have value 1) for smaller OTUs because the total count is distributed over several samples
- Small counts are more likely to be spurious, especially singletons, either because the OTU itself is spurious (e.g., an undetected chimera), or because of cross-talk

# Normalize OTU table by 16S copy number



## rrnDB

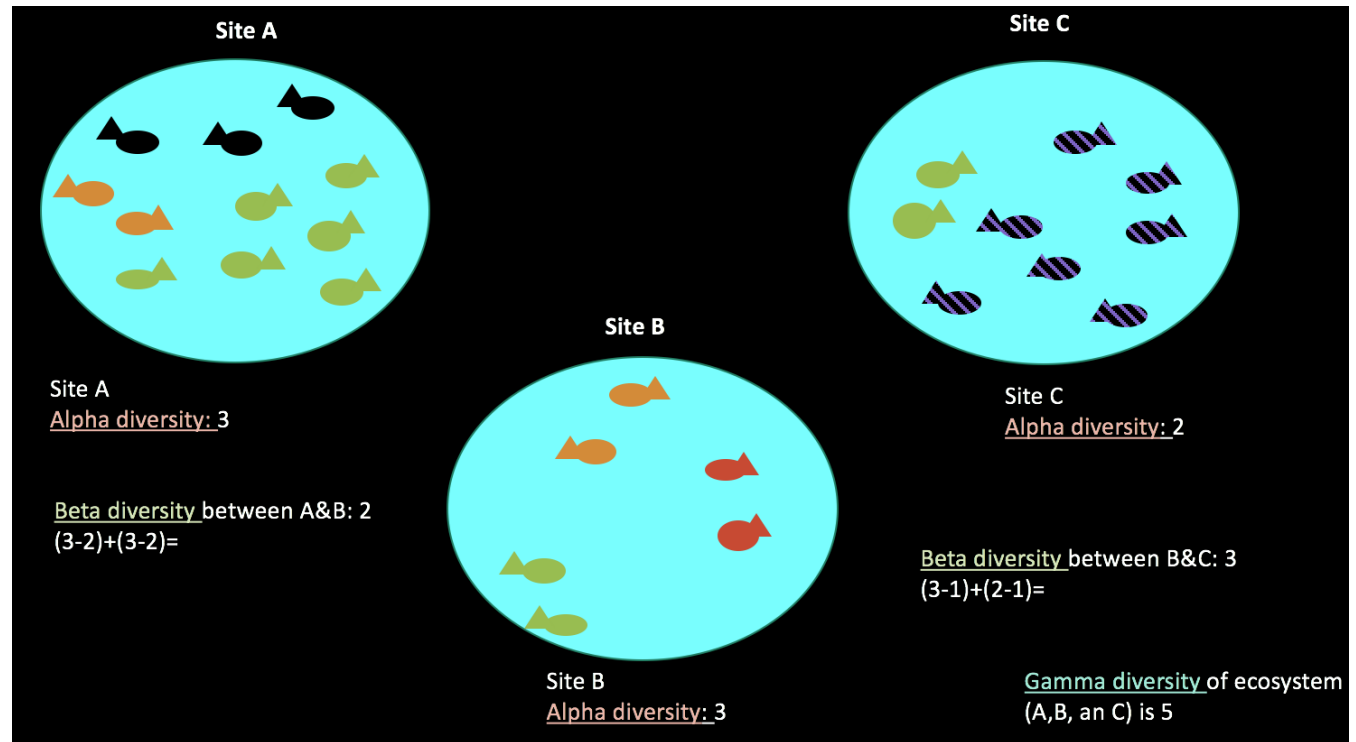
A searchable database documenting variation in ribosomal RNA operons (*rrn*) in Bacteria and Archaea. Find information such as the 16S gene copy number of an organism by looking up its name under the NCBI or RDP taxonomy or by full-text search of rrnDB's records.



## Abundance of bacterial 16S rRNA sequences, genomes and DNA in forest litter and soil.

Relative abundance of bacterial 16S rRNA sequences in the amplicon pool from *Picea abies* litter and soil (Baldrian et al., 2012), and estimates of the relative abundance of bacterial genomes and DNA. The estimates were calculated using the values of 16S rRNA copy numbers and genome sizes of the closest hits to each bacterial OTU.

# Estimation of diversity indices



- **Alpha-diversity:** diversity of organisms in one sample / environment
  - Shannon index
  - Chao1
  - Observed OTUs (Richness)
- **Beta-diversity:** differences in diversities across samples or environments
  - UniFrac (Lozupone et al, AEM, 2005) (phylogenetic)
  - Bray-Curtis dissimilarity measure (OTU abundance)
  - Jaccard similarity coefficient (OTU presence/absence)

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diversity indices

PROCESSING OF THE  
RESULTS

## Alpha diversity

### Shannon index (Shannon entropy)

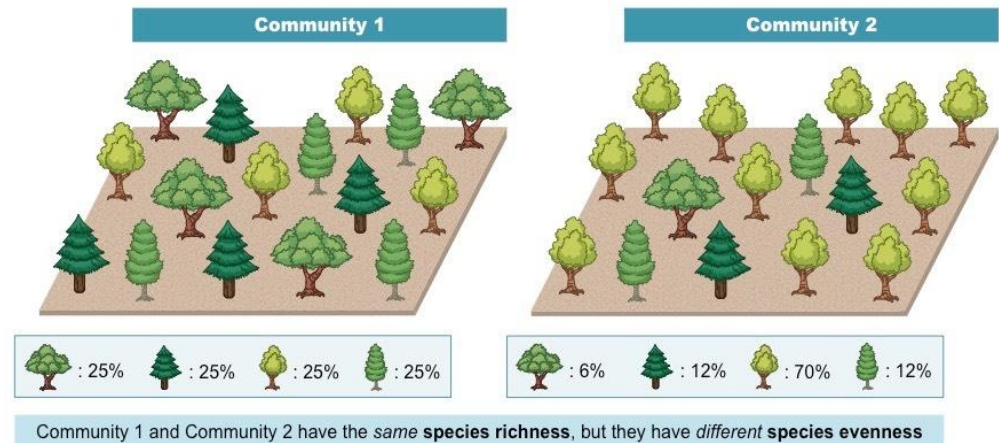
Then the Shannon entropy quantifies the uncertainty in predicting the species identity of an individual that is taken at random from the dataset.

### Species evenness

Species evenness refers to how close in numbers each species in an environment is. Mathematically it is defined as a diversity index, a measure of biodiversity which quantifies how equal the community is numerically.

$$H' = - \sum_{i=1}^S p_i \ln p_i$$

$p_i$  – proportion of the population made up of species  $i$   
 $S$  – number of species in sample



$$J' = \frac{H'}{H'_{\max}} \quad H'_{\max} = - \sum_{i=1}^S \frac{1}{S} \ln \frac{1}{S} = \ln S$$

### Chao1 index

Estimate diversity from abundance data (importance of rare OTUs)

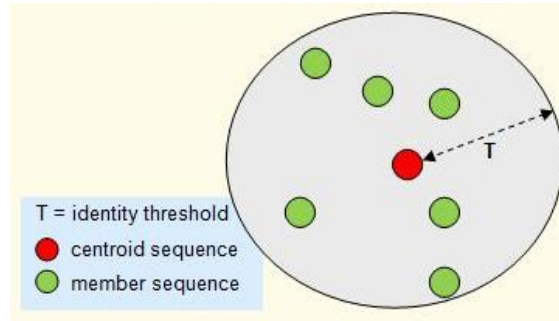
$$S_{est} = S_{obs} + \left( \frac{f_1^2}{2f_2} \right)$$

where  $S_{obs}$  is the number of species in the sample,  $f_1$  is the number of singletons and  $f_2$  is the number of doubletons.



# Getting of the representative sequences from the clusters

centroid



consensus

HM2XOCT01AL89G xy=136_1490	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01A4CCZ xy=342_1537	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01AFZ30 xy=65_510	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01A7wVS xy=383_166	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01BSJ9X xy=618_1043	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01ASEGw xy=206_1454	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01B0M39 xy=710_1143	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01ATS79 xy=222_1703	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01A26G7 xy=329_505	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01BTPEY xy=631_1112	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01ARHZZ xy=196_333	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01BZKDN xy=698_89	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
HM2XOCT01A8KCI xy=390_1905	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G
CONSENSUS	500	0	0	C	G	C	G	T	A	T	A	C	A	A	G	A	T	-	C	C	G	T	A	G	G	T	G	A	A	C	C	T	G	C	G	G	A	A	G	G

most abundant

M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000155	0	1	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
M03794:7:000000000153	0	1	T	C	A	A	A	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	
CL0141 MQSTABUN153	0	0	T	C	A	A	C	C	C	T	C	A	A	G	C	C	T	G	G	C	T	T	G	C	T	G	T	T	G	G	A	C	C	T	C	T	C	

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RESULTS

# Taxonomic classification of OTUs

## Similarity-based

Find homology or minimum alignment distance

Tools:

- local alignments (e.g. BLAST, MEGAN, METAXA2, RTAX)
- global alignments (e.g. GAST)
- overlap alignments (e.g. SINA)

Pro/Con:

- good accuracy for similar sequences
- performs less well on distant lineages
- can be slow on large reference databases

## Composition-based

Detect specific features

Tools:

- kmer searches (e.g. NBC/RDP, UTAX, SINTAX)
- hidden Markov models (e.g. PHYMMBL, C16S)

Pro/Con:

- computationally efficient and fast
- performs well on distant lineages
- training required
- limited resolution for shorter sequences

## Phylogeny-based

Evolutionary model to determine best placement

Tool:

- ML, NJ, Bayesian (e.g. PPLACER, EPA)

Pro/Con:

- great accuracy for similar sequences
- classification in its evolutionary context
- computationally complex
- requires accurate reference tree
- difficult for non-coding regions

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Estimation of  
diversity indices

PROCESSING OF THE  
RESULTS



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PROCESSING OF THE  
RESULTS

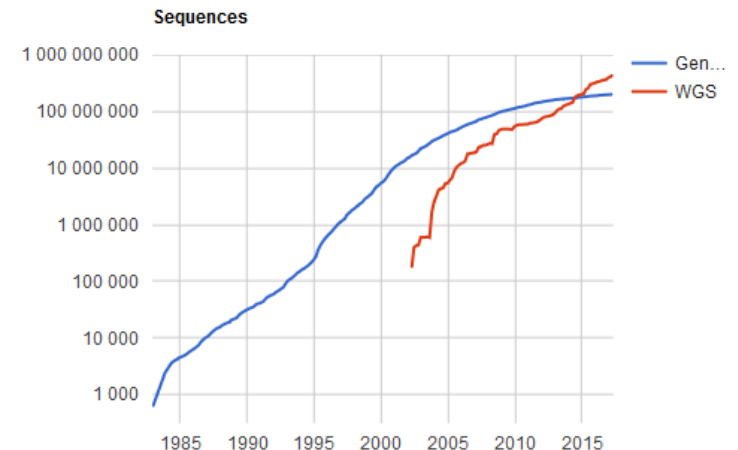
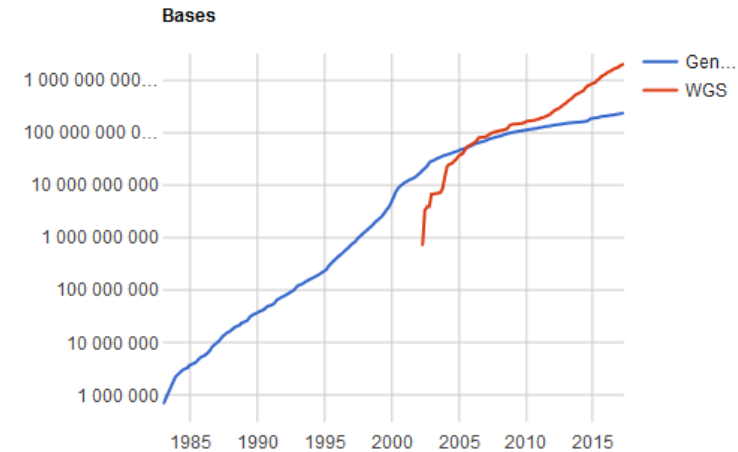


# Identification of OTUs

All genes

**GenBank - *genetic sequence database, an annotated collection of all publicly available DNA sequence***

- largest 😊
- many errors ☹️



<https://www.ncbi.nlm.nih.gov/genbank/>

Gen – GenBank  
WGS – whole genome sequences

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Identification of  
OTUs

Construction of  
OTU table

Estimation of  
diversity indices

PROCESSING OF THE  
RESULTS

## Identification of OTUs

Identification of bacteria

### RDP – *Ribosomal Database Project*

provides quality-controlled, aligned and annotated Bacterial and Archaeal 16S rRNA sequences, and Fungal 28S rRNA sequences, and a suite of analysis tools to the scientific community



<https://rdp.cme.msu.edu/>



<https://www.arb-silva.de/>

**SILVA** - provides comprehensive, quality checked and regularly updated datasets of aligned small (16S/18S, SSU) and large subunit (23S/28S, LSU) ribosomal RNA (rRNA) sequences for all three domains of life (Bacteria, Archaea and Eukarya).



**GREENGENES**  
The 16S rRNA Gene Database and Tools

<http://greengenes.secondgenome.com/>

RAW DATA - R1 & R2  
FASTQ

Joining of pair-end data

Quality filtering  
Sample determination  
Sequence trimming

Chimera removal

Preparing for clustering  
e.g.: fungal ITS  
extraction

Clustering to OTUs

Getting the  
representative  
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# Identification of OTUs

Identification of fungi

**UNITE - *Unified system for the DNA based fungal species linked to the classification***

unite  
community

Unified system for the DNA based fungal species linked to the classification Ver. 7.1



Current version: 7.2; Last updated: 2017-06-08 ([read more](#))

Number of ITS sequences (UNITE+INSID): 741 222; Number of UNITE fungal Species Hypotheses with DOIs at 1.5% threshold: 73 929 ([more statistics](#))

Threshold 1.5 %

Include All SH-s

Start typing taxon name here ...

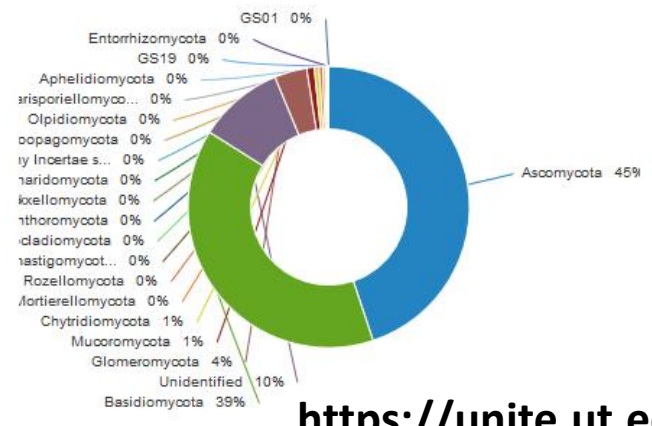
Go

Reset



- ▶ Ascomycota (33,051)
- ▶ Basidiomycota (28,743)
- ▶ Unidentified (7,240)
- ▶ Glomeromycota (2,743)
- ▶ Mucoromycota (636)
- ▶ Chytridiomycota (388)
- ▶ Mortierellomycota (322)
- ▶ Rozellomycota (198)
- ▶ Neocallimastigomycota (90)
- ▶ Blastocladiomycota (42)
- ▶ Entomophthoromycota (41)
- ▶ Kickxellomycota (39)
- ▶ Monoblepharidomycota (22)
- ▶ Fungi phy Incertae sedis (19)
- ▶ Zoopagomycota (15)
- ▶ Olpidiomyota (13)

SH graph: Fungi



<https://unite.ut.ee/>

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# PROCESSING OF THE RESULTS

## Introduction to multivariate data analysis (Iñaki Class)

