

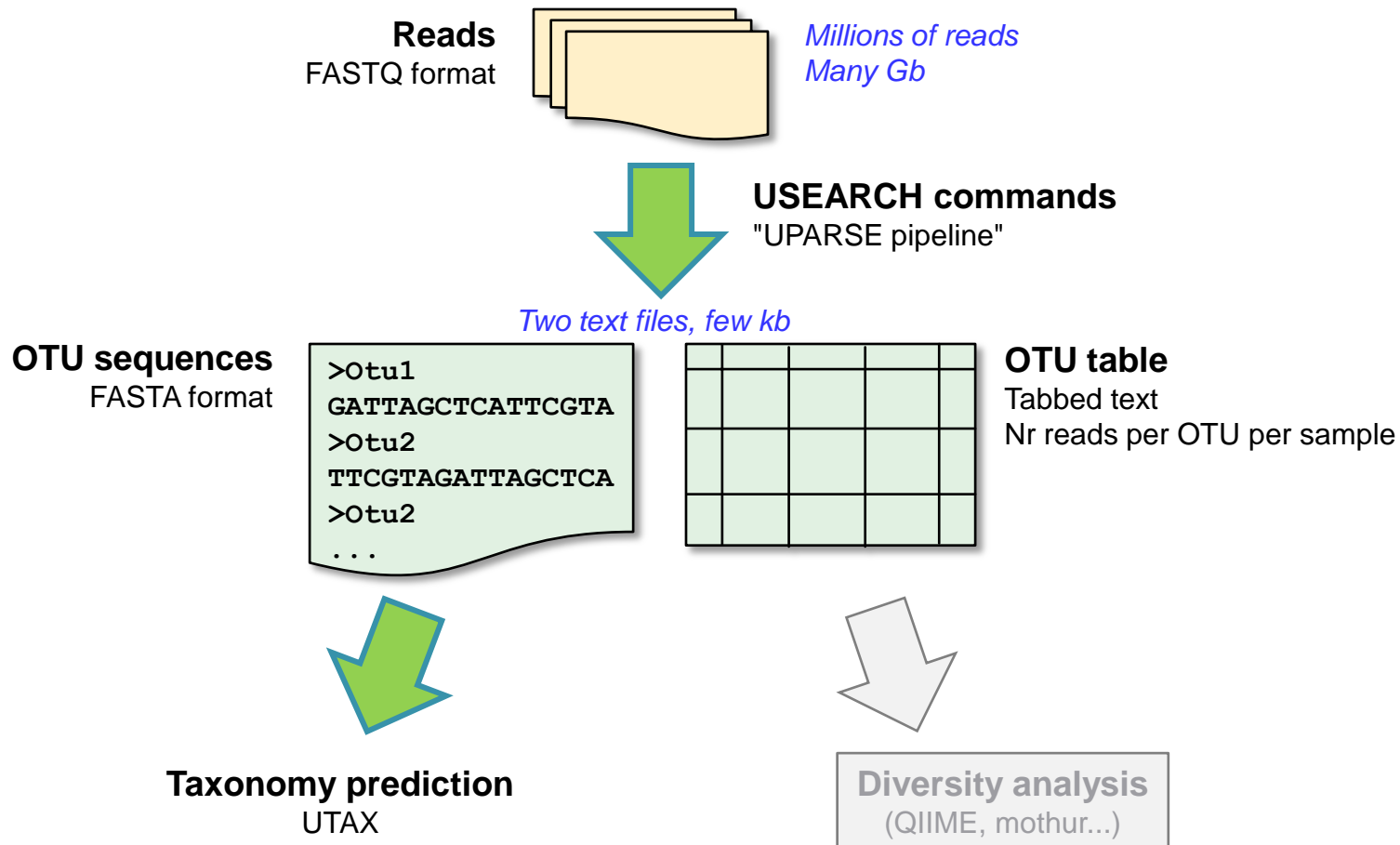
CLUSTERING

STAMPS 2016

Robert Edgar

Independent scientist
robert@drive5.com
www.drive5.com

OTU analysis



Naive clustering

- Mock community with 20 species
- Cluster reads at 97% using UCLUST
- Thousands of "OTUs"
 - terrible result...
 - clusters are **noise!**



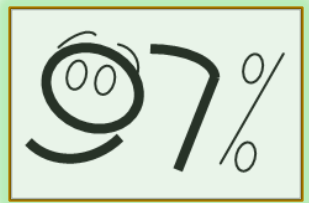
The magic number 97

Q. Why cluster at 97%?



a) Everybody does it

(true)



b) 97 is a happy prime

(true -- look it up!)



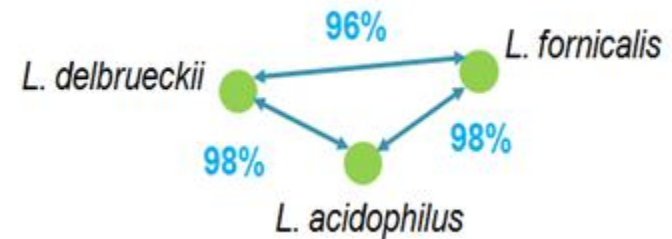
c) 97% clusters are species

(not true)

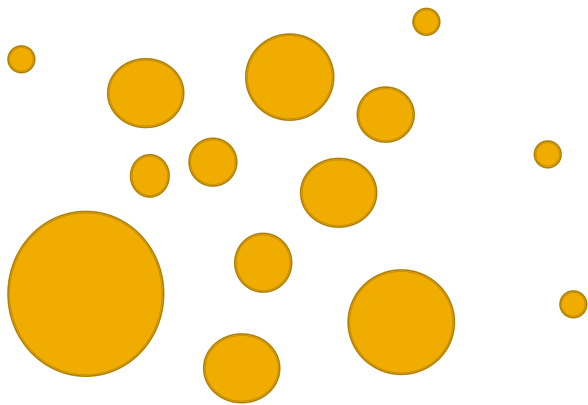
97% is not species



- Reasonable rule of thumb for **full-length** 16S
 - Paralogs in a single species usually >97%
 - But paralogs can be as low as 89%
 - Different strains usually >97%
 - Different species usually <97%
 - But not always, e.g. *Lactobacillus*
- Not so good for **short tags** like V₄
 - Different species often have identical V₄ tags
 - 10% genera in RDP14 have pair of identical V₄s



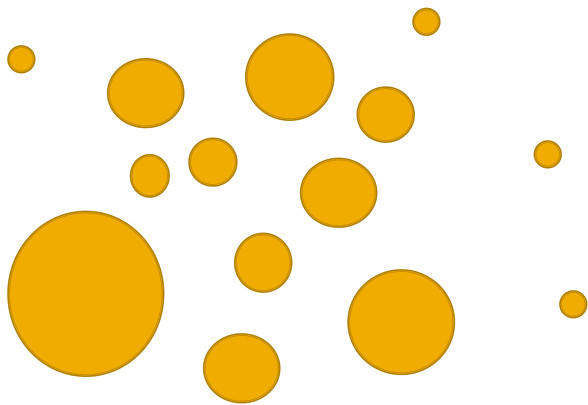
16S: reality vs. clusters



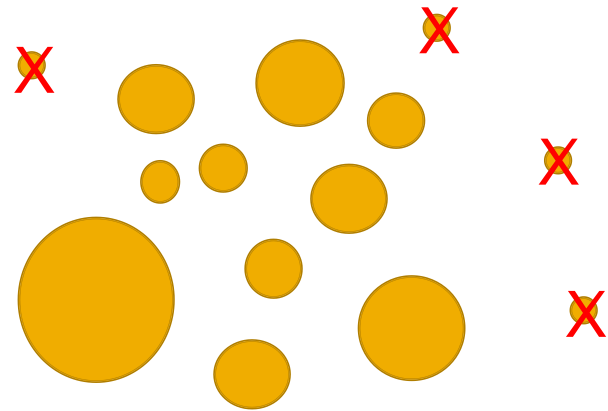
REALITY

Ecologically distinct strains,
size of blob = abundance

16S: reality vs. clusters

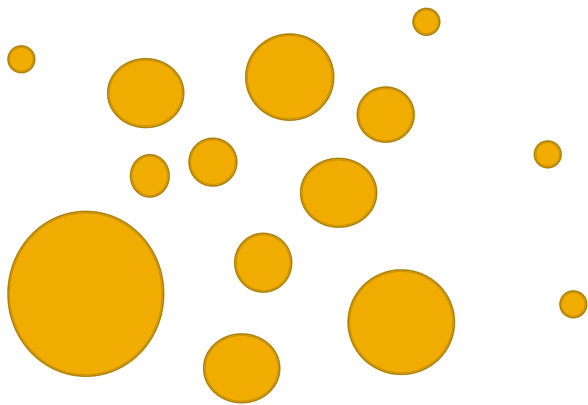


Reality

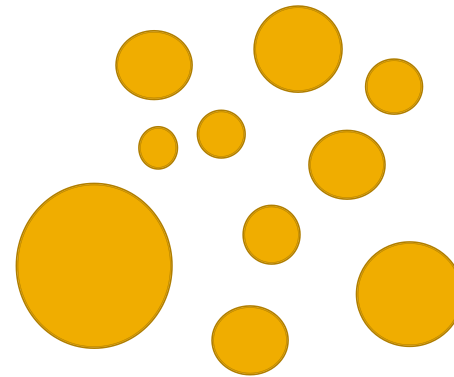


Rare strains not sampled

16S: reality vs. clusters

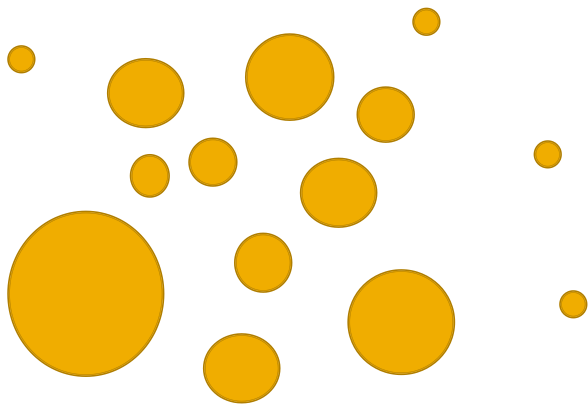


Reality

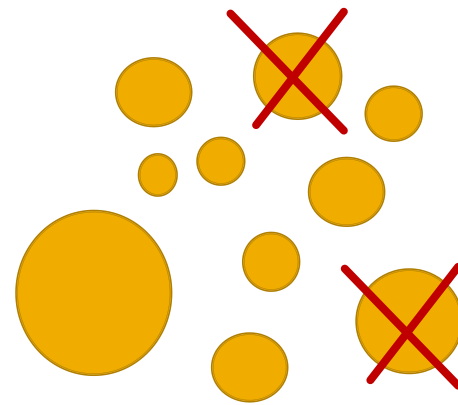


Rare strains not sampled

16S: reality vs. clusters

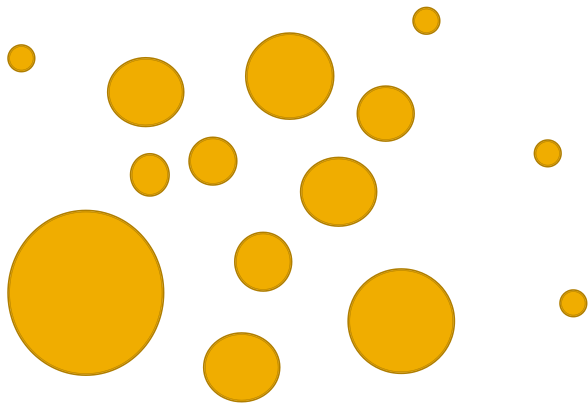


Reality

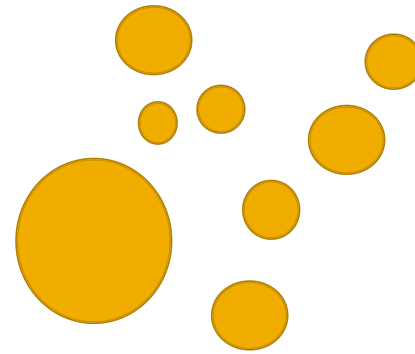


10-15% don't match
"universal primers"

16S: reality vs. clusters

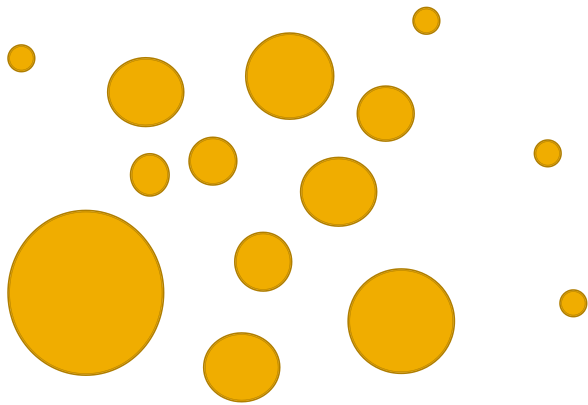


Reality

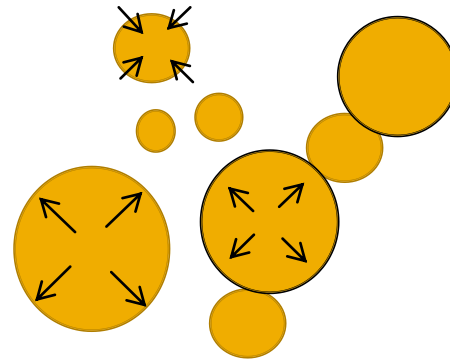


10-15% don't match
"universal primers"

16S: reality vs. clusters

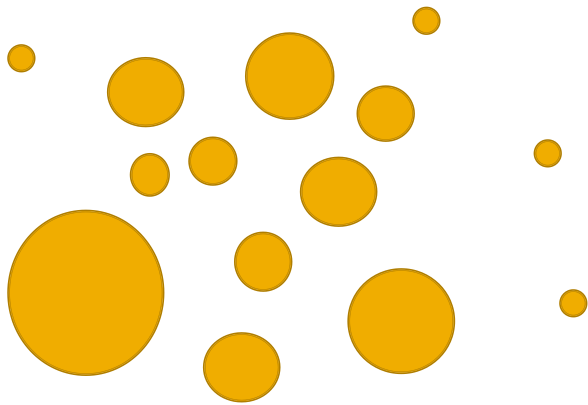


Reality

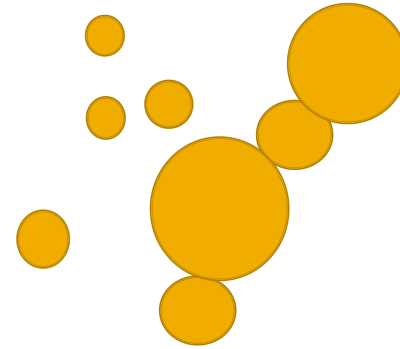


16S copy number varies
from 1 to 15 or so

16S: reality vs. clusters

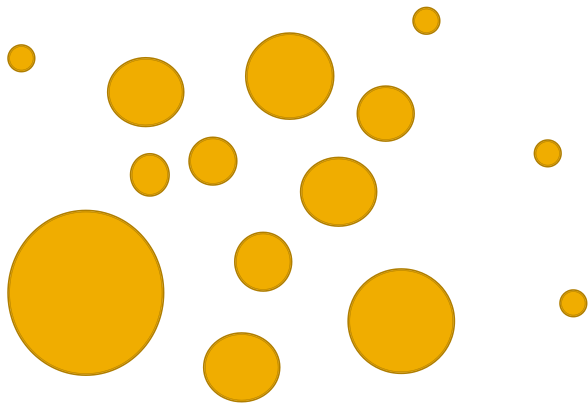


Reality

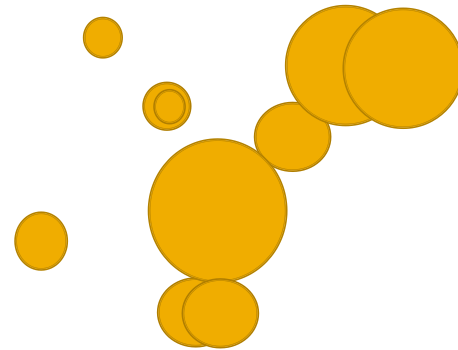


16S copy number varies
from 1 to 15 or so

16S: reality vs. clusters

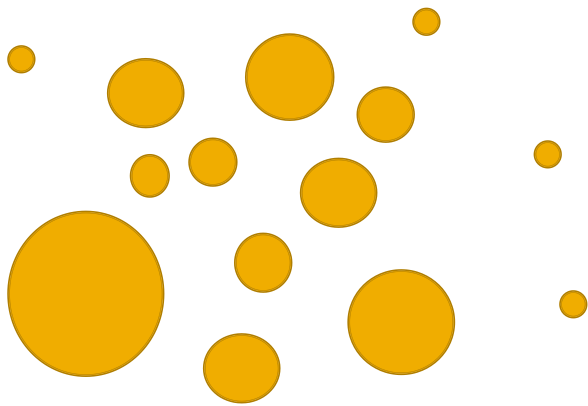


Reality

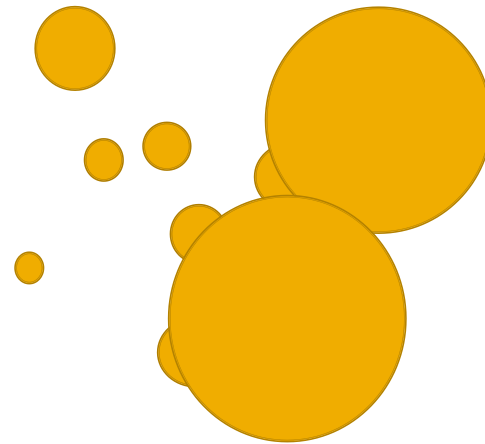


Clusters **split** (paralogs <97% similar)
and **merge** (species >97% similar)

16S: reality vs. clusters

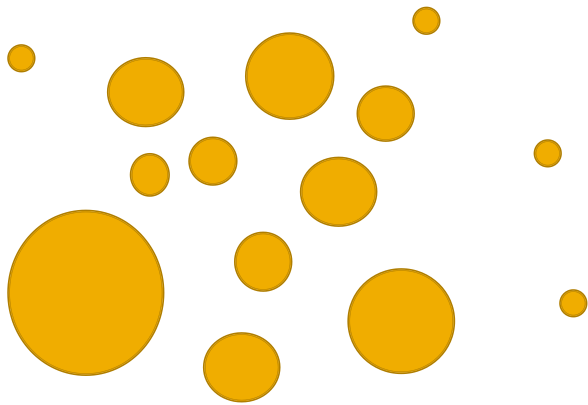


Reality

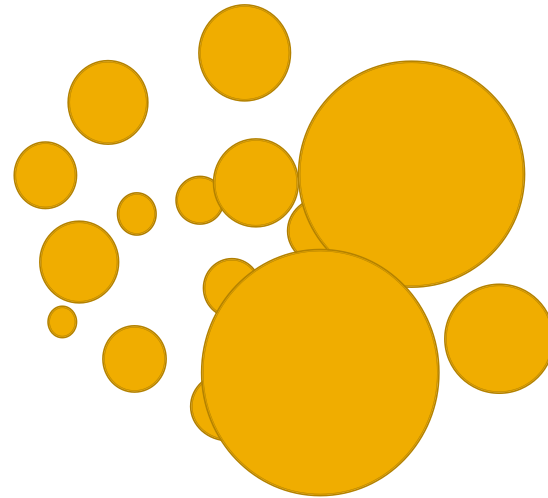


Amplification bias

16S: reality vs. clusters

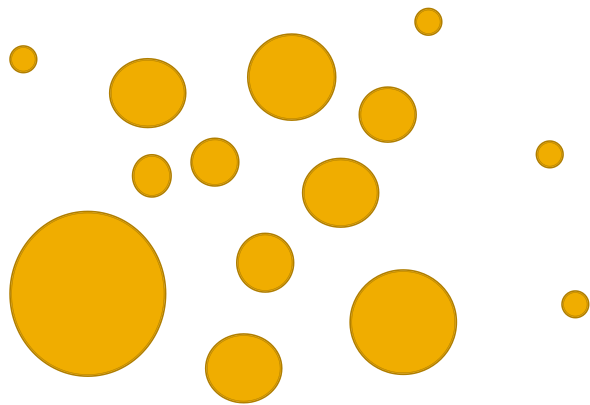


Reality

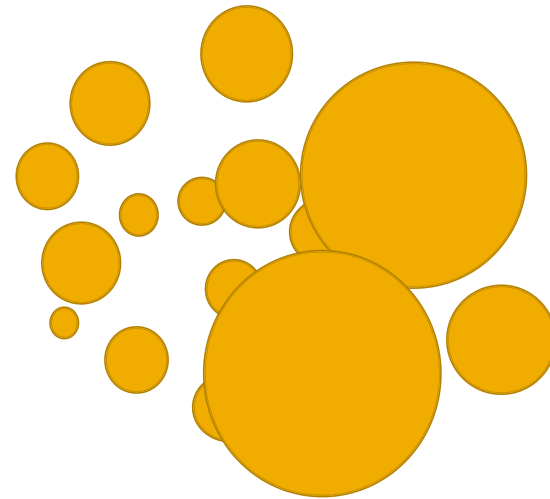


Polymerase errors,
chimeras, read errors,
contaminants

16S: reality vs. clusters



Reality



"OTUs"

Lump or split?

- One genome can contain many 16s genes
 - from one to 10+ typical
- Paralogs may be <100% identical
 - as low as 89%
- Any clustering %id will **lump** and **split**
 - Even in ideal scenario where no errors
- Clustering %id often motivated by "species"
 - I disagree

Lump or split?

- Lumping can obscure biological signals
- Splitting preserves information
 - e.g., better to distinguish strains than lump together
- Given all correct sequences
 - no reason to cluster
 - can estimate number of species from number of uniques
 - if needed, but usually not a very interesting or useful question
- Answer: **split!**
 - Resolve as many distinct genes as possible

Ideal analysis

- *Input:* Reads
- *Output:* Biological sequences
 - **All** biological sequences
 - **Nothing but** biological sequences

Achievable analysis

- Find subset of correct sequences $>3\%$
 - Because $\sim 3\%$ is practical limit for detecting errors
- Sane motivation for 97% clustering
- Should resolve as much detail as possible
 - For any gene 16S, ITS, COI...
 - Regardless of typical intra-species variation
 - Individuals, strains, species, genera... are all **informative**
 - ...and are **valid** OTUs!

Future is (almost) here!

- Denoising can resolve sequences to ~1 diff
 - DADA₂
 - UNOISE₂ (coming soon in USEARCH v9)
- Other high-resolution methods
 - "oligotyping" (Eren *et al.* ISME 2015)
 - "sub-OTU resolution" (derep.) (Tikhonov *et al.* ISME 2014)
- Denoising close to ideal analysis
 - **all** biological sequences, and **nothing but**

Reads → OTUs with USEARCH

- Pre-process reads
 - Paired read assembly (with updated Q scores)
 - Expected error filtering (suggest $E < 1$, $E^* = 0$)
 - Discard singletons (optional, but highly recommended)
 - Dereplicate -- find uniques & abundances
 - Sort uniques by decreasing abundance
- Clustering: UPARSE-OTU algorithm
 - Edgar *Nat. Meth.* 2013
 - **cluster_otus** command

UPARSE OTU clustering

UPARSE has been cited by
668 papers
[Google scholar](#)
Last updated 24 Jul 2016

Download USEARCH

Documentation

Support

Data analysis service

USEARCH

Ultra-fast sequence analysis



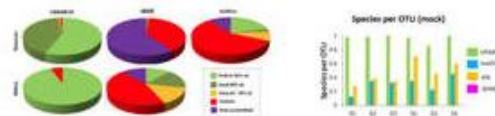
10 - 1,250x BLAST
1 - 1,000x CD-HIT

High-accuracy, high-throughput OTU clustering

UPARSE is a method for generating clusters (OTUs) from next-generation sequencing reads of marker genes such as 16S rRNA, the fungal ITS region and the COI gene. The clustering method itself is the [UPARSE-OTU algorithm](#), implemented as the [cluster_otus command](#) in [USEARCH](#). To run UPARSE in practice, you need to run a [pipeline](#) of scripts and USEARCH commands.

Benchmark tests

According to [results published in Nature Methods](#), UPARSE generates OTUs that are far superior to state-of-the-art methods including QIIME, mothur and AmpliconNoise on [mock community tests](#). OTU representative sequences are more accurate predictions of biological sequences, and the number of OTUs are much closer to the number of species.



UPARSE saved my PhD. I was struggling with spurious OTUs in my mock communities. I've tried QIIME and Amplicon Noise, with many different parameters, and I always got something like 100 OTUs. With UPARSE I get 24. Thank you!

Igor Stelmach Pessi
University of Liège

Reference

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

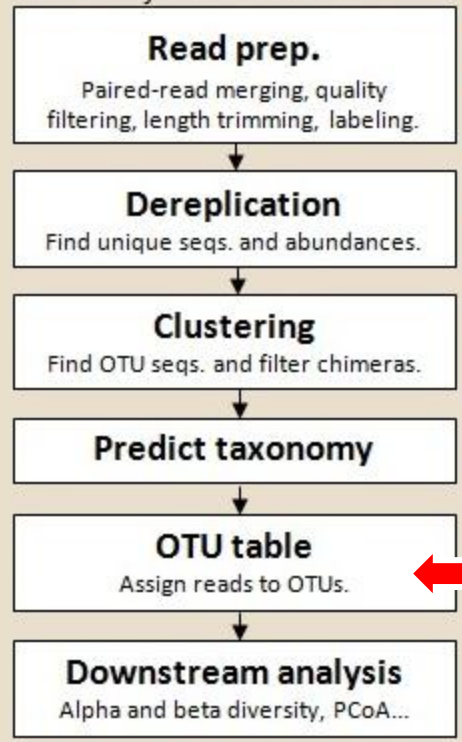


UPARSE OTU analysis pipeline

See also
[UPARSE home page](#)
[USEARCH home page](#)

Click boxes for documentation

Example pipelines



- illumina paired-end
- illumina unpaired
- Roche 454
- FASTA reads (no qual)
- Fungal ITS
- Tutorials and sample data
- Using QIIME and mothur
- Benchmarks and papers
- Data analysis service

Tutorials

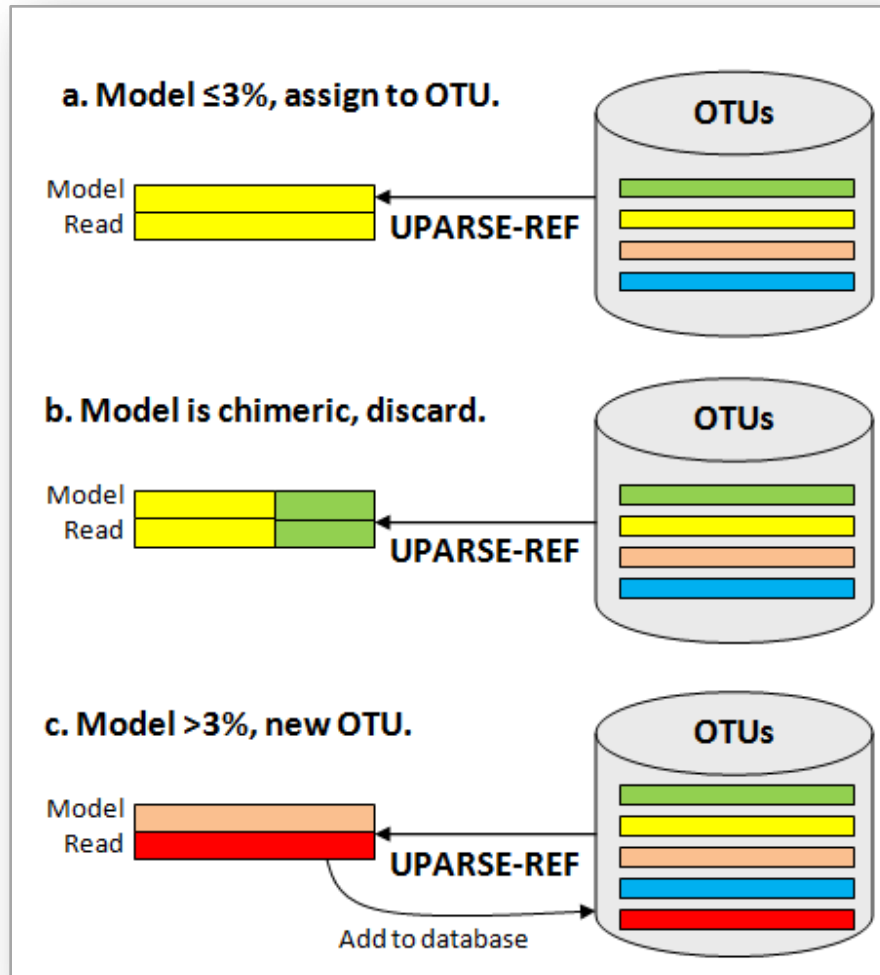
Click for more info

UPARSE-OTU

Process uniques
in decreasing
abundance order.

Compare each
sequence with
OTUs found so far.

Construct "model"
by max. parsimony
(fewest events)



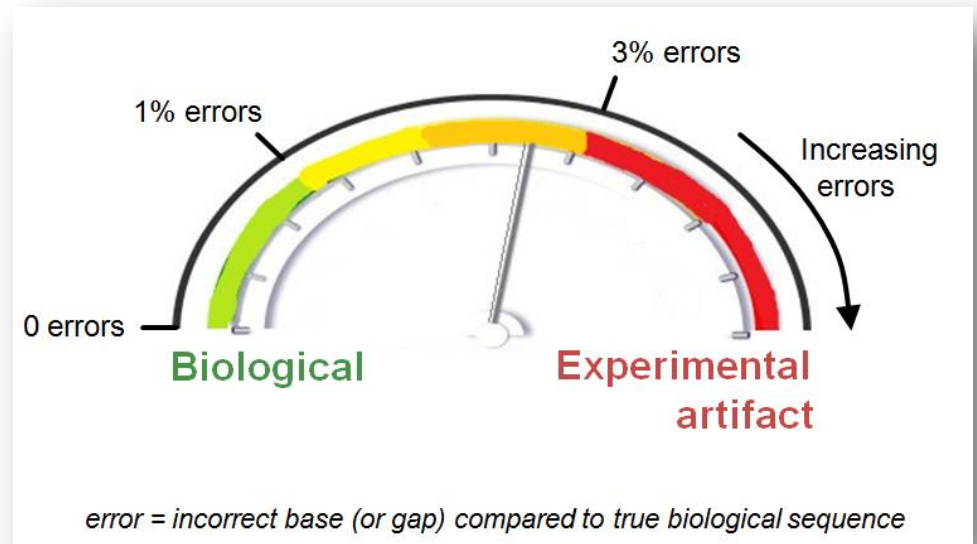
$\leq 3\%$ could be sequencing
error, chimera or correct --
don't need to distinguish.

Chimeras $> 3\%$ diverged
can be found accurately

Otherwise, new OTU

Benchmark test

- OTUs should be biological sequences
- Other criteria are possible, perhaps...
 - but should be clearly defined!
 - Nr. OTUs = nr. species popular but not valid



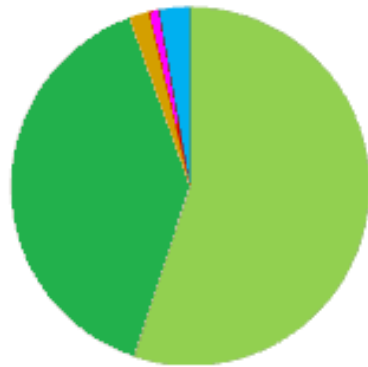
OTU classification

Color	Category	Description
Light Green	Perfect	100% identical to biological sequence.
Dark Green	Good	≥99% identical to biological sequence.
Yellow	Noisy	≥97% identical to biological sequence.
Red	Chimera	"Bad" chimera >3% from biological sequence
Grey	Contaminant	Sequence found in large ref. db.
Purple	Other	None of the above. Could be a novel contaminant, or -- much more likely -- have >3% errors.

16S mock community data

- HMP mock communities
- 21 species
- Even and Staggered mixes
- 454 Titanium and Illumina MiSeq 2x250
- Community & ref db. by Haas *et al.*
 - Chimeric 16S rRNA sequence formation and detection in Sanger and 454-pyrosequenced PCR amplicons. *Genome res.* (2011)

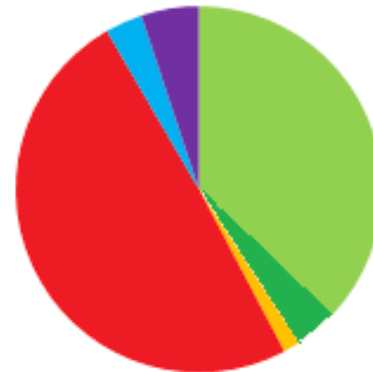
Results on HMP mock datasets



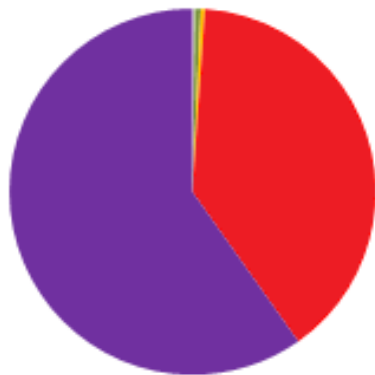
UPARSE 454



UPARSE Illumina



AmpliconNoise 454



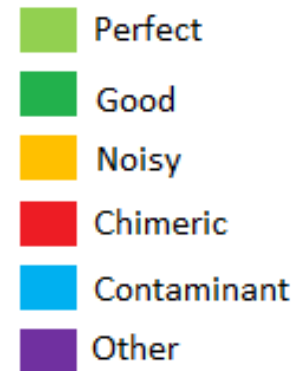
QIIME 454



QIIME Illumina



mothur 454



OTU table

- Matrix of OTUs vs. samples
- Value is nr. of reads

	Sample1	Sample2	Sample3	...
Otu1	1,023	455	992	...
Otu2	324	622	12	...
Otu3	871	29	321	...
....

QIIME "classic" tabbed text

Tab-separated text

Rows are OTUs, columns are samples

Simple, intuitive and convenient

Use cut, grep etc., load into spreadsheet...

```
#OTU ID F3D0      F3D141  F3D142  F3D143  F3D144  F3D145  F3D146  F3D147
OTU_6   749      535     313     372     607     849     493     2025
OTU_25  29       57      14      2       14      22      16      127
OTU_1   613     497     312     247     472     719     349     1720
OTU_8   426     378     255     237     382     627     330     1417
OTU_31  149     38      10      19      25      21      43      31
OTU_2   366     392     327     185     313     542     248     1367
OTU_7   196     370     92      107     48      155     74      105
OTU_10  46      169     87      109     171     209     120     864
OTU_80  26      6       0       1       4       8       18      11
```

mothur "shared" file

Tab-separated text

Rows are samples ("groups"), columns are OTUs

label	Group	numOtus	OTU_6	OTU_25	OTU_1	OTU_8	OTU_31	OTU_2	OTU_7	OTU_10	OTU_80
usearch	F3D0	9	749	29	613	426	149	366	196	46	26
usearch	F3D1	9	85	9	441	140	115	372	210	74	14
usearch	F3D141	9	535	57	497	378	38	392	370	169	6
usearch	F3D142	9	313	14	312	255	10	327	92	87	0
usearch	F3D143	9	372	2	247	237	19	185	107	109	1
usearch	F3D144	9	607	14	472	382	25	313	48	171	4
usearch	F3D145	9	849	22	719	627	21	542	155	209	8
usearch	F3D146	9	493	16	349	330	43	248	74	120	18
usearch	F3D147	9	2025	127	1720	1417	31	1367	105	864	11

BIOM v1 (JSON)

```
{
  "id":null,
  "format": "Biological Observation Matrix 0.9.1-dev",
  "format_url": "http://biom-format.org/documentation/format_versions/biom-1.0.html",
  "type": "OTU table",
  "generated_by": "QIIME revision 1.4.0-dev",
  "date": "2011-12-19T19:00:00",
  "rows":[
    {"id":"GG_OTU_1", "metadata":null},
    {"id":"GG_OTU_2", "metadata":null},
    {"id":"GG_OTU_3", "metadata":null},
    {"id":"GG_OTU_4", "metadata":null},
    {"id":"GG_OTU_5", "metadata":null}
  ],
  "columns": [
    {"id":"Sample1", "metadata":null},
    {"id":"Sample2", "metadata":null},
    {"id":"Sample3", "metadata":null},
    {"id":"Sample4", "metadata":null},
    {"id":"Sample5", "metadata":null},
    {"id":"Sample6", "metadata":null}
  ],
  "matrix_type": "sparse",
  "matrix_element_type": "int",
  "shape": [5, 6],
  "data":[[0,2,1],
          [1,0,5],
          [1,1,1],
```

Text, but complex
Hard to work with
in scripts
Can't use cut, grep,
awk...

BIOM v2 (HDF5)

- Totally unrelated to BIOM v1 format
- Not text, opaque binary format
- Motivation: huge OTU tables
 - e.g. Earth Microbiome Project

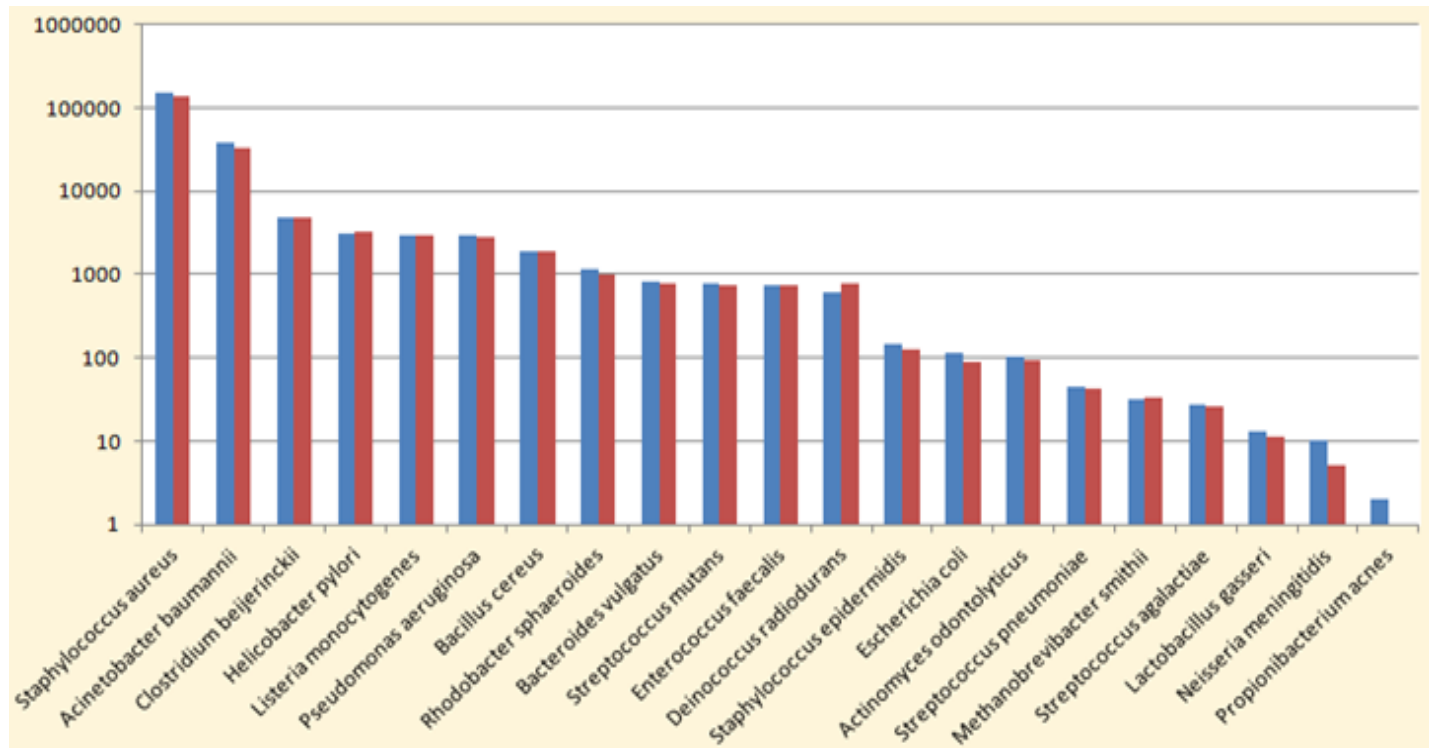


OTU table values

- Number of reads
 - "Raw"
 - Sub-sampled
 - e.g. to same number reads / sample
 - Rarefied
 - Normalized
- Frequencies
- No standards
 - Minimal software compatibility

Read abundance vs. cells

- Nr reads does **not** predict cell abundance



Read abundance for Even(!) mock community (Bokulich *et al.* 2013)

Metadata

- Taxonomy predictions
- Sample information
 - Healthy / diseased
 - Time / date, location...
 - Temperature, salinity, phase of moon...
- No standards, no software compatibility

Make OTU table with USEARCH

- Clustering gives one sequence for each OTU
 - "Representative sequence", "centroid"
- Align unfiltered reads to OTU sequences
 - database search (usearch_global command)
 - if $\geq 97\%$, assign to closest OTU
 - recovers most low-quality & singleton reads
 - almost all unmapped reads have many errors / chimeras
- Outputs one or more formats
 - QIIME classic, mothur shared and / or BIOM v1