







Ground Truth

Reproducibility

Standards





History does not repeat itself, but it rhymes.

Mark Twain

Core activities
Characterization (description)
Classification

Core resources

Identification

Literature and databases*

Reference materials

Tools

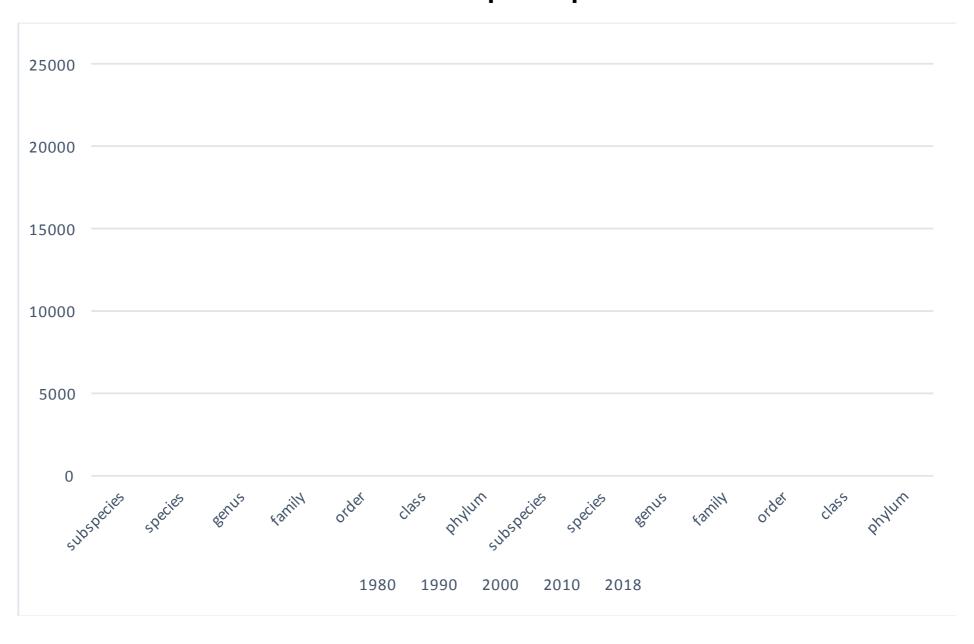
Hardware and software

SOPs and workflows





It started with a simple question



Would reannotation of taxonomic reference files be useful?





usearch guided reannotation

Manual review of fasta taxonomic annotations

Adjustment of taxonomic depth (seven levels)

Convert annotations to usearch format

Removal of eukaryote, plastid and cyanobacterial squences

Reassignment of sequence identity

Classification of relabeled sequences using usearch sintax function

NamesforLife type strain database as reference (April 2018 release)

Default cutoff value (pseudo-bootstrap of 80)

Sequence identified at all seven levels - reassign

Sequence identified at 5 or six levels – reassign if mean score > 80

Sequence identified at 4 or less levels – retain original identity

Correct reassigned names

Comparison of species level identity to NamesforLife nomenclature

Results

Eliminate virtually all taxa with multiple parents found in source files Increase number of correctly identified (high scoring) 16S sequences

However,

None of the reference databases cover >82.8% of the validly published bacteria and archaea





The microbiome experiment

Hypothesis – are OTU - OTU and OTU - taxon name consistent and meaningful when different reference taxonomies are applied?

Input data

eight diverse Illumina 16S (v4) metagenome samples

Software

mothur version 1.39, variation of Schloss' MiSeq SOP

Hardware

Mac Pro 3.7 GHz Quad Core Intel Xeon E5, 32GB RAM/SSD

Reference taxonomies

NamesforLife type strain database (May 17, 2018 release)

Silva nr_v.132 (trimmed, using both original annotation and reannotated)

Analysis – Principle Coordinate Analysis

Non-metric Multidimensional Scaling

Test for significance - Kolmogorov – Smirnov







Assemble contigs

Align

contigs

Screen contigs

Select unique sequences

Screen aligned contigs

Filter sequences

Select unique sequences Pre-cluster unique sequences

De-noise

Chimera check

Remove bad sequences

Classify sequences*

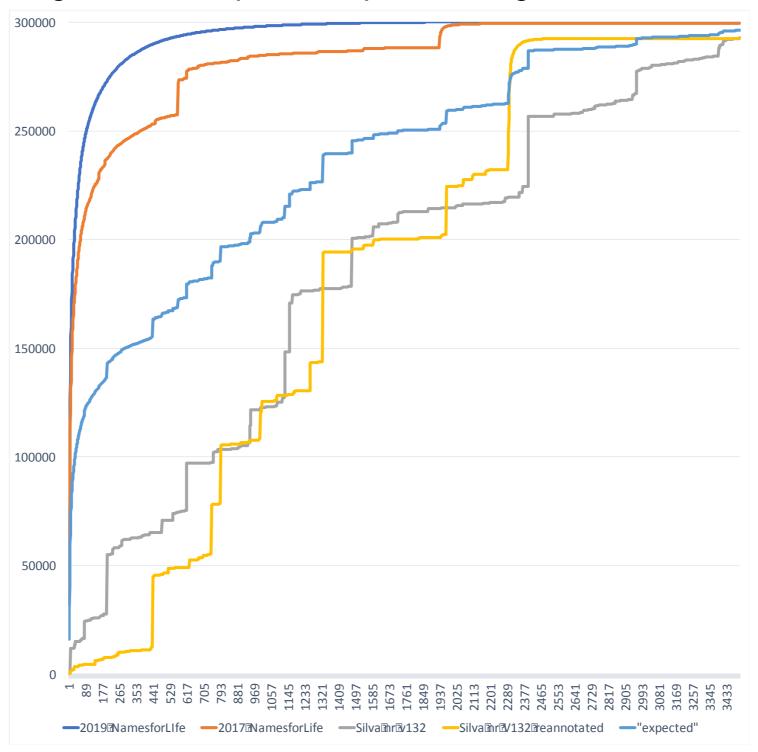
N4L 2018, 2017, 2016, 2013, 2010, 2000, 1990, 1980, Silva v132, Silva v132 reannotated Output consensus taxonomy, shared reads, OTU to unique sequence mapping

External analysis





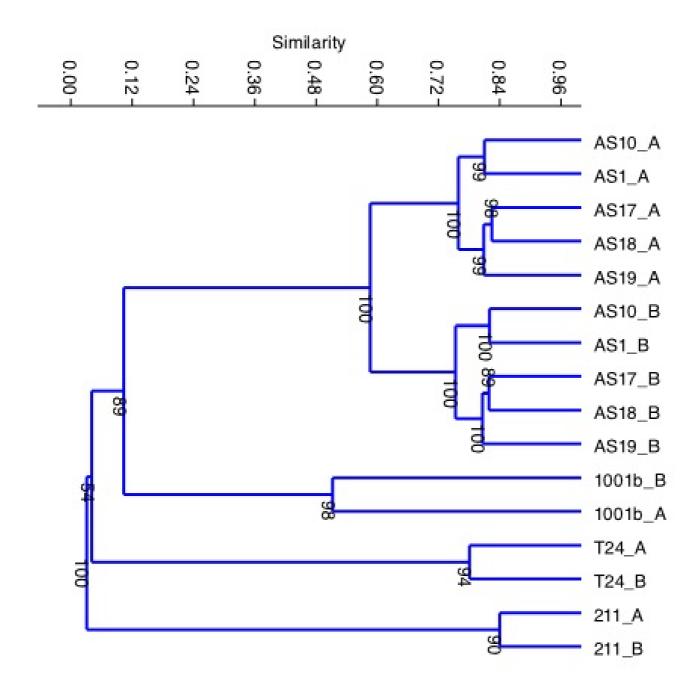
Cumulative taxonomic abundance, combined metagenome samples compared using four taxonomies







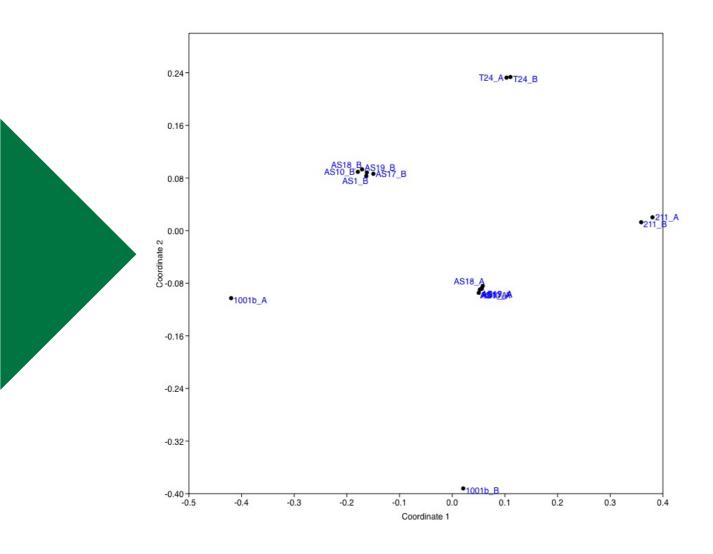
Bray-Curtis distance, 2018 vs 2017 taxonomy, UPGMA, bootstrapped 500 iterations

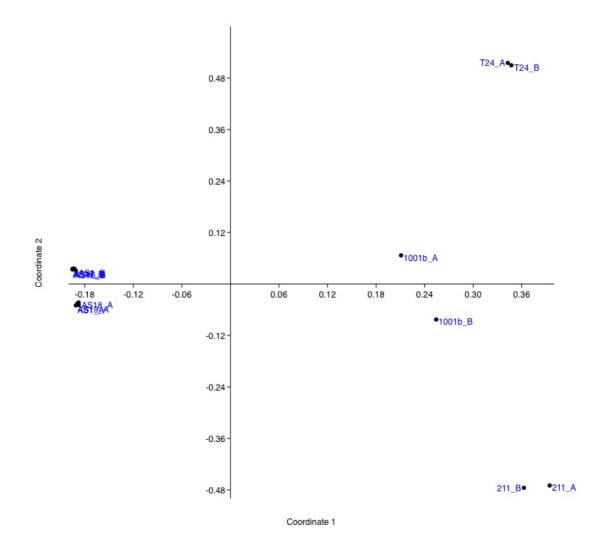






NMDS and PCoA of metagenome analysis using 2017 and 2018 taxonomies







The analysis

The goal – objective way of comparing two or more taxonomies applied to the same metagenome samples

 H_0 – no difference between taxonomies

 H_a – taxonomies different, comparison requires reannotation using same taxonomy

Nature of metagenome and taxonomic data – nonparametric, unbounded

The Kolmogorov-Smirnov Test (2 sample)

- 1. Test whether two samples come from the same/different distributions
- 2. Cumulative distribution of named reads are compared
- 3. KS test statistic is estimated

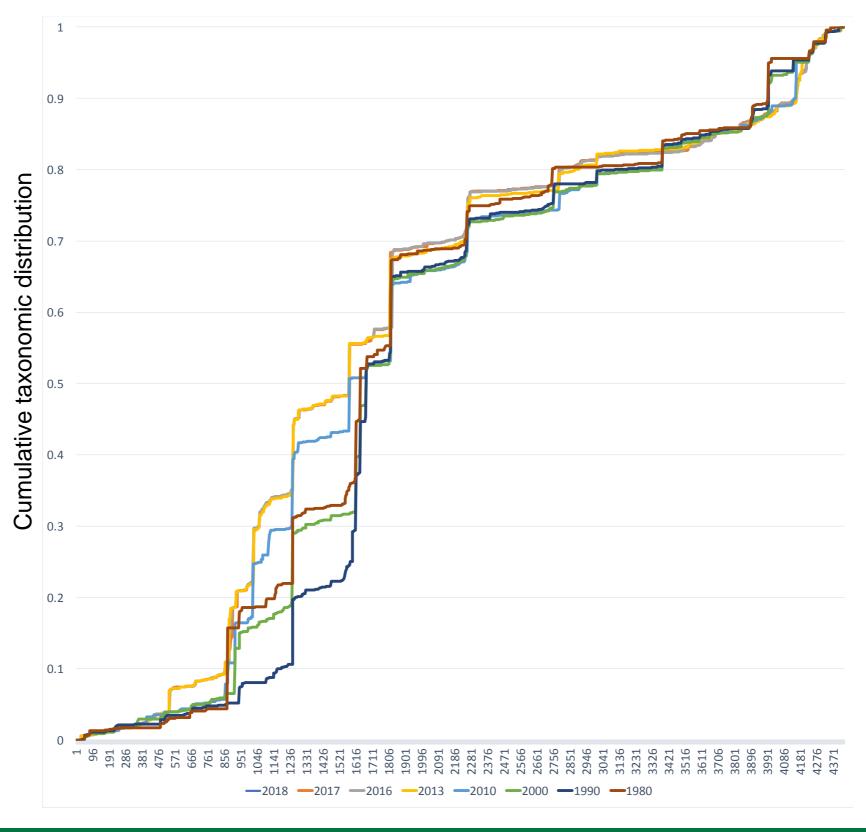
$$KS = \sqrt{|TD_{1,n} - TD_{2,n}|_{max}/n}$$

where

TD = cumulative taxonomic distribution measured for differences between paired taxon frequencies n = number of unique taxa in sample











What we found

Of the possible pairwise comparisons of taxonomic distributions for the amplicon metagenomics data, only two distributions were considered the same: 2018-2017. All others were significantly different from one another.

Comparison of samples required reanalysis with and against the same taxonomic file to ensure that any differences between samples are due to biological or environmental factors.

Results of analyses and any assertions of novel taxa or functions may not be meaningful if an out-of-date reference taxonomy is used.

Given the rate of change, taxonomic reference files more than one year old should be reannotated prior to use.

Reproducibility – it depends Replicability – it depends Generalizability – it depends





The ANI experiment

Objective – Reproduce previous comparative studies using ANI_M, ANI_B, ANI_{BH}, and extend to ANI_G and AAI

Hypothesis – closely related strains will have higher ANI/AAI scores. Same strain will have identical score.

 $ANI_{A\&B} = \sum (\% identity * alignment length) / \sum (length genes in genome A)$

 $AF_{A\&B} = \sum (length BBH genes) / \sum (length genes in genome A)$

Differences among methods regarding source/quality of genomes coding vs. non-coding

Use protein coding genes only vs. relaxed approach

Plasmid and other extrachromosomal genes removed

Self contained vs. external calls to 3rd party software or services

Output is % similarity (A-> B, B->A), but may not always be transitive.

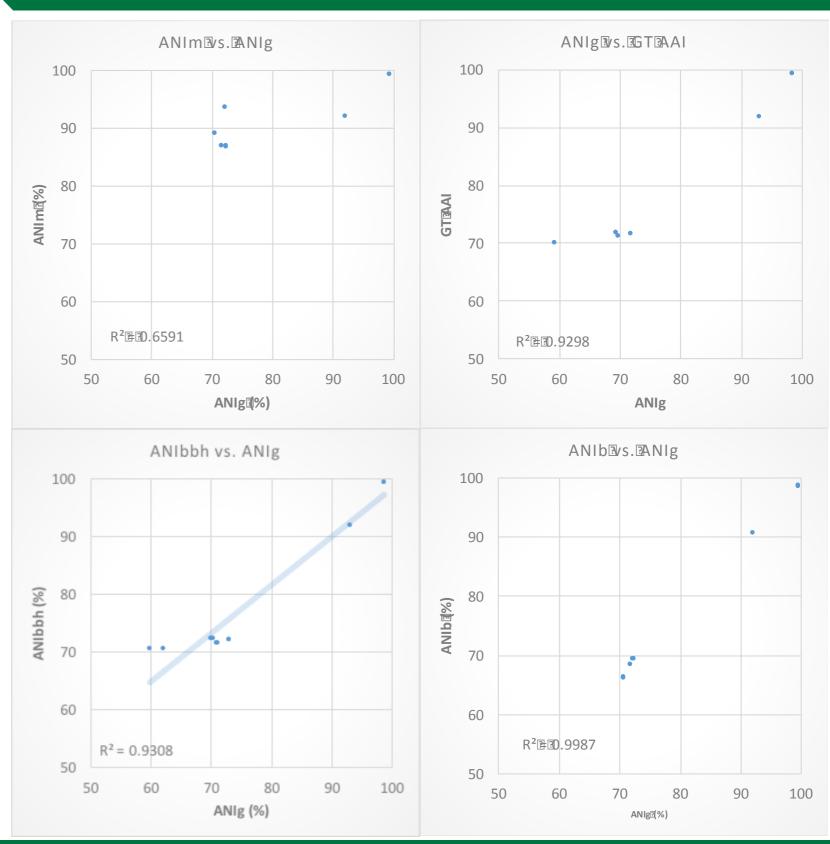
Establish thresholds for identifying species/subspecies

At this time the method may not yet be useful for identification and classification.





Comparison of major ANI algorithms, reanalysis of Krebs reference genome collection





Findings

Objective – Reproduce previous comparative studies using ANI_M, ANI_B, ANI_{BH}, and extend to ANI_G and AAI

Problems in establishing exact input sequences

Some results were ambiguous across methods

Idea of fixed cut-offs problematic but thresholds may be useful (e.g. MiSi) In ongoing studies comparing true replicate genome sequences, ANI methods rarely show identity.

Effect of sample preparation, sequencing, assembly and annotation methods likely to prove important or significant.

Current thoughts

Documentation of source materials and methods are frequently inadequate or lacking

Documentation of ANI method and version, date of web service used, any methodological or analytical variations needed to correctly interpret results

Reproducibility – sometimes Replicability – sometimes Generalizability – not yet





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