Greengenes 16S Ribosomal RNA Gene Database Update For 2011 HMP Reference Sets And Tools

#### CONTRIBUTORS

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## ABSTRACT

Due to global interest in the human microbiome's role in health and disease, a diverse community of international researchers from the medical, microbiological and computational fields have recently converged to address questions in microbial ecology. Activities such as describing community structure, portraying population dynamics, and depicting diagnostic test candidates all benefit from mapping assay data to high-quality reference sets with useful nomenclature. In popular workflows, ribosomal gene segments are hybridized to probes or sequenced with NGS technology. Annotating the matches to both cultured and vetted uncultured clades reveals trends overlooked by sole reliance on cultured references. The 2011 Greengenes 16S rRNA Taxonomy was created from Infernal-improved NAST alignments, FastTree-validated tree topology, nomenclature reconciliation with NCBI for cultured strains, and manual curation of thousands of yet-to-be cultured groups. The effort resulted in standardized or proposed names for >4000 hierarchical taxonomic nodes. From these relations, existing datasets from 454 (Roche), HiSeq (Illumina), and PhyloChip Assay (Second Genome) and expected datasets from PGM (Ion Torrent) and PacBioRS (Pacific Biosciences) can be compared. Files are available for download in their complete forms as well as subsets suitable for metagenomic pipeline tools. Furthermore, a chromatogram processing and capture tool has been established for those desiring to contribute to future reference sets.

## NOVEL TREE - RICH TAXONOMY

### **→ Methods**

▶Infernal alignment of full length sequences (Nawrocki, 2010)

- ▶ Hypervariable Lane mask (Lane, 1991) ▶Uchime chimera filter (Edgar, 2010)
- ▶ Chimera Slayer chimera filter (Haas, 2011)
- ▶ Dual Study Heuristic filter
- ▶ FastTree de-novo tree construction from 408K seqs (Price, 2010)

## ▶tax2tree node re-mapping (McDonald, in prep)

- ▶nomenclature mapping from NCBI
- ▶nomenclature mapping from Greengenes
- ▶ Precision/Recall name conflict resolution ▶Back-filling classifier for unnamed nodes
- ▶Back-propagation to collapse redundantly named nodes

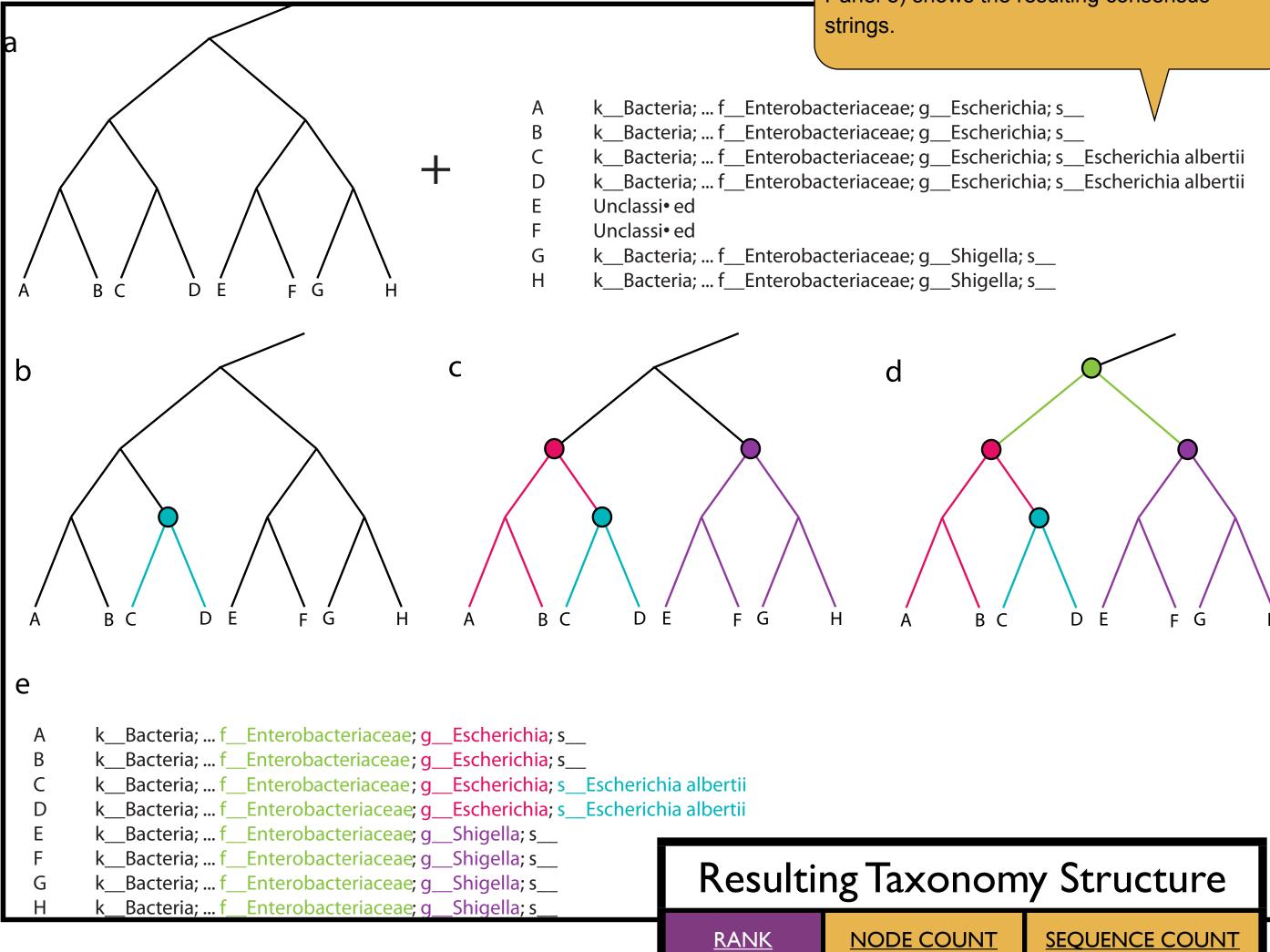
## Overview of the tax2tree workflow:

Panel a: Inputs are a phylogenetic tree + taxonomy strings for some (or all) of the tips in the tree including two Unclas-

sified tips (E,F). The taxonomy mapping can optionally have rank abbreviations already appended on (k\_\_\_, p\_\_\_, etc...). Note, the consensus strings are shortened (...) for brevity.

Panels b, c, and d: Assignment of species, genus and family nodes, respectively. When decorating genus, we are able to infer in this case that tips E and F are under g\_\_Shigella as the lowest common ancestor with tips G and H has >= 50% relative abundance of the genus name.

Panel e) shows the resulting consensus



#### **Examples of important** nomenclature updates:

Brevibacterium stationis Desulfomicrobium terraneus

Rhodoferax ferrireducens

Vibrio fischeri

Anaerocellum thermophilum

Marinibacillus marinus Clostridium orbiscindens Thermoanaerobacter tengcongensis Caldicellulosiruptor bescii Corynebacterium stationis Desulfomicrobium thermophilum

Species

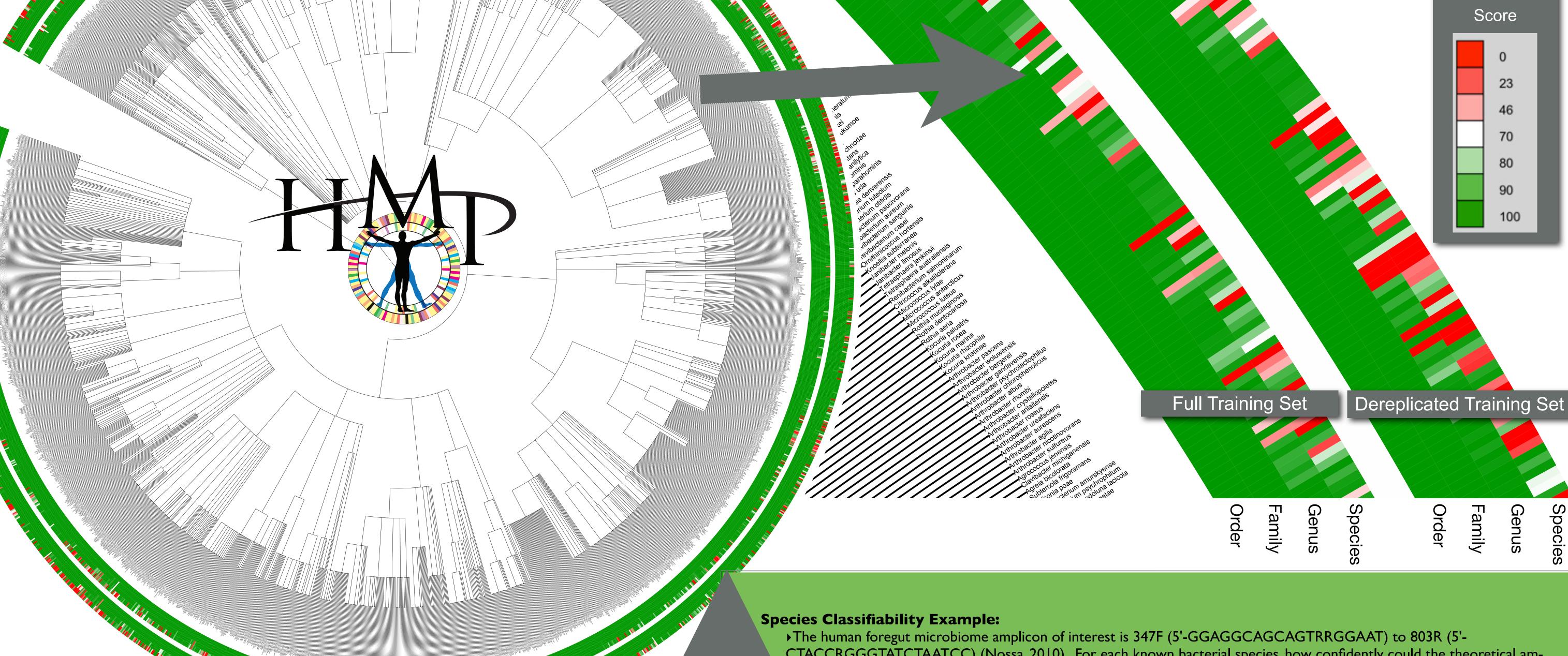
Jeotgalibacillus marinus Flavonifractor plautii

Albidiferax ferrireducens

Aliivibrio fischeri

Caldanaerobacter subterraneus

Frequently Asked Question: "Which species should I expect to confidently distinguish considering my subsection of the 16S rRNA gene sequenced from my specimens?"



## **CLASSIFIABILITY**

Test by classifying 1,200 sequence clusters from human feces against GG and other reference sets.

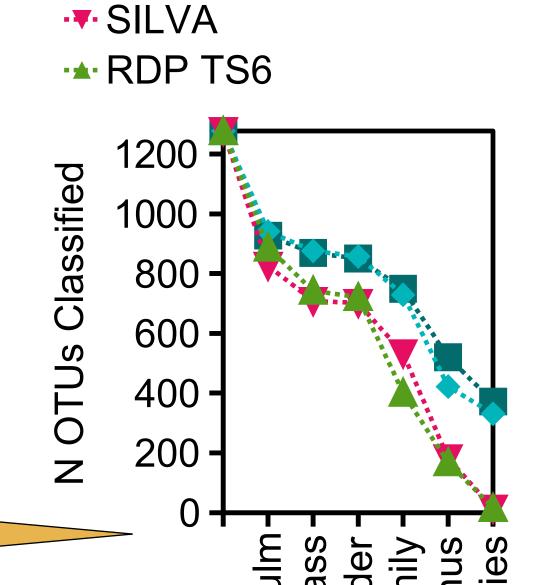
408K GG reference sequences

Dereplicate at 99% or 91.3%

Trim to amplicon span Train Bayesian classifier

Compare number of pyrosequenced OTUs classified at each rank against using GG or other refer-

New Greengenes taxonomy allows taxonomic nomenclature to be applied to a significantly greater number of NGS sequences.



-- GG99

**→** GG91.3

CTACCRGGGTATCTAATCC) (Nossa, 2010). For each known bacterial species, how confidently could the theoretical amplicon distinguish it from all other nodes (named or not)?

#### Method

The search pattern, GGAGGCAGCAGTRRGGAATJGGATTAGATACCCYGGTAG, was used in conjunction to the greengenes online sub-alignement locator (http://greengenes.lbl.gov/cgi-bin/nph-probe\_locator.cgi) to determine the coordinates of the theoretical amplicon at position 1882 to 4081 (inclusive of both primers). A subsequence from 1917 to 4048 gathers all the DNA sequence exclusive of the primers.

▶408K sub-sequences were obtained.

http://greengenes.lbl.gov/Download/Sequence\_Data/Fasta\_data\_files/

→347to803\_gg\_norm\_unaligned.fasta.gz Train the Mothur (Schloss, 2009) classifier using all 408K sequences (Full Training Set) or a non-redundant set of 176K sequences (Dereplicated Training Set).

Classify each sequence from the training set and record the confidence at each taxonomic rank.

Find average "classifiability" for each node as the mean confidence of classification to that node divided by the set of sequences belonging to that node.

▶ Plot the classifiability of the named species as a heat map using ITOL (Letunic, 2006).

## **Observations**

Family-level classification was highly confident throughout the tree.

▶Genus and species level confidence is dependent on the genus and species.

▶ Confidence was greater using the Full Training Set compared to the Dereplicated Training Set.

### For Further Discussion

▶Within the trained model the count of taxonomy-by-word intersections will not differ across the two methods but the priors will change.

▶ More redundant sequences in one node than another affects the model's view of the distribution of a given 8mer across taxonomic nodes.

## PHYLOCHIP<sup>TM</sup> ASSAY ANNOTATION

#### New Greengenes taxonomy facilitates PhyloChip results annotation.

1,016,064 probes

sity

Hybridizatio

314K

112K

- 59K OTUs, each tracked with multiple probe pairs
- Overcomes sampling effort problem encountered with NGS approaches typical I6S rRNA gene PCR yields 500 to 1,000 ng in a 20 uL volume
- /uL → 6E+11 total sequences → 600 billion sequences per PCR How many should we observe? 600? 60,000 (I out of every 10 mil-

1500bp @ 660 g/mole/bp  $\Rightarrow$  5E-14 moles / uL  $\Rightarrow$  3E+10 molecules

SECOND GENOME
THE MICROBIOME COMPANY

Days after treatment

OTU 46174

21

- Hybridize them all on to a PhyloChip ...

OTU 31902

Dominant populations do not occlude minority populations

Example output from a PhyloChip (Second Genome, Inc.) experiment tracking bacterial population dynamics in hindgut for 21 days after a specific medical treatment OTU annotations are mapped from the new Greengenes taxonomy:

### OTU 31902: p\_\_Cyanobacteria;

4C0d-2; o YS2; unclassified

## **OTU 46174**:

Bacteroidetes; Bacteroidia;

o Bacteroidales; Rikinellaceae;

# GREENGENES SUBSETS

- All named isolates
- All HMP Genome Strains
- Knight, Caporaso: QIIME-ready reference sets
- Current aligned Core Set for NAST templates 36,550 genes represents the known 16S diversity.
- Older Core Sets
- VISIT: <a href="http://greengenes.lbl.gov/Download/Sequence\_Data/Fasta\_data\_files">http://greengenes.lbl.gov/Download/Sequence\_Data/Fasta\_data\_files</a>

## GENOMES QC

Contig collections from genome projects finished or unfinished can be quality-assessed using the Greengenes Core Set. Surprisingly, over 50 genomes contain zero full-length 16S rRNA genes. Finishing effort will be needed to assemble 16S genes in these projects for use as references for future trees.

## ACKNOWLEDGMENTS

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