There is no evidence of a universal genetic boundary among microbial species

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Abstract

A fundamental question in studying microbial diversity is whether there is a species boundary and if the boundary can be delineated by a universal genetic discontinuity. To address this question, Jain et al. computed the pairwise average nucleotide identity (ANI) of 91,761 microbial (bacterial and archaeal) genomes (the 90K genome dataset) and found that the ANI values from the 8 billion comparisons follow a strong bimodal distribution[1]. The authors concluded that a clear genetic discontinuum and species boundary were evident from the unprecedented large-scale ANI analysis. As a result, the researchers advocated that an ANI of 95% can be used to accurately demarcate all currently named microbial species. While the FastANI program described in the paper is useful, we argue that the paper's conclusion of a universal genetic boundary is questionable and resulted from the substantial biased sampling in genome sequencing. We also caution against being overly confident in using 95% ANI for microbial species delineation as the high benchmarks reported in the paper were inflated by using highly redundant genomes.

Main Text

To demonstrate our point, we first show that biased within-species sampling can generate the bimodal distribution of ANI observed in the original paper even when the speciation rate is constant and the genetic diversity is continuous. We simulated a phylogenetic tree with 3,000 tips and a constant rate of diversification, estimated from a genome tree of 3,000 bacterial genomes that represents the phylogenetic diversity

within the 10,616 NCBI RefSeq complete bacterial genomes (the 10K genome dataset).
We calculated ANI between tips using a function that accurately captures the
relationship between ANI and the branch length in the real data (Supplementary Fig. 1).
Because the diversification rate is constant, the branch lengths follow an exponential
distribution expected from a Poisson process. As expected, the frequency of ANI
declined monotonically when ANI increased (Fig. 1a). However, when only 30 tips (1%
of the tips) were sampled with a protocol that emulated within-species sampling bias
(each tip has two very closely related genomes sequenced), the ANI distribution
became bimodal (Fig. 1b). The right peak increased in prominence when more tips were
sampled with bias (data not shown). Our simulations show that limited within-species
sampling bias alone can create the bimodal distribution when genetic diversity is
continuous and a universal species boundary does not exist.

In the original 90K genome dataset, 33% of named species have been
sequenced at least twice. As cultivation bias is widespread, next we show that there is
substantial within-species sampling bias in the genome datasets. For the model
organism *Escherichia coli*, its 602 complete genomes in the 10K genome dataset only
represent 22% of the diversity captured by the 16S rRNA gene in the GreenGene
database (Supplementary Fig. 2). For species with at least 100 genomes, on average
the first 75.8% [range: 50.9-97.7%] of dropped genomes contribute less than 5% of the
genetic diversity of the species as measured by the branch length (Fig. 2). As
demonstrated in our phylogenetic simulation, these highly redundant genomes can
create the bimodal distribution even when genetic diversity is continuous. Contrary to
the authors' claim, randomly subsampling (e.g., 5 genomes from the same species)

from a biased dataset will not correct for the pre-existing sampling bias. When two

genomes were sampled randomly from each of the 397 named species with >= 10

genomes in the 90K dataset, the bimodal distribution was evident (Fig. 1c). However,

when we reduced the within-species sampling bias by selecting two phylogenetic

representative genomes from the same dataset, the distribution flattened near the end

(Fig. 1d), further demonstrating that the bimodal distribution is caused by widespread

sampling bias within species. In fact, the within-species sampling bias has increased

over time when more strains were sequenced based on their medical and economical

relevance and not on their phylogenetic positions (Supplementary Fig. 3), thereby

producing the consistent bimodal distributions of ANI over different periods of time as

observed in the original study[1].

The authors indicated that only 0.2% of the 8 billion ANI values are between 83%

and 95% and suggested that such a wide gap is evidence of a clear genetic boundary in

microbial genomes. In our phylogenetic simulation of 3,000 genomes with continuous

genetic diversity, only 0.11% of ANI values span the same region. This is expected

because without biased sampling, the vast majority of all pairwise comparisons for a

genome will be with distantly related genomes, whose number will always be much

larger than the number of closely related taxa. As a result, the fraction of ANI values in

the intermediate range [83%-95%] will always be marginal for a decently sized tree (>

50 tips) and will decrease when the number of genomes increases (Supplementary Fig.

4). Our result shows that the low density of ANI values in the gap region does not

necessarily indicate a genetic boundary. Instead, it could simply reflect the hierarchical

structure of the phylogenetic relationship.

Previous studies have advocated using an ANI cutoff to demarcate bacterial

species[2–5]. Based on the unprecedented large-scale ANI analysis of the study, Jain et

al. claimed that using the 95% ANI criterion led to both high recall and precision rates

(>98.5%) in demarcating microbial species[1]. However, both benchmarks were

calibrated with ~78k named genomes that were highly redundant. For example, there

were more than 5,000 *E. coli* genomes in the dataset. As such, the recall and precision

rates can be misled by the overly sampled genomes. To better assess the performance

of using 95% ANI for species demarcation, we queried the 3,000 representative

genomes against the 10K dataset. We found the recall and precision rates were much

lower, at 73.4% and 83.3% respectively. The large impact of the extreme sampling bias

on the benchmarks is best illustrated by the authors' own finding that excluding the *E.*

coli vs. *Shigella spp.* comparison alone substantially increased their overall precision

from 93.1 to 98.7%. Plotting of the intraspecific ANI for intensely sequenced species in

the 90K dataset (61 species with >=100 genomes, representing 6 phyla) shows that a

universal boundary at 95% ANI clearly does not exist, as ANI values drop well below

95% with few exceptions (Fig. 3).

Although having no cultivation bias, metagenome assemblies do favor abundant

strains over rare ones. This bias can lead to the appearance of genetic clusters when

diverse but rare strains in the community are excluded from the assembled genomes.

However, a rare strain in one environment can be abundant in another. For example,

various *E. coli* strains spanning a continuum of diversity live in the gut, water, and soil[6].

Analyzing gut metagenomic data alone will most likely reveal a genetic cluster of *E. coli*

because it will miss the environmental *E. coli* strains that are either rare or absent in the

gut. Unless low abundance strains can be readily recovered or metagenomic

sequences from different types of environments are compared, there are also potential

pitfalls associated with demonstrating naturally occurring clusters using metagenome-

assembled genomes (MAG)[7]. Interestingly, several metagenomic studies have revealed

genetic continuum in nature[8,9].

There is much evidence against the existence of a universal genetic boundary for

microbial species. First, the rate of evolution is highly variable across species.

Secondly, selection and recombination are thought to be the main cohesive forces

driving the formation of genetic clusters. Although recombination rate can be influenced

by sequence similarity, there is no correlation between the recombination rate and ANI

in bacteria[10], as recombination can also be affected by physical and ecological barriers.

Microbes living in narrow ecological niches and with limited dispersal rate (e.g., obligate

intracellular bacteria) may develop genetic clusters. On the other hand, free living

microbes exploring different habitats and mixing by dispersal are more likely to exhibit a

genetic continuum[11]. Selection is unlikely to produce a universal genetic boundary

either, as microbial species are unique in nature, with each species subject to its own

evolutionary and ecological forces[12].

In summary, our study shows that the genetic boundary perceived in the original

paper can be explained by persistent within-species sampling bias from historic and

current genome sequencing efforts. A more balanced analysis of the present genomic

data shows that although genetic clusters may exist in individual species, there is no

universal genetic boundary among microbial species in the existing taxonomy.

Figures

Figure 1. Distribution of ANI values. Top panel: from the phylogenetic simulations a).

Comparisons between 3,000 taxa simulated using a phylogenetic tree with 3,000 tips

and a constant rate of diversification. b). Comparisons in the same dataset except that

30 of 3,000 taxa each have two very closely related genomes sequenced. Bottom

panel: from genomes subsampled from the 90K genome dataset. c). Two genomes

were randomly selected from each of the 397 named species with >= 10 genomes. d).

Two phylogenetic representative genomes were selected for each of the 397 named

species with >= 10 genomes.

Figure 2. Widespread within-species sampling bias revealed by the Treemmer analysis.

Treemmer iteratively drops genomes contributing the least amount of diversity within a

phylogenetic tree until only three genomes remain. For each species, the relative tree

length (RTL) is plotted against the proportion of genomes dropped at each iteration.

Colored lines are the 5 most sequenced species in the 10K genome dataset. Shaded

black lines are species with >=10 genomes. The bar chart shows species ranked by the

number of genomes sequenced.

Species

	Genome counts
Yersinia pestis	306
Yersinia enterocolitica	156
Xanthomonas oryzae	136
Xanthomonas axonopodis	110
Vibrio parahaemolyticus	691
Vibrio cholerae	577
Streptococcus suis	961
Streptococcus pyogenes	333
Streptococcus pneumoniae	7398
Streptococcus mutans	172
Streptococcus equi	251
Streptococcus agalactiae	892
Stenotrophomonas maltophilia	131
Staphylococcus haemolyticus	170
Staphylococcus epidermidis	387
Staphylococcus aureus	7575
Shigella sonnei	971
Shigella flexneri	156
Serratia marcescens	313
Salmonella enterica	5425
Pseudomonas syringae	208
Pseudomonas fluorescens	111
Pseudomonas aeruginosa	2284
Pasteurella multocida	113
Oenococcus oeni	203
Neisseria meningitidis	799
Neisseria gonorrhoeae	411
Mycobacterium tuberculosis	5181
Mycobacterium avium	113
Mycobacterium abscessus	746
Listeria monocytogenes	1312
Leptospira interrogans	289
Legionella pneumophila	550
Lactobacillus rhamnosus	113
Lactobacillus plantarum	184
Klebsiella pneumoniae	1637
Helicobacter pylori	649
Haemophilus influenzae	135
Francisella tularensis	107
Escherichia coli	5546
Enterococcus faecium	511
Enterococcus faecalis	484
Enterobacter hormaechei	269
Enterobacter cloacae	520
Enterobacter aerogenes	108
Clostridium botulinum	188
Clostridioides difficile	940
Chlamydia trachomatis	141
Campylobacter jejuni	934
Campylobacter coli	808
Burkholderia ubonensis	286
Burkholderia pseudomallei	672
Burkholderia cenocepacia	248
Brucella abortus	177
Bordetella pertussis	611
Bacteroides fragilis	115
Bacillus subtilis	129
Bacillus cereus	342
Bacillus anthracis	148
Acinetobacter pittii	105
Acinetobacter baumannii	1999

density

0.75
0.50
0.25

ANI

Figure 3. The distribution of ANI for 61 named species with >= 100 genomes in the 90K

genome dataset. Each block represents a bin of ANI values between pairs of genomes

within the same species and is colored by the ANI density in the bin. Species are

colored by phylum. The vertical dotted line represents the 95% ANI threshold. The high

ANI density on the right (above the 95% threshold) is due to comparisons between

highly redundant genomes from the sampling bias.

Supplementary Materials

Supplementary Figures

Supplementary Figure 1. The relationship between the branch length and ANI. Shaded

black points are pairwise comparisons between the 3,000 representative genomes. Blue

points are the median ANI for each branch length bin (bin width: 0.05 substitution/site).

Orange line is the fitted curve. Branch lengths greater than 2.5 were removed because

of the lack of data points.

Supplementary Figure 2. Incomplete and biased coverage of *E. coli* diversity by

complete genomes. Of 44 *E. coli* 99% OTUs in the phylogeny of the 16S rRNA gene,

only 15 (in black) are represented by the 602 complete *E. coli* genomes in the 10K

genome dataset, and only 22% of the total branch length in the phylogeny is covered by

these 15 OTUs. The branch length was square-root transformed to better show the

short branches in the figure.

Supplementary Figure 3. Relative tree length (RTL) decay for species with >= 10

sequenced genomes in the NCBI RefSeq database in four different time periods.

Supplementary Figure 4. The percentage of ANI values in the [83%-95%] range is

marginal and has a negative relationship with the number of genomes. A random

phylogeny with an exponential distribution of branch lengths was used to simulate

continuous genetic diversity across genomes. The pairwise ANI values were

calculated from the phylogenetic distances between genomes (see materials and

methods). The simulations were run with phylogenies of different sizes ranging from 50

to 3,000 tips.

Materials and Methods

Genome datasets

Two genome datasets were used in this study. The first is the 90K genome
dataset from the original paper. It contains both complete and draft bacterial and
archaeal genomes. The second dataset consists of 10,616 complete bacterial genomes
downloaded from the NCBI RefSeq database on September 6, 2018 (10K genome
dataset). From each genome in the 10K dataset, we identified 31 universal protein-
coding marker genes using AMPHORA2[13] and constructed a bacterial genome tree
based on the concatenated and trimmed protein sequence alignment of the marker
genes using FastTree[14]. Treemmer (version 0.3)[15] was used to choose 3,000
representative genomes that maximized the phylogenetic diversity in the 10K genome
dataset.

ANI

The ANI values for the 90K genome dataset were downloaded from the original
study. For the 10K genome dataset, the 3,000 representative genomes were compared
against the full 10K dataset using FastANI (version 1.2).

Modeling the relationship between branch length and ANI

The 3,000 representative genomes were used to model the empirical relationship
between ANI and branch length. The median of ANI was calculated across binned
branch lengths (bin width: 0.05 substitution/site) to use as the actual data to fit the
relationship between ANI and branch length l through the function $ANI(\%) = k +$

$\frac{\alpha \cdot (100-k)}{l^{s}+\alpha}$, where k, s and α are shape parameters to be estimated. Minimization of the

sum of squares error was performed using the *optim* function in R. The best fit

parameters for our data are $\alpha = 0.075$, $k = 73.94$, and $s = 0.63$. Branch lengths greater

than 2.5 substitutions/site were removed because of the lack of data points.

<u>Simulation of continuous genetic diversity and biased within-species sampling</u>

The *rtree* function in the *ape* package in R was used to simulate a random

phylogenetic tree of 3,000 tips, with its branch lengths following an exponential

distribution with a constant rate of 19.2, estimated from the genome tree of the 3,000

representative genomes. Using the formula described above, the ANI value between a

pair of genomes was computed from the branch length between them. To simulate

biased sampling within species, a random tip was chosen and two descendants were

added to the tip, with the branch length from the tip to the descendant sampled from the

same exponential distribution, but its value restricted to the bottom 1% of the

distribution. This procedure was repeated on the remaining 2,999 tips until n tips were

processed this way. Each simulation was run with 10 replicates.

<u>Assessing the within-species sampling bias</u>

Species with >= 10 genomes were selected from the 10K genome dataset. For

each species, a subtree compiling the respective genomes was extracted from the full

phylogeny of 10,616 genomes and Treemmer was used to iteratively remove one tip of

the tip-pair with the shortest branch length until 3 tips remained. The remaining total

branch length of the tree was divided by the total branch length of the initial tree to

calculate the relative tree length (RTL) at each iteration. *Rickettsia japonica* and

Chlamydia muridarum were removed from our analyses because their genomes have

identical marker sequences and branch lengths equal to zero.

To further evaluate the within-species sampling bias, we tested how much known

genetic diversity is recovered by the complete genomes, using *E. coli* as an example.

We extracted 868 unique 16S rRNA gene sequences with no ambiguous bases from

602 complete *E. coli* genomes in the 10K genome dataset and BLAST searched them

against 8,655 *E. coli* 16S rRNA gene sequences from the GreenGene 13.8 database. A

match was defined as a pair of sequences with 100% identity for their entire sequences.

The matched GreenGene 16S rRNA sequences were then mapped to the 99% OTUs

(operational taxonomic units) of the GreenGene database. The total branch length

covered by the mapped OTUs in the 16S rRNA tree of 44 *E. coli* 99% OTUs was

calculated to estimate the coverage of *E. coli* diversity by complete genomes.

<u>Subsampling of two genomes</u>

For named species with >= 10 genomes in the 90K dataset, two genomes were

sampled from each species either randomly or by selecting the pair with the lowest ANI

value. Among all pairwise comparisons within the species, the pair with the lowest ANI

best represents the phylogenetic diversity of the species.

<u>Benchmark the performance using 95% ANI for species demarcation</u>

Using the 3,000 representative genomes as the query, we ran FastANI against

the full 10K genome dataset. For each query genome, the subject genome with the

maximum ANI value was used to benchmark the performance of using the 95% ANI

threshold to demarcate bacterial species. A true positive is a query-subject pair

belonging to the same species and having an ANI >= 95%. A false positive is a genome

pair of different species with an ANI >= 95%, and a false negative is a genome pair of

the same species with an ANI of less than 95%. Precision was calculated by: the

number of true positive / (number of true positive + number of false positive) and recall

was calculated by: the number of true positive / (number of true positive + number of

false negative).

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Contributions

M.W. conceived the design of the response, C.S.M., Y.G. and M.W. conducted the

analysis, M.W. wrote the first draft and C.S.M., Y.G. and M.W. edited the final version.

Competing interests

The authors declare that there are no competing interests.