

Supplementary Data

riboSeed: leveraging prokaryotic genomic architecture to assemble across ribosomal regions

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Extended Methods

Reference Selection Recommendations

Using a recently-diverged reference sequence maximizes chances of a successful assembly. We have outlined two methods to select an appropriate reference for a given isolate: a robust method, and a quick method.

Method 1: Kraken

Kraken [11] is a k -mer-based phylogeny tool that can be used to identify the strains present in a metagenomic dataset; installation and usage instructions can be found here: <https://ccb.jhu.edu/software/kraken/>. After downloading and installing Kraken, along with the MiniKraken database from their website, Kraken can be run on an isolate's reads, generating the a taxonomy report.

The MiniKraken database was built from all the complete genomes from RefSeq, allowing the user to identify which strain in the database has the closest match to the sequenced isolate.

Method 2: Reads2Type and cgFind

Reads2Type [8] is also a k -mer-based phylogeny tool, but it relies on a lightweight, prebuilt database of 55-mers from a set of reference strains. This allows the analysis to be performed in the web browser, and it does not require the user to upload complete read files, allowing it to perform well when either speed or network access is limited. It works by taking one read at a time from the input file, generating 55-mers, and comparing to a prebuilt database. If there is not enough resolution information to identify the isolate by that read alone, additional reads are processed until a single taxonomy identification is achieved. This method works best on trimmed reads. Instructions and the webserver can be found at <https://cge.cbs.dtu.dk/services/Reads2Type/>

Given the genus and species from Reads2Type, users can make use of cgFind, a web tool we developed to provide easy access to downloadable genomes based on the complete prokaryotic genomes found in NCBI. The tool can be found at <https://nickp60.github.io/cgfind>.

Making the artificial chromosome

The artificial chromosome used for testing was constructed using the `makeToyGenome.sh` script included in the GitHub repository (<https://github.com/nickp60/riboSeed>) under the `scripts` directory. Briefly, the 7 rDNA regions from the *E. coli Sakai* genome were extracted with 5kb flanking sequence upstream and downstream; these sequences were then concatenated end to end to form a single, ~100kb sequence containing the 7 rDNAs as well as their flanking context.

Effect of reference sequence identity on riboSeed performance

The following range of substitutions were introduced into a artificial genome using the `runDegenerate.sh` script (included in the GitHub repository under the `scripts` directory), which facilitates the following procedure: 0.0, 0.0025, 0.0050, 0.0075, 0.0100, 0.0150, 0.0200, 0.0250, 0.0500, 0.0750, 0.1000, 0.1250, 0.1500, 0.1750, 0.2000, 0.2250,

0.2500, 0.2750, 0.3000. An artificial test genome is constructed (see above), and reads simulated using pIRS (100bp, 300bp inserts, stdev 10, 30-fold coverage, built-in error profile). Then, for each of a range of substitution frequencies, substitutions are introduced into the simulated genome, either just in the flanking regions or uniformly throughout. riboSeed is run on the reads using the mutated genome as the reference, and the results are evaluated with riboScore. This script was run 100 times, using a different random seed each time. As pseudo random number generation may differ between operating systems, comparable but not identical results can be expected.

Performance Across Prokaryotic Phyla

Performance on Archaeal Data

We assessed the effectiveness of riboSeed in assembling archaeal genomes. Most (~55%) archaeal genomes have only a single rDNA, and none has been observed to have more than four. As riboSeed requires a sequencing dataset and a reference genome, our ability to benchmark was limited; of the 104 entries in *rrnDB* with multiple rDNAs, only 7 had multiple entries at the species level. Among those, only 2 had publicly available short read data. We used riboSeed to re-assemble *Methanosa*cina *barkeri* *Fusaro DSMZ804* (SRR2064286) and *Methanobacterium formicicum* *st. JCM10132* (DRR017790). *Methanosa*cina *barkeri* *Fusaro DSMZ804* and *Methanobacterium formicicum* *st. BRM9* were the only isolates that were suitable for riboSeed, in that there was publicly available short read data, more than a single rDNA operon, and an appropriate complete reference genome at the species level. Results are shown in Table S1.

*Methanosa*cina *barkeri* *Fusaro DSMZ804* was sequenced using an Illumina HiSeq2000 with 101bp paired-end reads, with an average fragment length of 400bp. We downsampled to use 5% of the 19.4Gbp dataset with seqtk (<https://github.com/lh3/seqtk>). *Methanosa*cina *barkeri* *str. Wiesmoor* (CP009526.1) was used as a reference. The resulting riboSeed assembly showed correct assembly of 3 of 3 rDNAs, while *de novo* assemble failed to resolve any.

M. formicicum *st. JCM10132* was sequenced on an Ion Torrent PGM, generating 106.5Mbp of 89bp single-end reads. *M. formicicum* *BRM9* (CP006933.1) was used as a reference. While riboSeed with default parameters did not resolve any of the assembly gaps (final assembly *k*-mers 21, 33, 55, and 77), re-running the final assembly with *k*-mers of 21, 33, 55, 77, and 99 resulted in closing 2 of 2 rDNA gaps. We are unsure why the addition of 99-mers improved assembly with 89-bp reads, but we are actively investigating this. This represents the first application of riboSeed to Ion Torrent data.

Taken together, we show that given appropriate datasets and parameters, archaeal datasets can be processed in the same manner used for bacteria.

Performance Across Bacterial Phyla

In order to assess riboSeed's wider applicability, we selected additional datasets representing major bacterial phyla for those not already present in our analysis. In all cases, riboSeed improved the assemblies compared to the *de novo* with no missassemblies introduced (Table S1). Thus, riboSeed can be applied to a wide range of organisms.

Table S1: Comparison of *de novo* and riboSeed's *de fere novo* assemblies

Phylum	Class	Sequenced Strain Name	SRA	Reference Strain Name	Accession	<i>de novo</i>		<i>de fere novo</i>	
						rdDNAs	✓ –	✓ –	✓ –
5	Actinobacteria	Actinobacteria	SRR4271515	241	NC_016782.1	5	5	3	2
	Chlamydiae	<i>Corynebacterium diphtheriae</i> NCTC 13129	SRR5942978	434/Bu	NC_010287.1	2	0	2	0
	Firmicutes	<i>Chlamydia trachomatis</i> Population 1	SRR5942978	630	AM180355.1	11	0	9	2
	Proteobacteria	<i>Clostridioides difficile</i> C00005970	ERR251735	ATCC25416	NZ_CP012981.1	6	6	3	3
	Proteobacteria	<i>Burkholderia cepacia</i> DHQP2016-12-119	SRR6334321	DK_1622	NC_008095.1	4	0	4	0
	Proteobacteria	<i>Mycococcus xanthus</i> DSM 16526	SRR4236978	DRR090193	ATCC BAA-847	NC_020555.1	3	0	0
	Tenericutes	<i>Helicobacter cinaedi</i> MRY12-0051	ERR1938252	ATCC 23114	NC_013511.1	2	0	2	0
	Euryarchaeota	<i>Mycoplasma hominis</i> Australia	DRR017790	BRM9	CP006933.1	3	0	3	0
	Euryarchaeota	<i>Methanobacterium formicicum</i> JCM110132	SRR2064286	Wiesmoor	CP009526.1	2	0	2	0
	Methanomicrobia	<i>Methanosaarcina barkeri</i> Fusaro DSMZ804							

✓ correct assembly; – unassembled; × incorrect assembly

```

riboSeed (reference, riboSelect_clusters, reads, iters,
flanking_width)
  ref = reference;
  clusters = parse riboSelect_clusters;
  region = clusters + flanking_width;
  for i in iters do
    map reads to ref;
    for cluster in clusters do
      filter and extract reads region;
      subassemble;
      return pseudocontig;
    end
    assess subassembly;
    if success then
      make pseudogenome from pseudocontigs ;
      ref = pseudogenome ;
    end
  end
  run assembler with reads and pseudocontigs;
end

```

Figure S1: Pseudocode of riboSeed algorithm

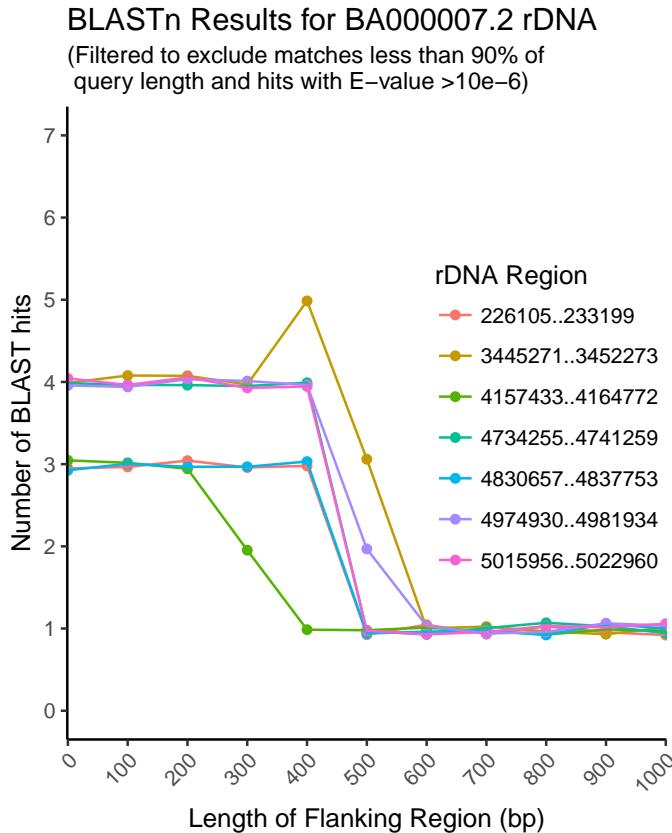


Figure S2: BLASTn was used to perform *in silico* DNA-DNA hybridization of all rDNA regions from *E. coli* Sakai with variable flanking lengths. The number of hits is a proxy for occurrences in the genome; increasing the flanking length increases the specificity. (Points are jittered to aid visibility for overlapping values.)

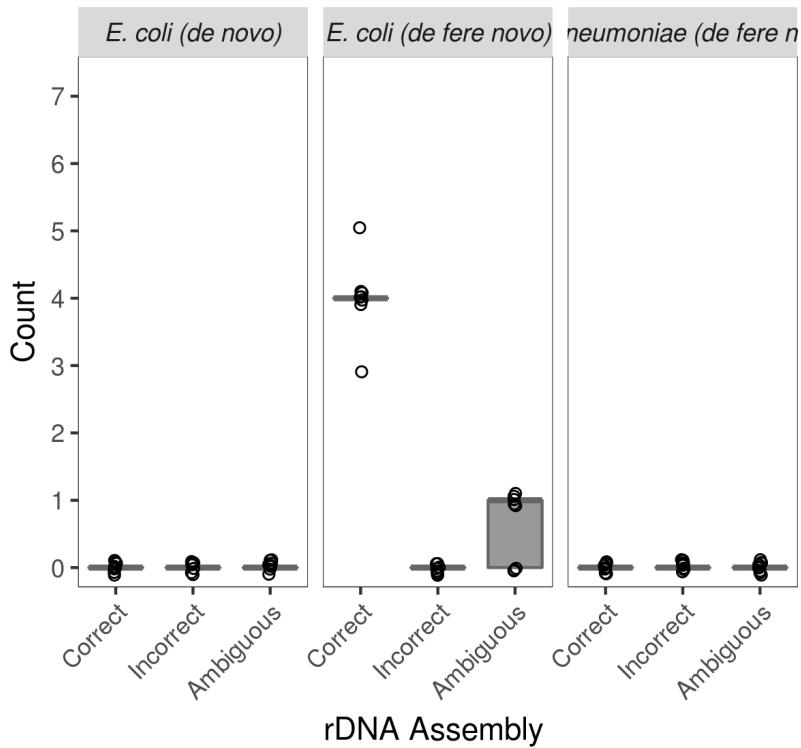
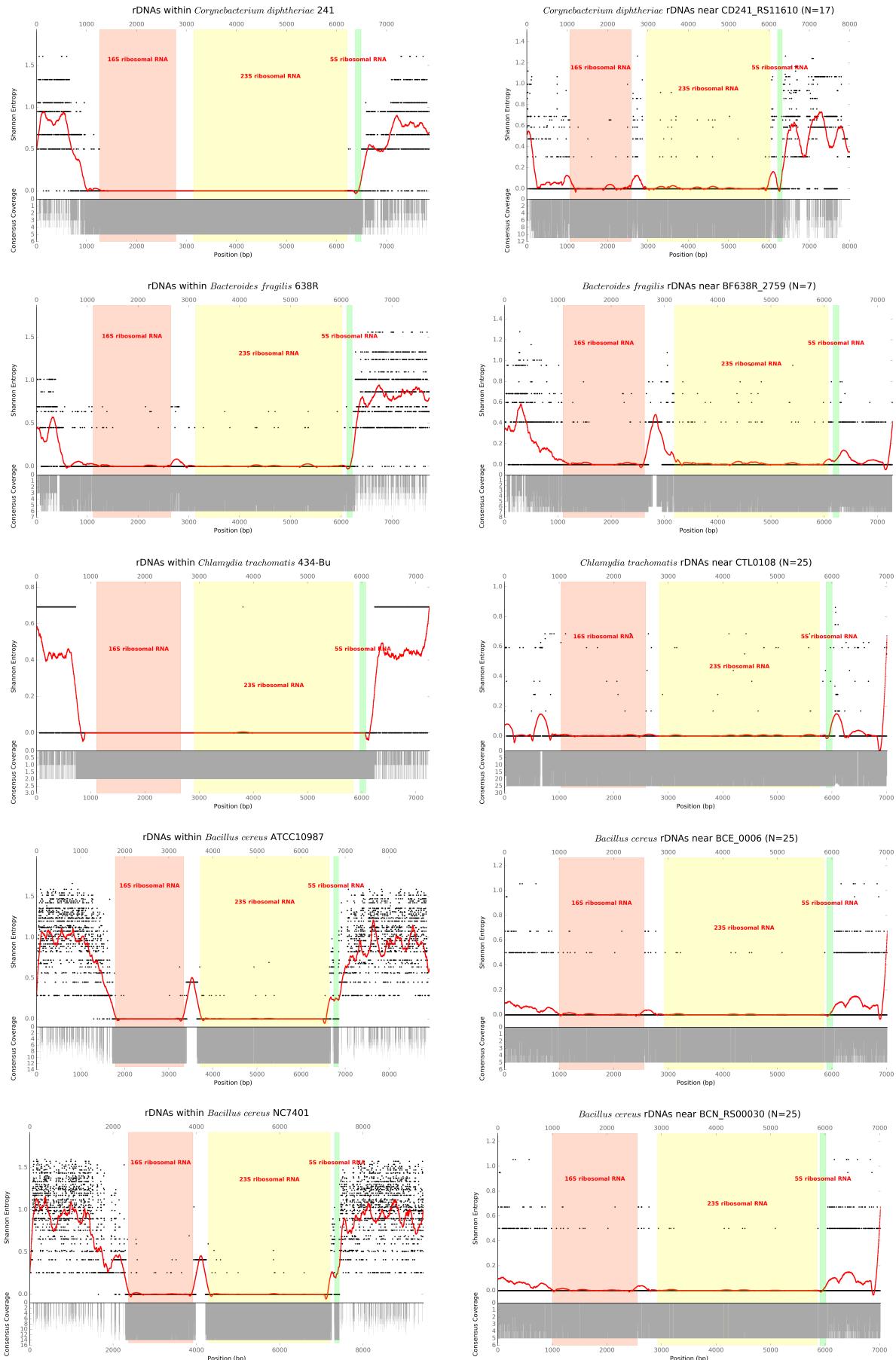
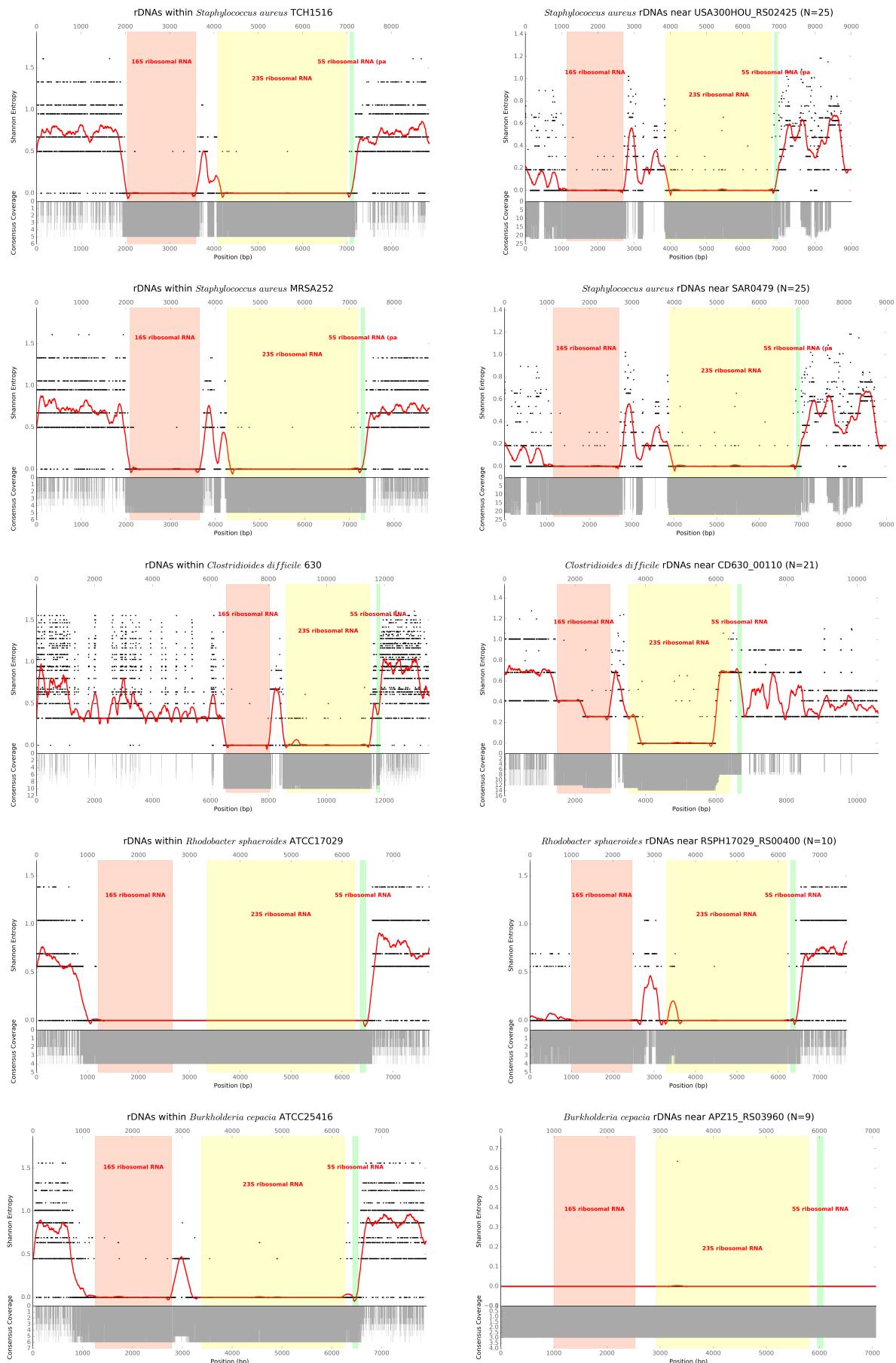
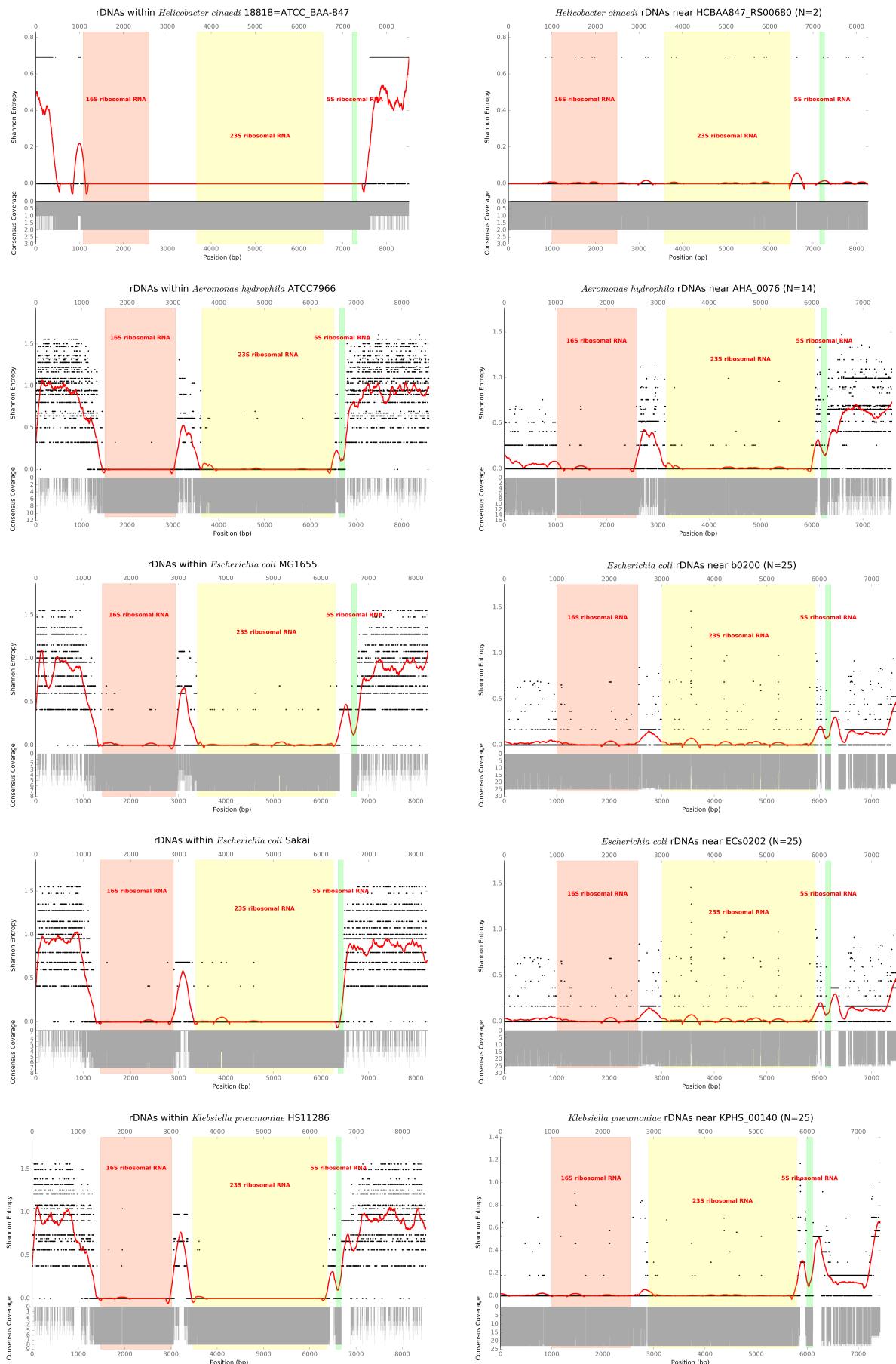
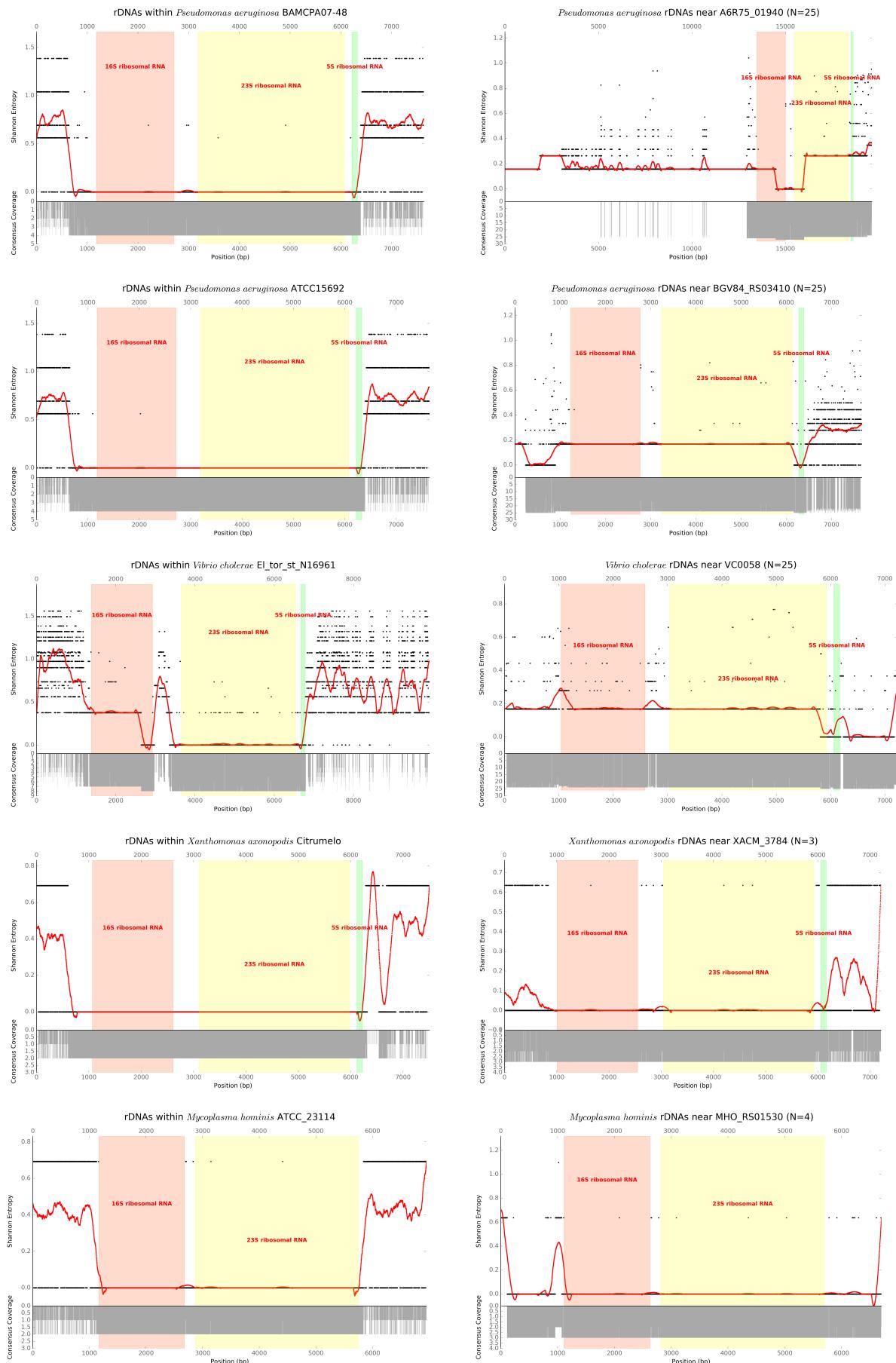


Figure S3: Assembly of artificial genome. *De fere novo* results in closure of 3-5 rDNAs with the correct reference; only 1-2 rDNAs are correctly assembled using *K. pneumoniae*. No rDNAs are assembled with *de novo* assembly. Scored with riboScore.py. N=8.









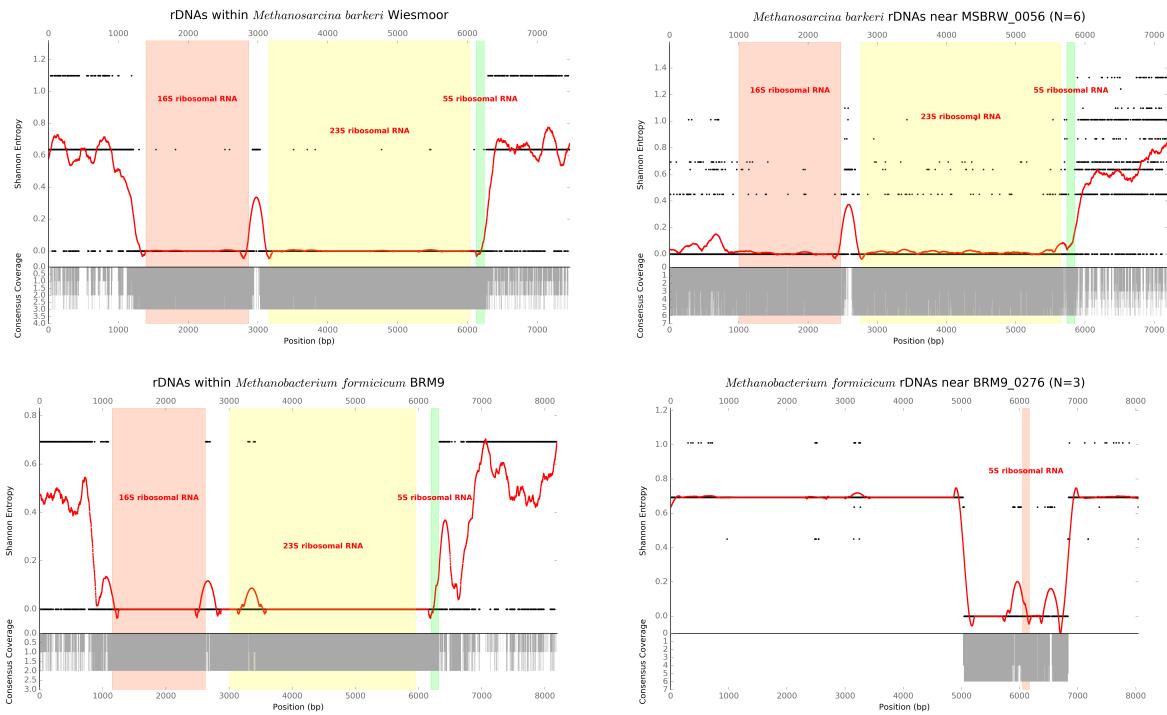


Figure S4: riboScan.py, riboSelect.py, and riboSnag.py were run on all the genomes used as references for *de novo* assemblies. Consensus alignment depth (grey bars) and Shannon entropy (black points, smoothed entropy as red line) for aligned rDNA regions. Similar to Figure 3 in the main text, for each genome, a gene neighboring the first rDNA operon was identified, and used to extract homologous rDNA operons from up to 25 other isolates at the species level. In most cases, the entropy is lower in homologous rDNAs than across all the rDNAs in a given genome.

Atypical rDNA operon structure

Ribosomal RNA coding regions are commonly arranged into operons consisting of a 16S rRNA, 23S rRNA, and one or more 5S rRNAs, often with various tRNAs interspersed. In the course of this study, we observed some taxa lacking this typical 16S–23S–5S rRNA operon. When rDNAs are not structured into operons, assemblies from short reads do not suffer from the issue of long repeats. We developed a module called **structure** for plotting rDNAs across a collection of genomes; this is available for riboSeed as of version 0.4.50. Figure S5 show the operon arrangement of a few examples of organisms exhibiting atypical operon structure. For comparison, the rDNAs in the reference strains used in this study are shown in Figure S6. Work to characterize this anomaly is underway in our lab.

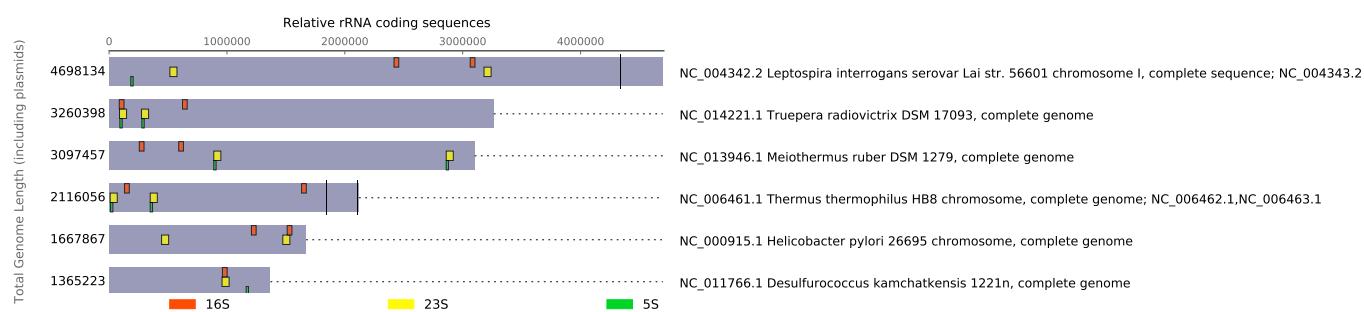


Figure S5: Atypical rDNA operon structure in select taxa. rRNA lengths are not shown to scale. Note that the NCBI record for *Helicobacter pylori* (NC_000915.1) shows a 5S rRNA not detected by Barrnap.

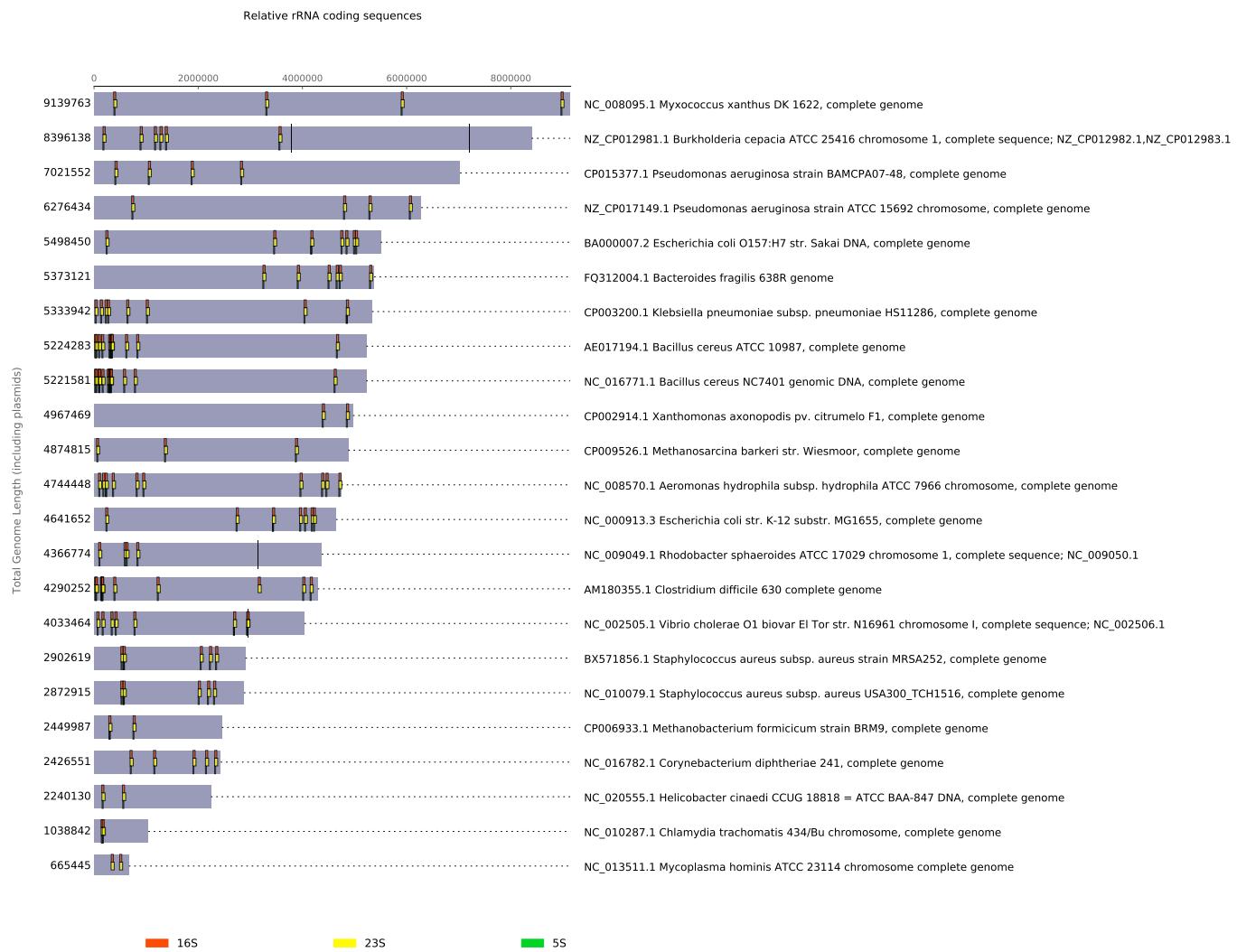


Figure S6: Typical rDNA structure for strains used in this study.

Key Parameters

--ref_as_contig

The assembly that results from including riboSeed’s “long reads” is sensitive to the manner in which they are incorporated into the *de novo* assembly. Here, for our analyses, we used the SPAdes assembler [1], as it has built-in ways to include contigs (using the “–trusted-contigs” or “–untrusted-contigs”) in FASTA format. Other assemblers could be used, but most require long reads to have a quality score associate with them, preventing direct use of riboSeeds long reads.

As mentioned in the Methods section, riboSeed uses the reference rDNA region in the initial subassembly; in subsequent subassemblies, the longest contig of the previous subassembly is used. These regions can be treated one of four ways using the `--ref_as_contig` argument: `trusted`, `untrusted`, `infer`, or `ignore`. Additionally, if the user is worried that the reference rDNA will too heavily influence the initial subassembly, they can enable the `--initial_consensus` flag to use a mapping consensus assembly instead of the de Bruijn graph based assembly from SPAdes.

The default manner in which rDNA regions (either from the reference or from the previous iteration’s subassembly) behaviour is to infer (`--ref_as_contig infer`): if the percent of reads mapping to the (whole) reference sequence is over 80%, than the rDNA region will be included as a trusted contig. If below 80%, the reads will be treated as untrusted.

If a user wishes to have the subassemblies only using the reads (true *de novo* assembly), they can use the `ignore` option. We only recommend this with very close references.

Further, if the user wishes to explicitly define the behaviour, `trusted` or `untrusted` can be provided to the `--ref_as_contig` argument.

--score_min

By default, the accepted alignment score for BWA mapping is $\frac{1}{2}$ the read length. If needed, this can be increased for greater stringency when dealing with more divergent references, or decreased to include more reads, which may be advantageous when assembling a low coverage dataset.

Table S2: Hits resulting from searching the SRA database for various sequencing technologies as of January, 2017

Search term	Hits	Percentage
illumina	2242225	94.27
pacbio	21131	0.89
ion	30560	1.28
roche	42445	1.78
oxford	12301	0.52
solid	29791	1.25
Total	2378453	100

Table S3: Accessions for 25 *E. coli* genomes used to calculate substitution rate

GCA_000005845.2_ASM584v2
GCA_000019385.1_ASM1938v1
GCA_000026245.1_ASM2624v1
GCA_000026345.1_ASM2634v1
GCA_000026545.1_ASM2654v1
GCA_000146735.1_ASM14673v1
GCA_000257275.1_ASM25727v1
GCA_000520055.1_ASM52005v1
GCA_000732965.1_ASM73296v1
GCA_001007915.1_ASM100791v1
GCA_001442495.1_ASM144249v1
GCA_001469815.1_ASM146981v1
GCA_001660565.1_ASM166056v1
GCA_001660585.1_ASM166058v1
GCA_001753565.1_ASM175356v1
GCA_001888075.1_ASM188807v1
GCA_001901025.1_ASM190102v1
GCA_001936315.1_ASM193631v1
GCA_002056065.1_ASM205606v1
GCA_002078295.1_ASM207829v1
GCA_002156825.1_ASM215682v1
GCA_002163935.1_ASM216393v1
GCA_002192275.1_ASM219227v1
GCA_002220265.1_ASM222026v1
GCA_900096795.1_Ecoli_AG100_Sample3_Doxycycline_Assembly

All available at <ftp://ftp.ncbi.nlm.nih.gov/genomes/all/GCA/>

Table S4: Strain names and accessions for reference genomes used in this study. For the full list, which includes strains used in the supplementary data, SRA accession numbers for reads, and more, please consult the supplementary file “strain_metadata.tab”

Strain Name	Accession
<i>E. coli MG1655</i>	NC_000913.3
<i>E. coli Sakai</i>	BA000007.2
<i>A. hydrophila ATCC 7966</i>	NC_008570.1
<i>B. cereus ATCC 10987</i>	AE017194.1
<i>B. cereus NC7401</i>	NC_016771.1
<i>B. fragilis 638R</i>	FQ312004.1
<i>K. pneumoniae</i>	CP003200.1
<i>R. sphaeroides ATCC 17029</i>	NC_009049.1, NC_009050.1
<i>S. aureus TCH1516</i>	NC_010079.1
<i>S. aureus MRSA252</i>	BX571856.1
<i>V. cholerae El Tor str. N16961</i>	NC_002505.1, NC_002506.1
<i>X. axonopodis pv. Citrumelo</i>	CP002914.1
<i>P. aeruginosa BAMCPA07-48</i>	CP015377.1
<i>P. aeruginosa ATCC 15692</i>	NZ_CP017149.1

Table S5: Software Versions

Tool	Version
Mauve	2015-02-13 build 0
BLAST+	2.2.28+
Barrnap	0.8
BWA	0.7.8-r455
samtools	1.4.1
MAFFT	v7.310
SPAdes	v3.9.0
QUAST	4.4
bedtools	2.17.0
EMBOSS	6.5.7
pIRS	2.0.2
seqtk	1.2-r94
Parsnp	v1.2

Table S6: QUAST [3] results of *P. aeruginosa* BAMCPA07-48 assemblies comparing *de fere novo* assembly, *de novo* assembly, and reference-based assembly (where the *P. aeruginosa* ATCC 15692 reference is included in the *de novo* assembly as a trusted contig). Blue and red highlight the best and worst results, respectively. riboSeed’s *de fere novo* assembly either outperforms or performs comparably to *de novo* assembly in all categories. Using the reference as a trusted contig results in longer assemblies but with a much higher rate of mismatches, indels, and misassemblies.

	<i>de fere novo</i>	<i>de novo</i>	reference-based
Genome fraction (%)	98.106	97.868	98
Duplication ratio	1.001	1.001	1.017
Largest alignment	630503	402463	757685
Total aligned length	6893293	6876715	6993532
NGA50	176510	176510	135376
LGA50	12	13	14
# misassemblies	2	2	9
Misassembled contigs length	212498	212498	2347560
# mismatches per 100 kbp	1.89	1.69	11.66
# indels per 100 kbp	2.48	2.44	2.94
# N’s per 100 kbp	0	0	0
# contigs	154	159	388
Largest contig	630503	402463	1103106
Total length	6893293	6876715	7237564
Total length (>= 1000 bp)	6865091	6848513	7130244
Total length (>= 10000 bp)	6687664	6663031	6617370
Total length (>= 50000 bp)	6242010	6168232	5534330

Table S7: Assembling *S. aureus* UAMS-1 with BugBuilder. We did not have access to critical information about the pipeline parameters used in the original assembly; this prevented exact recapitulation of the published results. Therefore, we approximated the settings based on notes from the publication. The performance of Pilon [9], GapFiller [2], or no finishing software was assessed with both the *de fere novo* and *de novo* assemblies. rDNA counts were visually determined using Mauve; all other metrics were generated with QUAST, using the scaffolds from the assemblies and the *S. aureus* MRSA252 reference. Misassembly/mismatch stats were removed, as the reference and sequenced strain have an average nucleotide identity of 97.62%, and the misassemblies cannot be differentiated from strain differences with QUAST. Blue and red highlight the best and worst results, respectively.

rDNAs	<i>de fere novo</i>			<i>de novo</i>		
	GapFiller	Pilon		Gapfiller	Pilon	
		–	–		–	–
# contigs (≥ 0 bp)	3	3	3	0	0	0
Total length	2773352	2781986	2763179	2768273	2770267	2752929
Reference length	2902619	2902619	2902619	2902619	2902619	2902619
GC (%)	32.78	32.80	32.78	32.72	32.73	32.71
Reference GC (%)	32.81	32.81	32.81	32.81	32.81	32.81
Unaligned length	20087	15391	19075	18980	16819	20141
Genome fraction (%)	94.736	94.941	94.508	94.487	94.604	94.105
Duplication ratio	1.001	1.004	1.000	1.002	1.003	1.000
# N's per 100 kbp	151.44	58.59	160.65	184.23	111.00	200.73
Largest alignment	403935	469521	403508	459120	459623	403508
Total aligned length	2753265	2766595	2744104	2749293	2753448	2732788
NA50	223096	195659	222384	205820	164646	222384
NGA50	176550	177158	222384	157488	164646	176039
LA50	5	5	5	5	6	5
LGA50	6	6	5	6	6	6

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