

The enveomics collection: a toolbox for specialized analyses of microbial genomes and metagenomes

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Genomic and metagenomic analyses are increasingly becoming commonplace in several areas of biological research, but recurrent specialized analyses are frequently reported as in-house scripts rarely available after publication. We describe the enveomics collection, a growing set of actively maintained scripts for several recurrent and specialized tasks in microbial genomics and metagenomics, and present a graphical user interface and several case studies. Our resource includes previously described as well as new algorithms such as Transformed-space Resampling In Biased Sets (TRIBS), a novel method to evaluate phylogenetic under- or over-dispersion in reference sets with strong phylogenetic bias. The enveomics collection is freely available under the terms of the Artistic License 2.0 at <https://github.com/lmrodriguezr/enveomics> and for online analysis at <http://enve-omics.ce.gatech.edu> .

1 The enveomics collection: a toolbox for specialized analyses of 2 microbial genomes and metagenomes

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9 ABSTRACT

10 Genomic and metagenomic analyses are increasingly becoming commonplace in several areas of
11 biological research, but recurrent specialized analyses are frequently reported as in-house scripts rarely
12 available after publication. We describe the enveomics collection, a growing set of actively maintained
13 scripts and bioinformatics algorithms for several recurrent and specialized tasks in microbial genomics
14 and metagenomics, and present a graphical user interface and several case studies. Our resource
15 includes previously described (e.g., Nonpareil, MyTaxa, ANI/AAI calculator) as well as new algorithms
16 such as Transformed-space Resampling In Biased Sets (TRIBS), a novel method to evaluate
17 phylogenetic under- or over-dispersion in reference sets with strong phylogenetic bias. The enveomics
18 collection is freely available under the terms of the Artistic License 2.0 at
19 <https://github.com/lmrdriguezr/enveomics> and for online analysis at <http://enve-omics.ce.gatech.edu>.

20 INTRODUCTION

21 Microbial genomics and metagenomics have become key components of several areas of modern
22 research including biomedicine, epidemiology, plant and animal pathology, environmental engineering
23 and science, microbial ecology, and evolutionary biology. Specialized computational analyses in these
24 areas are hence becoming commonplace for the non-expert, often resulting in the reimplementations of
25 scripts critical for understanding and reproducing the reported results, with varying levels of quality,
26 reproducibility, and availability. While the literature analysing *ad hoc* scripts is scarce, a 2004 survey on
27 the availability of URLs reported in MEDLINE found that 19% of 1,020 analysed pages were always
28 unavailable, and only 63% were always available (Wren, 2004). Moreover, we searched the manuscripts
29 available as full-text in PubMed Central with the terms “in-house script”, “in-house developed script”, “in-
30 house perl script” (other languages didn’t return additional results), or the same terms in plural, and found
31 1,929 matching articles (as of January 05, 2016). From these, 1,654 were related to genomics or
32 metagenomics and 449 to microbial genomics or metagenomics. We further explored the latter set of
33 manuscripts, and found that only 6% provided access to the source code (26/449), with an additional 1%
34 reporting websites no longer available (3/449) or not including the reported scripts (3/449). 3% of the
35 manuscripts explicitly indicated that the scripts were available upon request (13/449), but in only 3 cases
36 the authors provided the code within two months of the request. The large majority (90%) did not provide

37 any reference, provided references to previous publications of the same group not including the source
38 code, or referenced only the programming language in which the scripts were implemented. While this
39 survey is not a systematic analysis of availability of in-house scripts, nor does it provide quality
40 assessments, the results do underscore the prevalence of a phenomenon that undermines reproducibility
41 in studies applying microbial genomics or metagenomics. On one hand, individual tools are the basis for
42 complete and reproducible methods that are reported either in manuscripts, white papers, or standard
43 operating procedures (SOPs), but the abovementioned statistics showed that the tools infrequently
44 become available. On the other hand, a suitable alternative to providing developed tools for results
45 reproducibility is to make data on each step of the analyses publicly available, but this approach is also
46 rarely adopted, with the added issues of larger file sizes and heterogeneity, making the documentation of
47 data even more challenging than documenting code. Here, we present a growing collection of actively
48 maintained scripts for several recurrent and specialized tasks in microbial genomics and metagenomics,
49 together with comprehensive documentation, a graphical user interface, and some cases of use. Our
50 collection may also constitute a reference example for other researchers in the future, and an actively
51 maintained framework that could be collaboratively expanded.

52 **IMPLEMENTATION**

53 The enveomics collection is a multi-language set of over 70 independent scripts that accomplish
54 specialized tasks in genomics and metagenomics, including code in Ruby, Perl, AWK, Bash, and R. In
55 addition, the collection features reusable libraries that automate recurrent sub-tasks. For example, the
56 enveomics_rb Ruby library includes object-oriented representations of trees, read-placement results, sets
57 of orthologous genes, and complex (non-contiguous) sequence coordinates, together with methods for
58 accessing and downloading remote data. The R code is packaged into a single library (enveomics.R) to
59 simplify its distribution.

60 **Preferred file formats**

61 Format incompatibility between data sources and analysis tools is a common problem in bioinformatics,
62 and there are several tools and libraries dedicated to the translation between format specifications (Rice,
63 Longden & Bleasby, 2000). In order to mitigate the impact of this problem, the enveomics collection has
64 been designed to support only a reduced number of formats, with a wide range of alternative variations.
65 For example, sequence files are expected to be in FastA, but the scripts in the collection always support
66 multi-FastA and can parse variations of the definition lines and colon-lead comments (Suppl. Table S1).
67 Supported formats for other data types include tabular BLAST for similarity searches (including variations
68 with additional columns and comments), JPlace for phylogenetic read placement (Matsen et al., 2012),
69 and tables in raw text with tab-delimited columns.

70 **Access to remote servers**

71 Local data sources are often insufficient and commonly out-dated. In response, we have implemented
72 several utilities to simplify the automated access to remote databases using the Representational State
73 Transfer Application Program Interfaces (RESTful APIs) of the European Bioinformatics Institute of the
74 European Molecular Biology Laboratory (EMBL-EBI), the U.S. National Center for Biotechnology
75 Information (NCBI) E-Utilities, the Kyoto Encyclopedia of Genes and Genomes (KEGG), and the M5nr
76 (Kanehisa & Goto, 2000; Sayers et al., 2009; Wilke et al., 2012; Li et al., 2015). All the scripts using these
77 modules are categorized in Annotation/database mapping, and include additional documentation such as
78 informing the user that third-party software or database is used and thus, the latter resources should be
79 cited appropriately in any resulting publications.

80 **Enveomics-GUI**

81 The documentation and parameter descriptions for all the scripts are standardized into a set of JSON files
82 that allow the dynamic creation of Graphical User Interface (GUI) forms through the enveomics-GUI
83 package, including a set of examples and reference files (Fig. 1). The package is a collection of Ruby
84 libraries, including EnveGUI that implements graphical user interaction with Shoes 4
85 (<https://github.com/shoes/shoes4>). The JSON files meet the definitions of the ECMA-404 standard (Ecma
86 International, 2013), but their processing (implemented in the EnveJSON library) ignores object entries
87 with “_” key, that are utilized for comments, and implements external file inclusion using the object entries
88 with “_include” key. The package is distributed as source code (requires Shoes 4 and JRuby), as a stand-
89 alone OS-independent Java Archive (JAR), and as a bundled Mac OS X application.

90 **RESULTS**

91 **Reimplementations and novel algorithms**

92 The enveomics collection aims to simplify the use of novel and previously described algorithms for the
93 analysis of community (e.g., Chao1.pl, AlphaDiversi-ty.pl, Newick.autoprune.R) and population diversity
94 (e.g., BlastTab.recplot2.R, RecPlot2.find_peaks.R, CharTable.classify.rb), among other tasks in microbial
95 genomics and metagenomics. Here we describe representative modules (see also Suppl. Table S1)
96 including algorithms developed by our group.

97 *Reciprocal Best Match and Average Sequence Identity.* The detection of Reciprocal Best Matches (RBMs)
98 is a reliable method for the identification of orthology (Wolf & Koonin, 2012) that has been widely used in
99 genome-aggregate metrics of genetic relatedness (Konstantinidis & Tiedje, 2005; Goris et al., 2007).
100 Although phylogenetic reconstruction remains the gold standard for orthology detection, RBM provides a
101 fast alternative for high-throughput analyses such as genome-wide scanning. The enveomics collection
102 contains utilities for the detection of RBMs (rbm.rb) and the compilation of Orthology Groups (OGs;
103 ogs.mcl.rb), as well as the estimation of Average Nucleotide Identity (ANI; ani.rb; generally suitable for

104 comparisons of genomes assigned to the same genus) and Average Amino acid Identity (AAI; `aai.rb`;
105 suitable for comparisons of genomes assigned to different species).

106 *Transformed-space Resampling In Biased Sets (TRIBS)*. Environmental analyses often rely on pre-
107 existing reference databases as a proxy to the presence of features in query datasets. However,
108 databases seldom represent the source of the query sets uniformly, introducing sampling biases. TRIBS
109 is a novel algorithm that reduces the impact of biased sampling by uniformly resampling reference objects
110 in a transformed space generated by Multidimensional Scaling (MDS). This enables the testing of
111 differences between a dataset and a given subset for the detection of under- or over-dispersion of traits
112 (`TRIBS.test.R`, `TRIBS.plot-test.R`). The method was originally designed for the detection of phylogenetic
113 under-dispersion of traits in groups of genomes with strong phylogenetic bias (Suppl. Fig. S1;
114 `TRIBS.test.R`).

115 *Automated pruning of phylogenetic trees*. The *enveomics* collection also features a utility to automatically
116 prune trees keeping clade representatives (`Newick.autoprune.R`), a useful tool for the navigation of large
117 trees such as those produced from 16S rRNA gene databases. This script iteratively extracts the
118 cophenetic matrix from a tree and removes terminal nodes with at least one other node closer than a
119 target minimum distance (by default, the first quartile of all the paired distances in the initial tree). In some
120 cases, the complete cophenetic matrix is prohibitively expensive to estimate (in the initial iterations for
121 large trees); in those cases the script takes a random sample of terminal nodes and removes sister nodes
122 (or their children) closer than the target distance. An example of a pruned tree is presented in Fig. 2B-C.

123 **Case studies using the *enveomics* collection**

124 *Core genome phylogenies*. Whole-genome phylogenetic reconstruction is a powerful method for the
125 resolution of evolutionary relationships. The *enveomics* collection includes utilities to download genomes
126 of a given species, detect RBMs between pairs of genomes, identify OGs, and identify the genes shared
127 among all the genomes in the collection –the core genome– (`RefSeq.download.bash`, `rbm.rb`, `ogs.mcl.rb`,
128 `ogs.extract.rb`). After computing independent alignments of each core OG, a concatenated alignment can
129 be generated with the options of excluding invariable sites and keeping track of coordinates (`Aln.cat.rb`) to
130 generate robust phylogenies with OG-specific models. In addition, the OGs can be used to estimate
131 several gene-content properties (`ogs.stats.rb`) and the rarefied core and pan-genomes (`ogs.core-pan.rb`)
132 of the species. As a less expensive alternative to the entire core genome phylogeny, one could also
133 identify and analyse only the collection of 111 single-copy genes typically present in archaeal (often ~26
134 genes) and bacterial (often ~106 genes) genomes (`HMM.essential.rb`). We implemented a workflow using
135 the *enveomics* collection, together with the multiple alignment tool Clustal Omega (Sievers et al., 2011)
136 and the phylogenetic reconstruction tool RAXML (Stamatakis, 2014) and applied it to the 17 publicly
137 available complete genomes of *Xanthomonas oryzae* (Fig. 2). The resulting phylogeny identifies known
138 pathovars and the overall structure is consistent with a previous phylogenomic reconstruction (Rodriguez-

139 R et al., 2012). The complete analysis is fully automated, and the code is deposited in the enveomics
140 collection at Examples/essential-phylogeny.bash. The execution took 31.2 minutes using two 2.9 GHz
141 processors.

142 *Gene variants in a metagenome.* Characterizing the allelic diversity of genes in metagenomes allows
143 targeted analyses of specific traits and the exploration of population discreteness and intra-population
144 variations, independent of cultivation and amplification (Caro-Quintero & Konstantinidis, 2012; Rodriguez-
145 R & Konstantinidis, 2014a). We explored the intra-population diversity of a metagenomic-recovered bin
146 (LL-70.1) using the mapping of metagenomic reads (LL_1101B; SRR948448 (Tsementzi et al., 2014))
147 from a water sample in January 2011 at Lake Lanier (GA, USA). Read mapping was performed with
148 BLAST (Altschul et al., 1990), and results were analysed and visualized using BlastTab.catsbj.pl and
149 BlastTab.recplot.R (Fig. 3), revealing small gene-content variations (panels 2 and 4), but a large allelic
150 variation and the presence of closely related organisms at about 90% ANI (panels 1 and 3). However, a
151 clear genetic discontinuity exists separating this species, as evidenced by the gap around 95% identity, a
152 phenomenon further discussed in (Caro-Quintero & Konstantinidis, 2012; Rodriguez-R & Konstantinidis,
153 2014a). The enveomics collection also includes utilities for the normalization (BlastTab.topHits_sorted.rb,
154 BlastTab.sumPerHit.pl, BlastTab.seqdepth_ZIP.pl), characterization (Chao1.pl, AlphaDiversity.pl,
155 TRIBS.test.R), and visualization (Table.barplot.R, TRIBS.plot-test.R, BlastTab.recplot2.R) of reference
156 allele distributions in a metagenome using read mapping. Additionally, the allelic diversity of a particular
157 gene of interest can be explored beyond known variants using phylogenetic read placement (Matsen,
158 Kodner & Armbrust, 2010; Berger, Krompass & Stamatakis, 2011), that can be visualized in the
159 interactive Tree of Life (iTOL) (Letunic & Bork, 2007), and further explored to characterize distances to
160 known variants or ancestral nodes (JPlace.to_iTOL.rb, JPlace.distances.rb) as in (Rodriguez-R et al.,
161 2015).

162 **Availability**

163 The source code for all the scripts and additional documentation are deposited and maintained at
164 <https://github.com/lmrodriguezr/enveomics>. The enveomics-GUI is maintained at
165 <https://github.com/lmrodriguezr/enveomics-gui>. In addition, we have made available a server with online
166 interfaces for select tools at <http://enve-omics.ce.gatech.edu/>, including the ANI and AAI calculators, and
167 previously reported tools like Nonpareil (Rodriguez-R & Konstantinidis, 2014b), a tool to estimate the level
168 of coverage in metagenomic samples, and MyTaxa (Luo, Rodriguez-R & Konstantinidis, 2014), a
169 taxonomic classification tool for sequence fragments.

170 **DISCUSSION**

171 The enveomics collection offers a wide array of tools implementing specialized recurrent tasks in
172 microbial genomics and metagenomics and is aimed for users with or without expertise in bioinformatics.
173 The collection features (i) a web-based interface for select tools and the complete documentation of all

174 the tools, **(ii)** a comprehensive graphical user interface (GUI), **(iii)** a command-line interface (CLI) that
175 allows integration with development platforms and automation, and **(iv)** Ruby and R application interfaces
176 (API) for developers. In addition, the collection has a language-agnostic design, allowing the
177 implementation of different tools in the most convenient language depending on available libraries or
178 other considerations. To allow this heterogeneity, all the tools are integrated using a standardized JSON-
179 based documentation scheme, allowing the incorporation of additional tools into the collection for the
180 different interfaces. Finally, examples of input data and parameters are provided to encourage the quick
181 use of the tools without dauntingly extensive user manuals.

182 A few of the scripts in our collection, in particular those implementing the most simple tasks, are
183 overlapping with those developed by others (e.g., (Rice, Longden & Bleasby, 2000; Stajich et al., 2002;
184 Cock et al., 2009)). Our goal here was not to perform exhaustive comparisons to previously published
185 scripts. As explained above, these scripts were frequently not available for comparisons. Rather, the goal
186 was to put together a resource that offers easy-to-use tools for the non-bioinformatician and is
187 comprehensive with respect to recurrent tasks in microbiome research. As such, we hope that the
188 scientific community will find this resource useful, and will provide feedback on the scripts and algorithms,
189 and suggestions for further improvements.

190 **SUPPLEMENTARY DATA**

191 Supplementary data are available at NAR Online.

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273
274

Figure 1(on next page)

Screen captures of the enveomics GUI in Mac OS X.

(A) Initial (home) screen with search bar, listing all categories and subcategories, and highlighting a few selected and randomly picked scripts. (B) Complete list of scripts per category. (C) Task form for aai.rb pre-filled with an example. (D) Result of the aai.rb analysis. All screen captures correspond to v0.1.2. Future versions may differ.

A

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Home All tasks Examples Update Website About

Welcome to the Enveomics collection!

Sequence similarity search

Statistics 7 scripts

Manipulation 11 scripts

Execution 3 scripts

Sequence analyses

Statistics 4 scripts

The most popular

- ani.rb**
Calculates the Average Nucleotide Identity between two genomes.
- aal.rb**
Calculates the Average Amino acid Identity between two genomes.
- FastQ.split.pl**
Splits a FastQ file into several FastQ files interleaved with interposed sister reads using any even number of reads.
- Taxonomy.silva2ncbi.rb**
Re-formats Silva taxonomy into NCBI format.
- AlphaDiversity.pl**
Calculates various alpha diversity metrics on a set of sequences.

B

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Sequence similarity search

Statistics

- BlastPairwise.AAsubs.pl**
Counts the different AA substitutions in the best hit blast alignments, from a BLASTP pairwise format output (-outfmt 0 in BLAST+, -m 0 in legacy BLAST).
- BlastTab.advance.bash**
Calculates the percentage of a partial BLAST result. The value produced slightly subestimates the actual advance, due to un-flushed output and trailing queries that could be processed but generate no results.
- BlastTab.seqdepth.pl**
Estimates the sequencing depth of subject sequences.

subject sequences. The values reported by this script may differ from those reported by BlastTab.seqdepth.pl if this script uses the aligned length of the read while BlastTab.seqdepth.pl uses the read length of the subject sequence.

depth of subject sequences (genes or contigs) assuming a Zero-Inflated Negative Binomial distribution for non-covered positions. It uses the corrected method of moments.

Running aai.rb

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Test example:
Average Amino acid Identity (AAI) between Mycoplasma genitalium (Bacteria) and Nanoarchaeum equitans (Archaea).
Parameters highlighted in blue have been changed from the default.

aai.rb

Calculates the Average Amino acid Identity between two genomes.

If you use this script, also cite (Konstantinidis & Tiedje, 2005, J. Bacteriol. 187: 2653-2662).

See also: [ani.rb](#) [rbm.rb](#)

Hide optional parameters

Seq1*

FastA file containing the genome 1 (proteins).

Open file

Alternatively, you can supply the GI of a genome (nucleotides) with the format: `gi|1234567|refseq|NC_012345.1`

Seq2*

FastA file containing the genome 2.

Open file

Alternatively, you can supply the GI of a genome (nucleotides) with the format: `gi|1234567|refseq|NC_012345.1`

C

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aai.rb

Command:

```
ruby /Users/lmr3/.enveomics/enveomics-master/Scripts/aai.rb --seq1 /Users/lmr3/.enveomics/enveomics-master/Tests/Mgen_M2288.faa --seq2 /Users/lmr3/.enveomics/enveomics-master/Tests/Nequ_Kin4M.faa --
```

Start time: Sun Dec 27 17:32:29 2015
Log: /var/folders/jt/_hggq8d5sv7wssf7bgmqm0000gp/T/enveomics20151227-39351-1rpleb
 Temporal directory: /var/folders/jt/_hggq8d5sv7wssf7bgmqm0000gp/T/d20151227-39388-1cm7hhc.
 Creating databases.
 Reading FastA file: /Users/lmr3/.enveomics/enveomics-master/Tests/Mgen_M2288.faa
 File contains 493 sequences.
 Reading FastA file: /Users/lmr3/.enveomics/enveomics-master/Tests/Nequ_Kin4M.faa
 File contains 536 sequences.
 Running one-way comparisons.
 ! One-way AAI 1: 32.78% (SD: 7.50%), from 492 proteins.
 ! One-way AAI 2: 33.04% (SD: 7.89%), from 530 proteins.
 ! Two-way AAI : 31.40% (SD: 5.97%), from 178 proteins.

Running time: 36.723s.

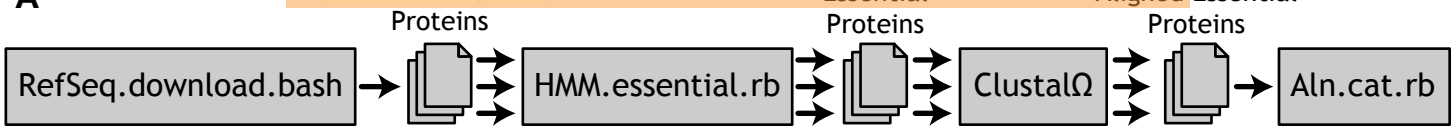
D

Figure 2(on next page)

Example of a complete workflow primarily using tools from the enveomics collection applied to *Xanthomonas oryzae* genomes.

(A) The workflow uses the enveomics collection, Clustal Omega, and RAxML, to generate a phylogenetic tree based on the concatenated alignment of 105 single-copy essential genes. (B) In the resulting phylogeny, two clades form naturally corresponding to the pathovars *oryzae* (*Xoo*, left) and *oryzicola* (*Xoc*, right). Note that the tree is un-rooted, but the rooting point is suggested (vertex) based on phylogenomics of the genus (Rodriguez-R et al., 2012). The invariable sites were removed using Aln.cat.rb (35,386 sites removed) and the phylogeny was reconstructed using the remaining 552 informative sites. From the 105 detected essential genes, 22 were identical across all genomes and were excluded from the analysis. (C) A simplified version of the tree was produced by automatically pruning terminal nodes at a distance lesser than 0.01, resulting in a tree with 7 genomes (out of 17) with similar structure.

A



B

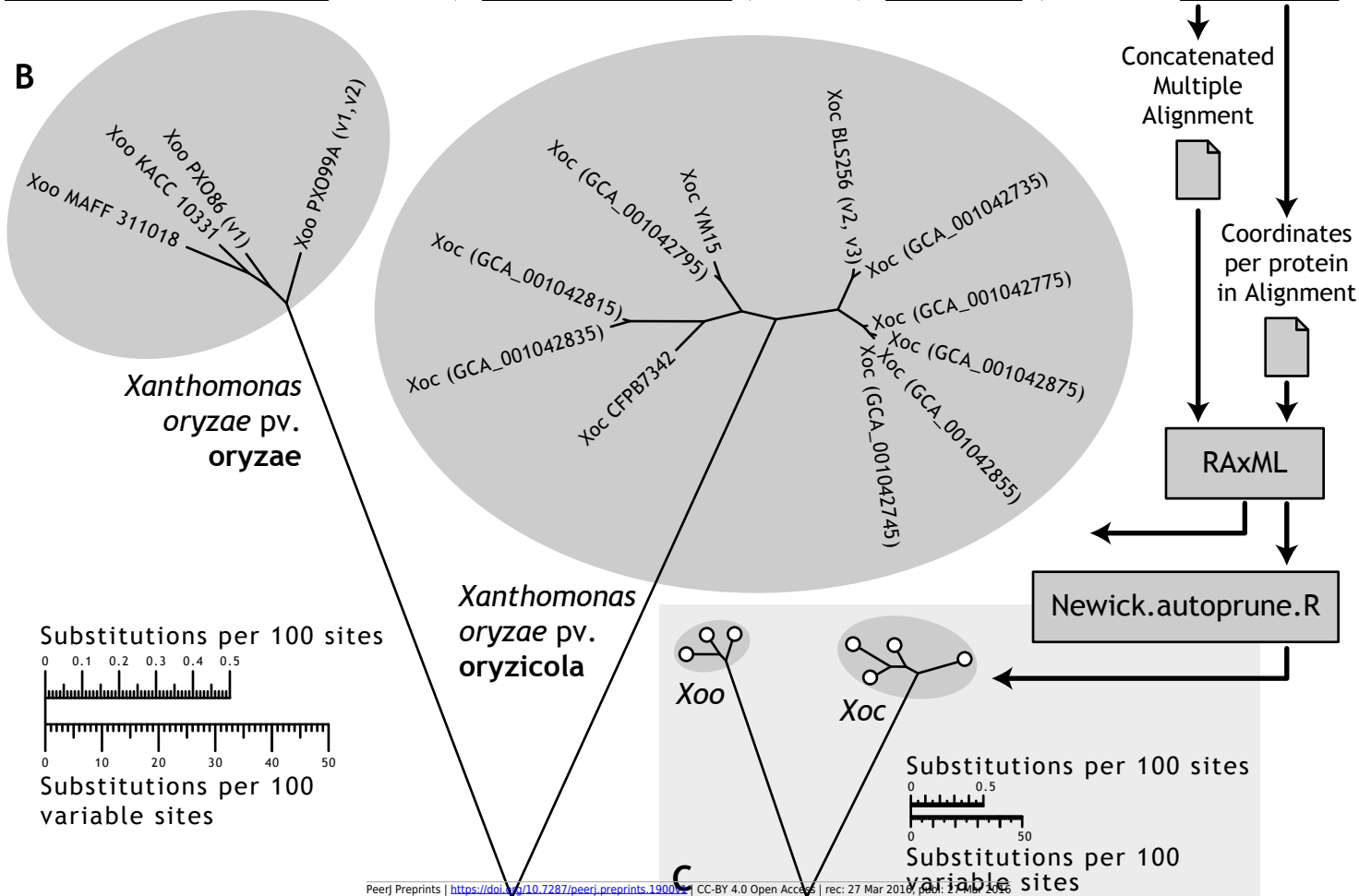


Figure 3(on next page)

Example of a fragment recruitment plot.

This figure showcases the result of processing a BLAST search of metagenomic short sequencing reads (150 bp long in this case; each matching read is represented by a dot in main panel 1) against a population genome sequence assembled/binning from the same metagenome (X-axis). The tabular BLAST result was parsed using BlastTab.catsbj.pl, and graphical representation was generated with the BlastTab.recplot2.R. The circled numbers 1 through 5 denote the distinct panels of the layout: **(1)** Main panel representing the reads recruited, placed by location (X-axis) and identity (Y-axis). **(2)** Sequencing depth across the reference, in logarithmic scale. Bars at the bottom represent regions without mapping reads (sequencing depth of zero). **(3)** Identity histogram of mapping reads (light gray) and smoothed spline (black), in logarithmic scale. **(4)** Sequencing depth histogram. Peaks from values above 95% identity are automatically identified as skewed normal distributions (red), with centrality measures, percentage of the reference length, and fit error (bottom-right legend) reported for each peak (marked in the right edge). **(5)** Color scale for the number of stacked reads per 2-dimensional bin in panel 1. The background of panels 1 and 3, and the line colors in panels 2 and 4, correspond to matches with identity above (dark blue) and below (light blue) a user-defined cutoff. By default, the identity cutoff is set to 95%, corresponding to the species boundary (Konstantinidis & Tiedje, 2005). See also (Rodriguez-R & Konstantinidis, 2014a) for additional discussion.

