

Metagenomics:

Contig Binning, Taxonomic Assignment & Validation

STAMPS – Day 7

July 25, 2024

Michael Nute & Todd Treangen

mike.nute@gmail.com

mn56@rice.edu

Approximate Agenda

9:00 to 9:15	(15 min.)	Intro/kickoff (Todd)
9:15 to 9:25	(10 min.)	Advertisement: Emu (Mike, 6 slides)
9:25 to 9:50	(25 min.)	Binning lecture (Mike, 17 slides)
9:50 to 10:10	(20 min.)	Binning tutorial (using contigs from monday)
10:10 to 10:35	(25 min.)	Break - Group Photo (by Lillie?)
10:35 to 10:45	(10 min.)	Phylogenetics + MSA lecture (Mike)
10:45 to 11:25	(40 min.)	Parsnp/strain analysis lecture (Mike)
11:25 to 11:45	(20 min.)	Parsnp tutorial (Todd)
11:45 to 12:00	(15 min.)	Parsnp tutorial discussion (Todd)

Metagenomic Contig Binning

You have completed a de novo metagenomic assembly...

...we're sorry, but the insights are in another castle.

Review: Metagenome Assembly

- Metagenome assembly produces contigs.
- Example (right):
 - SRA Run ID: SRR27117388
 - Human Stool Sample
 - 49,897,298 paired-end reads
 - 300bp per read (NovaSeq 6000)
- Assembly Results (MEGAHIT):
 - 179,415 contigs
 - (96k at right → min-length=500)
 - For contigs>500bp: Mean-length=2,758

*Decent depth
& length*

*A LOT of
contigs!*

Now What?

Combined reference | 32 364 850 bp | 10 references | 55 fragments

Genome statistics ≡ final.contigs

Genome fraction (%)	60.665
Duplication ratio	1.201
Largest alignment	73 347
Total aligned length	19 479 654
NGA50	...
LGA50	...

Misassemblies

# misassemblies	933
Misassembled contigs length	4 463 218

Mismatches

# mismatches per 100 kbp	1998.16
# indels per 100 kbp	53.58
# N's per 100 kbp	0

Statistics without reference

# contigs	96 118
Largest contig	609 215
Total length	265 146 582
Total length (>= 1000 bp)	228 488 467
Total length (>= 10000 bp)	135 933 934
Total length (>= 50000 bp)	62 609 535

[Extended report](#)

[illegible]




...frankly a decent description of it.

Things we want to know...



- Which of these contigs are part of the same genome?
- What organism does each of those genomes belong to?
- (other things)

Combined reference | 32 364 850 bp | 10 references | 55 fragments




Genome statistics ≡ final.contigs

Genome fraction (%) 	60.665
Duplication ratio 	1.201
Largest alignment 	73 347
Total aligned length 	19 479 654
NGA50 	...
LGA50	...


Misassemblies

# misassemblies 	933
Misassembled contigs length 	4 463 218

Mismatches

# mismatches per 100 kbp 	1998.16
# indels per 100 kbp 	53.58
# N's per 100 kbp 	0

Statistics without reference

# contigs 	96 118
Largest contig	609 215
Total length	265 146 582
Total length (\geq 1000 bp)	228 488 467
Total length (\geq 10000 bp)	135 933 934
Total length (\geq 50000 bp)	62 609 535

[Extended report](#)

Which contigs go together

- How do we know?
- Clues:
 - Read coverage
 - K-mer composition
 - Alignment to reference genomes
 - Paired-end read linkage
 - Assembly graph properties
 - ...*among others*

Brief Communication | Published: 14 September 2014

Binning metagenomic contigs by coverage and composition

[Johannes Alneberg](#), [Brynjar Smári Bjarnason](#), [Ino de Bruijn](#), [Melanie Schirmer](#), [Joshua Quick](#), [Umer Z Ijaz](#), [Leo Lahti](#), [Nicholas J Loman](#), [Anders F Andersson](#) ✉ & [Christopher Quince](#) ✉

[Nature Methods](#) **11**, 1144–1146 (2014) | [Cite this article](#)

26k Accesses | 1045 Citations | 97 Altmetric | [Metrics](#)

JOURNAL ARTICLE

COCACOLA: binning metagenomic contigs using sequence COmposition, read CoverAge, CO-alignment and paired-end read LinkAge FREE

[Yang Young Lu](#), [Ting Chen](#), [Jed A Fuhrman](#), [Fengzhu Sun](#) ✉

Bioinformatics, Volume 33, Issue 6, March 2017, Pages 791–798,

<https://doi.org/10.1093/bioinformatics/btw290>

Published: 02 June 2016 **Article history** ▼

Binning Clues: Coverage & Composition

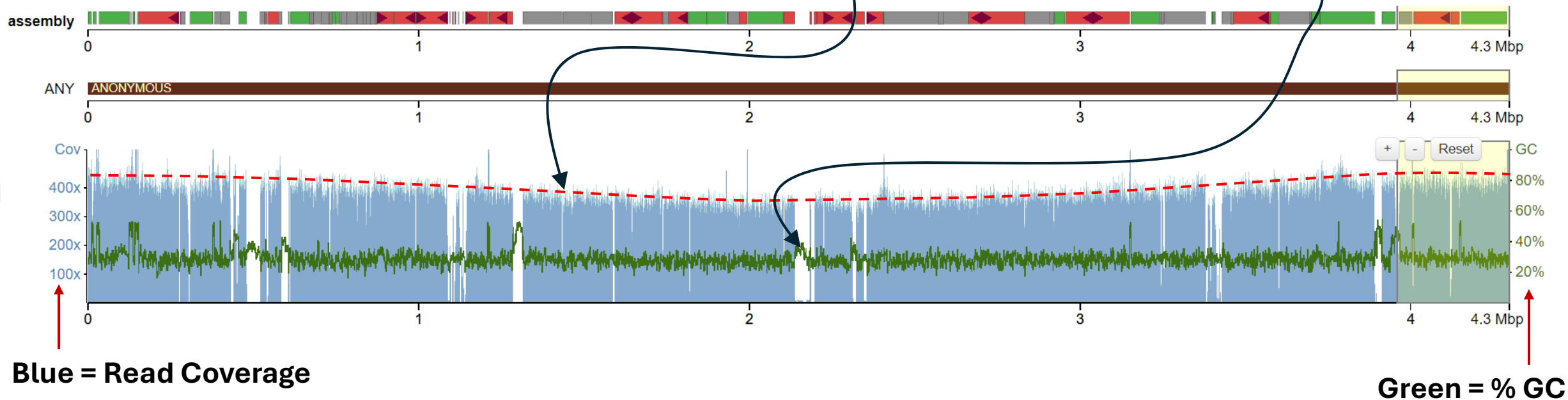
- *Example (below):*
 - *C. Difficile* clinical isolate
 - WGS de-novo assembly.
 - Comparison to CD630 reference

Read coverage is approximately constant over the length of the genome...

- Except for $\pm 5\%$ variation in sinusoidal pattern (WHY?)

GC % in reference is consistent EXCEPT for regions where clinical coverage is zero.

- Possible contamination or misassembly in reference?



Binning Clues: Reference Co-Alignment & Pair-End Linkage

- **Reference co-alignment:**

- Contigs align to the same reference genome
- Pretty good sign, especially if they don't overlap
- Note:
 - *This means that really, any taxonomic classifier can be a “contig binning” algorithm if you use it that way, but that doesn't mean it's a very good one...*

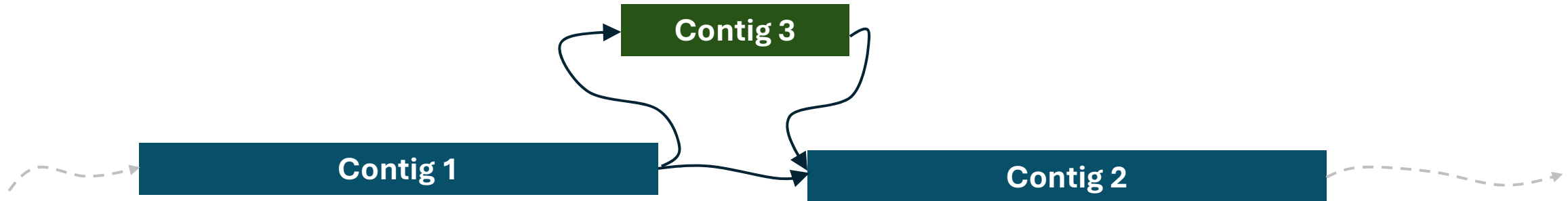
- **Paired-End Linkage:**

- Read pair-mates map to two different contigs
- More = more likely the same genome

Should be self-explanatory why these would indicate contigs come from same organism...

Binning Clues: Assembly Graph

Example: Bubble



What does this tell us?

*Should Contigs 1, 2 & 3 be in the same bin?
Two different bins?*

Contig Binning & Validation

Brass Tacks

Contig Binning Software List (Partial)

Algorithms [\[edit \]](#)

Binning algorithms can employ previous information, and thus act as [supervised classifiers](#). Many, of course, do both. The classifiers are often performing [alignments](#) against [databases](#), and try to separate sequence based on [DNA](#),^[9] like [GC-content](#).

Some prominent binning algorithms for metagenomic datasets obtained through NGS are PhyloPythia, SORT-ITEMS, and DiSBIN among others [10].

TETRA [edit]

TETRA is a statistical classifier that uses the frequency of nucleotides in **DNA**, therefore there are four nucleotides called tetramers. TETRA works by taking the **alignments** and **scores** are then calculated, which are then compared to expected by looking to individual nucleotides and vectors corresponding to different sets of nucleotides from the sample are. It is expected that the results will be trained with DNA k-mers from known sequences.

MEGAN [edit]

In the DIAMOND^[12]+MEGAN^[13] approach, the resulting alignments are mapped to a node in the NCBI taxonomy that lies above the taxon deemed "significant", if its bit score lies above a threshold. In other words, say, of the best score seen for that reference sequence, is that current DNA reference sequence in that environment.

Phylopythia [\[edit \]](#)

Phylopythia is one supervised class trained with DNA k-mers from know

Sort-ITEMS [\[edit \]](#)

SORT-ITEMS^[14] is an alignment-based binning algorithm developed by Innovations Labs of Tata Consultancy Services Ltd., India. Users need to perform a similarity search of the input metagenomic sequences (reads) against the reference database using BLASTx search. The generated BLASTx output is then taken as input by the

uses a range of BLAST align-
can be assigned. An ortholog
alignment-based binning algo
DiScRIBinATE,^[15] ProViDE [

DiScRIBinATE [edit]

DiScRIBinATE [15] is an align (TCS) Ltd., India. DiScRIBin/ Incorporating this alternate s and specificity of assignment overall misclassification rate.

ProViDE [\[edit \]](#)

ProViDE [16] is an alignment tool developed by the University of Illinois at Chicago, Ltd. for the estimation of viral sequence divergence. It is based on the Sort-ITEMS for the taxonomic classification of BLAST parameter threshold. It uses a sequence divergence and the

Remark on this Wikipedia list:

this is a goofy list. The only methods here I'd heard of here are MEGAN and Phylopythia which are really just taxonomic classifiers. Nearly all of these are quite old.

PCAHIER [[edit](#)]

PCAHER,^[18] another binning algorithm developed by the Georgia Institute of Technology., employs nucleotide frequencies as the features and adopts a hierarchical classifier (PCAHER) for binning short metagenomic reads. PCAHER uses principal component analysis to reduce the high dimensionality of the feature space. The effectiveness of PCAHER was demonstrated through comparisons against a non-hierarchical classifier, and two existing binning algorithms (TETRA and Phylopythia).

SPHINX [edit]

SPHINX,^[17] another binning algorithm developed by the Innovation Labs of Tata Consultancy Services, employs a hybrid strategy that achieves high binning efficiency by utilizing the principles of both 'composition'- and 'alignment'-based binning algorithms. The approach was designed with the objective of analyzing metagenomic datasets using composition-based approaches, but nevertheless with the accuracy and specificity of alignment-based approaches. It was observed to classify metagenomic sequences as rapidly as composition-based algorithms. In addition, it was observed that the accuracy and specificity of assignments) of SPHINX was observed to be comparable with reference alignment-based algorithms.

INDUS and TWARIT [edit]

Represent other composition-based binning algorithms developed by the Innovation Labs of Tata Consultancy Services Ltd. These algorithms utilize a range of oligonucleotide compositional (as well as statistical) parameters while maintaining the accuracy and specificity of taxonomic assignments.^{[19][20]}

References [\[edit \]](#)

1. [^] Maguire, Finlay; Jia, Baofeng; Gray, Kristen L.; Lau, Wing
12. [^] Buchfink, Benjamin; Xie, Chao

More modern list:

- MetaBAT2 (2019)
- CONCOCT (2014)
- COCACOLA (2017)
- VAMB (2022)
- MyCC (2016)
- MaxBin 2.0 (2015)
- MetaWatt-3.5 (2012)
- MetaWRAP (2018)
- Autometa (2019)
- MetaBinner (2023)
- UltraBinner (2023)

...and probably a dozen more since 2020

Remarks on More Modern Binner

More modern list:

- MetaBAT2 (2019)
- CONCOCT (2014)
- MyCC (2016)
- MaxBin 2.0 (2015)
- MetaWatt-3.5 (2012)
- COCACOLA (2017)
- MetaWRAP (2018)
- Autometa (2019)
- VAMB (2022)
- MetaBinner (2023)
- UltraBinner (2023)

Very different core algorithm from original MetaBAT (2016), added assembly graph & other features

*All largely based primarily on:
1) Coverage
2) Kmer composition*

Added co-alignment & paired-end read linkage

Deep-Learning method that got a lot of attention

Heavy use of single-copy marker genes to refine bins

MetaBinner + CONCOCT + MetaBAT2

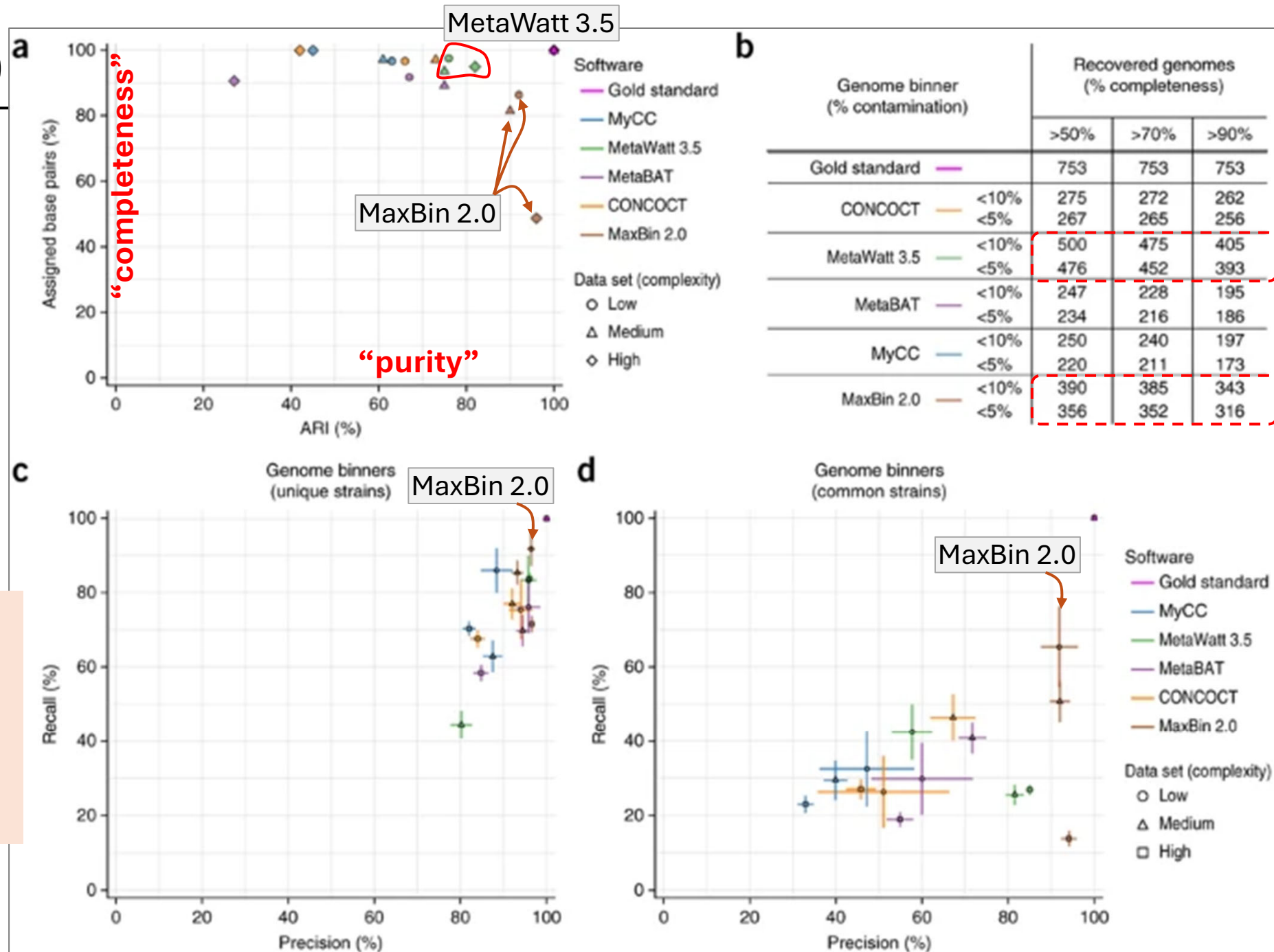
CAMI 1 (2017)

CAMI: Critical Assessment of Metagenome Interpretation
(Sczyrba, et al., 2017)

ARI = adjusted Rand Index. (% of contig pairs correctly co-assigned).

...hard not to conclude that MaxBin was the winner here, with honorable mention to MetaWatt for best sensitivity...

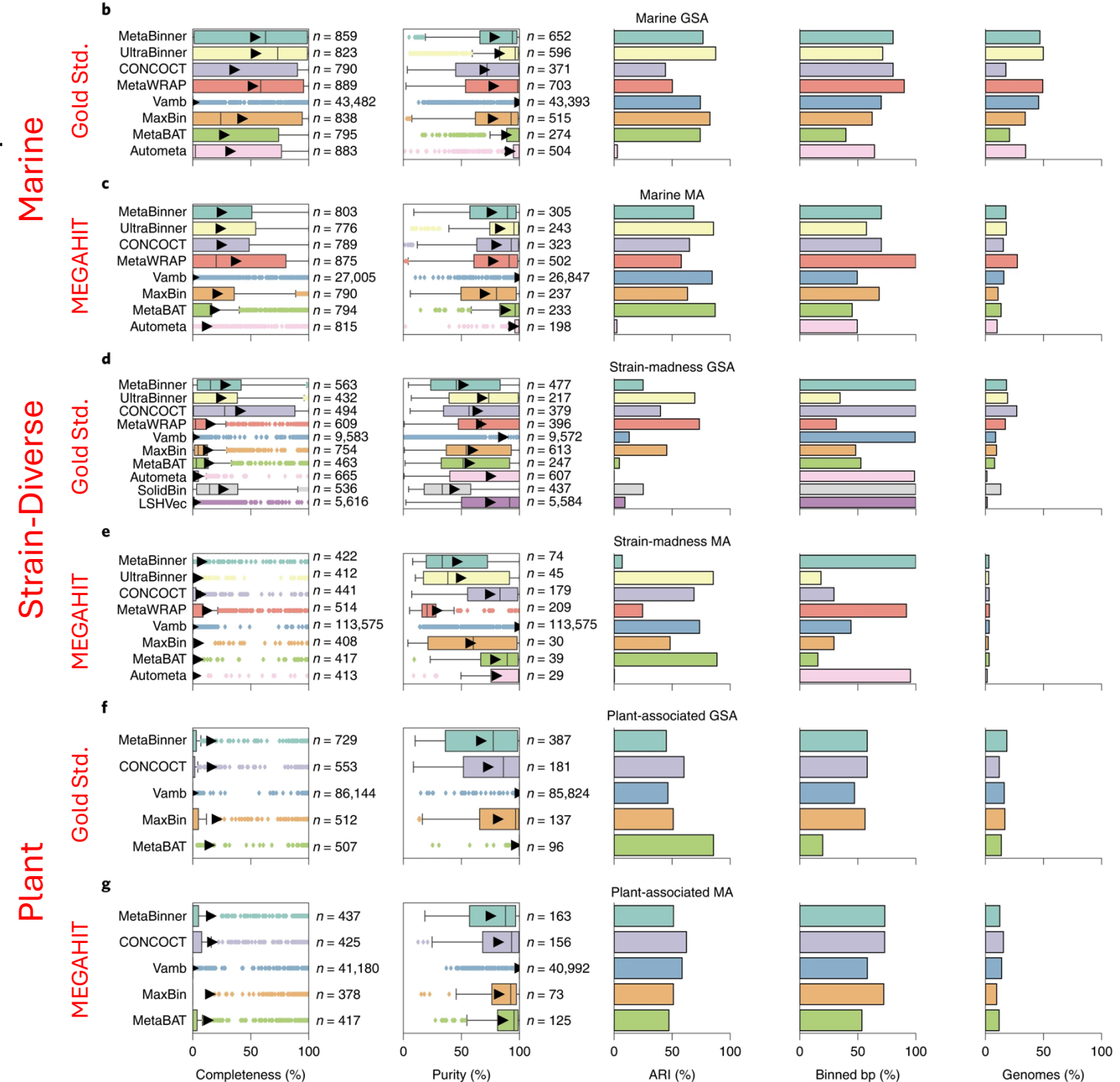
Fastest: MetaBAT



CAMI 2 (2022)

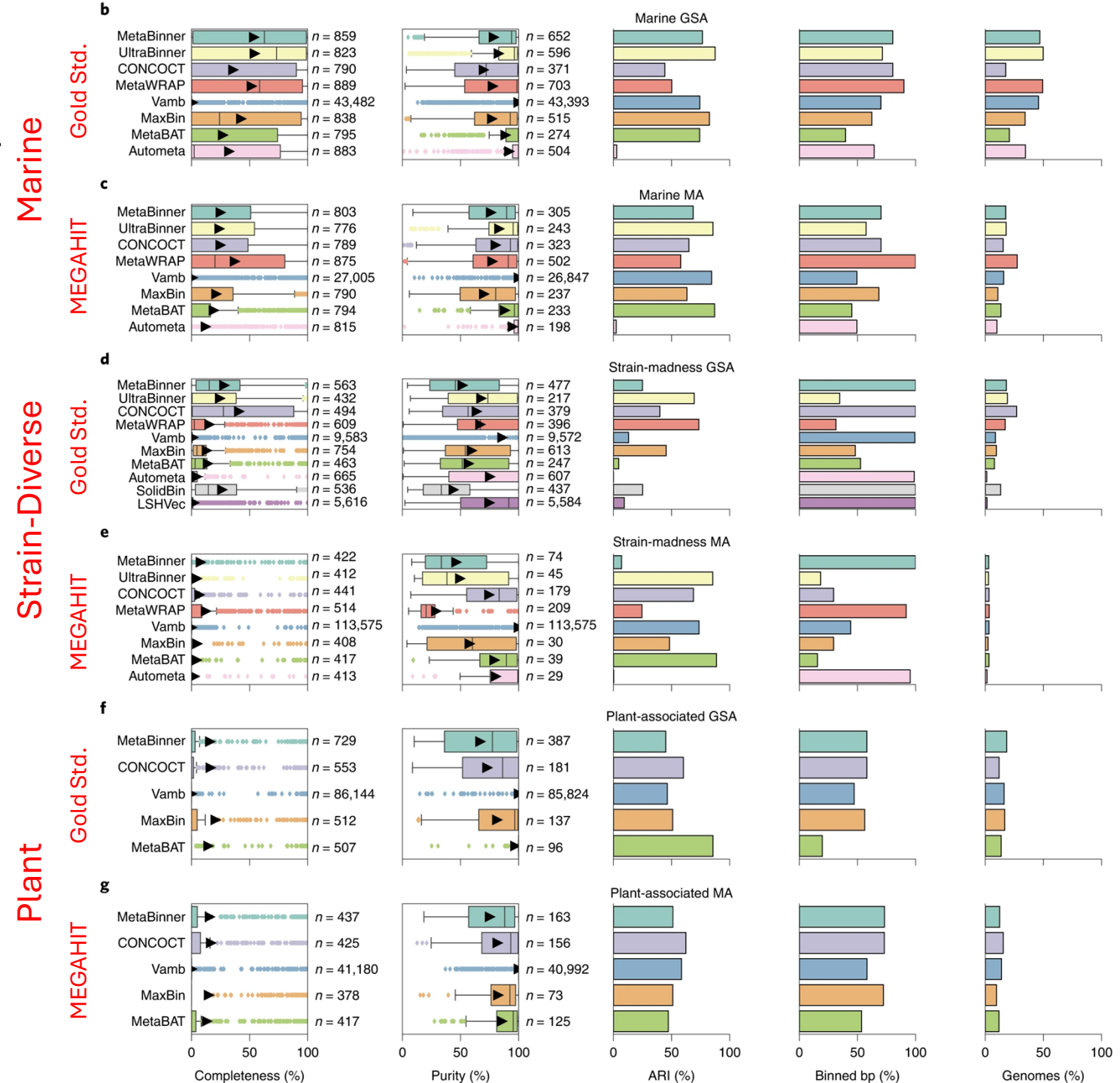
CAMI-2

- 3 different datasets reflecting “modern” challenges
 - novel microbes (Plant)
 - strain diversity
 - high synteny (i.e. short contigs) (Marine)
- 2 Types of contigs
 - “gold-standard”
 - short+long read assembly
 - “megahit”
 - short reads only

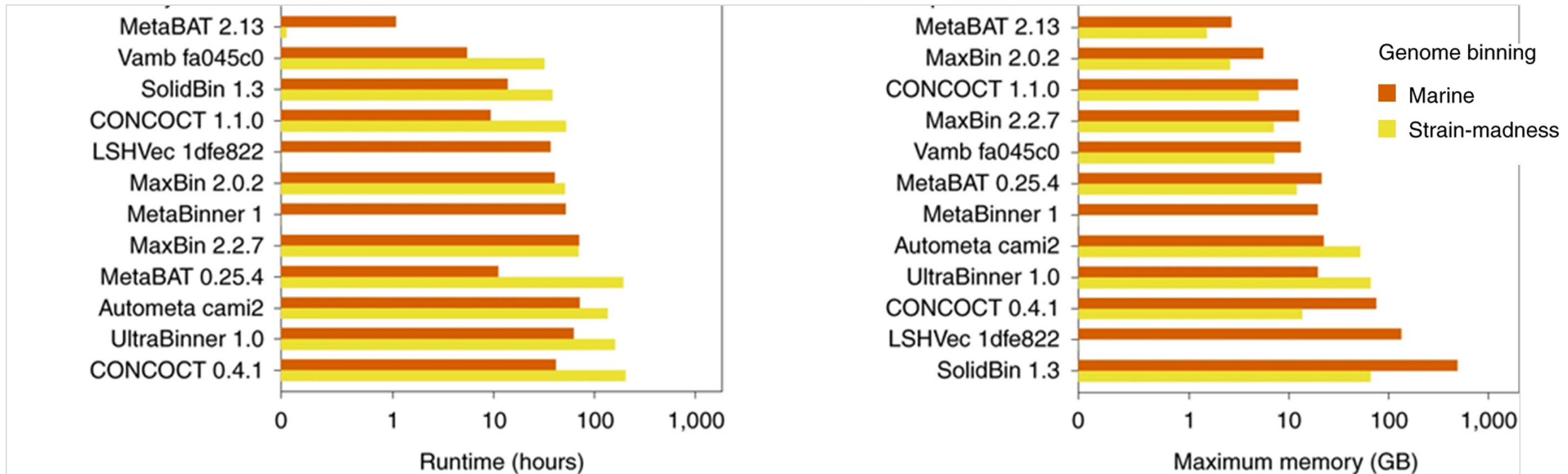


CAMI 2 (2022)

- Far less clear which method is optimal overall
- MetaBinner/UltraBinner seem to do reliably well (“ensemble” methods)
 - CONCOCT also quite well.
- VAMB is an outlier:
 - Very high purity, Very low completeness
 - “Partial” binning
 - Recall the bubble example
- MetaBAT:
 - Does reasonably well, not best
 - By far lowest compute cost (time/RAM)
- Several methods unable to run on larger datasets



CAMI 2 (2022): Runtime & Memory Usage



Contig Binning Software: MetaBAT2

- Inputs:
 - Contig file
 - Coverage file
- Parameters (default value):

[PeerJ](#). 2015; 3: e1165. PMCID: PMC4556158
Published online 2015 Aug 27. doi: [10.7717/peerj.1165](https://doi.org/10.7717/peerj.1165) PMID: [26336640](https://pubmed.ncbi.nlm.nih.gov/26336640/)

MetaBAT, an efficient tool for accurately reconstructing single genomes from complex microbial communities

[Dongwan D. Kang](#),^{1,2} [Jeff Froula](#),^{1,2} [Rob Egan](#),^{1,2} and [Zhong Wang](#)^{✉1,2,3}

[PeerJ](#). 2019; 7: e7359. PMCID: PMC6662567

Published online 2019 Jul 26. doi: [10.7717/peerj.7359](https://doi.org/10.7717/peerj.7359) PMID: [31388474](https://pubmed.ncbi.nlm.nih.gov/31388474/)

MetaBAT 2: an adaptive binning algorithm for robust and efficient genome reconstruction from metagenome assemblies

[Dongwan D. Kang](#),¹ [Feng Li](#),² [Edward Kirtan](#),¹ [Ashleigh Thomas](#),¹ [Rob Egan](#),¹ [Hong An](#),² and [Zhong Wang](#)^{✉1,3,4}

Argument	Definition	Default Value	Notes
-m	Minimum size of a contig for binning (must be ≥ 1500)	2500	CAMI-2 ran both 1500, 2500 and found little difference
--maxP	% of 'good' contigs considered for binning	95%	Greater \Rightarrow more sensitivity
--minS	Minimum score of a edge for binning (from 1 to 99)	60	Greater \Rightarrow more specificity
--maxEdges	Maximum number of edges per node	200	Greater \Rightarrow more sensitivity
--pTNF	4-mer probability cutoff for building 4-mer graph	0	Used to skip a preparation step, for speed.
-x --minCV	Minimum mean coverage of a contig in each library for binning.	1	Could be adjusted based on expectations/needs vis-à-vis depth
-s --minClsSize	Minimum size of a bin as the output	200kbp	

Contig Bin Validation: CheckM2

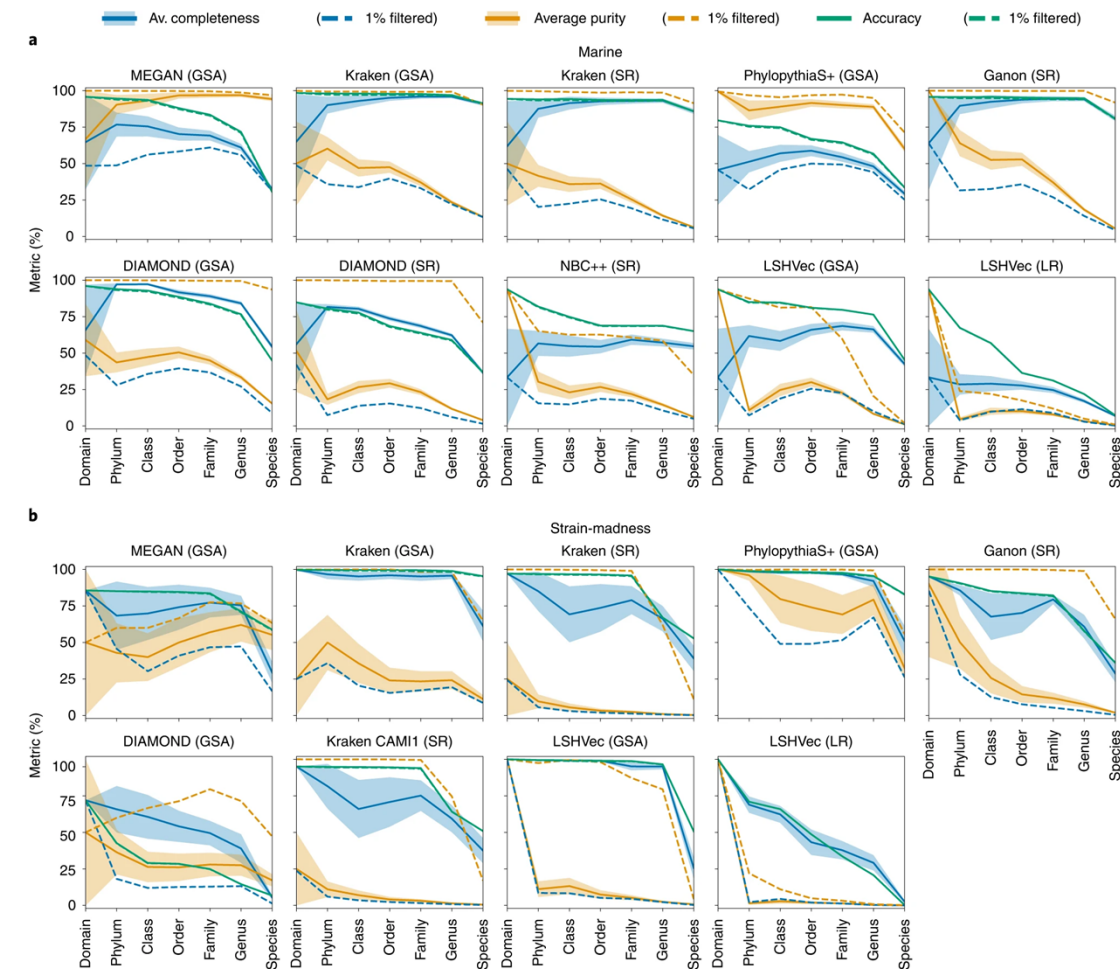
- Original Idea (CheckM, i.e. version 1):
 - Use *lineage-specific, single-copy marker genes* (à la MetaPhlAn) to estimate how complete the bin is.
 - Issue: only well-studied lineages have good database of markers
- New Idea (CheckM2):
 - Use a fancier machine learning algorithm to predict bin quality
- Output:

Bin #	Completeness (%)	Contamination (%)	Coding Density	Contig N50	Average Gene Length	Genome Size	GC Content	Total Coding Sequences	Total Contigs	Max Contig Length
1	45.5	0.0	0.885	609,215	336.2	896,313	53%	788	3	609,215
2	10.7	0.0	0.859	34,272	324.0	273,576	60%	242	13	56,825
3	75.6	21.3	0.878	13,874	296.7	2,692,053	58%	2,660	247	100,008
4	73.0	0.0	0.907	123,875	351.4	2,083,794	38%	1,796	31	196,080
5	92.2	3.6	0.886	182,793	363.5	2,626,142	60%	2,136	32	355,138
6	24.2	0.0	0.865	4,241	272.0	475,717	64%	506	113	14,380
7	10.5	0.0	0.903	8,236	302.3	247,455	61%	247	34	17,322
8	58.6	0.8	0.907	4,656	268.2	1,372,490	64%	1,550	302	13,961
9	88.0	5.5	0.883	28,581	311.1	3,020,618	58%	2,861	146	107,535
10	38.7	0.2	0.889	81,426	327.6	885,202	63%	802	16	111,004
11	61.0	0.0	0.899	162,638	331.1	1,125,146	54%	1,020	8	265,330

Taxonomic Assignment

- Note that neither MetaBAT nor CheckM2 assign taxonomy to each of our bins.
- Taxon assignment is pretty simple, many ways to do it:
 - BLAST
 - Sourmash
 - Kraken2
 - ...any of the other tools evaluated in CAMI-2:
 - MEGAN
 - DIAMOND
 - Ganon
 - PhyloPythiaS+
 - NBC++

Nota Bene: This step has not been included in the tutorial because most of these programs require a large database to be downloaded or built in advance. It should be simple to do this using Sourmash, however, based on that material from earlier in the course...



Any Questions?

Things to think about:

- What is contig binning and what is the goal?
- What information can we use to bin contigs together?
- How many contig binning methods are there?
- What is the best contig binner?
- Is VAMB any good?
- What does CheckM2 produce?
- Why didn't I add taxon assignment to the tutorial?



Binning & Validation Tutorial

- Tasks:
 1. Find the outputs from the QC/Assembly tutorial:
 - a) MEGAHIT Contigs
 - b) Coverage information (.bam file)
 - c) MetaBAT bins
 2. Run CheckM2 on the bins from (c)
 3. Run MetaBAT2 on the contigs/coverage, see if you get the same # of bins
 4. Run CheckM2 on the new bins & compare to (4)
 5. Run CheckM2 on some bins from a Human stool sample (provided)
- Left as an exercise:
 6. Assign taxonomy to one of these bins using Sourmash
- Other bash-scripting concepts:
 - Setting & using bash variables
 - Absolute vs. Relative paths

Thank You

Please don't hesitate to follow up with me after STAMPS if I can help or go through any of this material again.