

Detecting new miRNAs from deep sequencing data: A field study in worm

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25th TBI Winterseminar



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Outline

- Data
- Pre-Analysis
- miRNA prediction
 - Classifier
 - miRDeep
 - miRanalyzer
- Results

Data

C. elegans small RNA sequencing

Set 1 (MDC - set)

Data

- SOLiD sequenzer
- 6,080,238 reads
- 18-27 nt in length

Mapping

- 3,288,430 reads mapped with BWA (54%)
- remapping by segemehl (377,594 tags)

C. elegans small RNA sequencing

Set 2 (Bartel - set)

Data(GSE5990)

- mixed-stage C. elegans
- 454 sequenzer
- 850,870 reads (181,668 tags)
- 18-36 nt in length

Mapping

- 138,868 tags mapped by segemehl (76%)

C. elegans small RNA sequencing

Set 3 (Berezikov - set)

Data(GSE15169)

- mixed-stage C. elegans
- 454 sequenzer
- 181,849 reads (23,327 tags)
- 18-38 nt in length

Mapping

- 22,277 tags mapped by segemehl (95%)

Pre-Analysis

ncRNA mapping results (MDC – set)

ncRNA type	# of annotated ncRNAs	# of annotated ncRNAs with read support	% of annotated ncRNAs with read support
all	6606	4137	62.6
microRNA	174	165	94.8
tRNA	631	626	99.2
snoRNA	133	114	85.7
snRNA	90	84	93.3
rRNA	25	25	100
21U-RNA	5356	2839	53.0
SL2 splice leader	8	8	100
others	189	149	78.8

ncRNA mapping results (MDC – set)

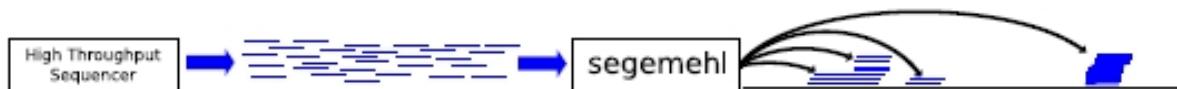
ncRNA type	# of tags	# of reads	% of all tags	% of all reads
all	68,385	1,045,428	18,1	31,8
miRNAs	6,625	857,689	1,8	26,1
tRNAs	10,93	93,789	2,9	2,9
snoRNAs	1,163	3,774	0,3	0,1
snRNAs	2,186	5,652	0,6	0,2
rRNAs	33,519	80,152	8,9	2,4
21U-RNAs	3,524	6,181	0,9	0,2
others	6,946	1,626	1,8	0,0

miRNA prediction

Classifier

Data preparation

1. Map small RNAs to the human genome, using `segemehl`



2. Cluster hits based on their genomic location (distance <100nt)



3. Divide consecutive reads into blocks, using `blockbuster`



4. Discard clusters with <2 blocks and/or <10 reads (small information content)

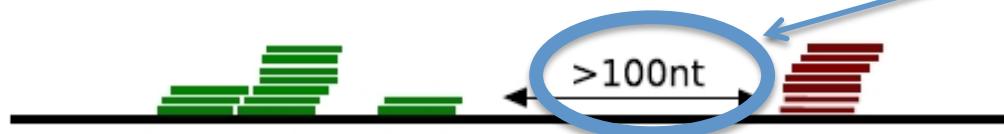


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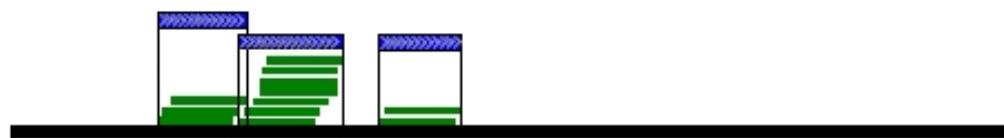
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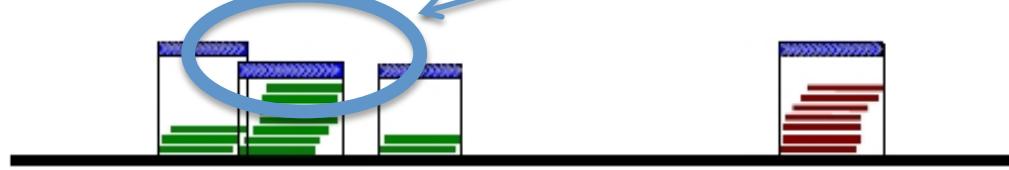
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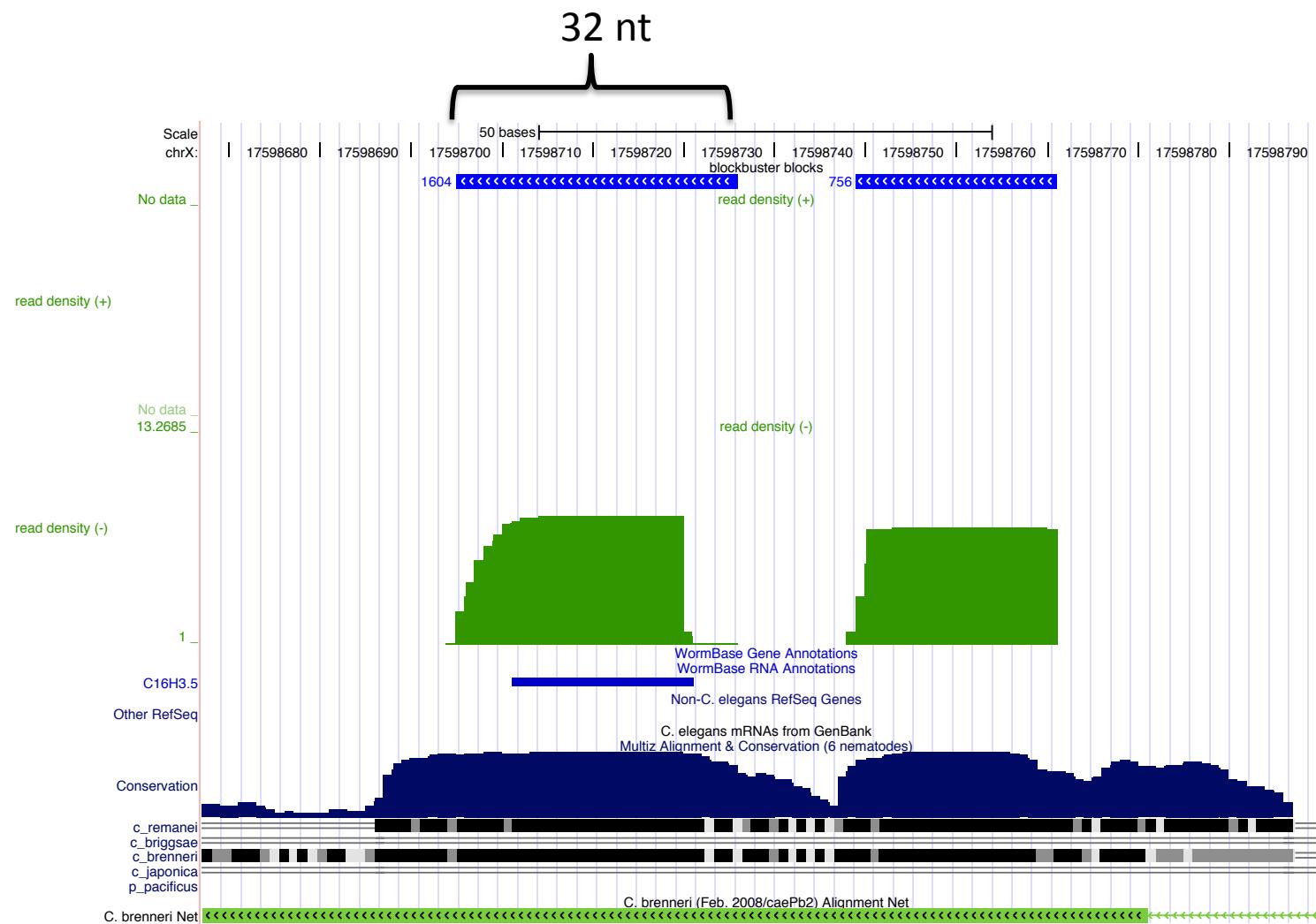
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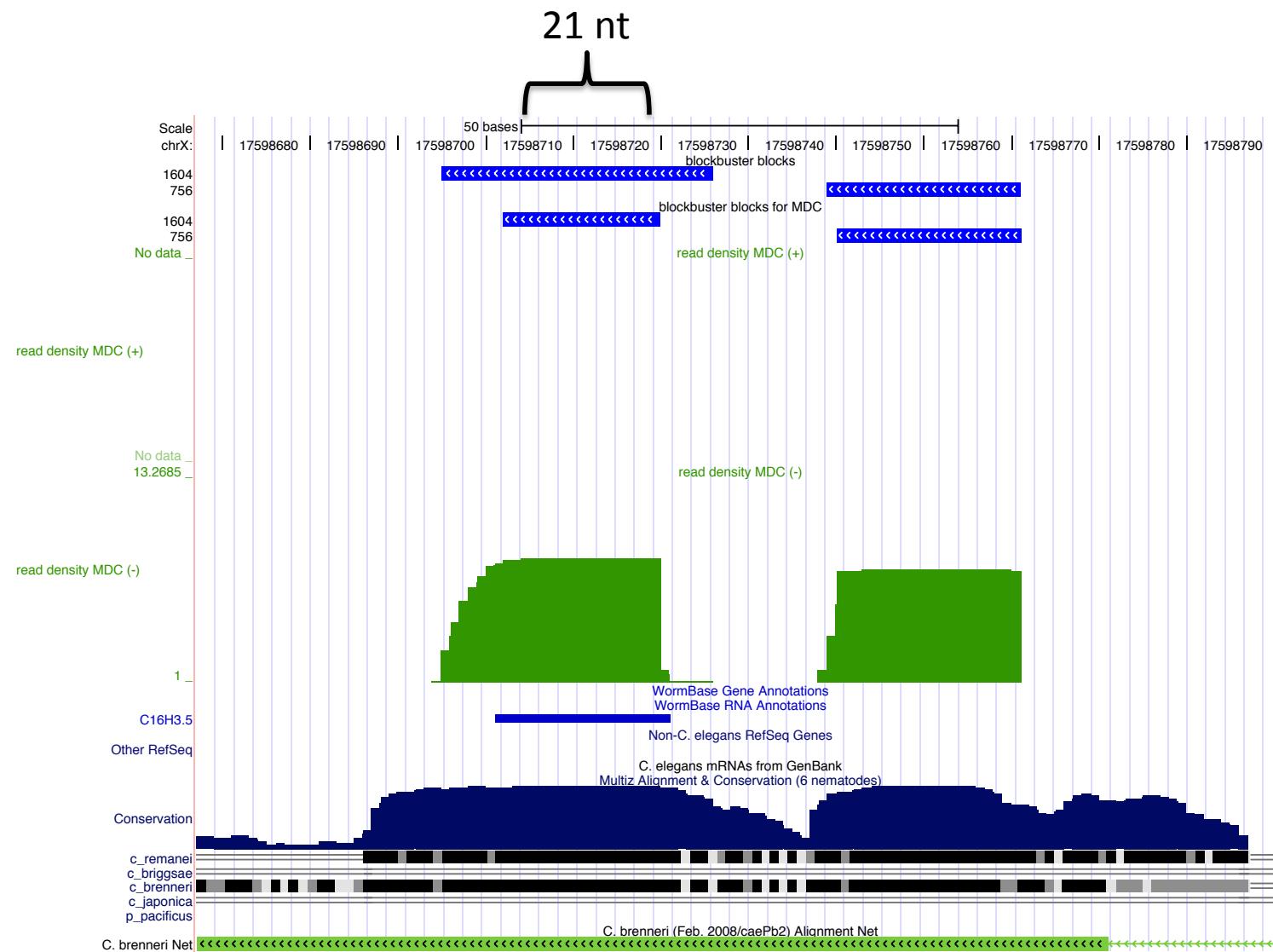
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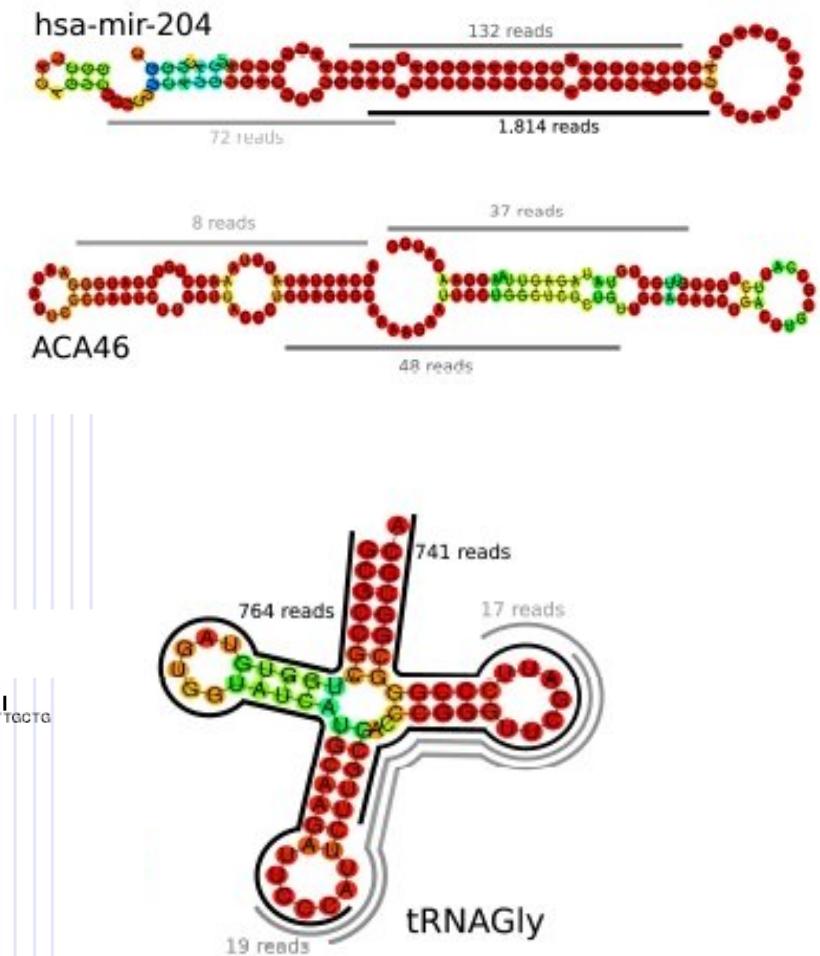
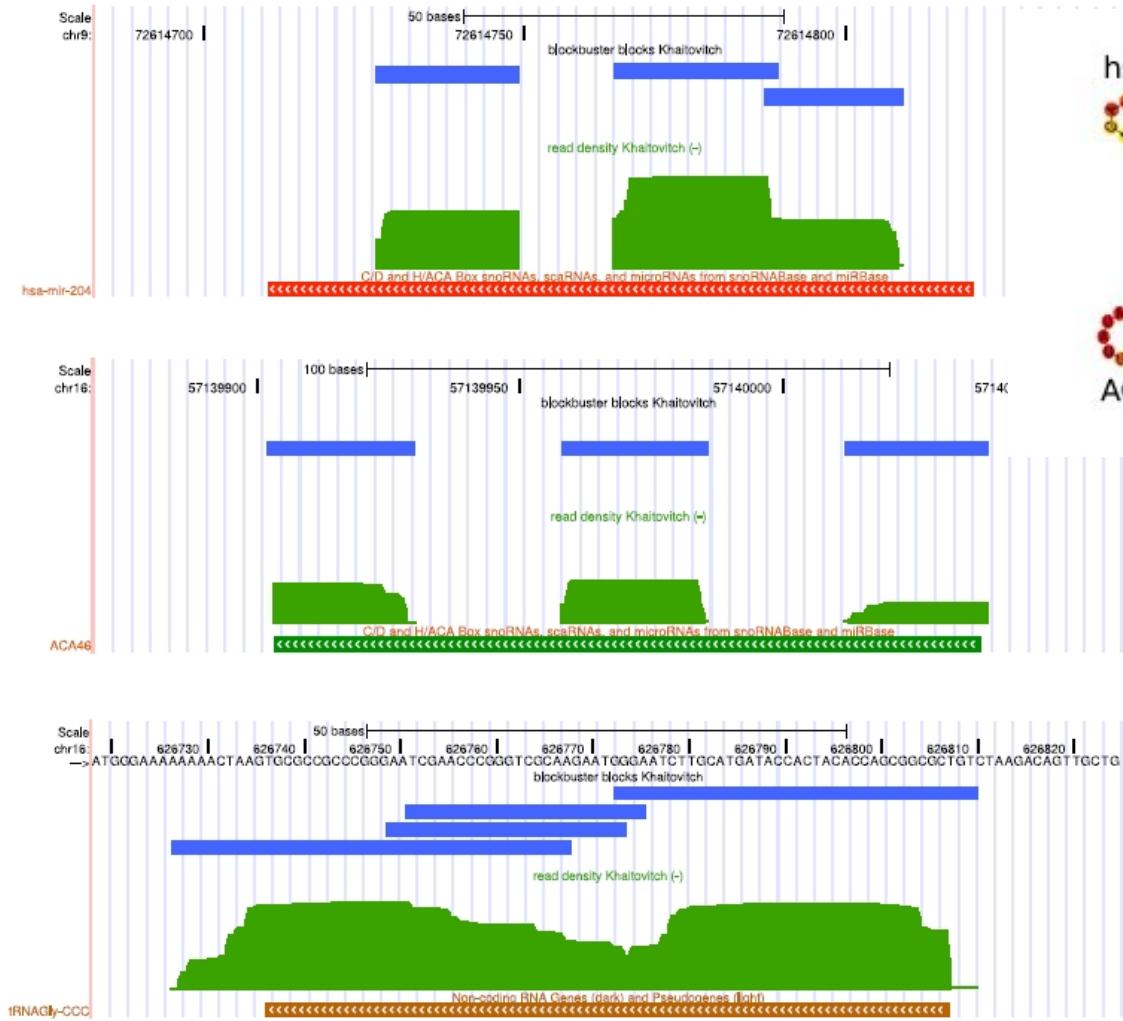
Data preparation



Data preparation



Different ncRNAs have different read patterns



Classifier (random forest)

- Trained on human reads mapping to human ncRNAs
(khaitovitch – brain data – 454 sequenzer)
- Trainingset:
 - 243 miRNAs
 - 19 snoRNAs (H/ACA)
 - 116 snoRNAs (C/D)
 - 336 tRNAs
 - 273 other ncRNAs

Confusion Matrix (10-fold cross validation)

miRNA	snoRNA (H/ACA)	snoRNA (C/D)	tRNA	other	<- classified as
232 (95%)	3	2	4	2	miRNA
8	3 (2%)	1	3	4	snoRNA (H/ACA)
2	2	71 (61%)	16	25	snoRNA (C/D)
3	0	5	291 (87%)	37	tRNA
9	1	15	53	195 (71%)	other

Classification of worm ncRNAs

	Recall			PPV
	miRNA	snoRNA	tRNA	
MDC set	61% (49 of 80)	0% (0 of 28)	76% (438 of 573)	94%
Bartel set	85% (61 of 71)	5% (1 of 8)	76% (438 of 573)	95%
Berezikov set	79% (26 of 33)	0% (0 of 0)	31% (52 of 169)	96%

miRDeep

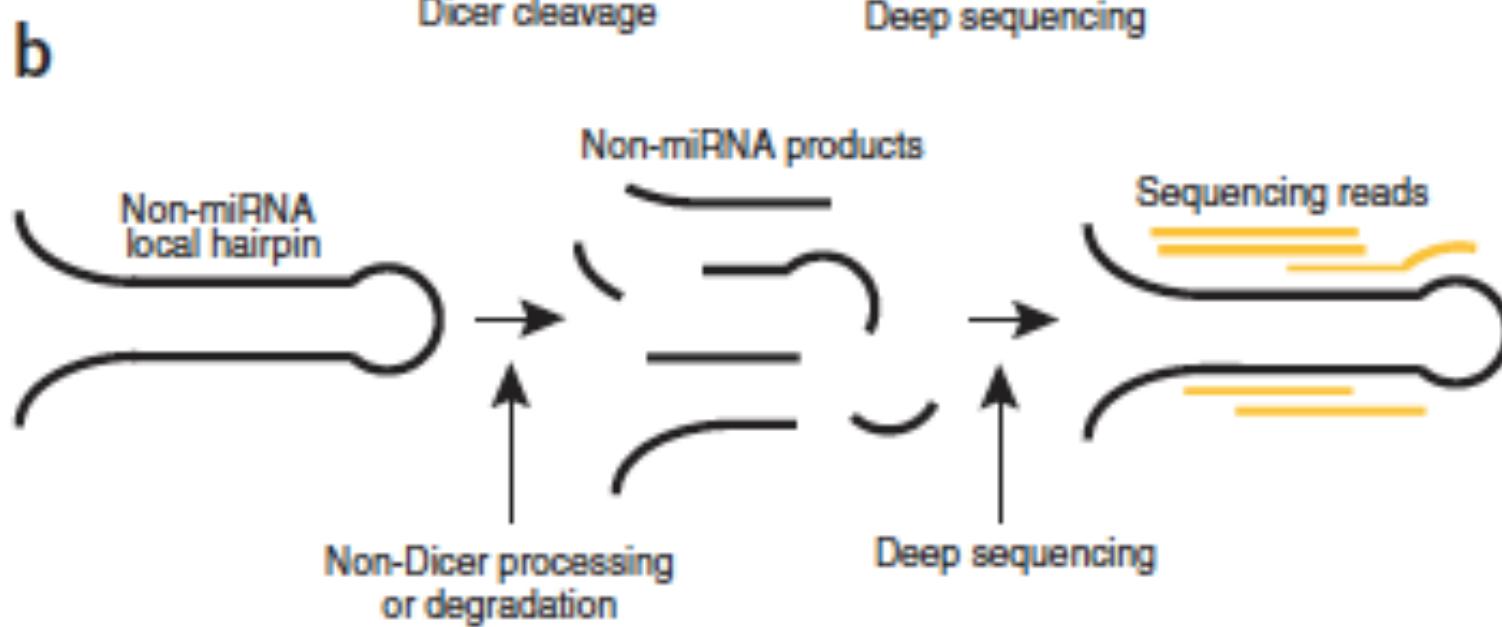
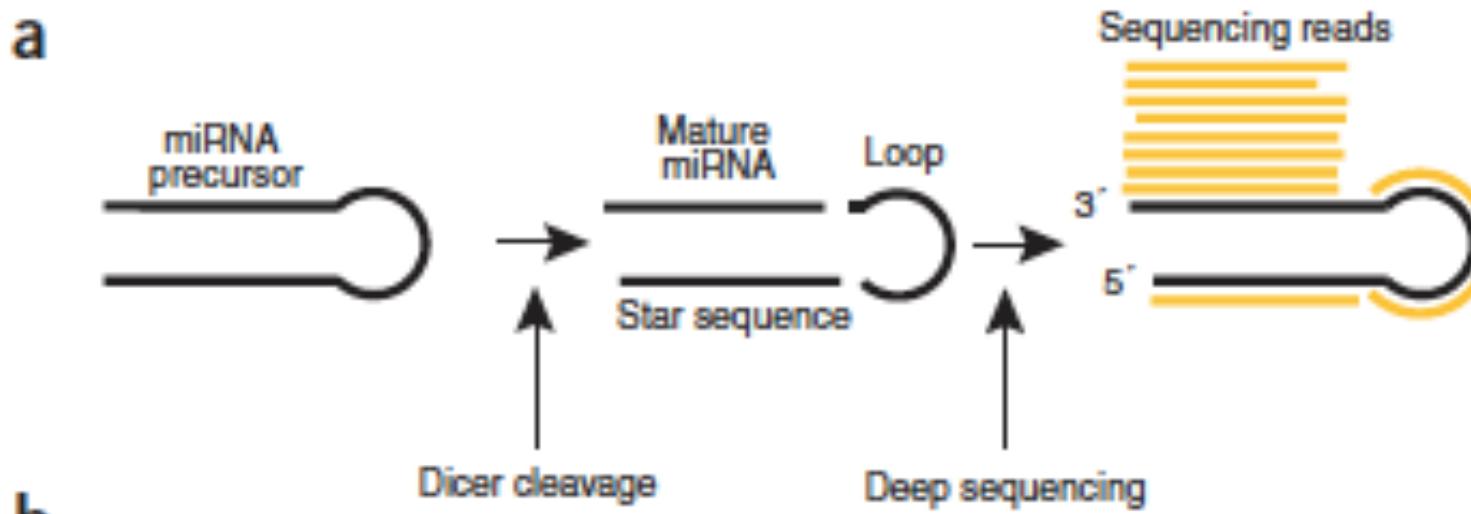
Discovering microRNAs from deep sequencing data using miRDeep

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<http://www.nature.com/naturebiotechnology>

The capacity of highly parallel sequencing technologies to detect small RNAs at unprecedented depth suggests their value in systematically identifying microRNAs (miRNAs). However, the identification of miRNAs from the large pool of sequenced transcripts from a single deep sequencing run remains a major challenge. Here, we present an algorithm, miRDeep, which uses a probabilistic model of miRNA

and 454 Life Sciences/Roche, can sequence DNA orders of magnitude faster and at lower cost than Sanger sequencing and are evolving so rapidly that increases in sequencing speed by at least another order of magnitude seem likely over the next few years. Although the Solexa/Illumina system can produce ~32 million sequencing reads in one run, read length is currently limited to 35 bp. In contrast, the current 454 platform yields reads up to 200 bases each, although the number of reads



miRanalyzer

miRanalyzer: a microRNA detection and analysis tool for next-generation sequencing experiments

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Juan Manuel Falcón-Pérez⁵ and Ana M. Aransay^{1,*}

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²Institute for Bioinformatics and Systems Biology, German Research Center for Environmental Health, Ingolstädter Landstrasse 1, D-85764 Neuherberg, ³Department of Genome-Oriented Bioinformatics, Wissenschaftszentrum Weihenstephan, Technische Universität München, 85350 Freising, ⁴Bioinformatics Group, Department of Computer Science, University of Leipzig, Haertelstr. 16-18, D-04107 Leipzig, Germany and ⁵Metabolomics Unit, CIC bioGUNE, CIBERehd, Technology Park of Bizkaia, 48160 Derio, Bizkaia, Spain

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ABSTRACT

Next-generation sequencing allows now the sequencing of small RNA molecules and the estimation of their expression levels. Consequently, there will be a high demand of bioinformatics tools to cope with the several gigabytes of sequence data generated in each single deep-sequencing experiment. Given this scene, we developed

INTRODUCTION

The recent years witnessed a profound change in our understanding of the regulation of gene expression. Small non-coding RNA especially came into focus as it became clear that they are key players in many cellular processes by post-transcriptionally regulating gene expression via either degradation, translational repression, or both (1,2). MicroRNAs, belonging to the family of small non-coding RNAs are endogenous in many animal and

Results

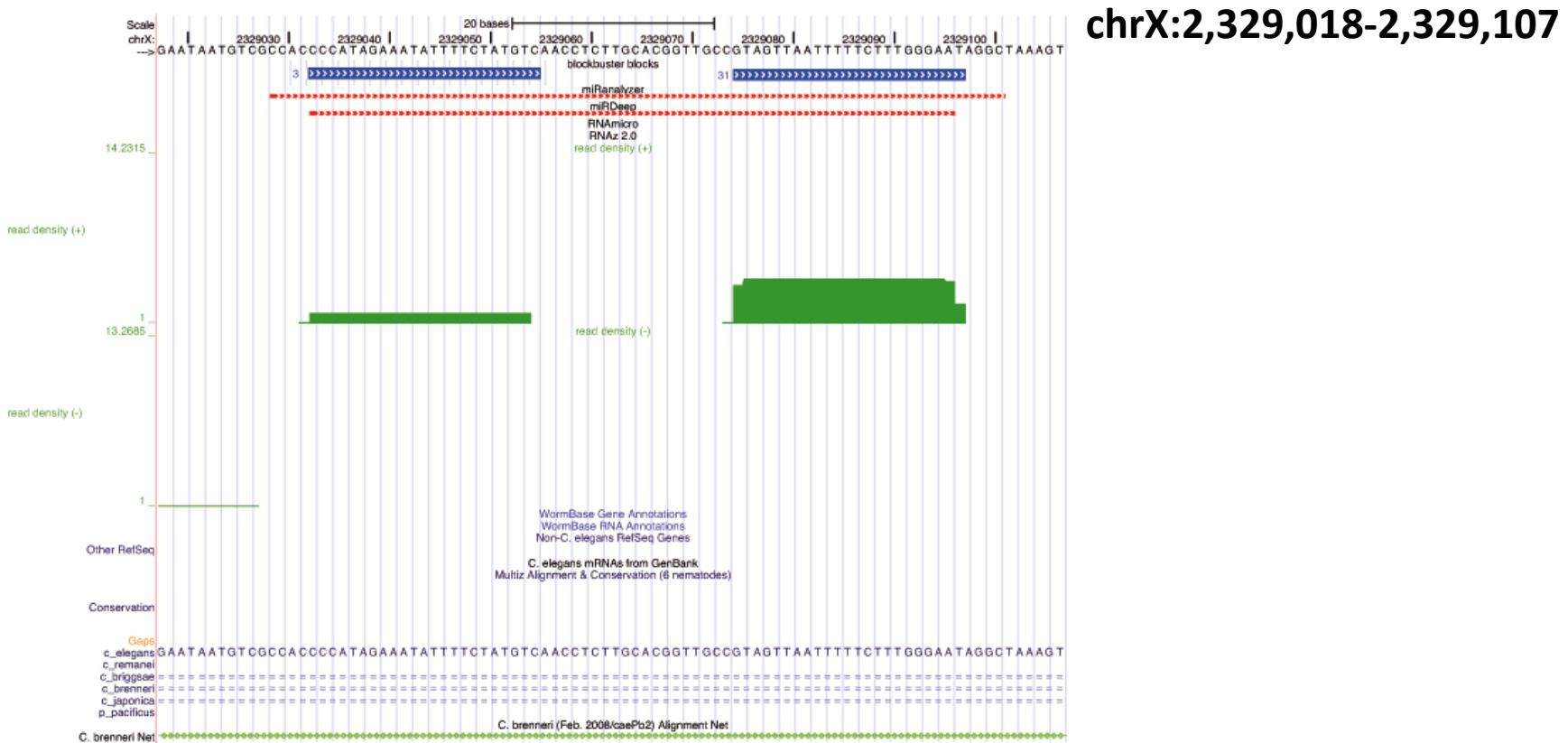
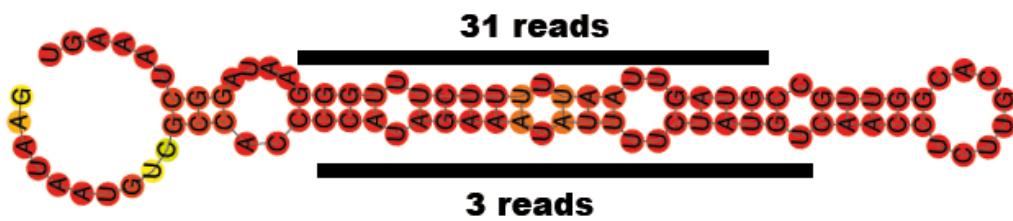
miRNA prediction in worm

MDC	miRDeep	miRanalyzer	Classifier
miRDeep	335	50	3
miRanalyzer	50	650	4
Classifier	3	4	5

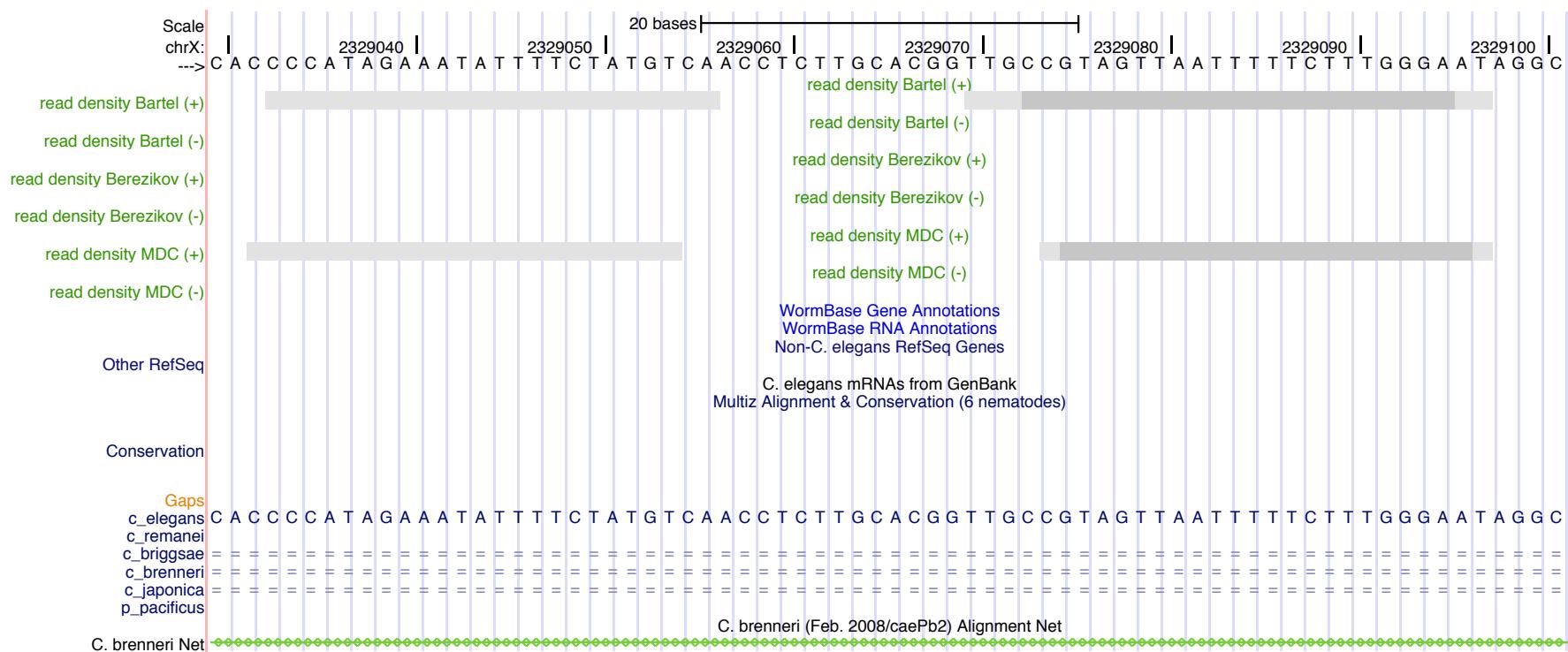
Bartel	miRanalyzer	Classifier
miRanalyzer	67	1
Classifier	1	7

Berezikov	miRanalyzer	Classifier
miRanalyzer	27	0
Classifier	0	1

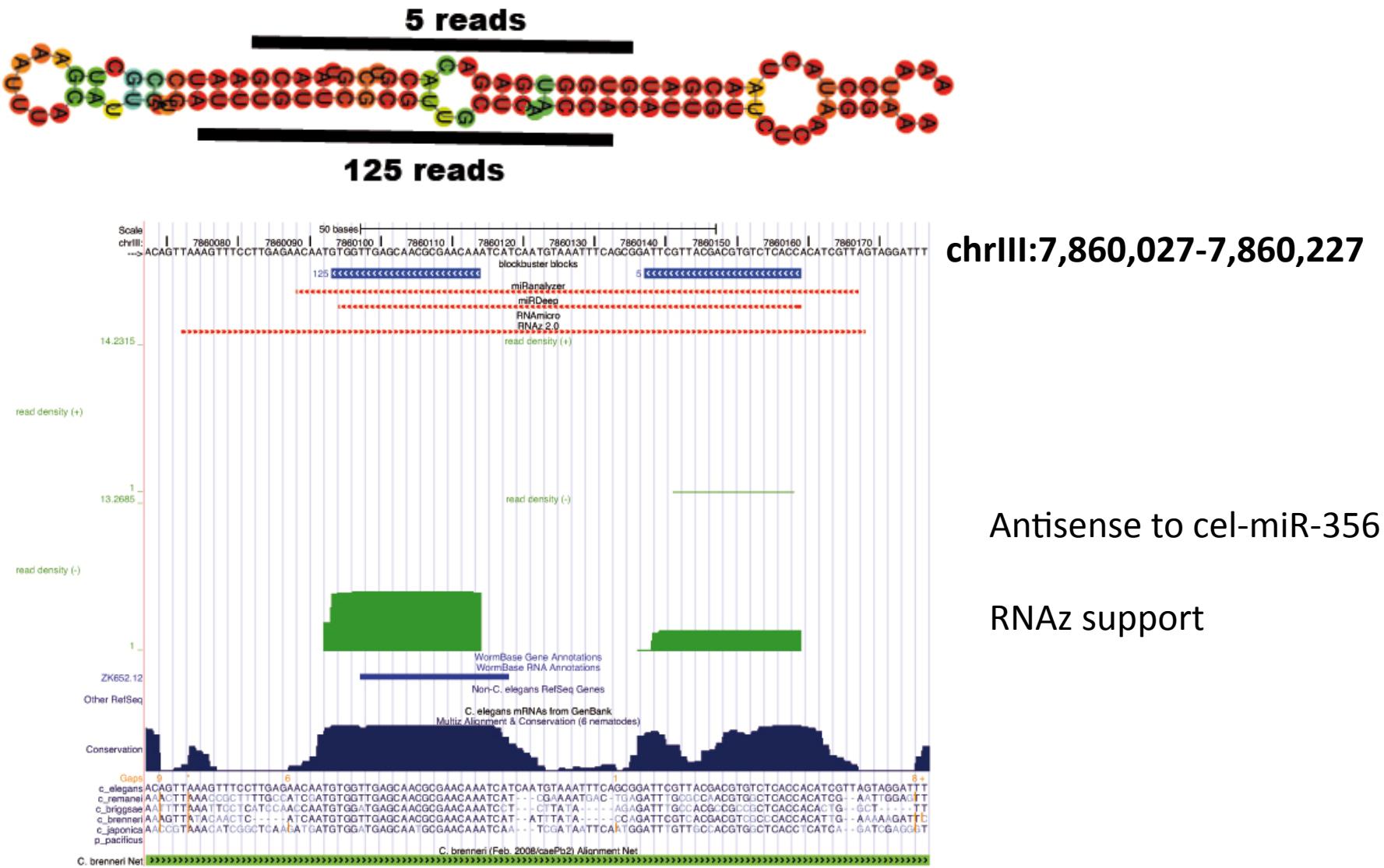
3 candidates found by the classifier, miRanalyser and miRDeep (MDC – set)



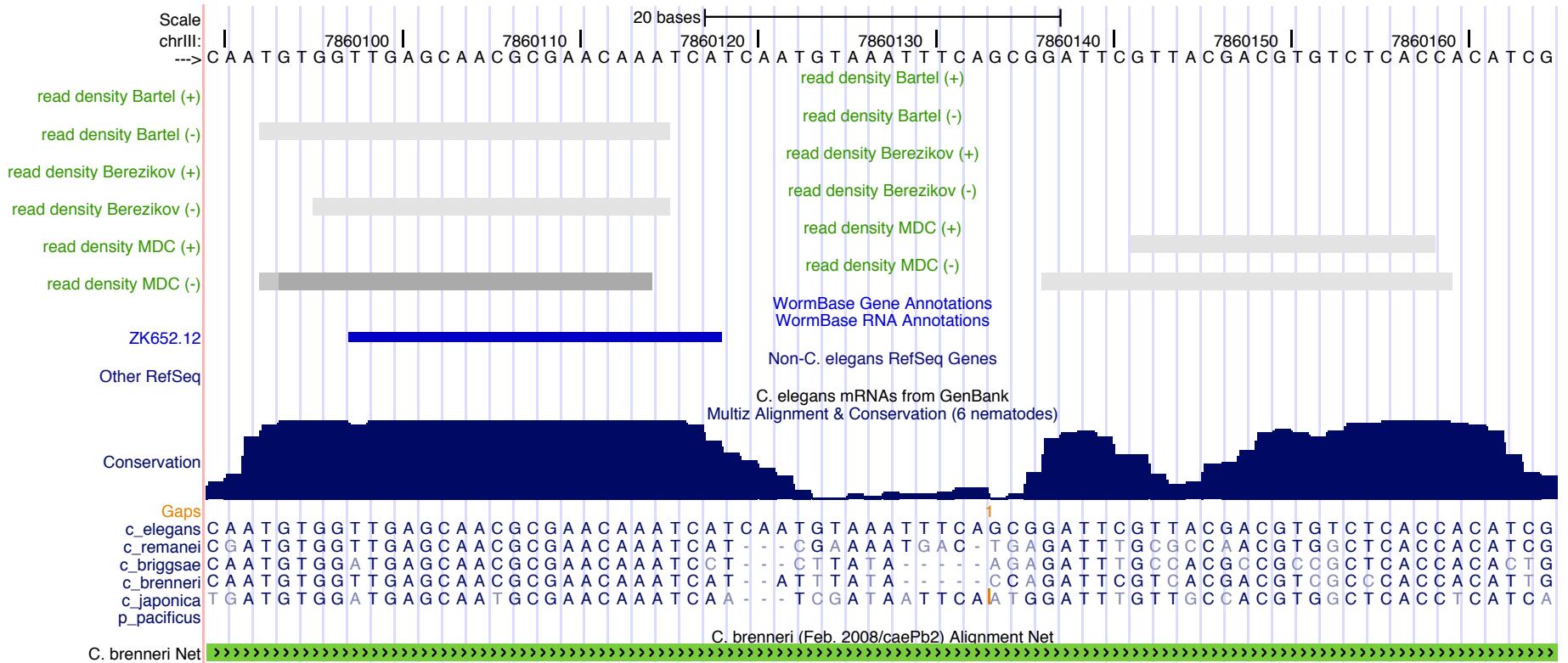
3 candidates found by the classifier, miRanalyser and miRDeep (all sets)



3 candidates found by the classifier, miRanalyser and miRDeep (MDC – set)

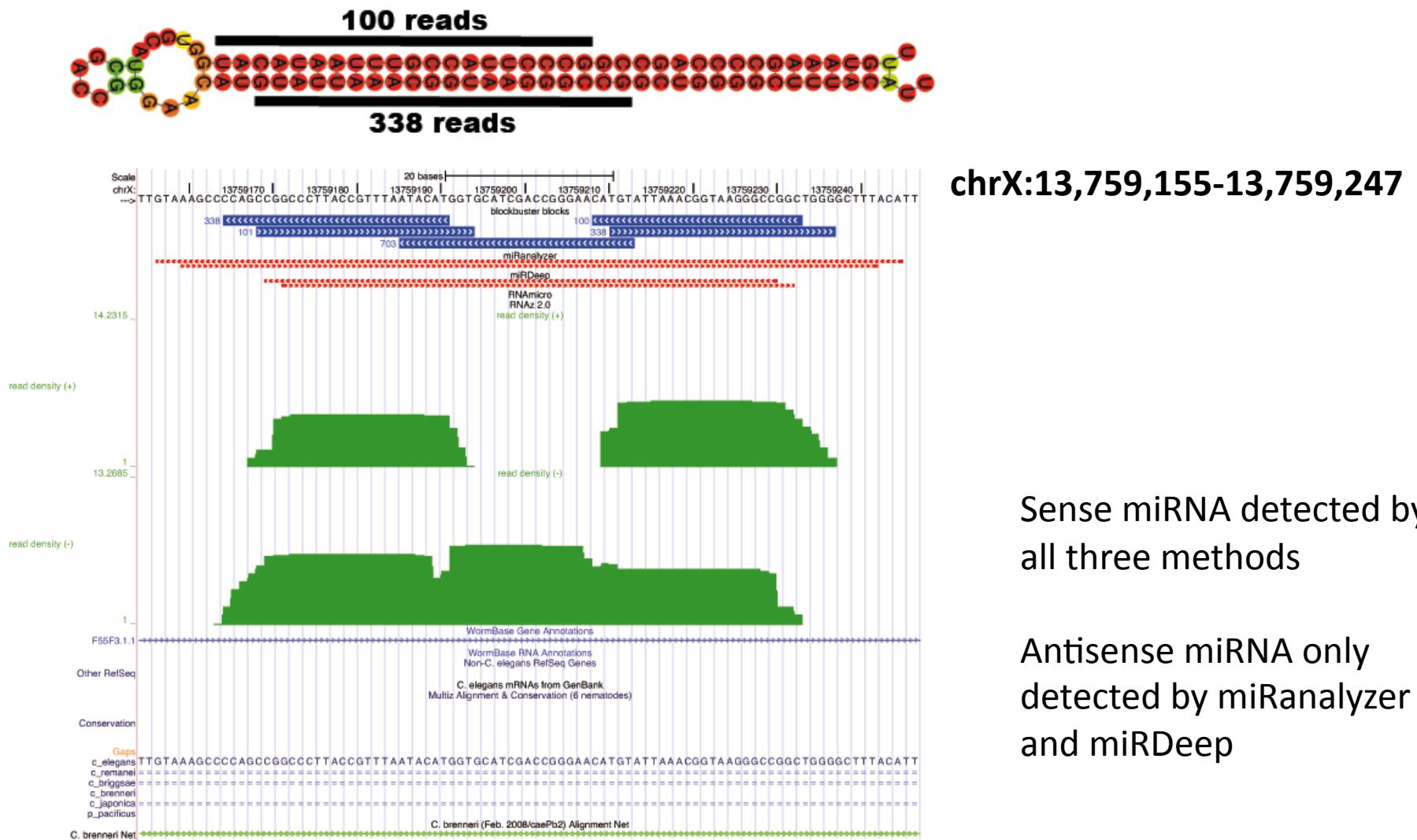


3 candidates found by the classifier, miRanalyser and miRDeep (all sets)



- miRanalyser predictions for all datasets
- No classification (only one block in other datasets)

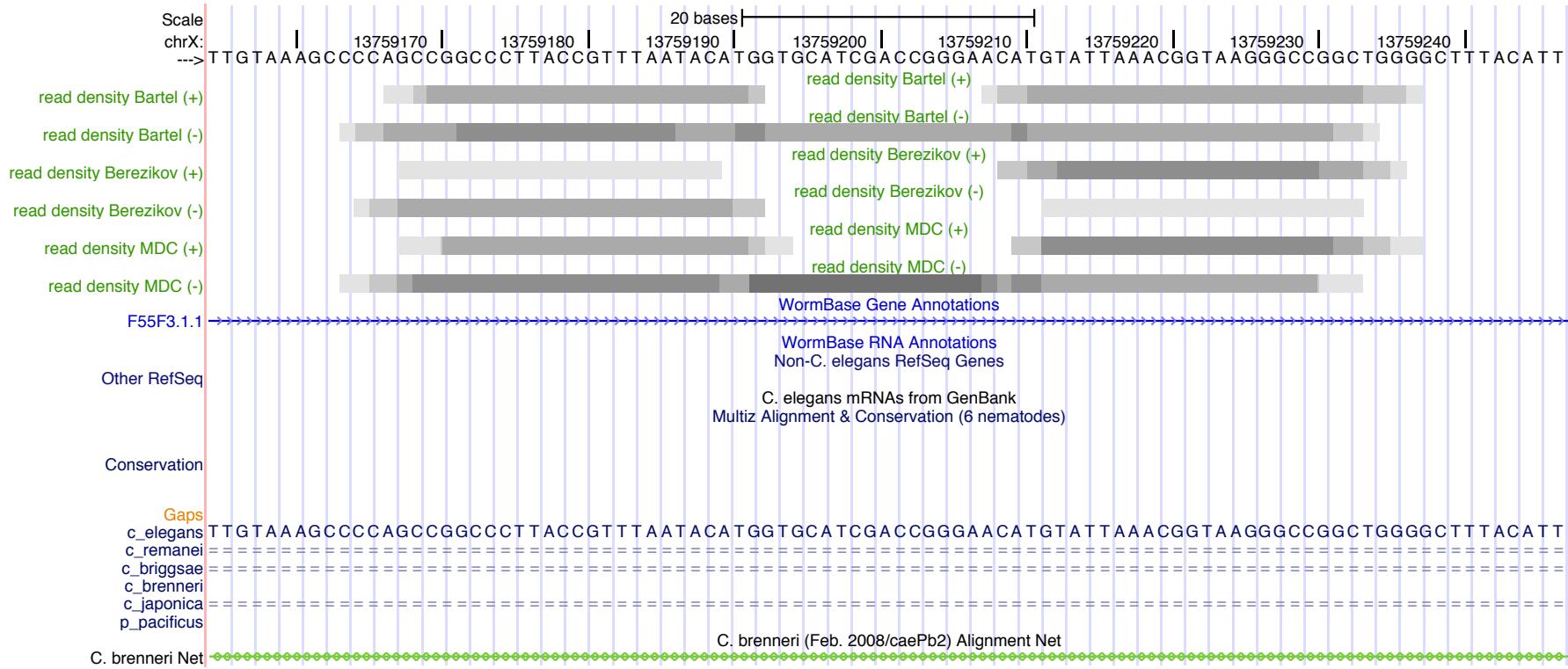
3 candidates found by the classifier, miRanalyser and miRDeep (MDC – set)



Sense miRNA detected by all three methods

Antisense miRNA only detected by miRanalyser and miRDeep

3 candidates found by the classifier, miRanalyser and miRDeep (all sets)



- miRanalyser predictions for all datasets
- classification for sets from Bartel and MDC (no classification for Berezikov)

Conclusion

- There are a lot of not annotated ncRNAs (amongst others microRNAs) waiting to be found by using deep sequencing data
- Classifier can be used between species (patterns seem to be conserved)
- Different experiments (454, Solid, Solexa) generate comparable patterns

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