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### circRNA detection tool benchmarking

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circular RNAs are formed through back-splicing unique circRNA feature: backsplice junction (BSJ)



# circRNAs have the potential to serve as cancer biomarkers



#### circRNA detection based on RNA sequencing data

- step 1: mapping
- step 2: circRNA detection in unmapped reads



### which algorithm to use for circRNA detection?

- > 40 circRNA prediction tools  $\Rightarrow$  lack of extensive orthogonal validation
- differences in
  - mapping tool
  - candidate-based VS segmented read-based approach

and Peter F Stadler<sup>1,2,3,4,9,10,11\*</sup>

- filtering
- ...

#### Resource CirComPara: A Multi-Method Comparative **Bioinformatics Pipeline to Detect and Study** Diverse alternative back-splicing and alternative circRNAs from RNA-seq Data splicing landscape of circular RNAs **Open Access** RESEARCH CrossMar Enrico Gaffo<sup>1</sup>, Annagiulia Bonizzato<sup>2</sup>, Geertruy te Kronnie<sup>2</sup> and Stefania Bortoluzzi<sup>1,\*</sup> Statistically based splicing detection reveals METHOD **Open Access** nt and tissue-specific inducti Circular RNA profile in gliomas revealed by luring human fetal developen CIRI: an efficient and unbiased algorithm for de identification tool UROBORUS han J. Palpant<sup>3</sup>, Peter L. Wang<sup>1</sup>, Nastaran Afari<sup>2</sup>, Chuan Jiar nt<sup>2\*</sup> and Julia Salzman<sup>1</sup> novo circular RNA identification Xiaofeng Song<sup>1</sup>, Naibo Zhang<sup>2</sup>, Ping Han<sup>3</sup>, Byoung-San Moon<sup>2</sup>, Rose K. Lai<sup>4,\*</sup>, Kai Wang<sup>5,\*</sup> and Wange Lu<sup>2,\*</sup> Yuan Gao<sup>1,2†</sup>, Jinfeng Wang<sup>1†</sup> and Fangging Zhao<sup>1\*</sup> Acfs: accurate circRNA Gene expression **Open Access** METHOD identification and quantification A multi-split mapping algorithm for circular circtools—a one-stop software solution for from RNA-Seg data RNA, splicing, trans-splicing and fusion circular RNA research Xintian You<sup>1</sup> & Tim OF Conrad<sup>1,2</sup> detection Tobias Jakobi<sup>1,2,\*</sup>, Alexey Uvarovskii<sup>1,2</sup> and Christoph Dieterich<sup>1,2,\*</sup> marieke.vromman@ugent.be Steve Hoffmann<sup>1,2,3</sup>, Christian Otto<sup>1,2,3</sup>, Gero Doose<sup>1,2,3</sup>, Andrea Tanzer<sup>4</sup>, David Langenberger<sup>1,2,3</sup>, Sabina Christ<sup>5</sup>, Manfred Kunz<sup>6</sup>, Lesca M Holdt<sup>3,7</sup>, Daniel Teupser<sup>3,7</sup>, Jörg Hackermüller<sup>2,5,8</sup>

### large-scale collaborative RT-qPCR benchmarking study of circRNA prediction tools



#### different tools predict a 40-fold difference in nr of circRNAs



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## different circRNA count distributions and filtering for different tools



### large-scale collaborative RT-qPCR benchmarking study of circRNA prediction tools



#### circRNA selection

- random selection of 100 circRNAs per tool
  - 20 circRNAs with BSJ count < 5
  - 80 circRNAs with BSJ count  $\geq$  5
- 1560 seleted circRNAs





### large-scale collaborative RT-qPCR benchmarking study of circRNA prediction tools



#### step 1: qPCR validation



#### step 2: RNase R validation

- validate that the qPCR target is a true circRNA, and not linear RNA
- RNase R degrades linear RNAs
- validation =  $Cq_{treated} Cq_{untreated} > 3$



#### step 2: RNase R validation

validation = 
$$Cq_{treated} - Cq_{untreated} > 3$$



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#### combination of qPCR and RNase R validation rates





#### going back to the first slide

#### extrapolation to total nr of predicted circRNAs

theoretical nr of true positive circRNAs = total nr of detected circRNAs \* combined validation rate



#### conclusions

- work in progress
  - amplicon sequencing data-analysis
  - overlap between tools
  - combination of tools

- circRNA detection tools generate vastly different numbers of circRNAs
- RT-qPCR and RNase R validation rates are generally quite good

