

# Quantification

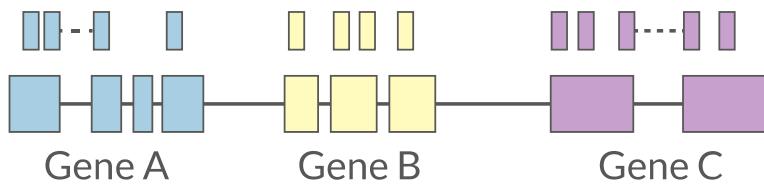
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RNA-seq data analysis

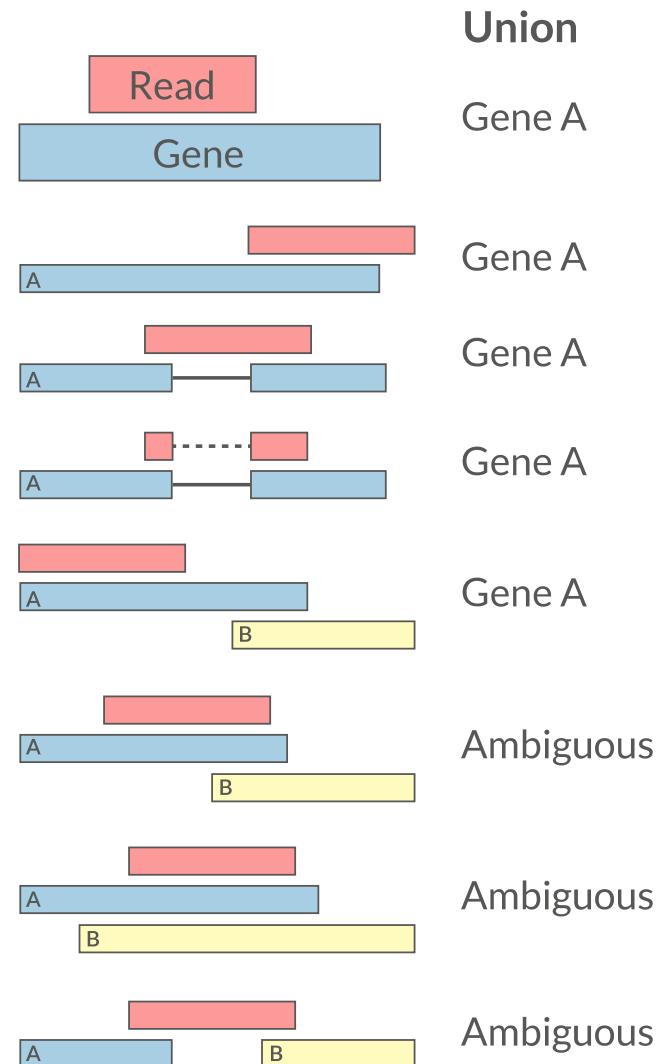
Johan Reimegård | 30-November-2020

# Count the reads

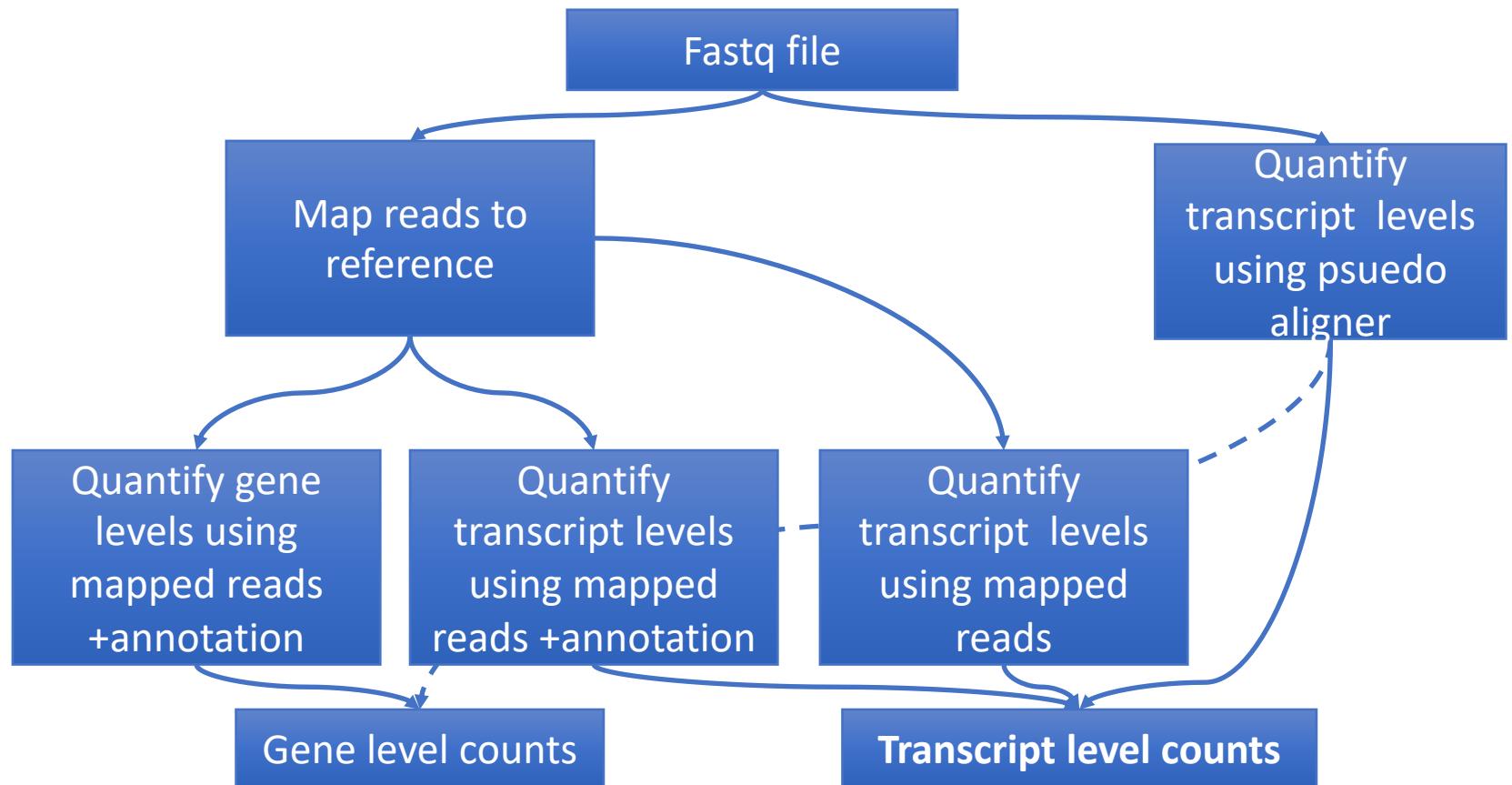
- Read counts = gene expression
- Reads can be quantified on any feature (gene, transcript, exon etc)
- Intersection on gene models
- Gene/Transcript level



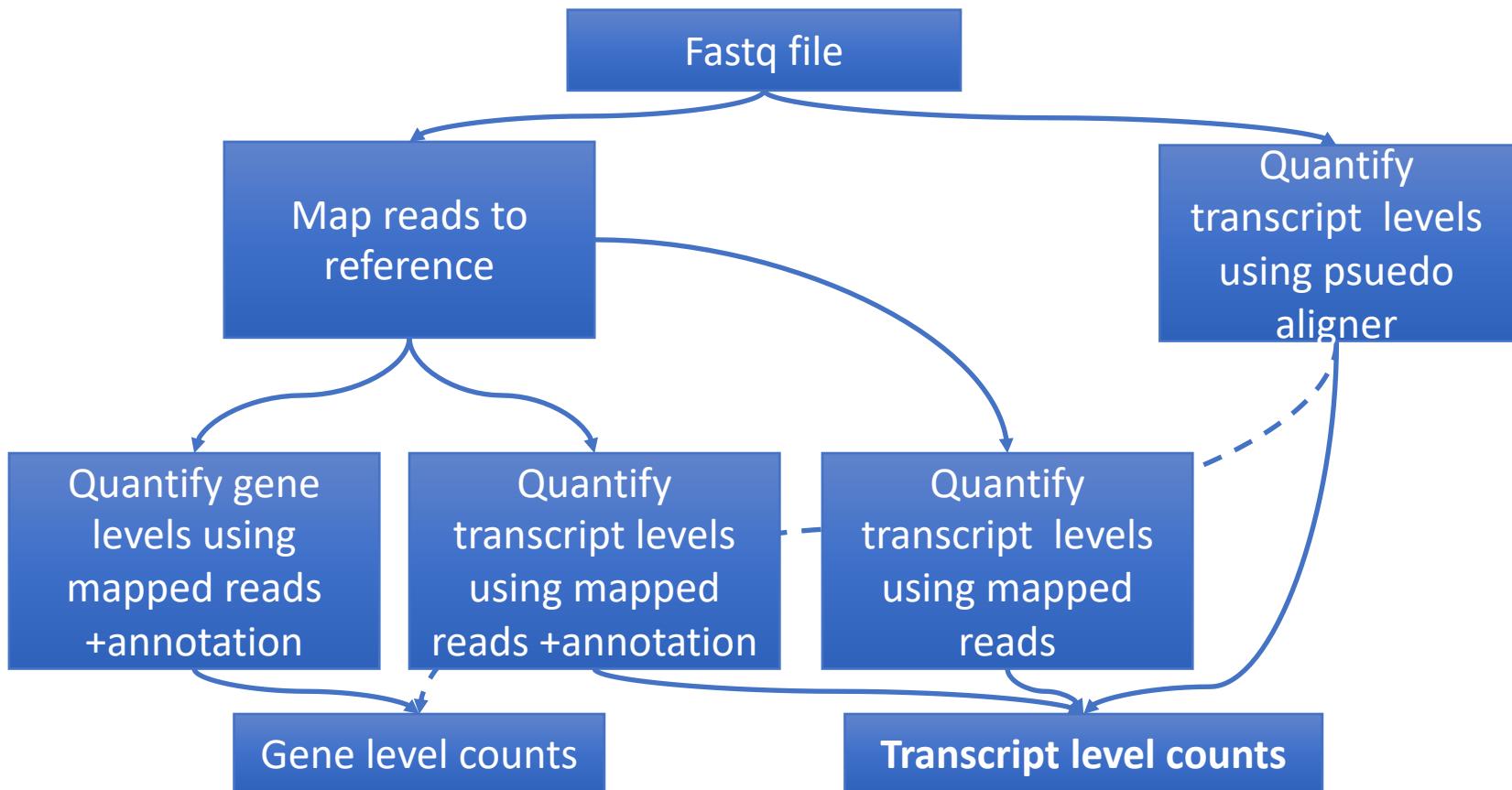
[featureCounts](#), [HTSeq](#)



# Different paths to get a count table



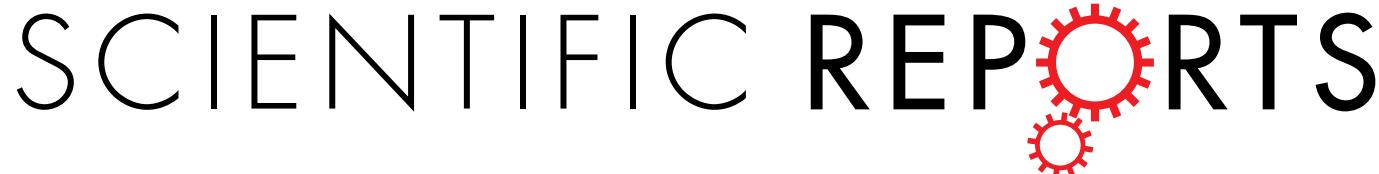
# Good news is that they are all working very well!!



# Gene expression estimates

- Expression estimates on gene level
- Expression estimates on transcript level

# Gene level analysis



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## Benchmarking of RNA-sequencing analysis workflows using whole-transcriptome RT-qPCR expression data

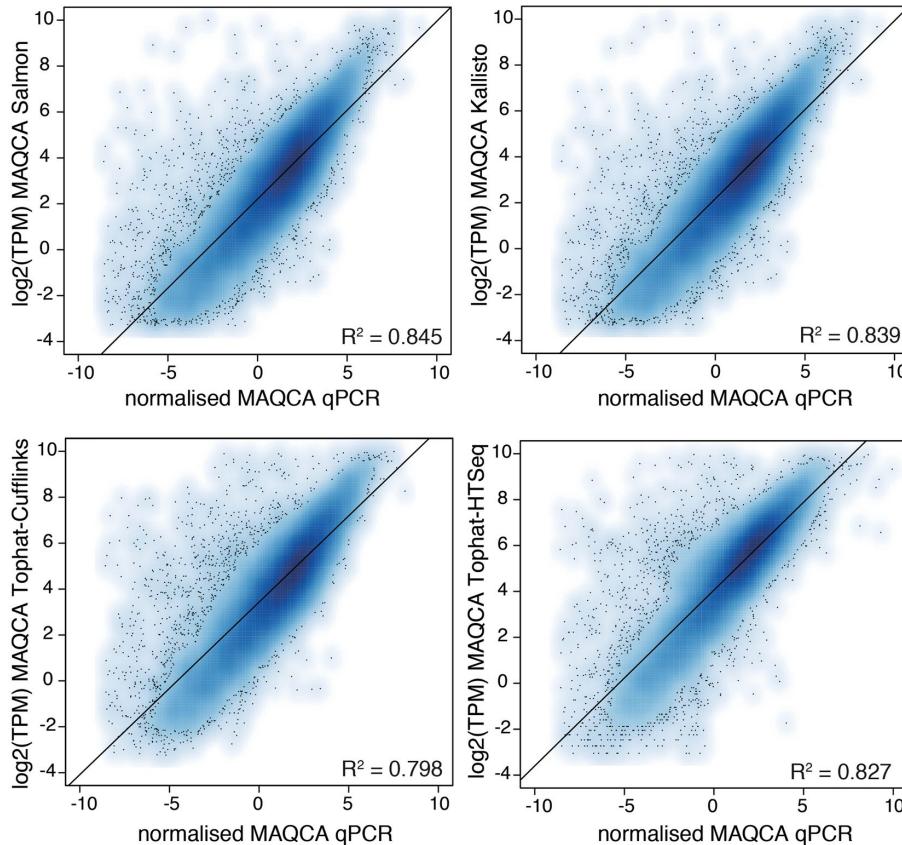
Received: 18 July 2016

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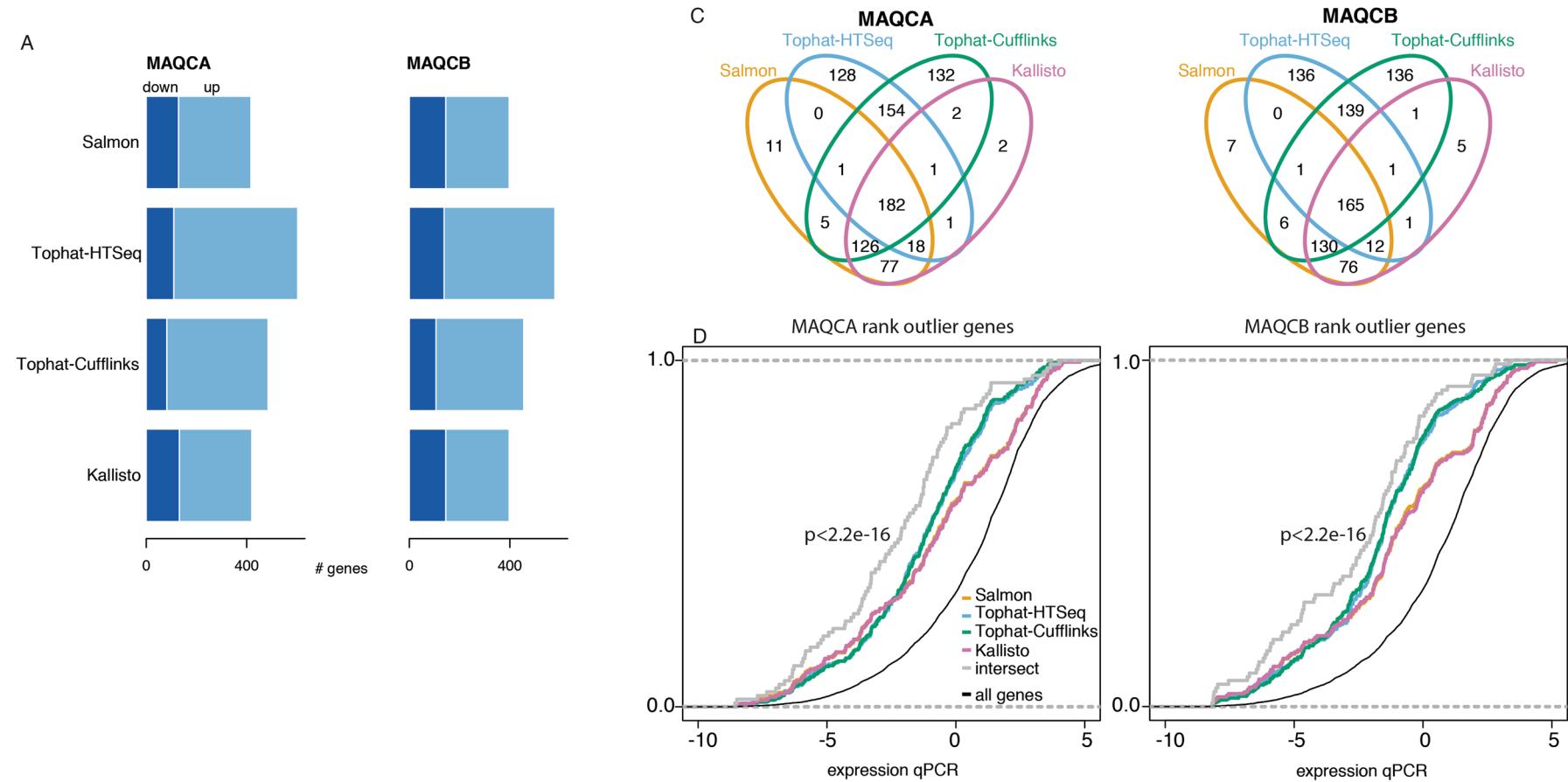
Celine Everaert<sup>1,2,3</sup>, Manuel Luypaert<sup>4</sup>, Jesper L. V. Maag <sup>5</sup>, Quek Xiu Cheng<sup>5</sup>, Marcel E. Dinger <sup>5</sup>, Jan Hellemans<sup>4</sup> & Pieter Mestdagh<sup>1,2,3</sup>

# Expression levels are similar between RT-qPCR and RNA-seq data

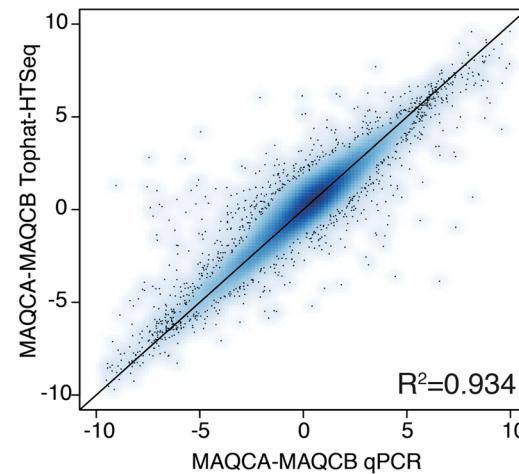
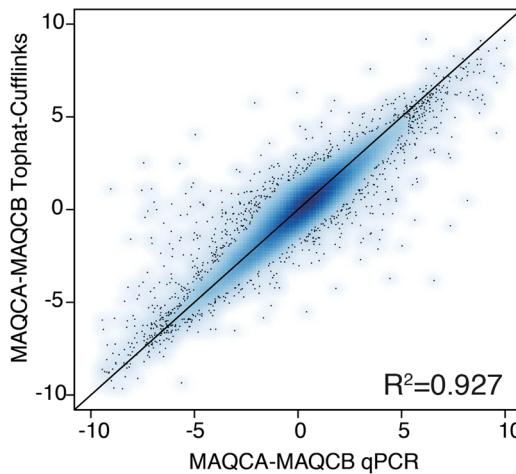
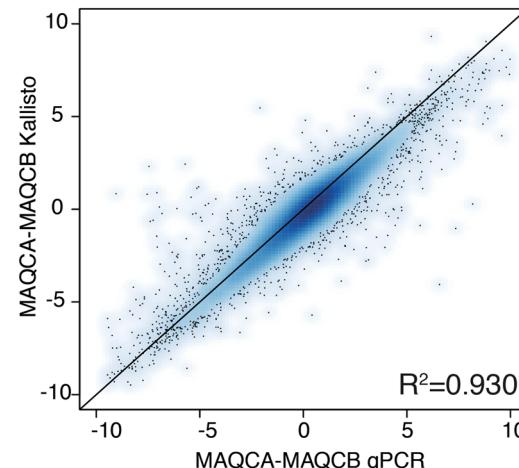
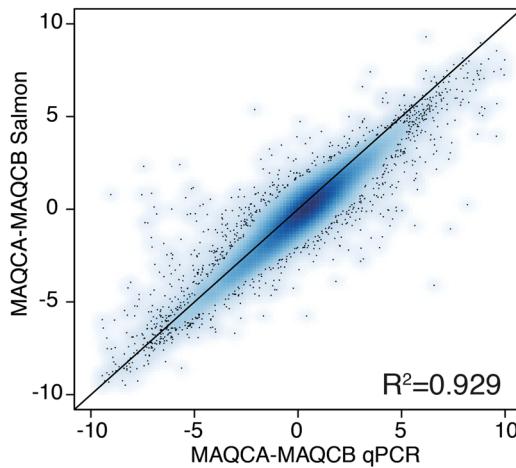


**Figure 1.** Gene expression correlation between RT-qPCR and RNA-seq data. The Pearson correlation coefficients and linear regression line are indicated. Results are based on RNA-seq data from dataset 1.

# Lowly expressed genes are more problematic to identify using RNA seq



Most problems are consistent so they disappear when you do diff-exp analysis



# Transcript level analysis

Zhang *et al.* BMC Genomics (2017) 18:583  
DOI 10.1186/s12864-017-4002-1

BMC Genomics

RESEARCH ARTICLE

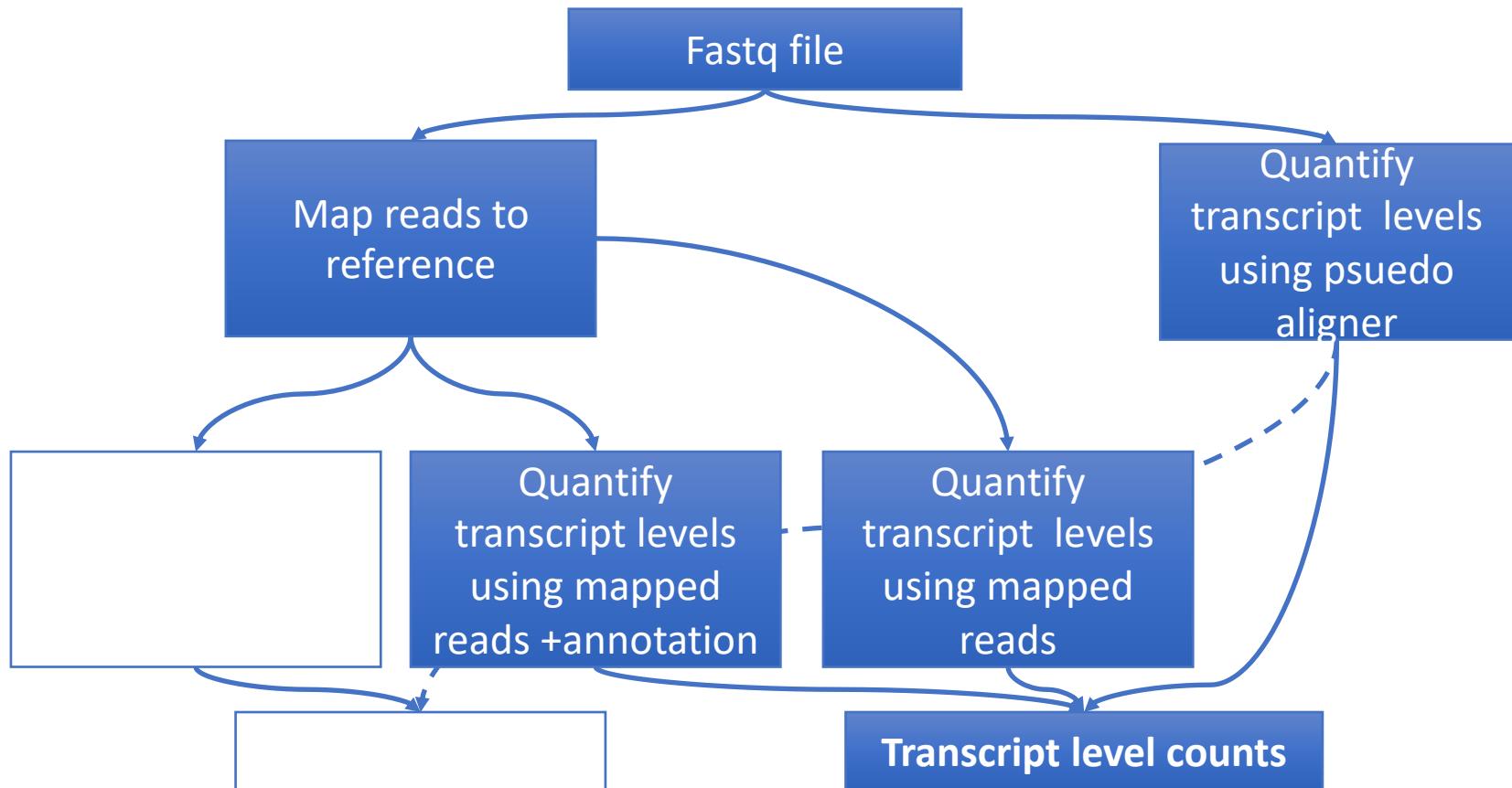
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## Evaluation and comparison of computational tools for RNA-seq isoform quantification

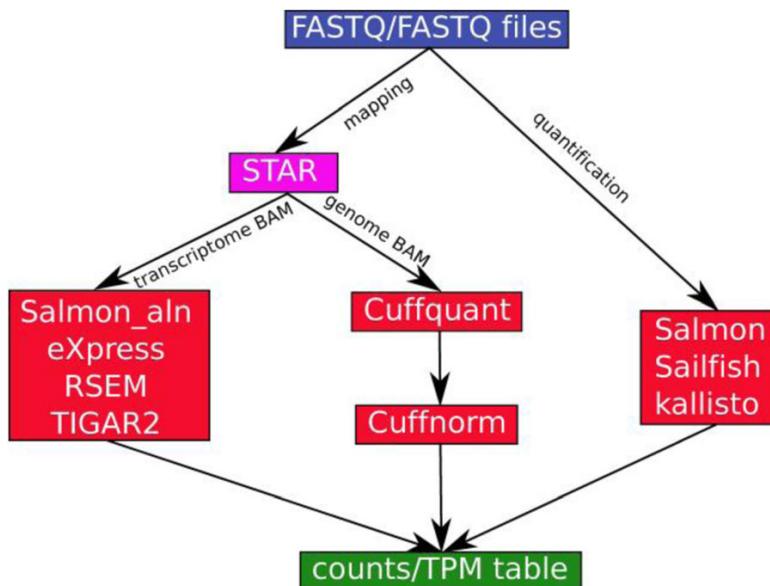


Chi Zhang<sup>1</sup>, Baohong Zhang<sup>1</sup>, Lih-Ling Lin<sup>2</sup> and Shanrong Zhao<sup>1\*</sup>

# Transcript level analysis



# Methods used in paper

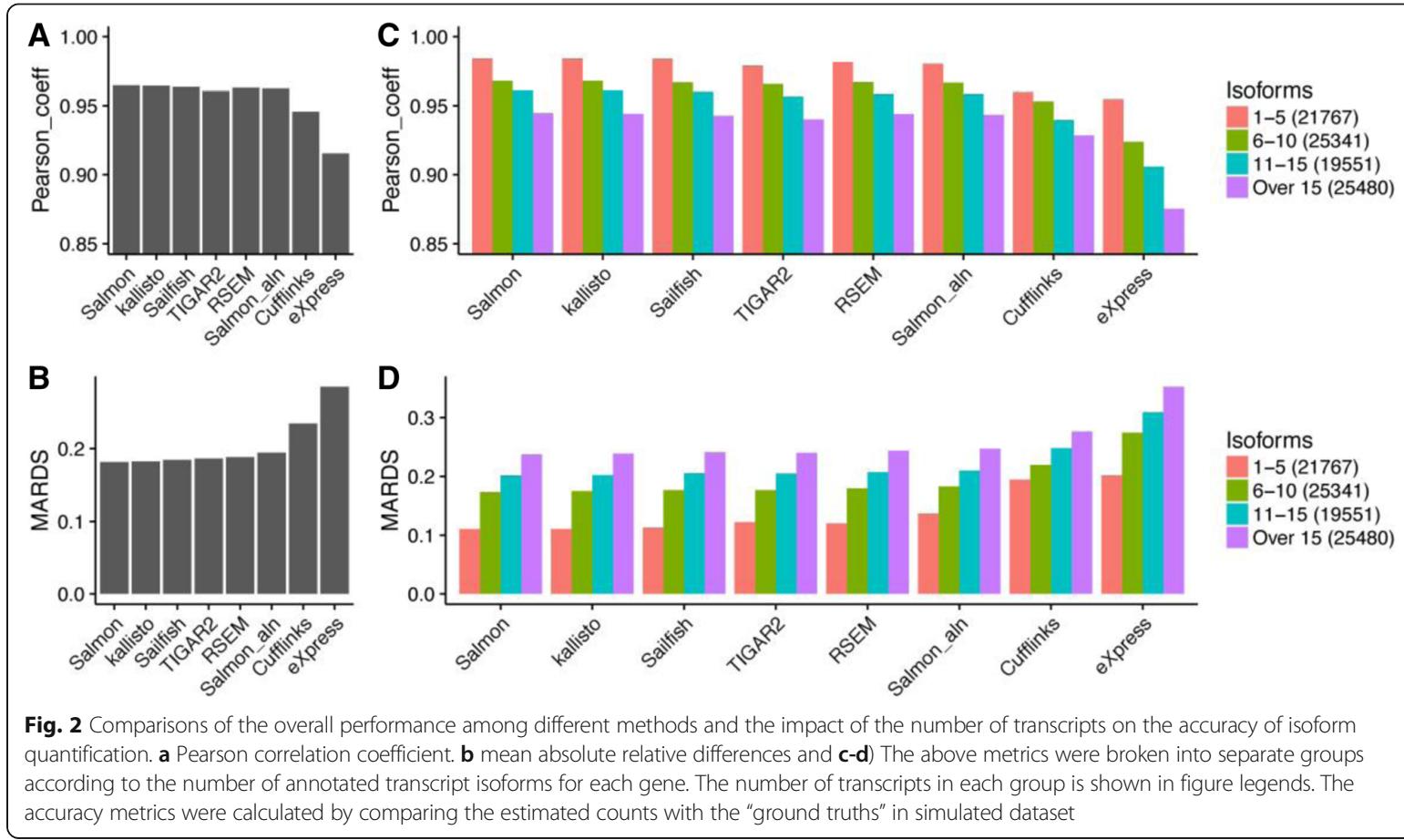


**Table 1** Run time metrics of each method on 50 million paired-end reads of length 76 bp in an high performance computing cluster

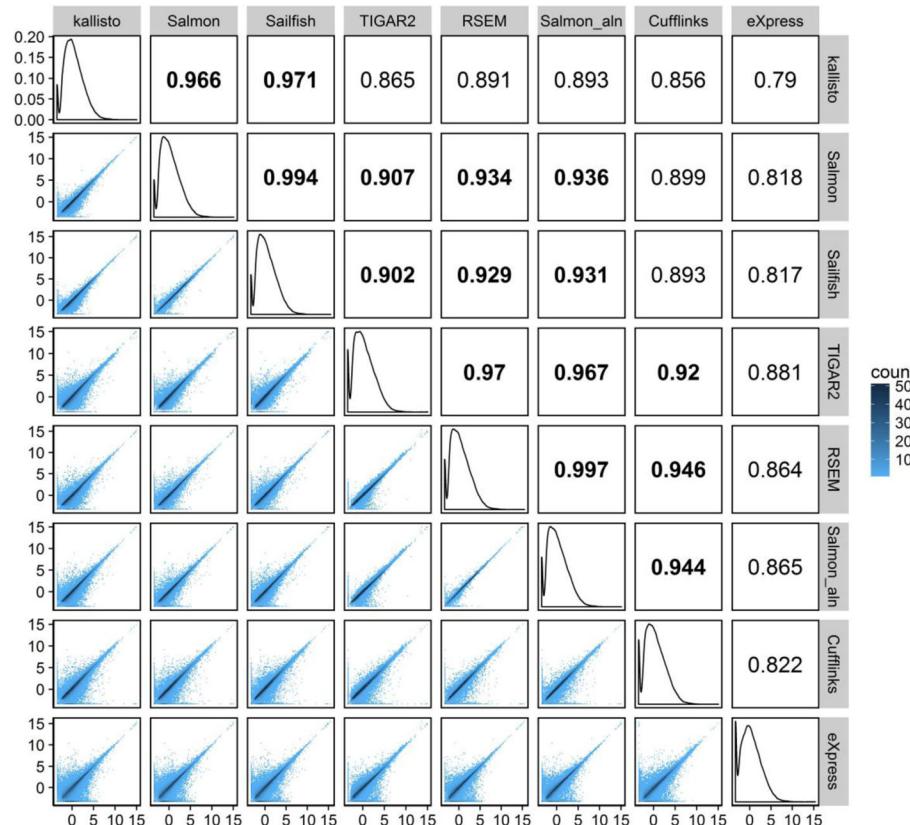
	Memory (Gb)	Run time (min)	Algorithm	Multi-thread
Cufflinks	3.5	117	ML	Yes
RSEM	5.6	154	ML	Yes
eXpress	<u>0.55</u>	30	ML	No
TIGAR2	<b>28.3</b>	<b>1045</b>	VB	Yes
kallisto	3.8	7	ML	Yes
Salmon	6.6	6	VB/ML	Yes
Salmon_aln	3	7	VB/ML	Yes
Sailfish	6.3	<u>5</u>	VB/ML	Yes

For methods that support multi-threading, eight threads were used. For alignment-free methods (Kallisto, Salmon and Sailfish), a mapping step was included. The best performer in each category is underlined and the worst performer is in bold  
ML Maximum Likelihood, VB Variational Bayes

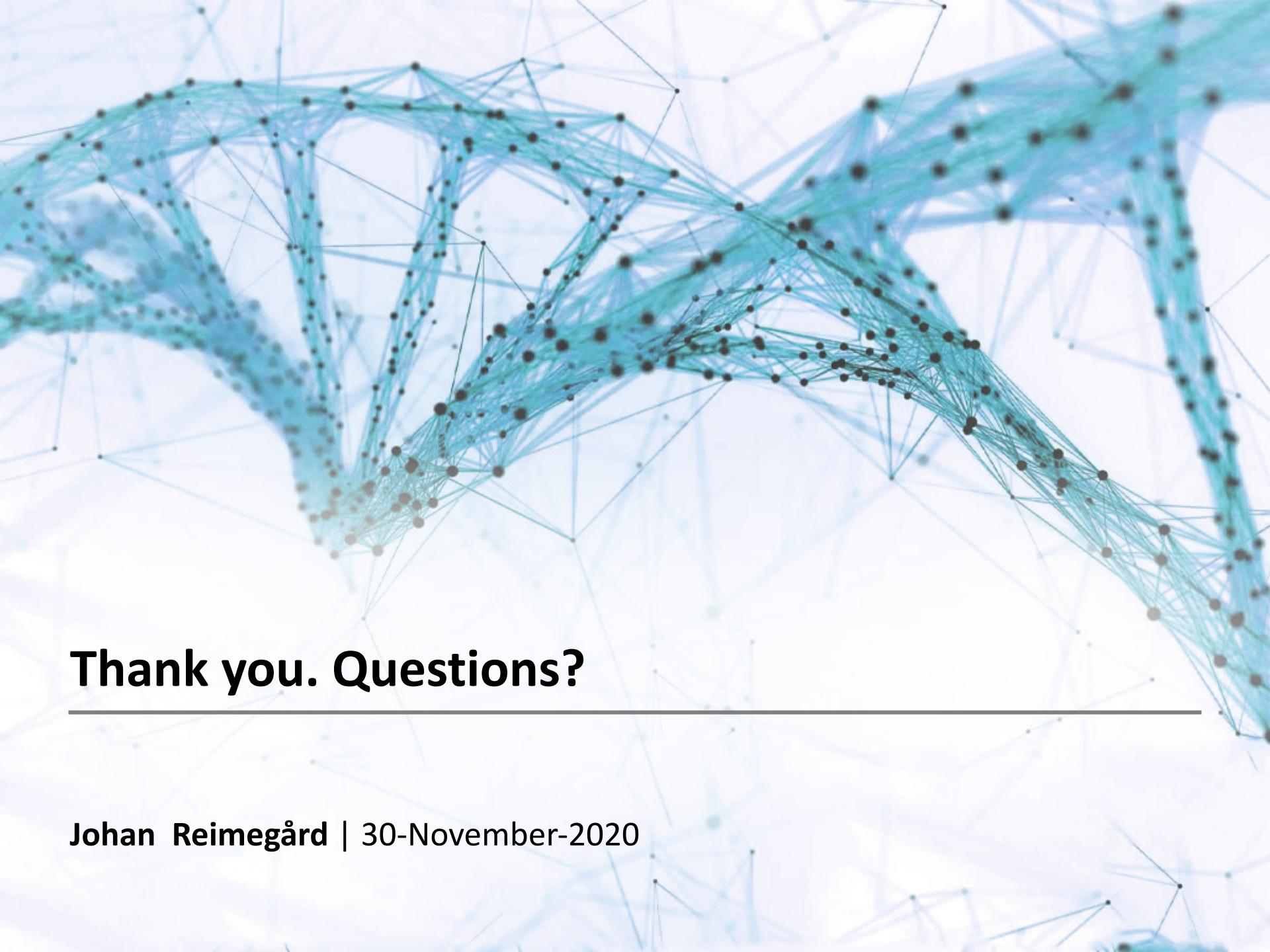
# Isoform quantification problematic for genes with many isoforms



# Results are very similar between methods



**Fig. 5** Pairwise correlation of estimated TPM values for all transcripts between methods for the HBRR-C4 sample. The distribution of transcripts' TPMs from each method was plotted on the diagonal panels. Pairwise density plots and  $R^2$  values are shown in the lower and upper triangular panels, respectively.  $R^2$  values over 0.9 are in *bold*. Methods are grouped using hierarchical clustering

The background of the slide features a complex, abstract network graph. It consists of numerous small, dark brown dots representing nodes, connected by a dense web of thin, translucent blue lines representing edges. The graph is highly interconnected, forming several large, irregular clusters that cover most of the slide's surface.

**Thank you. Questions?**

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Johan Reimegård | 30-November-2020