# Piquant: a pipeline for assessing the performance of transcriptome quantification tools

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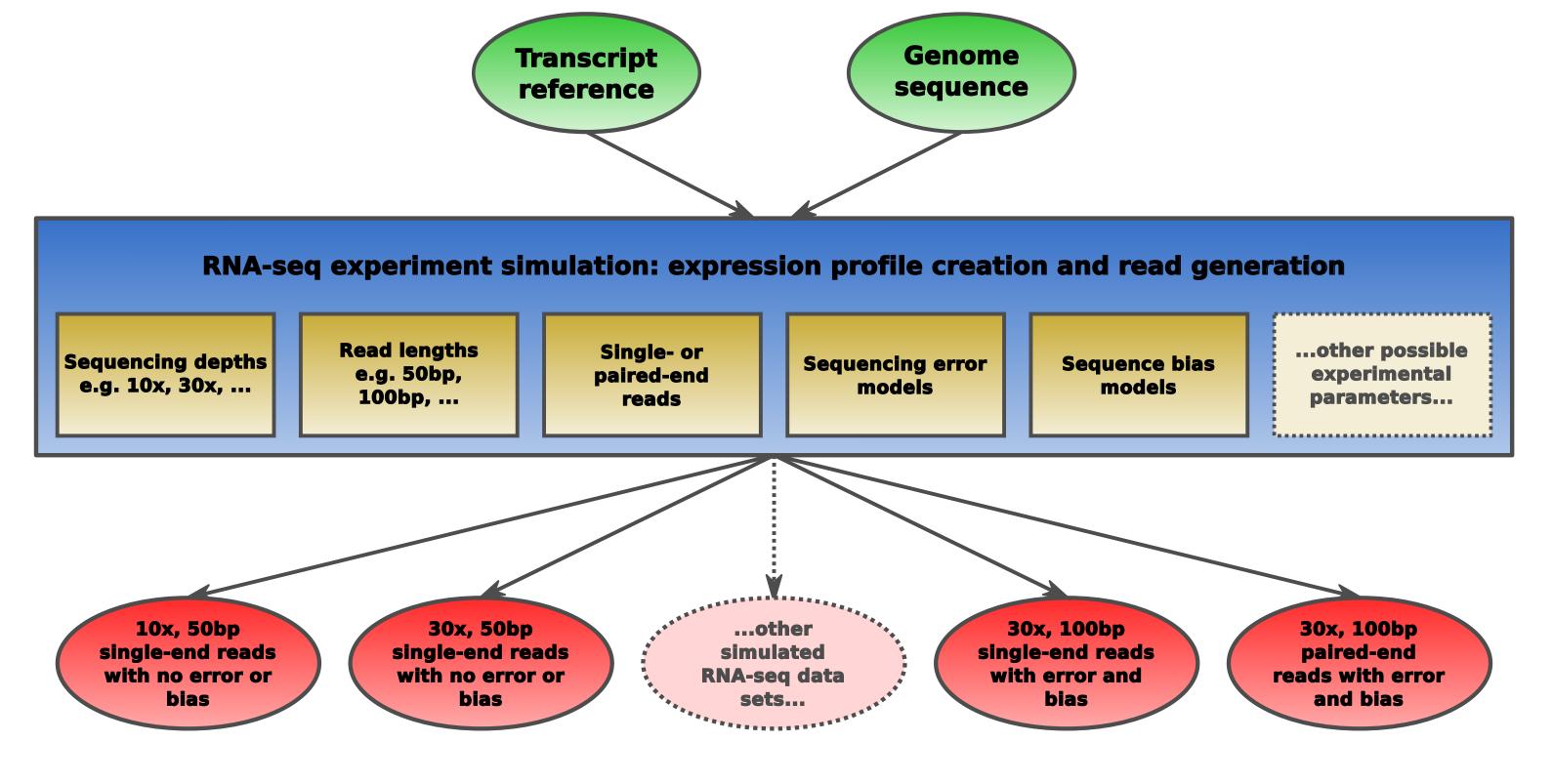
RNA-sequencing has become an important technique in cellular biology for characterising and quantifying the transcriptome, and many computational methods have been developed to reconstruct transcripts from RNA-seq data and then estimate their abundances. Gene expression estimates calculated by these methods have been shown to be relatively robust. However, for transcript quantification, complications arising from the ambiguous origin of short RNA-seq reads and from bias in their sequence composition are compounded, and thus estimates of isoform abundance may be less accurate. It is therefore important to be able to assess the conditions under which different transcriptome quantification tools perform well or more poorly, and how the many optional parameter choices available for each tool may affect their performance.

piquant is a pipeline of Python scripts to help assess the accuracy of transcriptome quantification by such tools. In its first stage, RNA-seq reads are simulated from a starting set of transcripts under specified combinations of sequencing parameters: for example, different read lengths and sequencing depths, single- and paired-end reads, reads with or without sequencing errors, and reads with or without sequence bias. In the second stage, a number of transcriptome quantification tools (or the same tool with different optional parameter choices) estimate isoform abundances for each set of simulated reads. Finally, the isoform expression estimates calculated by each tool for each RNA-seq data set are compared to the known transcript abundances used to generate the reads. The comparative accuracy of estimates calculated by each tool can then be assessed as sequencing parameters are changed, or for different groups of transcripts segregated by particular transcript classification measures, via a range of automatically generated statistics and graphs.

#### Stages of the *piquant* pipeline

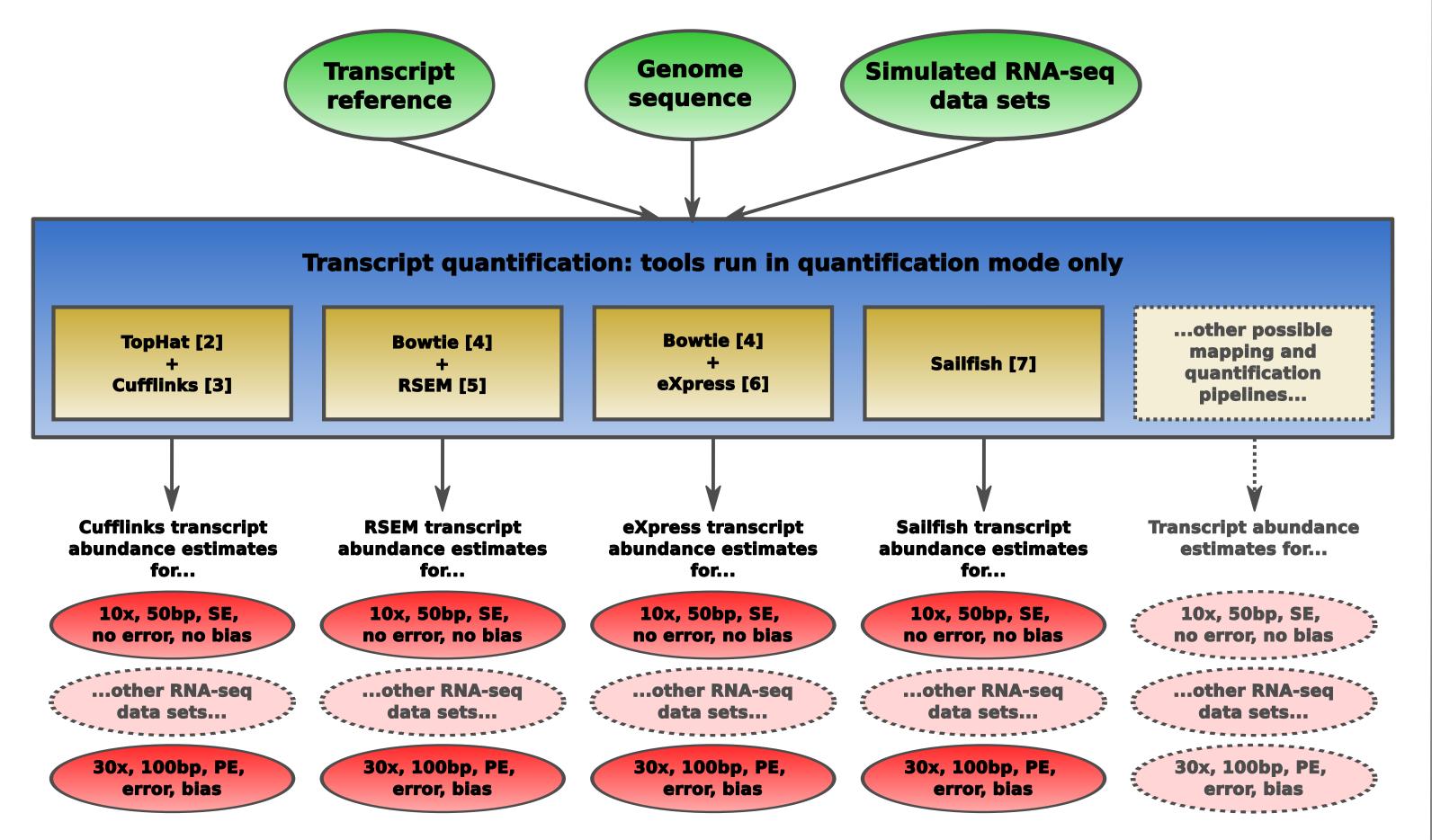
#### 1. Simulation

piquant uses the Flux Simulator [1] RNA-seq experiment simulator to generate sets of RNA-Seq data.



#### 2. Quantification

For each set of RNA-seq data, simulated reads are mapped to the genome or transcriptome, and quantification tools estimate transcript abundances.



# 3. Analysis

References

Biology, 12:R22 (2011).

2. Kim et al., *Genome Biology*, **14**:R36 (2013).

4. Langmead et al., Genome Biology, 10:R25 (2009).

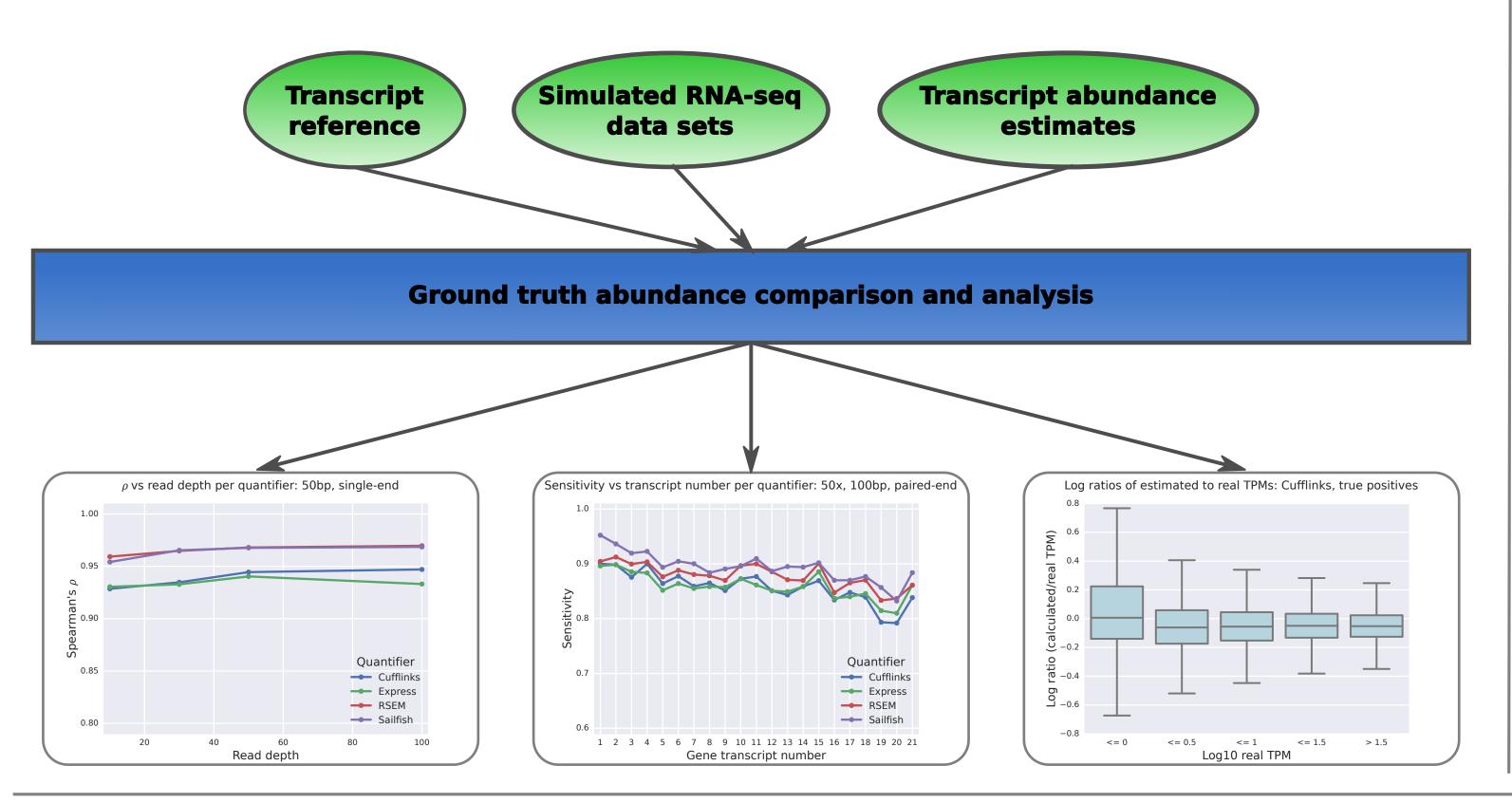
5. Li and Dewey, BMC Bioinformatics, 12:323 (2011).

6. Roberts and Pachter, Nature Methods, 10, 71-73 (2013).

7. Patro et al., *Nature Biotechnology*, **32**, 462–464 (2014).

8. Shi et al., *Nature Biotechnology*, **24**, 1151–1161 (2006).

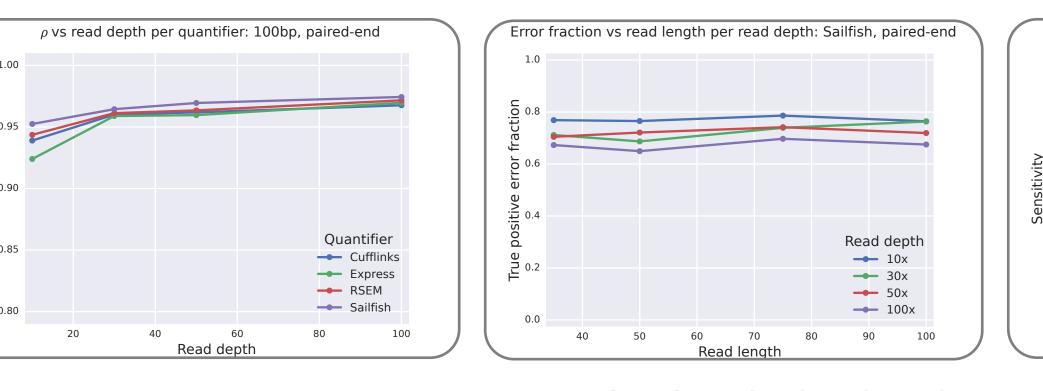
For each set of RNA-seq data and each quantification tool, estimated transcript abundances are compared with the ground truth values for accuracy, and a rich collection of statistics and graphs are automatically generated.



3. Trapnell et al., Nature Biotechnology, 28, 511–515 (2010); Roberts et al., Genome

#### **Statistics**

piquant calculates a range of statistics and draws graphs to aid the assessment of transcript quantification performance. Additional statistics, and their resultant graphs, are easily added with just a few lines of code.



**Spearman's**  $\rho$ : the rank correlation coefficient between estimated and ground truth abundances

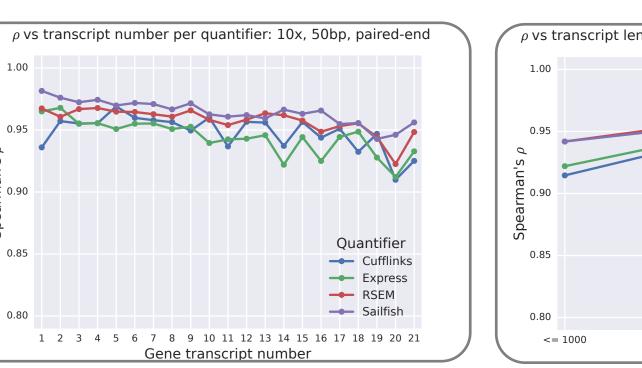
Sensitivity vs read depth per read length: RSEM, paired-end

**Error fraction: the fraction of** transcripts whose estimated abundance is >10% higher or lower than the ground truth value

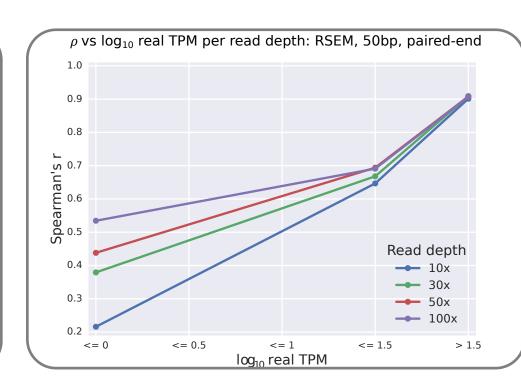
Sensitivity and specificity: the proportion of transcripts correctly identified as being present or not

#### **Classification of transcripts**

Classifiers split the whole set of input transcripts into discrete groups sharing similar properties. Such divisions allow quantification performance to be assessed across different types of transcript. Again, additional classifiers are added with a few lines of code.



**Transcript length:** Quantification accuracy is better for long transcripts than for



shorter isoforms

**Transcript prevalence: Quantification accuracy is better** for isoforms with high abundance than rare, lowly-expressed transcripts

# **Quantification tools**

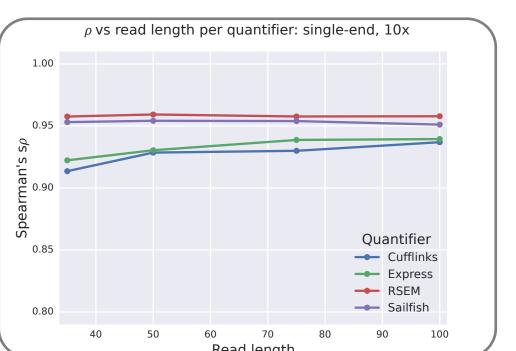
**Isoforms per gene:** 

**Quantification accuracy decreases** 

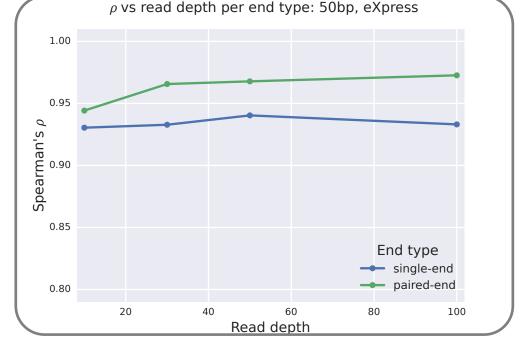
as the number of transcripts per

gene grows

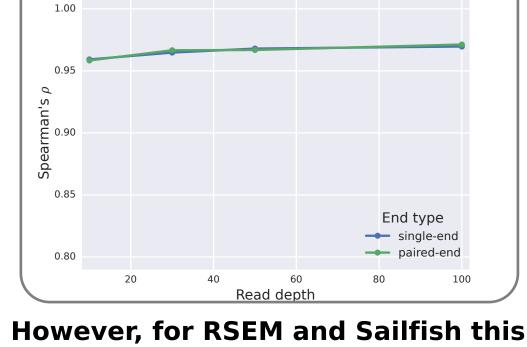
Out of the box, piquant runs four mapping and quantification pipelines: TopHat+Cufflinks, Bowtie+RSEM, Bowtie+eXpress, and the alignment-free Sailfish algorithm. Additional pipelines are easy to add, or the current configurations edited to test alternate optional parameter choices.



piquant allows the relative strengths of quantification pipelines to be assessed as sequencing parameters change



For Cufflinks and eXpress, pairedend reads improve quantification accuracy over single-end reads



benefit is less consistent or not present

# **Challenges and next steps**

An abiding risk when using simulated data is its potential failure to truly capture the characteristics of real experiments: here, the analyses produced by *piquant* will be of limited value if their verdicts about quantification tool performance fail to translate to the real world. An additional problem in the assessment of transcript quantification from RNAseq is the relative paucity of gold-standard experimental data by which to tests our conclusions. One such data set, produced as part of the MAQC project [8], consists of ~1000 gene abundances measured by TaqMan PCR for two reference RNA samples.

While noting that the relation between quantification accuracy at the level of genes and transcripts is itself not necessarily straightforward, we compared the performance of quantification tools on both simulated and TaqMan-validated RNA-seq data. In consequence, we will next investigate the effect on quantification of the following improvements to the RNA-seq simulation model:

- a biologically-realistic distribution of gene abundances
- reads arising from transcripts not present in the reference
- reads arising from chimeric transcripts

#### Links 1. Griebel et al., Nucleic Acids Research, 40 (20): 10073-10083 (2012).

piquant code repository



# piquant documentation

# **Neuroregulatory Genomics**





http://piquant.readthedocs.org/en/latest/index.html

http://neuroregulatorygenomics.org