

Genome analysis

TPMCalculator: one-step software to quantify mRNA abundance of genomic features

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Abstract

Summary: The quantification of RNA sequencing (RNA-seq) abundance using a normalization method that calculates transcripts per million (TPM) is a key step to compare multiple samples from different experiments. TPMCalculator is a one-step software to process RNA-seg alignments in BAM format and reports TPM values, raw read counts and feature lengths for genes, transcripts, exons and introns. The program describes the genomic features through a model generated from the gene transfer format file used during alignments reporting of the TPM values and the raw read counts for each feature. In this paper, we show the correlation for 1256 samples from the TCGA-BRCA project between TPM and FPKM reported by TPMCalculator and RSeQC. We also show the correlation for raw read counts reported by TPMCalculator, HTSeq and featureCounts.

Availability and implementation: TPMCalculator is freely available at https://github.com/ncbi/ TPMCalculator. It is implemented in C++14 and supported on Mac OS X, Linux and MS Windows.

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Supplementary information: Supplementary data are available at Bioinformatics online.

1 Introduction

Next-Generation Sequencing technologies are changing the way we analyze biological systems. RNA sequencing (RNA-seq) has become a standard procedure. Most RNA-seq experiments measure and compare transcript abundance between samples, which is a critical step for analyzing gene expression profiles under varying experimental conditions.

In 2008, Mortazavi et al. introduced a normalization method designed to measure mRNA abundance and named the measure 'reads per kilobase of exon model per million mapped reads' (RPKM) (Mortazavi et al., 2008). This was later modified by Trapnell et al. by an alternative method named 'fragments per kilobase of transcript per million fragments sequenced' (FPKM) (Trapnell et al., 2010). Although both RPKM and FPKM offer practical ways to quantify mRNA abundance while comparing genomic features in the same sample, they may present biased values when comparing multiple samples (Wagner et al., 2012). Due to this inconsistence, and the fact that RNA-Seq data analysis is more useful when comparing multiple samples from different experimental conditions, Wagner et al. introduced an alternative quantity to RPKM and FPKM named 'transcripts per million' (TPM) that corrects the inconsistences while comparing the RNA-seq abundance among independent samples.

Despite the theoretical and empirical demonstration that the units of mRNA abundance in terms of RPKM or FPKM differ between samples (Wagner et al., 2012), the most popular computational tools used by the research community still quantify the RNA-Seq abundance in terms of RPKM or FPKM. Although, there are emergent computational tools integrating TPM calculations in their pipelines, such as Salmon (Patro et al., 2017), the application is limited only to transcripts and cannot be used to estimate abundance of any other genomic features. Researchers, who would like to use

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TPM for other quantifications, need to implement their own scripts to calculate TPM values from raw read counts. This process requires the use of third-party software to calculate the raw read counts by introducing an additional step in the workflow pipeline. This is prone to errors due to inconsistences on read assignment models and the changing definition of genomic features in annotated databases such as GenBank and RefSeq (Coordinators, 2018).

Considering the value of the RNA-Seq abundance quantification and the lack of computational tools to process BAM files and calculate accurate TPM values directly from the alignments, we have developed a software package named TPMCalculator.

2 Materials and methods

TPMCalculator quantifies mRNA abundance directly from the alignments by parsing BAM files. The input parameters are the same gene transfer format (GTF) file used to generate the alignments, and one or multiple input BAM file(s) containing either single-end or paired-end sequencing reads. The TPMCalculator output is comprised of five files per sample reporting the TPM values and raw read counts for genes, transcripts, exons and introns.

The model to describe the genomic features used for a gene is created from the GTF provided by the user. TPMCalculator performs two transformations which are executed on the genomic coordinates generating regions for the genes that include the exons and 'pure' intron regions as shown in Supplementary Figure S1. The first transformation creates overlapped exons for all alternative spliced forms of the gene. A single gene model is generated with unique exons and introns which includes the sequence of all exonic regions. The second transformation creates a list of pure intron regions that replace those generated by the first transformation. We should emphasize that only the intron regions included are from regions that are not from overlapping exons of other genes. Reporting TPM values for these unique introns permits further identification of alternative splicing events such as intron retention. Additionally, a set of non-overlapped gene features (exons and introns) are generated and used for TPM calculation.

3 Results

TPMCalculator is a one-step software package to quantify mRNA abundance for several genomic features including genes, transcripts, exons and introns. The program processes RNA-Seq alignments in BAM file format producing text files with TPM values, raw read counts and feature lengths for each genomic feature.

To validate our software, we calculate the Pearson correlation coefficient between TPM and FPKM for normalized expression values using RNA-Seq data of 1256 samples from the TCGA-BRCA project (Koboldt *et al.*, 2012). The FPKM values were calculated using the RSeQC package (Wang *et al.*, 2012) as described in the Supplementary Material.

Figure 1 shows the correlation coefficients obtained in this comparison where 98.6% (1238 samples) correlated above 0.8 for the MAPQ = 0. Though, for the rest of MAPQ values, all samples were correlated with a correlation coefficient above 0.8 with the exception of the MAPQ = 255 where one sample correlated with 0.69.

Additionally, the correlation coefficients were calculated for the raw reads counts reported by TPMCalculator, HTSeq (Anders *et al.*, 2015) and featureCounts (Liao *et al.*, 2014). The correlation coefficient between raw read counts between TPMCalculator and HTSeq

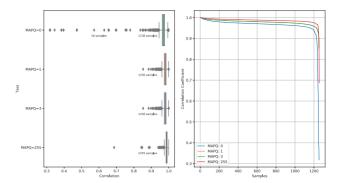


Fig. 1. Correlation coefficient calculated for 1256 samples between TPMCalculator (ExonTPM) and RSeQC FPKM

was above 0.9 for 99.2% (1246) samples. Only 10 samples showed no correlation with a coefficient below 0.2. TPMCalculator and featureCounts correlation coefficients were above 0.99 for 99.9% of the samples. Only one sample showed no correlation with correlation coefficient = 0.1. Samples with low correlation are shown in Supplementary Table S3.

TPMCalculator reports, in one single analysis, raw read counts and TPM values for genes, transcripts, exons and introns. Currently available tools are unable to generate a complete set of data for all genomic features. Additionally, TPMCalculator reduces the compute time and the resource requirements of RNA-Seq pipelines by eliminating several steps. TPMCalculator processes BAM files of size 7.0 GB in $\sim\!20$ min requiring only 4 GB of RAM. This is an open source project available on the NCBI GitHub repository at https://github.com/ncbi/TPMCalculator.

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Conflict of Interest: none declared.

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