Pre-Processing for the Zebrafish RNA-Seq Gene-Level Counts

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This vignette describes the pre-processing steps that were followed for the generation of the gene-level read counts contained in the *Bioconductor* package *zebrafishRNASeq*.

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1 Sample preparation and sequencing

Olfactory sensory neurons were isolated from three pairs of gallein-treated and control embryonic zebrafish pools and purified by fluorescence activated cell sorting (FACS) [1]. Each RNA sample was enriched in poly(A)+ RNA from 10–30 ng total RNA and 1 μ L (1:1000 dilution) of Ambion ERCC ExFold RNA Spike-in Control Mix 1 was added to 30 ng of total RNA before mRNA isolation. cDNA libraries were prepared according to manufacturer's protocol. The six libraries were sequenced in two multiplex runs on an Illumina HiSeq2000 sequencer, yielding approximately 50 million 100bp paired-end reads per library.

2 Read alignment and expression quantitation

We made use of a custom reference sequence, defined as the union of the zebrafish reference genome (Zv9, downloaded from Ensembl [2], v. 67) and the ERCC spike-in sequences (http://tools.invitrogen.com/downloads/ERCC92.fa). Reads were mapped with TopHat [3] (v. 2.0.4), with the following parameters,

--library-type=fr-unstranded -G ensembl.gtf --transcriptome-index=transcript --no-novel-juncs where ensembl.gtf is a GTF file containing Ensembl gene annotation.

Gene-level read counts were obtained using the htseq-count python script [4] in the "union" mode and Ensembl (v. 67) gene annotation.

After verifying that there were no run-specific biases, we used the sums of the counts of the two runs as the expression measures for each library.

3 Loading the zebrafish data into R

To load the gene-level read counts into R, simply type

```
library(zebrafishRNASeq)
data(zfGenes)
head(zfGenes)
##
                      Ctl1 Ctl3 Ctl5 Trt9 Trt11 Trt13
## ENSDARG0000000001
                       304
                            129
                                  339
                                       102
                                              16
                                                   617
## ENSDARG00000000002
                       605
                            637
                                  406
                                        82
                                             230
                                                  1245
## ENSDARG0000000018 391
                            235
                                  217
                                       554
                                             451
                                                   565
## ENSDARG0000000019 2979 4729 7002 7309
                                            9395
                                                  3349
## ENSDARG0000000068
                        89
                            356
                                   41
                                       149
                                              45
                                                    44
## ENSDARG00000000069 312 184 844
                                       269
                                             513
                                                   243
```

The ERCC spike-in read counts are in the last rows of the same matrix and can be retrieved in the following way.

```
spikes <- zfGenes[grep("^ERCC", rownames(zfGenes)),]</pre>
head(spikes)
##
                 Ctl1
                        Ct13
                               Ct15
                                       Trt9
                                             Trt11 Trt13
## ERCC-00002 97227
                       38556
                              68367 148331 169360 100974
                                             21184 21841
## ERCC-00003 10925
                        6240
                              11156
                                     36652
## ERCC-00004 379182 179870 256130 679783 529085 311169
                 2452
                        1183
                                1042
                                       1895
                                               3520
## ERCC-00009
                                   0
                                          0
## ERCC-00012
                    0
                           0
                                                  0
## ERCC-00013
                   89
                           8
                                   0
                                        205
                                                21
                                                         3
```

The typical use of this dataset is the indentification of differentially expressed genes between control (Ctl) and treated (Trt) samples. For additional details, exploratory analysis, and normalization of the zebrafish data see [5, 6]. The data are used as a case study for the *Bioconductor* package *RUVSeq*.

4 Session info

toLatex(sessionInfo())

- R version 3.3.0 (2016-05-03), x86_64-pc-linux-gnu
- Locale: LC_CTYPE=en_US.UTF-8, LC_NUMERIC=C, LC_TIME=en_US.UTF-8, LC_COLLATE=C, LC_MONETARY=en_US.UTF-8, LC_MESSAGES=en_US.UTF-8, LC_PAPER=en_US.UTF-8, LC_NAME=C, LC_ADDRESS=C, LC_TELEPHONE=C, LC_MEASUREMENT=en_US.UTF-8, LC_IDENTIFICATION=C
- Base packages: base, datasets, grDevices, graphics, methods, stats, utils
- Other packages: zebrafishRNASeq 0.106.2
- Loaded via a namespace (and not attached): BiocStyle 2.0.2, evaluate 0.9, formatR 1.4, highr 0.6, knitr 1.13, magrittr 1.5, stringi 1.0-1, stringr 1.0.0, tools 3.3.0

References

[1] T. Ferreira, S. R. Wilson, Y. G. Choi, D. Risso, S. Dudoit, T. P. Speed, and J. Ngai. Silencing of odorant receptor genes by G Protein $\beta\gamma$ signaling ensures the expression of one odorant receptor per olfactory sensory neuron. *Neuron*, 81:847–859, 2014.

- [2] P. Flicek, M. R. Amode, D. Barrell, K. Beal, S. Brent, D. Carvalho-Silva, P. Clapham, G. Coates, S. Fairley, S. Fitzgerald, et al. Ensembl 2012. *Nucleic Acids Research*, 40(D1):D84–D90, 2012.
- [3] C. Trapnell, L. Pachter, and S. L. Salzberg. TopHat: discovering splice junctions with RNA-Seq. *Bioinformatics*, 25(9):1105–1111, 2009.
- [4] S. Anders, P. T. Pyl, and W. Huber. HTSeq A Python framework to work with high-throughput sequencing data. bioRxiv preprint, 2014. doi:10.1101/002824.
- [5] D. Risso, J. Ngai, T.P. Speed, and S. Dudoit. Using controls for the normalization of RNA-Seq data. *Nature Biotechnology*, 2014. Accepted.
- [6] D. Risso, J. Ngai, T.P. Speed, and S. Dudoit. The role of spike-in standards in the normalization of RNA-seq. In D. Nettleton and S. Datta, editors, *Statistical Analysis of Next Generation Sequence Data*. Springer, 2014.