



Annotated Example of using RNA-Seq Pipeline for LINCS

Introduction

This pipeline describes a reproducible analysis of RNA-Seq data using R Markdown. We use an example data set from Grifford et al., that contains seven libraries. We took two samples of HCC1419 cells that were untreated and another two treated with Lapatinib for only 9 days before harvesting (drug tolerant persisters, DTPs). The raw data (GSE62074) is downloaded from the Sequence Read Archive (SRA) database.

In this document, all necessary softwares and R packages are already installed in the system and all the FASTQ files are downloaded, unzipped, and saved in the input directory. In this particular example, feature is gene, species is Homo Sapiens and we consider single-end reads. The following steps are used in this pipleine to analyze the raw sequence data.

LINCS RNA-Seq Pipeline

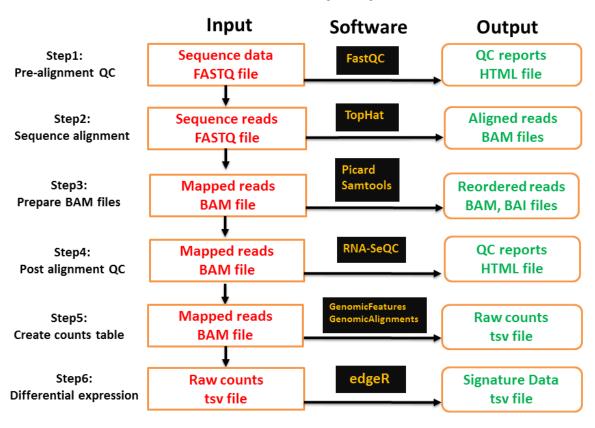


Figure 1:





Directories and some parameters

• Directory of the FASTQ files:

```
inputDirectory= "./inputDirectory"
```

• Download FASTQ files to the inputDirectory

```
dir("inputDirectory/")
## [1] "SRR1600265_1_1m.fastq" "SRR1600266_1_1m.fastq" "SRR1600267_1_1m.fastq"
```

• Directory of the results

[4] "SRR1600268_1_1m.fastq"

```
outputDirectory= "./outputDirectory"
```

• Sample metadata file

```
sampleInfo= read.table(file="./sampleInfo.tsv",header=TRUE, sep="\t")
sampleInfo
```

Type of sequencing and species

```
PE <- FALSE
Species= "Hs"
```

• Gene annotation file. This file contains information of the genes and is downloaded from UCSC with the genome version "hg19" (hg19, GRCh37 Genome Reference Consortium Human Reference 37 (GCA_000001405.1)).

Hs.genome.gtf <- "/opt/ga4raid/pipeline/opt/iGenomes-2015-08-15/Homo_sapiens/UCSC/hg19/Annotation/Genes</pre>

• Genome index file. This index file is built using Bowtie2 in order to align the reads quickly.





Hs.genome.index <- "/opt/ga4raid/pipeline/opt/iGenomes-2014-05-28/Homo_sapiens/UCSC/hg19/Sequence/Bowti</pre>

• Human genome reference file. We consider human reference hg19 and associated FASTA file is downloaded from UCSC along with the gene annotation.

Hs.genome.fa <- "/opt/ga4raid/pipeline/opt/iGenomes-2014-05-28/Homo_sapiens/UCSC/hg19/Sequence/WholeGen

• Location of Picard which is a set of command line tools for manipulating high-throughput sequencing data and formats such as SAM/BAM.

```
picard.path="/opt/ga4raid/pipeline/opt/picard/trunk/dist/"
```

• Location of FastQC software for post alignment QC

```
fastqc.path="/opt/ga4raid/pipeline/opt/FastQC/fastqc"
```

• Location of RNA-SeQC software for post alignment QC

```
rna_seqc.path="/opt/ga4raid/pipeline/opt/RNA-SeQC_v1.1.7.jar"
```

STEP 1: Pre-alignment QC

We start with pre-alignment quality control for each of the FASTQ files using FastQC (version: v0.9.4).

STEP 2: Sequence alignment

Step 2 runs sequence alignment using TopHat (version: v2.0.3) software and creates BAM files. TopHat has dependencies on Bowtie2 (version: 2.2.9) and Samtools (version: 1.3.1). The output includes:

- 1. accepted_hits.bam: A list of read alignments in SAM format
- 2. junctions.bed: A UCSC BED track of junctions reported by TopHat.
- 3. insertions.bed and deletions.bed: UCSC BED tracks of insertions and deletions reported by TopHat.





```
lapply(fastq.files, function(ff) {
    # '-p' indicate number of threads to run which is 1 here. '-G' indicate the
    # gtf file location. Library type is fr-unstranded which means reads from
    # the left-most end of the fragment (in transcript coordinates) map to the
    # transcript strand and the right-most end maps to the opposite strand.
    cmd = paste("nohup ", tophat.path = "tophat", " -p 1 -G ", genome.gtf = Hs.genome.gtf,
        " --library-type fr-unstranded --keep-fasta-order --no-novel-junc -o ",
        outputDirectory, "/tophat_out", sub(".fastq", "", basename(ff)), " ",
        genome.index = Hs.genome.index, " ", ff, sep = "")
    system(cmd)
})
```

STEP 3: Prepare BAM files

Step 3 reorders and add read group information to the bam files using Picard (version: 1.59) and indexes them using samtools (version: 1.3.1) softwares.

```
bam.files = list.files(outputDirectory, "accepted_hits.bam$", full = TRUE, recursive = T)
lapply(bam.files, function(s) {
    # reorder bam files (accepted_hits_sorted.bam) according to reference genome
    # which is needed to run RNA-SeQC in step 4.
    cmd = paste("java -jar ", picard.path, "/ReorderSam.jar ALLOW_INCOMPLETE_DICT_CONCORDANCE=true I=",
        s, " O=", sub(".bam", "_sorted.bam", s), " R=", genome.fa = Hs.genome.fa,
       sep = "")
    system(cmd)
    # The following command add read group information to the bam files. In the
    # simple case where a single library preparation derived from a single
    # biological sample that was run on a single lane of a flowcell, all the
    # reads from that lane run belong to the same read group. When multiplexing
    # is involved, then each subset of reads originating from a separate library
    # run on that lane will constitute a separate read group.
    cmd = paste("java -jar ", picard.path, "/AddOrReplaceReadGroups.jar I=",
        sub(".bam", "_sorted.bam", s), " 0=", sub(".bam", "_added.bam", s),
        " SORT_ORDER=coordinate RGID=1 RGLB=1 RGPL=illumina RGPU=barcode RGSM=1",
        sep = "")
    system(cmd)
    # The following command creates an index file (.bai) corresponding to the
    # bam files. This file acts like an external table of contents and allows
    # programs (e.g., IGV) to jump directly to specific parts of the bam file
    # without reading through all of the sequences.
    cmd = paste(samtools.path = "samtools", " index ", sub(".bam", "_added.bam",
        s), sep = "")
    system(cmd)
})
```





STEP 4: Post alignment QC

Step 4 generates post alignment QC reports using RNA-SeQC (version: 1.1.7) which includes read count metrics, correlation analysis, coverage metrics, mean coverage etc.

```
bam.files = list.files(outputDirectory, "accepted_hits_added.bam$", full = TRUE,
   recursive = T)
sampleid = sapply(strsplit(bam.files, "/"), function(x) x[length(x) - 1])
sampleid = sub("tophat_out", "", sampleid)
note = sapply(strsplit(sampleid, "_"), function(x) x[1])
dat = data.frame(`Sample ID` = sampleid, `Bam File` = bam.files, Notes = note,
    stringsAsFactors = F)
write.table(dat, file = paste0(outputDirectory, "/sample.file"), col.name = T,
   row.name = F, quote = F, sep = "\t")
if (PE == TRUE) {
   single.or.paired <- ""
} else {
    single.or.paired <- "-singleEnd"
system(paste0("mkdir ", outputDirectory, "/rna_seqc"))
cmd = paste("nohup java -jar ", rna_seqc.path, " -o ", paste0(outputDirectory,
    "/rna_seqc"), " -r ", genome.fa = Hs.genome.fa, " -s ", samplefile = paste0(outputDirectory,
    "/sample.file"), " ", single.or.paired, " -t ", genome.gtf = Hs.genome.gtf,
    sep = "")
system(cmd)
```

STEP 5: Counting reads and create counts table

Step 5 generates raw counts table using R packages GenomicAlignments (version: 1.6.3) and GenomicFeatures (version: 1.22.13).

```
library(GenomicFeatures)
library(GenomicAlignments)
# Download refGene as a TranscriptDb(TxDb) object.
refGene <- makeTxDbFromUCSC(genome = "hg19", tablename = "refGene")
# Download exons by gene
exonRangesList <- exonsBy(refGene, "gene")
bam.files = list.files(outputDirectory, "accepted_hits_added.bam$", full = TRUE,
    recursive = T)
countsTable <- NULL
for (i in bam.files) {
    if (PE == TRUE) {
        params <- ScanBamParam(flag = scanBamFlag(isUnmappedQuery = FALSE),</pre>
            tag = "NH")
        aligns <- readGAlignmentPairs(i, format = "BAM", param = params)</pre>
        # read genomic alignments for paired-end
    } else {
        params <- ScanBamParam(flag = scanBamFlag(isUnmappedQuery = FALSE),</pre>
            tag = "NH")
```





```
aligns <- readGAlignments(i, index = i, param = params)
        # read genomic alignments for single-end
    strand(aligns) = "*"
    unique_hits <- aligns[mcols(aligns)$NH == 1] #remove multimapping reads
    # Counting reads using summarizeOverlaps with mode='Union'. In this mode,
    # reads that overlap any portion of exactly one feature are counted. Reads
    # that overlap multiple features are discarded. Also, inter.feautre is a
    # logical indicating if the counting mode should be aware of overlapping
    # features. When TRUE, reads mapping to multiple features are dropped.
    counts <- assays(summarizeOverlaps(exonRangesList, unique_hits, mode = "Union",</pre>
        inter.feature = TRUE, SingleEnd = (!PE)))$counts
    countsTable <- cbind(countsTable, counts)</pre>
}
colnames(countsTable) <- as.vector(sampleInfo[, 2])</pre>
countsTable <- data.matrix(data.frame(entrez_geneID = rownames(countsTable),</pre>
    countsTable, stringsAsFactors = FALSE))
rownames(countsTable) <- NULL</pre>
write.table(countsTable, file = paste0(outputDirectory, "/countsTable.tsv"),
    quote = FALSE, sep = "\t", row.names = FALSE)
head(countsTable)
```

```
entrez_geneID SRR1600265 SRR1600266 SRR1600267 SRR1600268
##
## [1,]
                   1
                                0
                                           5
## [2,]
                   10
                                0
                                           0
                                                       0
                                                                   0
## [3,]
                  100
                                0
                                           0
                                                       0
                                                                   0
                                0
                                           0
                                                                   0
## [4,]
                 1000
                                                       0
## [5,]
                10000
                                            1
                                                                   0
## [6,]
                                            Λ
                                                                   0
            100008586
```

STEP 6: Differential expression analysis for two group comparison

Step 6 analyze the raw count data cretaed in step 5 using R package edgeR (version: 3.12.1), locfit (version: 1.5.9.1), and "org.Hs.eg.db" (version: 3.2.3).





```
# than the smaller sample size between two groups are retained for the
# subsequent analyses and other genes are filtered out.
o <- order(rowSums(y$counts))</pre>
y \leftarrow y[o,]
keep <- rowSums(cpm(y) > 1) >= min(summary(factor(Group)))
y <- y[keep, ]
dim(y)
## [1] 12835
# Recompute library sizes
y$samples$lib.size <- colSums(y$counts)
y$samples
##
                       group lib.size norm.factors
## SRR1600265
                     Control 333860
## SRR1600266
                     Control
                               314452
                                                  1
## SRR1600267 drugTreatment
                               295435
                                                  1
## SRR1600268 drugTreatment
                               297628
                                                  1
# Normalization: to get effective library size; default method: trimmed mean
# of M-values(TMM)
y <- calcNormFactors(y)</pre>
# estimate dispersion and test for DE
y <- estimateDisp(y)</pre>
## Design matrix not provided. Switch to the classic mode.
fit <- exactTest(y, pair = levels(Group))</pre>
# sort the result table by the p-values.
top_degs <- topTags(fit, n = nrow(fit$table))</pre>
signaturesData <- top_degs$table
rownames(signaturesData) <- NULL</pre>
# Get gene symbol
signaturesData$Gene <- sapply(mget(as.character(signaturesData$genes), get(paste0("org.",</pre>
    Species, ".egSYMBOL")), if not found = NA), function(x) x[1])
# signaturesData table
signaturesData <- signaturesData[, c(6, 2, 3, 4)]</pre>
colnames(signaturesData) <- c("Gene", "Sig1_LogFoldChange", "Sig1_LogCountPerMillion",
    "Sig1_PValue")
# SignaturesMetaData table
SignaturesMetaData <- data.frame(SignatureID = "Sig1", TreatmentSamples = paste(sampleInfo[-which(samples = paste)])
    "untreated"), 2], collapse = ", "), ControlSamples = paste(sampleInfo[which(sampleInfo$perturbagen =
    "untreated"), 2], collapse = ", "), CellLine = unique(sampleInfo$cell_line))
# save the results
```





```
write.table(signaturesData, file = pasteO(outputDirectory, "/signaturesData.tsv"),
    quote = FALSE, sep = "\t", row.names = FALSE)
write.table(SignaturesMetaData, file = pasteO(outputDirectory, "/SignaturesMetaData.tsv"),
    quote = FALSE, sep = "\t", row.names = FALSE)
head(signaturesData)
        Gene Sig1_LogFoldChange Sig1_LogCountPerMillion Sig1_PValue
## 1
       MUC19
                       6.006471
                                              11.151981 3.495780e-54
## 2 IGFBP5
                       4.466220
                                              9.801600 2.300321e-28
## 3
       PSCA
                      4.013605
                                             10.568195 4.928881e-23
## 4 SYNPO2
                                              8.163808 1.539999e-19
                      3.405706
## 5 SLC7A11
                                              6.646949 1.893714e-18
                      -5.614975
## 6
                                              11.777374 2.741182e-16
         SCD
                      -3.402830
head(SignaturesMetaData)
     SignatureID
                       TreatmentSamples
                                                ControlSamples CellLine
            Sig1 SRR1600267, SRR1600268 SRR1600265, SRR1600266 HCC1419
sessionInfo()
## R version 3.3.0 (2016-05-03)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Ubuntu 14.04.4 LTS
##
## locale:
## [1] LC CTYPE=en US.UTF-8
                                   LC NUMERIC=C
## [3] LC_TIME=en_US.UTF-8
                                   LC_COLLATE=en_US.UTF-8
## [5] LC_MONETARY=en_US.UTF-8
                                   LC_MESSAGES=en_US.UTF-8
  [7] LC_PAPER=en_US.UTF-8
                                   LC_NAME=C
   [9] LC_ADDRESS=C
                                   LC_TELEPHONE=C
## [11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C
## attached base packages:
## [1] parallel stats4
                           stats
                                     graphics grDevices utils
                                                                   datasets
## [8] methods
                base
## other attached packages:
## [1] org.Hs.eg.db_3.2.3
                             RSQLite_1.0.0
                                                  DBI 0.4-1
## [4] AnnotationDbi_1.32.3 IRanges_2.4.8
                                                  S4Vectors 0.8.11
## [7] Biobase_2.30.0
                             BiocGenerics_0.16.1 locfit_1.5-9.1
## [10] edgeR_3.12.1
                             limma_3.26.8
##
## loaded via a namespace (and not attached):
                        lattice_0.20-33 digest_0.6.9
## [1] Rcpp_0.12.5
                                                        grid_3.3.0
   [5] formatR_1.4
                        magrittr_1.5
                                        evaluate_0.9
                                                        stringi_1.1.1
## [9] rmarkdown_1.0 splines_3.3.0
                                                        stringr_1.0.0
                                       tools_3.3.0
## [13] yaml_2.1.13
                       htmltools_0.3.5 knitr_1.13
```