

Oct 15, 2019

Version 9

Stranded Transcript Count Table Generation from Long Reads V.9



Version 1 is forked from Transcript Coverage Analysis from Long Reads

DOI

dx.doi.org/10.17504/protocols.io.778hrrw

David A Eccles¹

¹Malaghan Institute of Medical Research (NZ)



David A Eccles

GrinGene Bioinformatics

Create & collaborate more with a free account

Edit and publish protocols, collaborate in communities, share insights through comments, and track progress with run records.

Create free account

OPEN ACCESS



DOI: https://dx.doi.org/10.17504/protocols.io.778hrrw

Protocol Citation: David A Eccles 2019. Stranded Transcript Count Table Generation from Long Reads. **protocols.io**https://dx.doi.org/10.17504/protocols.io.778hrrw

License: This is an open access protocol distributed under the terms of the **Creative Commons Attribution License**, which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited



Protocol status: In development

We are still developing and optimizing this protocol

Created: October 15, 2019

Last Modified: October 15, 2019

Protocol Integer ID: 28640

Keywords: stranded transcript count table generation from long read, stranded transcript count table generation, transcript reference fasta file, transcript table, transcript, protocol preparing reads for stranded mapping, different samples at the transcript level, using long read, transcript level, protocol preparing read, oriented fastq file, fastq file, long read, gene name, stranded mapping, annotation file

Abstract

This protocol is for comparing different samples at the transcript level, using long reads that are mapped to transcripts.

Input(s): demultiplexed and oriented fastg files (see protocol Preparing Reads for Stranded Mapping), transcript reference fasta file, annotation file

Output(s): transcript table, sorted by differential coverage, annotated with gene name / description / location

Troubleshooting

Before start

Obtain a transcript fasta file, and an annotation file. For the mouse genome, I use the following files:

- 1. Transcript [CDS] sequences from Ensembl; this file was the most current when I last checked.
- 2. Annotation file obtained from Ensembl BioMart (Ensembl Genes → Mouse Genes) as a compressed TSV file with the following attribute columns:
- Transcript stable ID
- Gene description
- Gene start (bp)
- Gene end (bp)
- Strand
- Gene name
- Chromosome/scaffold name



Demultiplex Reads

1

Demultiplex and orient reads as per the protocol <u>Preparing Reads for Stranded Mapping</u>. It is expected that these demultiplexed reads will be split up in the current directory, and coupled with a 'barcode_counts.txt' file. If that's the case, the following should work:

```
for bc in $(awk '{print $2}' barcode_counts.txt);
  do ls oriented/${bc}_reads_dirAdjusted.fq.gz;
done
```

Example expected output:

```
oriented/BC03_reads_dirAdjusted.fastq.gz
oriented/BC04_reads_dirAdjusted.fastq.gz
oriented/BC05_reads_dirAdjusted.fastq.gz
oriented/BC06_reads_dirAdjusted.fastq.gz
oriented/BC07_reads_dirAdjusted.fastq.gz
oriented/BC08_reads_dirAdjusted.fastq.gz
```

If the 'barcode_counts.txt' file is not present, this error will appear:

```
awk: fatal: cannot open file `barcode_counts.txt' for reading (No
such file or directory)
```

If one or more of the oriented read files is missing, it will look something like this:

```
oriented/BC03_reads_dirAdjusted.fastq.gz
oriented/BC04_reads_dirAdjusted.fastq.gz
ls: cannot access 'oriented/BC05_reads_dirAdjusted.fastq.gz':
  No such file or directory
ls: cannot access 'oriented/BC06_reads_dirAdjusted.fastq.gz':
  No such file or directory
oriented/BC07_reads_dirAdjusted.fastq.gz
oriented/BC08_reads_dirAdjusted.fastq.gz
```

Index Preparation

2 Prepare transcript index (see Guidelines for data sources)



```
lastdb Mus_musculus.GRCm38.cds.all.fa <(zcat</pre>
Mus_musculus.GRCm38.cds.all.fa.gz)
```

Transcriptome Mapping

3 Reads are mapped to the transcriptome with LAST.

> The results of that mapping can be piped through *last-map-probs* to exclude unlikely hits, then through 'maf-convert -n tab' to convert to a one-line-per-mapping CSV format. This CSV format is further processed to make sure that there is only one mapping per transcript-read pair.

```
mkdir -p mapped
for bc in $(awk '{print $2}' barcode_counts.txt);
  do echo "** ${bc} **";
  lastal -Q 1 -P 10 Mus_musculus.GRCm38.cds.all.fa <(pv
oriented/${bc}_reads_dirAdjusted.fq.gz | zcat) | \
    last-map-probs | maf-convert -n tab | cut -f 2,7,10 | sort | \
    uniq >
mapped/trnMapping_LAST_${bc}_vs_Mmus_transcriptome.txt.gz;
done
```

4 The result is then aggregated to sum up counts per transcript:

```
for bc in $(awk '{print $2}' barcode_counts.txt);
  do echo "** ${bc} **";
 zcat mapped/trnMapping_LAST_${bc}_vs_Mmus_transcriptome.txt.gz |
    awk -F'\t' -v "bc=\{bc\}" '{print bc,\{1, \{3\}\}' | sort | uniq -c |
    qzip >
mapped/trnCounts_LAST_${bc}_vs_Mmus_transcriptome.txt.gz;
done
```

Note: I've split this up into two steps (compared to previous versions of this protocol) so that an intermediate count of the total number of mapped transcripts per barcode can be done:



```
for bc in $(awk '{print $2}' barcode_counts.txt);
 do echo -n "${bc} ";
 zcat mapped/trnMapping_LAST_${bc}_vs_Mmus_transcriptome.txt.gz |
   awk '{print $2}' | sort | uniq | wc -l;
done
```

Annotation and Result generation

5



count_analysis.r

Transcript counts are merged with ensembl gene annotation, then converted into wide format (one line per transcript) using an R script.

The transcript annotation in this case is from ensembl BioMart (see Guidelines for more details).



```
#!/usr/bin/env Rscript
library(dplyr);
library(tidyr);
## load ensemble transcript metadata (including gene name)
ensembl.df <-
as.tbl(read.delim('ensembl_mm10_geneFeatureLocations.txt.gz',
  col.names=c('transcript','Description','Start','End',
              'Strand', 'Gene', 'Chr'),
  stringsAsFactors=FALSE));
ensembl.df$Description <- sub(' \\</pre>
[.*$','',ensembl.df$Description);
ensembl.df$Description <- sub('^(.
{50}).+$','\\1...',ensembl.df$Description);
ensembl.df[,1:7] <- ensembl.df[,c(1,7,5,3,4,2,6)];
colnames(ensembl.df)[1:7] <- colnames(ensembl.df)</pre>
[c(1,7,5,3,4,2,6)];
options(scipen=15); ## don't show scientific notation for large
positions
## load used barcode identifiers
bcNames <- read.table("barcode_counts.txt",</pre>
stringsAsFactors=FALSE)[,2];
## load count data into 'narrow' array (one line per count)
trn.counts <- tibble(); for(bc in bcNames){</pre>
  trn.counts <-
    bind_rows(trn.counts,
      as.tbl(read.table(
sprintf('mapped/trnCounts_LAST_%s_vs_Mmus_transcriptome.txt.qz',
bc),
        col.names=c('count', 'barcode', 'transcript', 'dir'),
        stringsAsFactors=FALSE)));
}
## remove revision number from transcript names (if present)
trn.counts$transcript <- sub('\\.[0-</pre>
9]+$','',trn.counts$transcript);
## convert to wide format (one line per transcript)
trn.counts.wide <- spread(trn.counts, barcode, count) %>%
 mutate(dir = c('+'='fwd', '-'='rev')[dir]);
for(bd in colnames(trn.counts.wide[,-1])){
  trn.counts.wide[[bd]] <- replace_na(trn.counts.wide[[bd]],0);</pre>
## merge ensembl metadata with transcript counts
```



```
gene.counts.wide <- inner_join(ensembl.df, trn.counts.wide,</pre>
by='transcript');
gene.counts.wide <- gene.counts.wide[order(-</pre>
rowSums(gene.counts.wide[,-(1:8)])),];
## write result out to a file
write.csv(gene.counts.wide,
file='wide_transcript_counts_LAST.csv',
  row.names=FALSE);
```

Downstream Workflows

6 A downstream workflow is still in development. I have had reasonable succes doing transcript-level differential expression analysis using **DESeq2**.

I would like to emphasise that batch effects should be considered for nanopore sequencing, given how frequently the technology changes. Make sure that at least the sequencing library (i.e. samples prepared in tandem on the same day from the same kit) is added into the statistical model, and try to make sure that sequencing libraries are fairly heterogeneous - replicates from a sample with skewed transcript distributions could influence the outcome of statistical tests.