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Stranded Transcript Count Table Generation from Long Reads V.11

Forked from Transcript Coverage Analysis from Long Reads

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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 fastq 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

BEFORE STARTING

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

1. Transcript sequences from Ensembl; this should be the union of cDNA, CDS, and ncRNA sequences (e.g. from This directory).
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 name
- Gene description
- Chromosome/scaffold name
- Gene start (bp)
- Gene end (bp)
- Strand

A recent version of these files can be obtained from This Zenodo Repository

Demultiplex Reads

1

Demultiplex and orient reads as per the protocol Preparing Reads for Stranded Mapping. 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 a substitution matrix for barcode mapping. The default substitution matrix is swayed too much by INDELs in the barcode sequences, so here's one that I've developed using a combination of trial & error and last-train:

```
#last -Q 0
#last -a 13
#last -A 13
#last -b 4
#last -B 4
#last -S 1
# score matrix (query letters = columns, reference letters = rows):
       A      C      G      T
A      5    -18     -7    -18
C    -18      6    -18    -12
G     -7    -18      5    -18
T    -18    -12    -18      6
```

cDNA.mat

*[note: this is a **different** matrix from that used for demultiplexing and read orientation]*

3 Prepare transcript index (see Guidelines for data sources). Following Martin Frith's recommendation, the '-uNEAR' seeding scheme is used to slightly increase sensitivity. This will generate seven additional files of the form <index name>.XXX:

```
lastdb -uNEAR Mus_musculus.GRCm38.cds.all.fa <(zcat Mus_musculus.GRCm38.cds.all.fa.gz)
```

Transcriptome Mapping

4 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 "** ${bc} **";
  lastal -P 10 -p cDNA.mat Mus_musculus.GRCm38.cds.all.fa <(pv
oriented/${bc}_reads_dirAdjusted.fq.gz | zcat) | \
    last-map-probs | maf-convert -n tab | cut -f 2,7,10 | sort | \
    uniq | gzip > mapped/trnMapping_LAST_${bc}_vs_Mmus_transcriptome.txt.gz;
done
```

5 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,$3}' | sort | uniq -c | \
    gzip > mapped/trnCounts_LAST_${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

6 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$Description <- sub(' \\[.*$','',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)[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", stringsAsFactors=FALSE)[,2];
## load count data into 'narrow' array (one line per count)
trn.counts <- tibble(); for(bc in bcNames){
  trn.counts <-
    bind_rows(trn.counts,
      as.tbl(read.table(
        sprintf('mapped/trnCounts_LAST_%s_vs_Mmus_transcriptome.txt.gz', bc),
        col.names=c('count','barcode','transcript','dir'),
        stringsAsFactors=FALSE)));
}

## remove revision number from transcript names (if present)
trn.counts$transcript <- sub('\\.[0-9]+$','',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);
}
## merge ensembl metadata with transcript counts
gene.counts.wide <- inner_join(ensembl.df, trn.counts.wide, by='transcript');
gene.counts.wide <- gene.counts.wide[order(-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

7 Here is a downstream workflow that carries out transcript-level differential expression analysis using DESeq2:

- Creating Differential Transcript Expression Results with 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.