Analysis of QuantSeq FWD UMI 3' RNA-seq data

Full length RNA-seq vs 3' RNA-seq



- > Full length RNA-seq: reads cover whole transcripts
- > 3' RNA-seq: reads cover only the 3' ends of transcripts
 - Not possible to detect transcript isoforms
 - Sufficient for gene-level quantitation

QuantSeq 3' mRNA-seq data

- Reads come from the 3' end, near polyA
 - polyA read-through to adapters is common, needs to be trimmed
- Just one fragment per transcript is produced
 - Transcript length does not affect read counts
- Use only R1 reads
 - R2 reads start with poly(T) and have low quality
- Option to remove PCR duplicates using unique molecular identifiers (UMIs)
 - Useful for low-input and formalin-fixed, paraffin-embedded (FFPE) samples*

QuantSeq workflow



Reverse transcription

- oligodT priming
- contains the R2 linker, so R2 reads start with poly(T)
- \rightarrow R2 reads have low quality
- \rightarrow use only R1 reads
- Second strand synthesis with random priming
 - Errors in first nucleotides due to non-specific hybridization of the random primer to the cDNA template.
 - \rightarrow Use aligner with soft-clipping

Unique molecular identifiers (UMIs)

 \succ Reduce amplification bias \rightarrow more accurate quantitation



www.lexogen.com/rna-lexicon-what-areunique-molecular-identifiers-umis-andwhy-do-we-need-them/

CSC

QuantSeq FWD UMI 3' mRNA-seq data

- Reads contain 6 nt unique molecular identifiers (UMIs)
- Located at the start of the read, need to be removed and stored in the read name before alignment to genome
- Deduplication: Reads which map to the same genomic location and have the same UMI are grouped together, and only one representative read is kept

QuantSeq 3' mRNA-seq data analysis steps

- > Alignment / STAR or HISAT2 \rightarrow BAM
- ➢ Preprocessing / Deduplicate aligned QuantSeq reads → BAM
- ➢ Quality control / RNA-seq quality metrics with RseQC → pdf
- ➢ RNA-seq / Count aligned reads per genes with HTSeq → tsv
- ➢ Utilities / Define NGS experiment → tsv

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- ➢ RNA-seq / Differential expression using DESeq2 → tsv
- ➤ Utilities / Annotate Ensembl identifiers → tsv

UMIs from sequence to read names

Use the tool Preprocessing / Extract UMIs from QuantSeq reads

- Extracts the 6-base UMI and stores it in read's name
- Removes the TATA spacer in position 7-10
- · log.txt file tells how many reads contained TATA and were processed
- Based on the extract tool of the UMI-Tools package

@A00464:250:HW3NWDRXX:1:2236:5565:26412 1:N:0:GCATGG+CTAACT

@A00464:250:HW3NWDRXX:1:2236:5565:26412_AGCGGG 1:N:0:GCATGG+CTAACT

+

Trim polyA, adapters and low quality ends

Tool Preprocessing / Trim QuantSeq reads with BBDuk

- Detects and removes polyA tails and Illumina TruSeq adapters
- Trims low-quality bases from read ends
- · Removes reads that are too short after trimming
- Based on the BBDuk tool of the BBTools package

Trim QuantSeq reads using BBDuk		×
Parameters		S Reset All
Kmer length for detecting adapters and polyA Contaminants shorter than k will not be found, k must be at least 1.	13	\$
Trimming direction from a kmer match Once a polyA or Truseq adapter kmer is matched in a read, that kmer and all the bases to this direction will be trimmed.	Trim to the right	~
Minimum length of kmers to report at read tips Look for polyA or Truseq adapter kmers down to this length at read tips. 0 means disabled.	5	Ŷ
Should read ends be trimmed based on quality After looking for kmers, remove low quality bases from read ends. Set the quality threshold with the next parameter.	Trim both ends	~
Threshold for quality trimming Regions with base quality below this Phred score will be trimmed, if quality trimming is selected. Can be a floating-point number like 7.3.	10	\Diamond
Minimum length for reads to be kept after trimming Reads shorter than this after trimming will be discarded	20	$\hat{\cdot}$

Base composition plot

% Reads

Raw reads

> After extracting UMIs and TATA

After trimming polyA and adapters



Remove amplification bias using UMIs

Tool Preprocessing / Deduplicate aligned QuantSeq reads

- Identifies which reads have the same mapping position.
- Groups those reads which have the same/similar UMI.
- Two grouping methods
 - **Unique:** reads must have exactly the same UMI sequence. Fast but doesn't allow for sequencing errors.
 - **Directional:** builds networks where nodes are UMIs and edges connect UMIs with an edit distance </= 1. Identifies clusters of UMIs. Slow, allows for errors.
- Keeps a single representative read
 - lowest number of mapping coordinates
 - highest mapping quality. Note that base quality is not considered
- Output is a deduplicated BAM file and optional statistics files
 - average edit distance between the UMIs at each position
 - counts for unique combinations of UMI and position
 - UMI-level summary statistics
- Based on the dedup tool of the UMI-Tools package
 - https://umi-tools.readthedocs.io/en/latest/reference/dedup.html

Adapter/polyA removal and deduplication reduce the number of reads



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Pathway analysis

Pathway analysis – why?

- Statistical tests can yield thousands of differentially expressed genes
- It is difficult to make "biological" sense out of the result list
- Looking at the bigger picture can be helpful, e.g. which pathways are differentially expressed between the experimental groups
- Databases such as KEGG, GO, Reactome and ConsensusPathDB provide grouping of genes to pathways, biological processes, molecular functions, etc

Gene set enrichment analysis

- 1. Perform a statistical test to find differentially expressed genes
- 2. Check if the list of differentially expressed genes is "enriched" for some pathways



ConsensusPathDB

One-stop shop: Integrates pathway information from 32 databases covering

- biochemical pathways
- protein-protein, genetic, metabolic, signaling, gene regulatory and drug-target interactions
- Developed by Ralf Herwig's group at the Max-Planck Institute in Berlin
- ConsensusPathDB over-representation analysis tool is integrated in Chipster
 - runs on the MPI server in Berlin