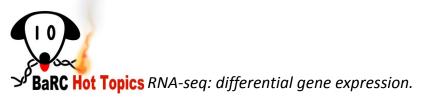
#### RNA-seq: A practical guide to the analysis of differential gene expression

December 1st, 2011

1



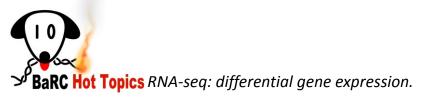
## **RNA-seq Applications**

#### Annotation

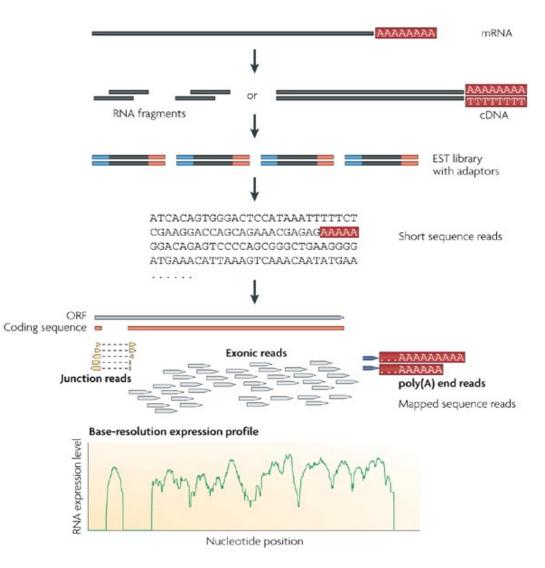
Identify novel genes, transcripts, exons, splicing events, ncRNAs.

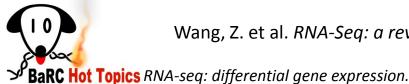
- Detecting RNA editing and SNPs.
- Measurements: RNA quantification and differential gene expression

Abundance of transcripts between different conditions



### **RNA-Seq Experiment**

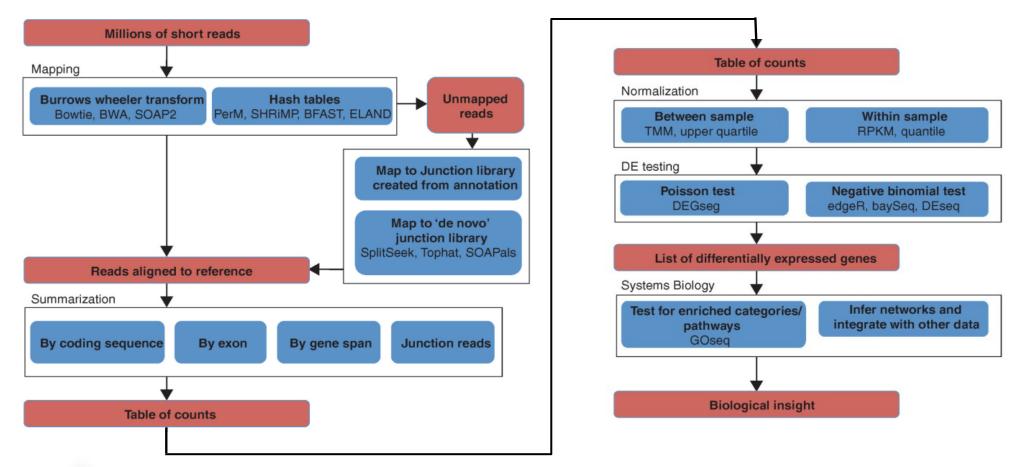




Wang, Z. et al. RNA-Seq: a revolutionary tool for transcriptomics Nature Reviews Genetics (2009)

Nature Reviews | Genetics

# Overview of the RNA-seq analysis pipeline for detecting differential expression



Oshlack et al., From RNA-seq reads to differential expression results, Genome Biology 2010.

BaRC Hot Topics RNA-seq: differential gene expression.

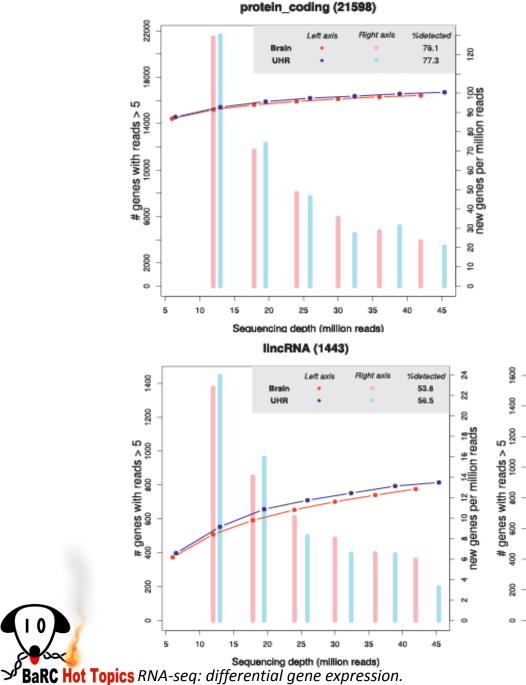
# Steps involved on RNA-seq analysis for detecting differential expression

- Experimental design
- Preprocess
  - Split by barcodes
  - Quality control and removal of poor-quality reads
  - Remove adapters and linkers
- Map the reads
- Count how many reads fall within each feature of interest (gene, transcript, exon etc).
- Remove absent genes
- Add offset (such as 1)
  - Prevent dividing by 0
  - Moderate fold change of low-count genes
- Identify differentially expressed genes.

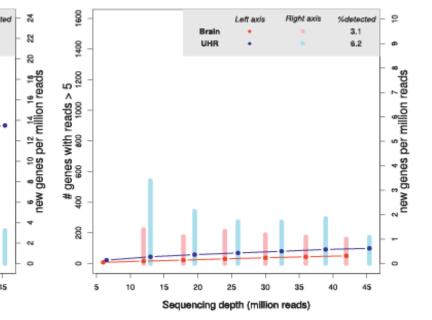
## **Experimental design**

- Include replicas in your experiment.!!!!!! Conclusions drawn from a single RNA-seq experiment can be very misleading.
- Number of reads needed for an experiment.
   Depends on the organism and the level of the differences you want to detect.
  - Evaluation of the coverage and depth of transcriptome by RNA-Seq in chickens. Wang et al. BMC Bioinformatics 2011, 12(Suppl 10):S5 <u>http://www.biomedcentral.com/1471-</u> 2105/12/S10/S5
  - *Differential expression in RNA-seq: A matter of depth. Genome Res. 2011.* PMID: 21903743.

#### Number of reads needed for an experiment



Differential expression in RNAseq: A matter of depth. Genome Res. 2011. PMID: 21903743.



snoRNA (1596)

#### Preprocess

#### • Useful tools for preprocessing

Fastx Toolkit : http://hannonlab.cshl.edu/fastx\_toolkit

FastQC: http://www.bioinformatics.bbsrc.ac.uk/projects/fastqc/

#### 1. Split by barcodes

- If the barcode is still on the sequence use "fastx\_barcode\_splitter.pl"
- cat sequence.txt | fastx\_barcode\_splitter.pl --bcfile mybarcodes.txt --bol mismatches 2
- --bcfile: File containing the barcodes
- --bol: beginning of the sequence
- If the facility has removed the barcode and added to the header or the sequence like @HWI-ST333\_0165\_FC:4:1101:1494:2176#ACCTGAAT/1 ATACATTGTTTCCTTTTTAGAAATATTCTGTTACTATTAT use "splitReadsByBarcodesInDescriptionLines.pl" script in /nfs/BaRC\_Public/BaRC\_code/Perl

splitReadsByBarcodesInDescriptionLines.pl sequence.txt Barcodes.txt splittedSeq

#### Preprocess

#### 2. Quality control and removal of poor-quality reads

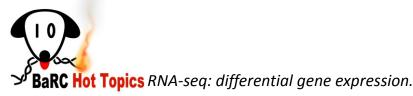
(http://iona.wi.mit.edu/bio/education/hot\_topics/QC\_HTP/QC\_HTP.pdf)

fastqc s\_1\_seq.txt s\_2\_seq.txt

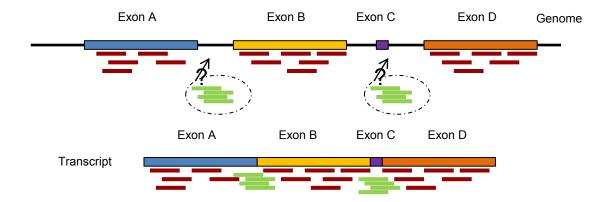
fastq\_quality\_filter -q 20 -p 80 -i s\_1\_seq.txt -o
s\_1\_seq.fastx\_trim

- -q = Minimum quality score to keep
- -p = Minimum percent of bases that must have [-q] quality
- 3. Remove adapters and linkers

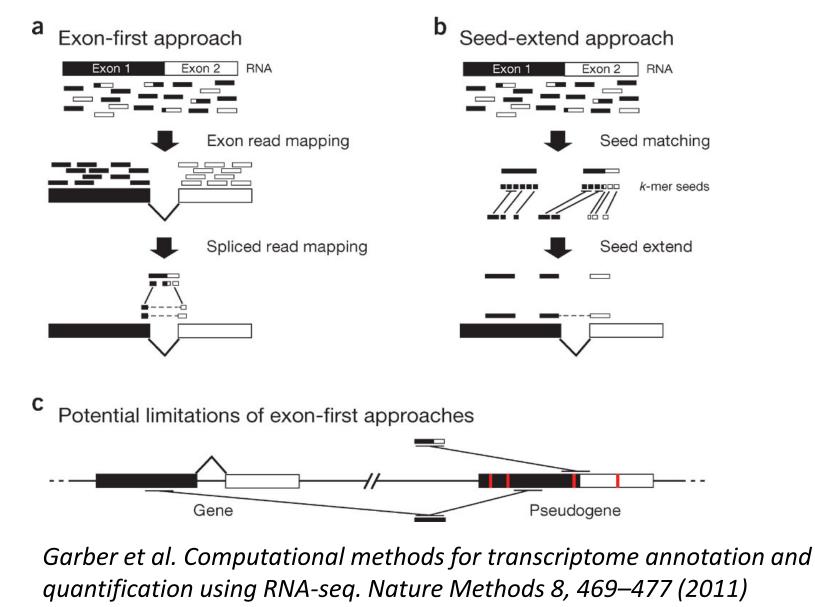
fastx\_clipper -a GATCGGAAGAGCACACGTCTGAACTCCAGTCAC -i
s\_1\_seq.fastx\_trim -1 22 -o s\_1\_seq.fastx\_trim\_clippedAdap



#### Map the reads to the genome



## Map the reads to the genome



Section 24 BaRC Hot Topics RNA-seq: differential gene expression.

## Map the reads to the genome

- Tophat (http://tophat.cbcb.umd.edu/)
  - include gff file that will be used in the counting
  - allow mapping to several places, the redundancy can be removed later.
- Sample commands:
  - 1. Run tophat

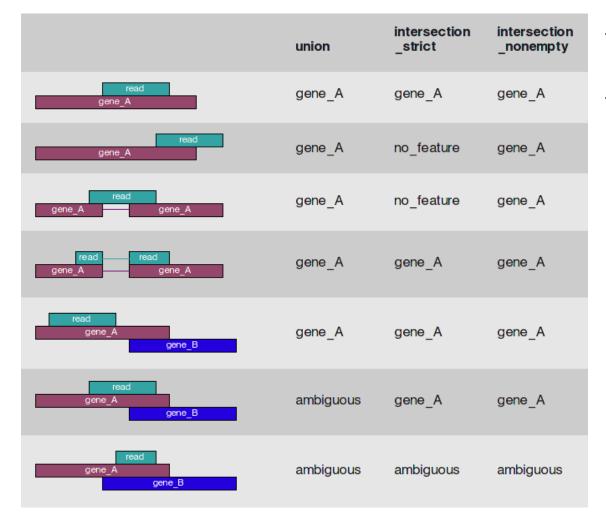
```
tophat -G
/nfs/genomes/human_gp_feb_09/gtf/hg19.refgene.gtf -o
OutputName --solexa-quals
/nfs/genomes/human_gp_feb_09_no_random/bowtie/hg19
seq.fastq
```

2. Convert BAM to SAM output (since SAM is required for htseq-count).

samtools view -h -o accepted\_hits.sam accepted\_hits.bam

### Count reads with Htseq-count Htseq-count

(http://www-huber.embl.de/users/anders/HTSeq/doc/count.html)



✓ Options

Quantifying genes
 versus quantifying
 transcripts

#### Count reads with Htseq-count

(http://www-huber.embl.de/users/anders/HTSeq/doc/count.html)

htseq-count is run for each sample :

```
bsub "htseq-count -m intersection-strict -s no
UHR1 accepted hits.sam
/nfs/genomes/human_gp_feb_09/gtf/hg19.refgene.gtf >brain1_htseq-
count.out "
```

-s: whether the data is from a strand-specific assay (default: yes)

-m: mode to handle reads overlapping more than one feature (default: union)

#### head UHR1\_htseq-count.out

```
67
A1BG
A1CF
         31
A2BP1
         3
head UHR2 htseq-count.out
         67
A1BG
         39
A1CF
A2BP1
         1
and then counts are combined (it can also be done with excel):
paste UHR1 htseq-count.out UHR2 htseq-count.out brain1 htseq-
count.out brain2 htseq-count.out | awk -F "\t" '{print
$1"\t"$2"\t"$4"\t"$6"\t"$8 }' > All htseqCounts.txt
```

A1BG	67	67	20	53
A1CF	31	39	0	0
A2BP1	3	1	542	532

# Remove absent genes, add 1 pseudocount

- Remove absent genes (zero counts in all samples). It reduces the number of tests and the false discovery rate correction.
- Add 1 pseudocount (prevent dividing by 0).
- Remove the rows at the bottom with descriptions like no\_feature, ambiguous, etc.

awk -F "\t" '{if (\$2>0 || \$3>0 || \$4>0 || \$5>0 ) print \$1"\t"\$2+1"\t"\$3+1"\t"\$4+1"\t"\$5+1 }' All\_htseqCounts.txt |

```
grep -v no_feature | grep -v ambiguous | grep -v
too_low_aQual | grep -v not_aligned | grep -v
alignment_not_unique > All_Counts_nozero_1pseudocount.txt
```

Note that these steps can be done with excel.

Jpdate: Pseudocounts can be added for display purposes or other manual processing, but a counts matrix used for statistics should NOT include pseudocounts.

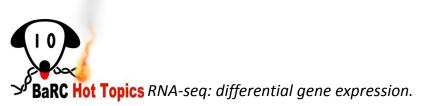
#### Add header

cat header.txt All\_Counts\_no0\_1pc.txt >
 All\_Counts\_no0\_1pc\_Header.txt

#### head All\_Counts\_no0\_1pc\_Header.txt

ID	UHR_1	UHR_2	brain_1	brain_2
A1BG	68	68	21	53
A1CF	32	40	1	1
A2BP1	4	2	543	533
A2LD1	11	11	4	7
A2M	1772	1901	407	450

File ready to use for running software to find DE genes



Finding differentially expressed genes: Need for Normalization

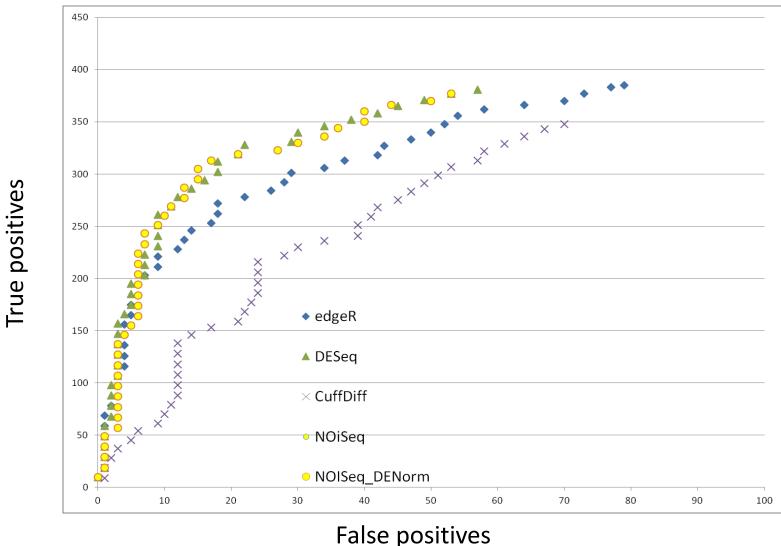
- More reads mapped to a transcript if it is
   i) long
  - ii) at higher depth of coverage
- Normalize such that i) *features* of different lengths and ii) total sequence from different conditions can be compared

#### Finding differentially expressed genes

Method	Normalization	Needs replicas	Input	Statistics for DE	Other
EdgeR	Library size and TMM (trimmed mean of M values)	Yes	Raw counts	Empirical Bayes estimation and exact tests based on the negative binomial distribution.	PMID: 19910308
DESeq	Library size	No	Raw counts	Negative binomial distribution.	Uses different coefficients of variation for different expression strengths. PMID: 20979621
baySeq	Library size	Yes	Raw counts	Empirical Bayesian methods using the negative binomial distribution.	PMID: 20698981
NOISeq	Several options: counts per million read, RPKM, Upper Quartile.	No	Raw or normalized counts	Compares replicates within the same condition to estimate noise distribution of M (log-ratio ) and D (absolute value of the difference). A feature is considered to be differentially expressed if its corresponding M and D values are likely to be higher than noise values.	PMID: 21903743
0 Di					18

## BaRC RNA-seq bakeoff

Differential expression between brain and human universal reference RNA (UHR) from the MicroArray Quality Control (MAQC) Project. (*BMC Bioinformatics. 2010 Feb 18;11:94, Nat Biotechnol. 2006 Sep;24(9):1151-61.*)



**BaRC Hot Topics** RNA-seq: differential gene expression.

# Finding differentially expressed genes with NOISeq

#### Sample code for running NOISeq

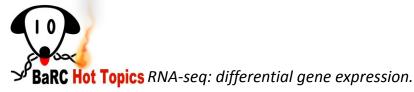
source("NOISeq.r")

mydata <- readData(file = "All\_Counts\_no0\_1pc\_Header.txt", cond1 = c(2:3), cond2 = c(4:5), header = TRUE)
myresults <- noiseq(mydata[[1]], mydata[[2]], repl = "bio", q = 0.9, nss = 0)
write.table(cbind (myresults\$Ms[myresults\$deg],myresults\$probab[myresults\$deg] ),
 file="genesDE\_FCUHRvbrain.txt", quote=F, sep="\t")</pre>

nss = 0 If the experiment didn't include replicas the number of replicates to be simulated is provided by nss parameter

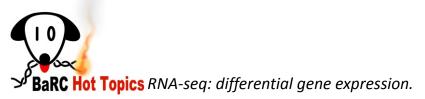
repl = "bio" indicates that the experiment includes biological replicates

- **q** = **0.9** indicates that the probability cut off for considering a gene differentially expressed is 0.9
- Sample code for DESeq, EdgeR and NOISeq: \wi-files1\BaRC\_Public\BaRC\_code\R



## References

- Tophat: http://tophat.cbcb.umd.edu/
- Htseq-Count: http://wwwhuber.embl.de/users/anders/HTSeq/doc/count.html
- EdgeR: PMID: 19910308 http://www.bioconductor.org/packages/release/bioc/html/edgeR.html
- DESeq: PMID: 20979621 http://www.bioconductor.org/packages/release/bioc/html/DESeq.html
- baySeq: PMID: 20698981 http://www.bioconductor.org/packages/release/bioc/html/baySeq.html
- NOISeq: PMID: 21903743 http://bioinfo.cipf.es/noiseq/doku.php?id=tutorial
- *From RNA-seq to differential expression results*. Oshlack A et al. Genome Biol. (2010). PMID: 21176179.

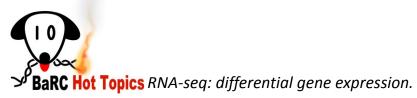


#### Resources

- BaRC Standard Operating Procedures (SOPs) https://gir.wi.mit.edu/trac/wiki/barc/SOPs https://gir.wi.mit.edu/trac/wiki/barc/SOPs/rna-seq-diff-expressions
- BaRC R scripts
   \wi-files1\BaRC\_Public\BaRC\_code\R
- BaRC Hot Topic: Assessing Sequence and Microarray Data Quality http://jura.wi.mit.edu/bio/education/hot\_topics/QC\_HTP/QC\_HTP.pdf
- BaRC Short Course: Introduction to Bioconductor microarray and RNA-Seq analysis

http://jura.wi.mit.edu/bio/education/R2011/slides/Intro\_to\_Bioconductor\_HotTo pics\_Oct\_2011.color.pdf

 BaRC Hot Topic: RNA-Seq: Methods and Applications http://jura.wi.mit.edu/bio/education/hot\_topics/RNAseq/RNA\_Seq.pdf



## Contact BaRC for help

• Please stop by with questions or if you would like to analyze your own RNA-seq data

wibr-bioinformatics@wi.mit.edu

