

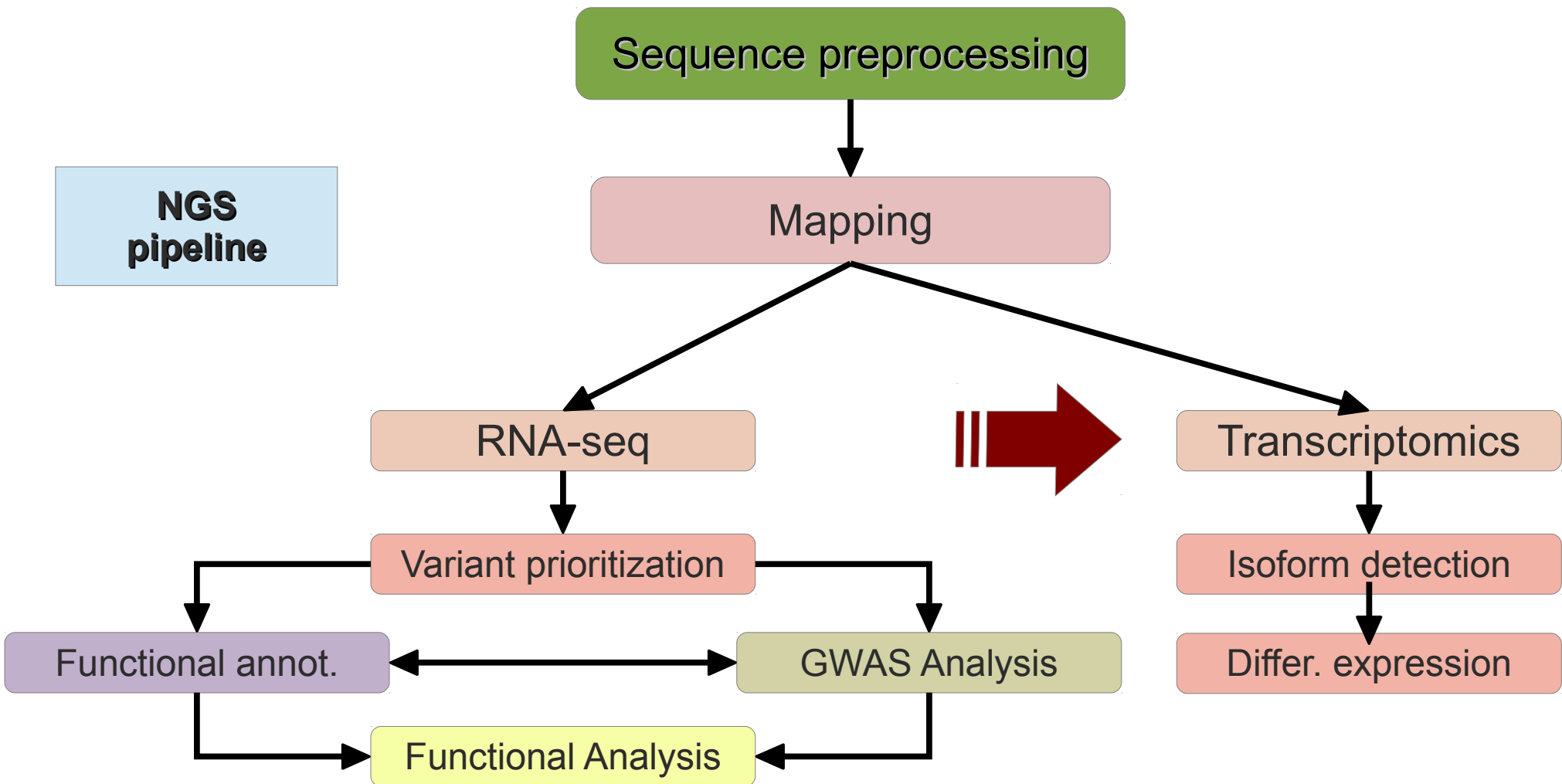


IX International Course of Massive Data Analysis FOR GENOMICS

Course Presentation



Where are we?



General context

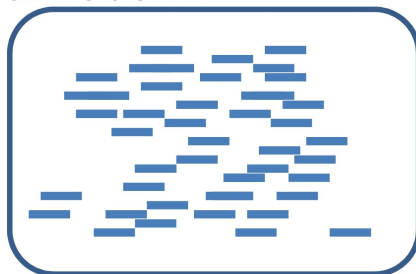
- Reference genome
 - Yes: map against it
 - No: try to assemble transcripts and then map
- Reference transcriptome
 - Yes: estimate known transcripts abundance & discover new transcripts
 - No: find expressed regions & their transcript combination

General context

Sequencing Reads

Reference Genome

Individual A



Sequencing depth

reads

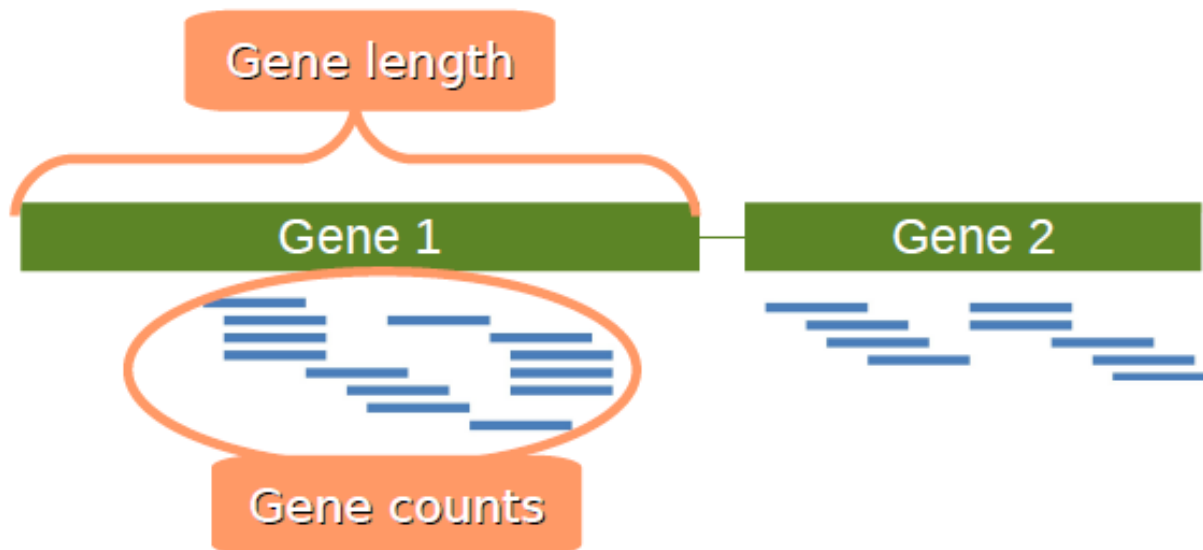


Gene length

Gene 1

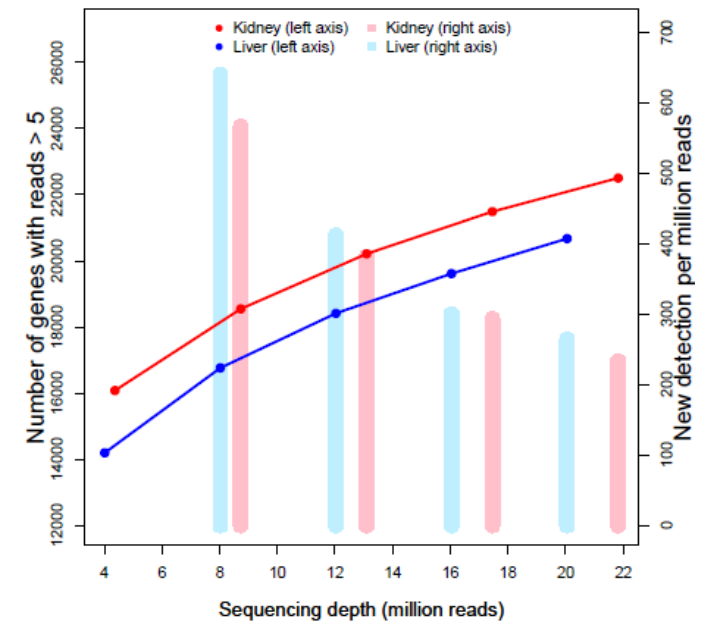
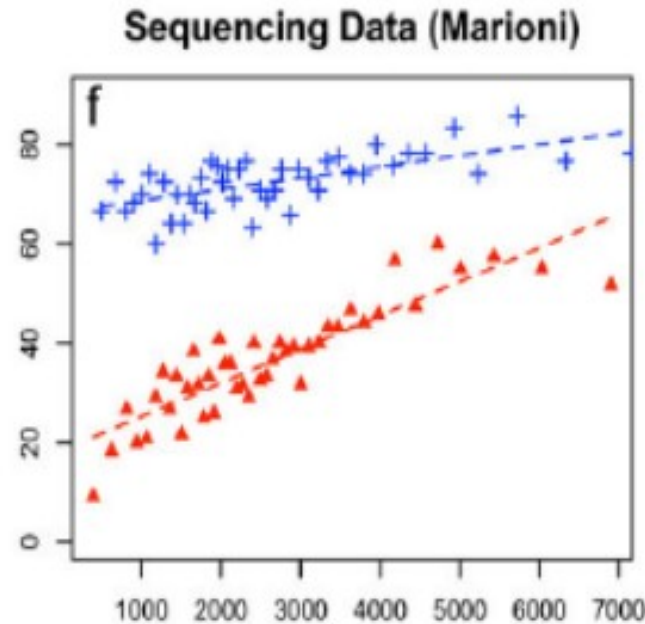
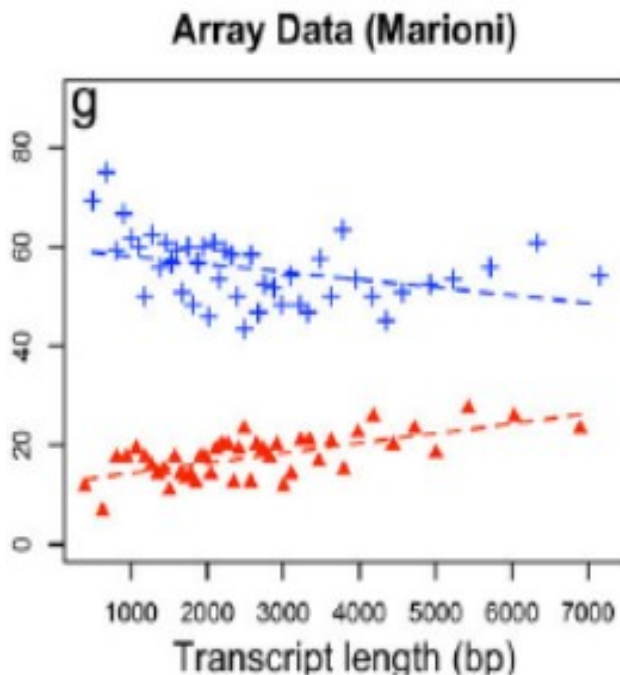
Gene 2

Gene counts



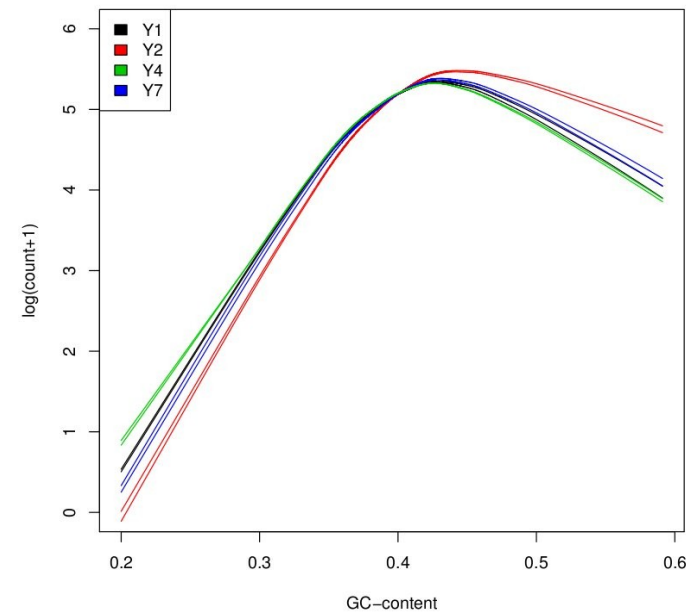
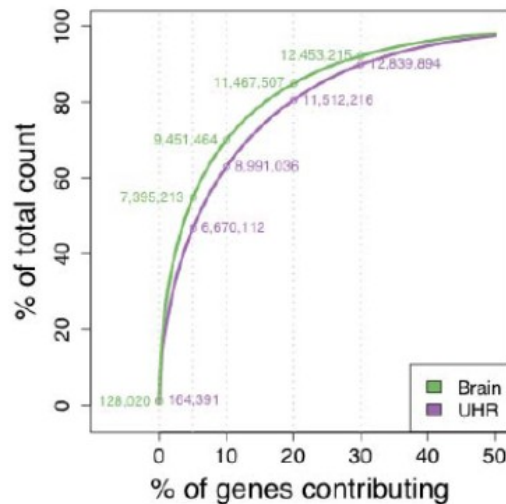
Gene/transcript length dependence

- Counts are proportional to...
 - the transcript length
 - the mRNA expression level.



Count Normalization

- Transcript length: *within* library
- Library size: *between* libraries
- Many other biases ...
 - Differences on the read count distribution among samples.
 - GC content of the gene affects the detection of that gene (Illumina)
 - sequence-specific bias is introduced during the library preparation



Count Normalization

- **RPKM**: Reads Per Kilobase of the transcript per Million mapped reads

$$RPKM = 10^9 \times \frac{C}{N * L}$$

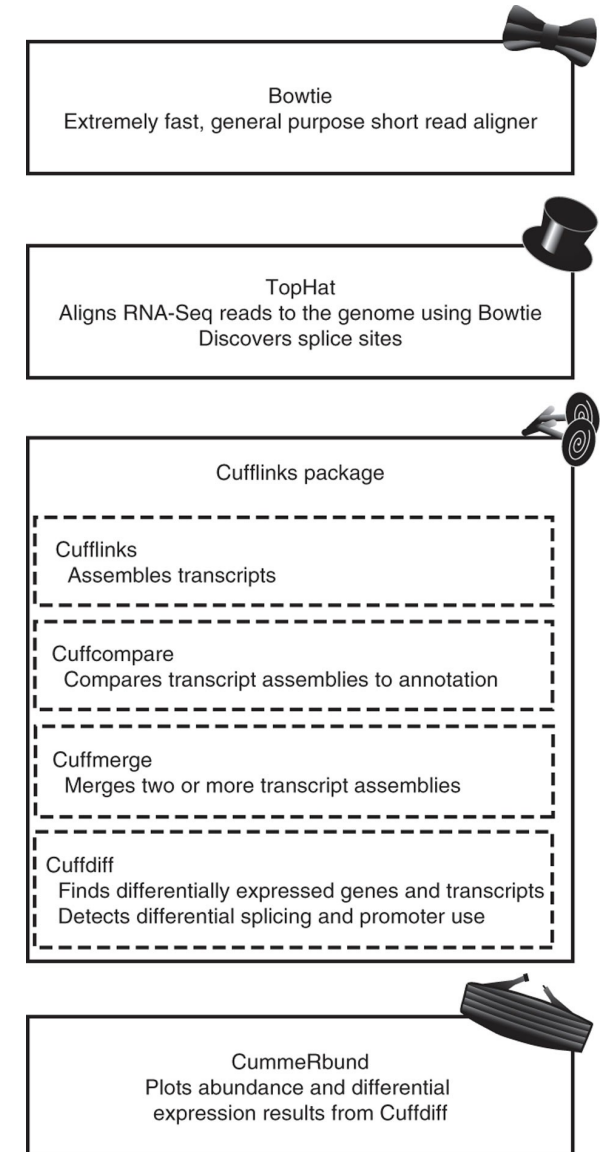
- **C** is the number of mappable reads mapped onto the gene's exons.
- **N** is the total number of mappable reads in the experiment.
- **L** is the total length of the exons in base pairs.
- Fragments Per Kilobase of exon per Million fragments mapped (FPKM),

Count Normalization

- **RPKM** (Mortazavi et al., 2008)
- **TC**: Gene counts are divided by the sequencing depth associated to that sample and multiplied by the average of the total counts across all the samples. Gene counts are divided by the gene length (kb) times the total number of millions of mapped reads.
- **Upper-quartile** (Bullard et al., 2010): Gene counts are divided by the upper quartile of counts for genes with at least one read.
- **Median** (Bullard et al., 2010): Gene counts are divided by the median of counts for genes with at least one read.
- **Quantile** (Irizarry et al., 2003): This method matches the gene count distributions across samples.
- **TMM** (Robinson & Oshlack, 2010): Trimmed Mean of M-values. Based on the hypothesis that most of the genes are not DE. A sample is taken as the reference. For each sample, the scaling factor is the weighted mean of log ratios between this sample and the reference after removing the most expressed genes and those with the largest log ratios. Gene counts are divided by this factor re-scaled by the mean of the normalized library sizes.
- **DESeq** (Anders & Huber, 2010): Based on the hypothesis that most of the genes are not DE. The scaling factor for a given sample is the median of the ratio, for each gene, of its counts over its geometric mean across all the samples. Gene counts are divided by this scaling factor.
- **FPKM** (Trapnell et al., 2010): implemented in Cufflinks, is similar to RPKM .

Software

- **Cufflinks:**
 - assembly
 - compare with the known transcripts
- **Cuffdiff:** differential expression
 - estimate fragment length distribution
 - calculate transcript abundances FPKM
 - sequence bias correction



Software

TopHat

- **Cufflinks:**
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CummeRbund:

an R package to handle results

