RNA Sequencing

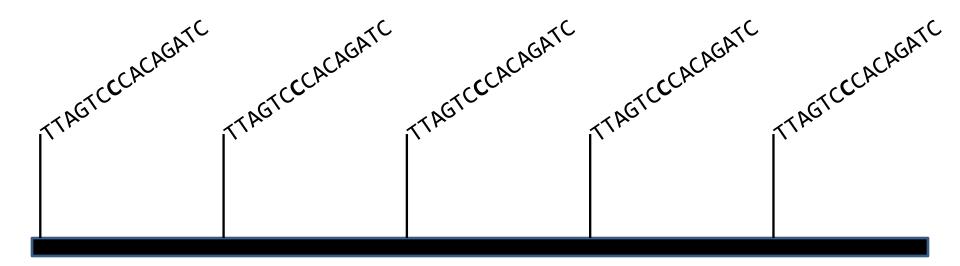
Elisha Roberson, Ph.D.

Depts. of Medicine & Genetics eroberson@wustl.edu 2016-December-06

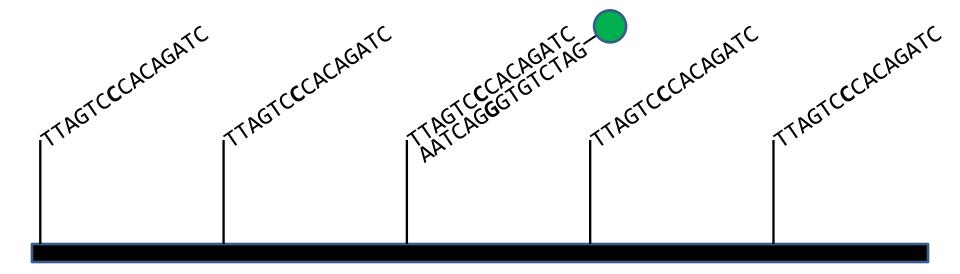




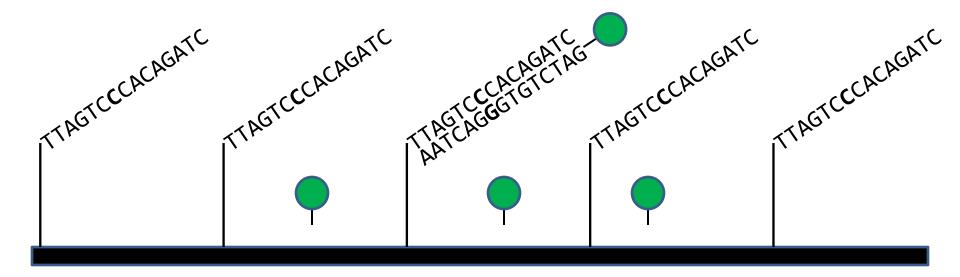
Comparing RNA-Seq and microarrays



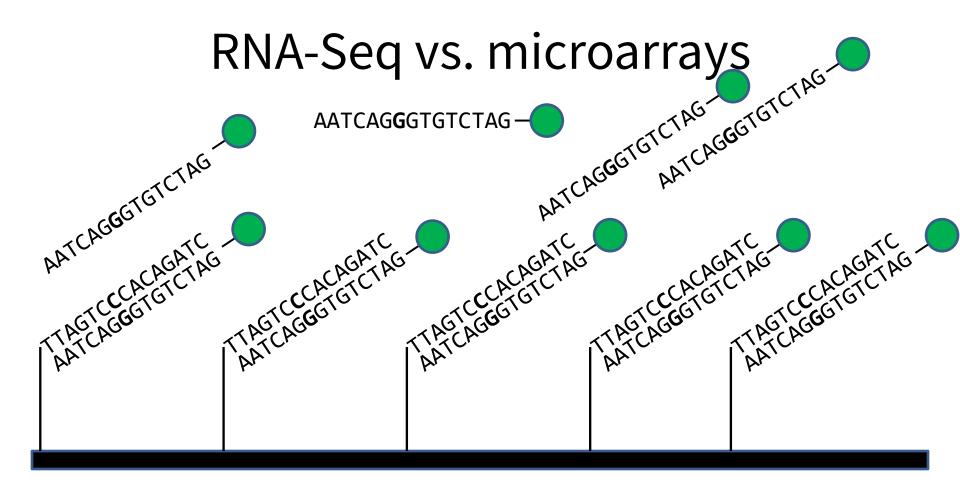
• Hybridization requires known targets



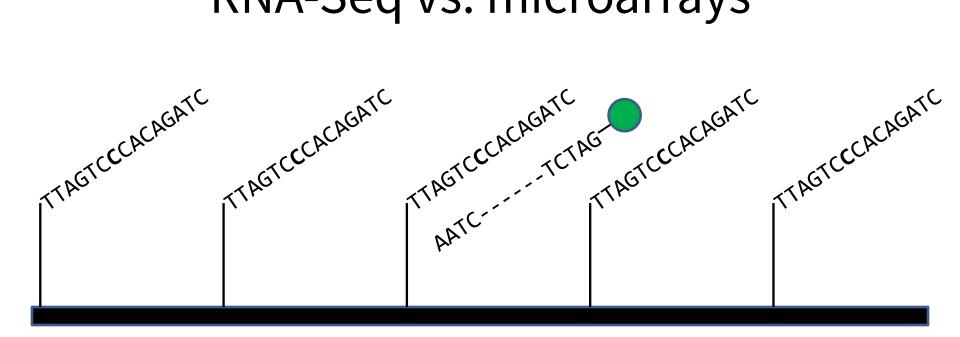
• Expression detected by fluorescence



• Low-level expression can be difficult to detect compared to background



• High-level expression can saturate probes



• Variation in the subject's RNA sequence can affect binding kinetics

RNA-Seq differences

- Counting no probes to saturate (though some sequences can predominate the library) and can always sequence more to get low-expressors
- Doesn't *require* a known gene sequence
- Not affected by variation, as long as it doesn't affect transcript stability

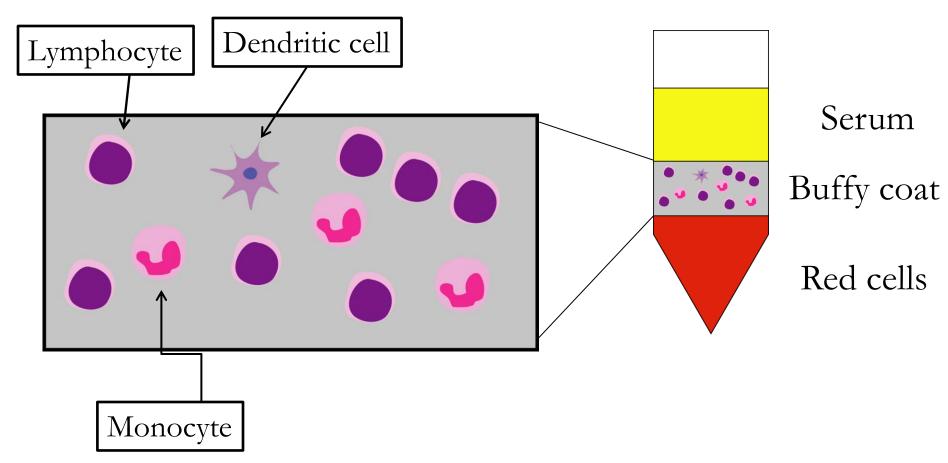
Isoform A Isoform B

Sample prep is *critical*

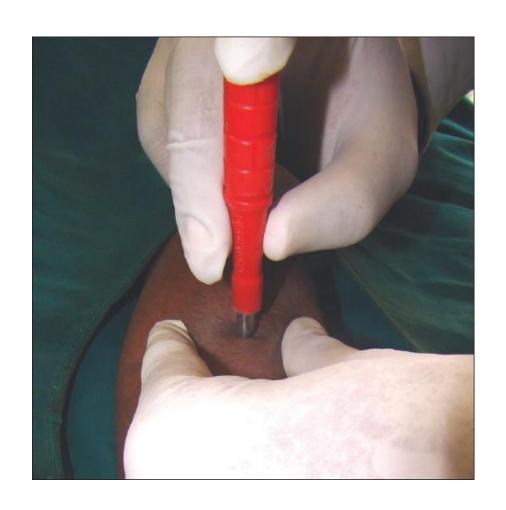


 Blood is low-risk, easy to access, and frequently used

• **But** strongly enriched for red cells. What transcripts will predominate?

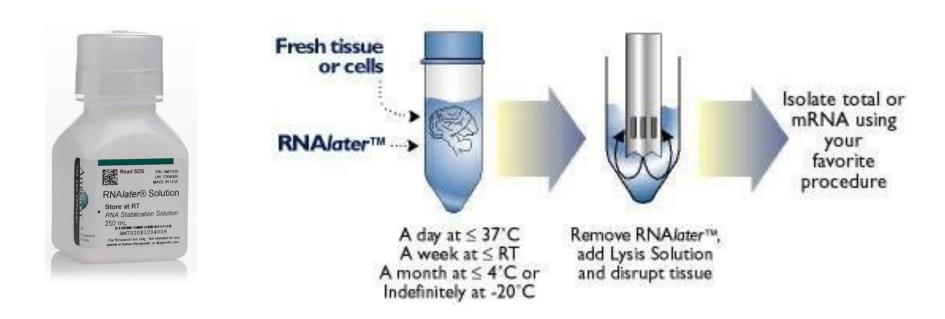


- Freeze PBMCs directly
- Lyse directly in Trizol

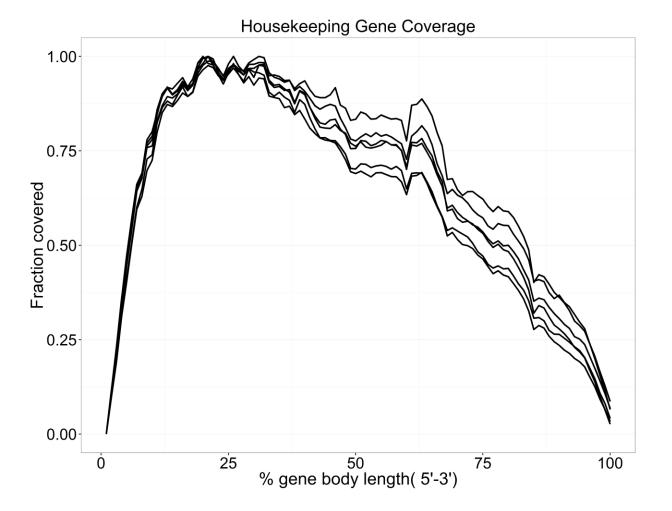


Tissues

- Biopsies, whole organs, etc.
- Unlike blood, the complex structure and embedded connective fibers make tissues more difficult to process and preserve.



• Some solutions, like RNAlater, can preserve tissue RNA for short-term storage at RT or long-term storage at -20°C

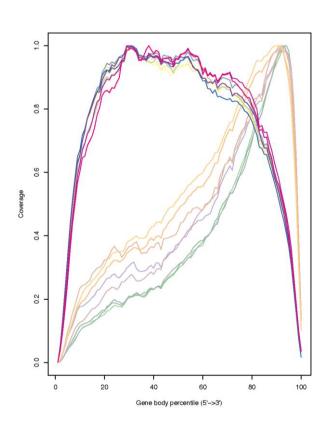


- PBMCs shipped on (too little) dry ice
- Degraded 3'-5'
- PolyA kit fails

Library prep / cDNA methods

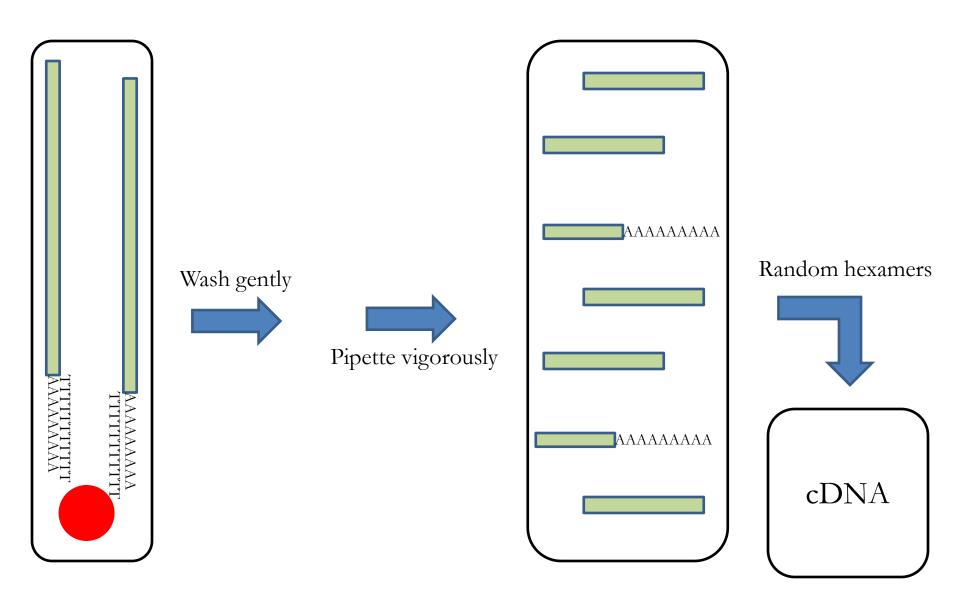
Polyadenylation (polyA) preps





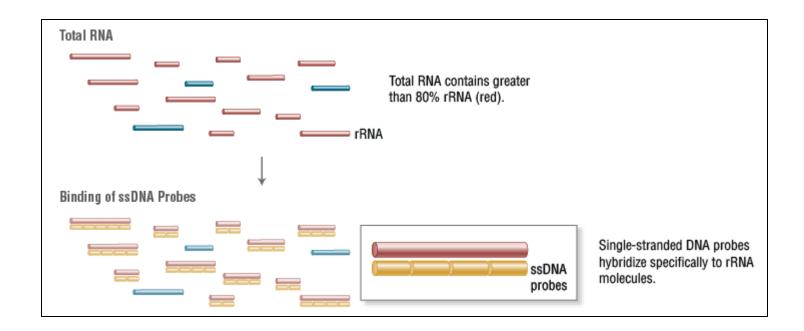
- polyT primed firststrand synthesis works
- **But** can lead to 3' bias and only captures polyadenylated transcripts.

Polyadenylation (polyA) preps

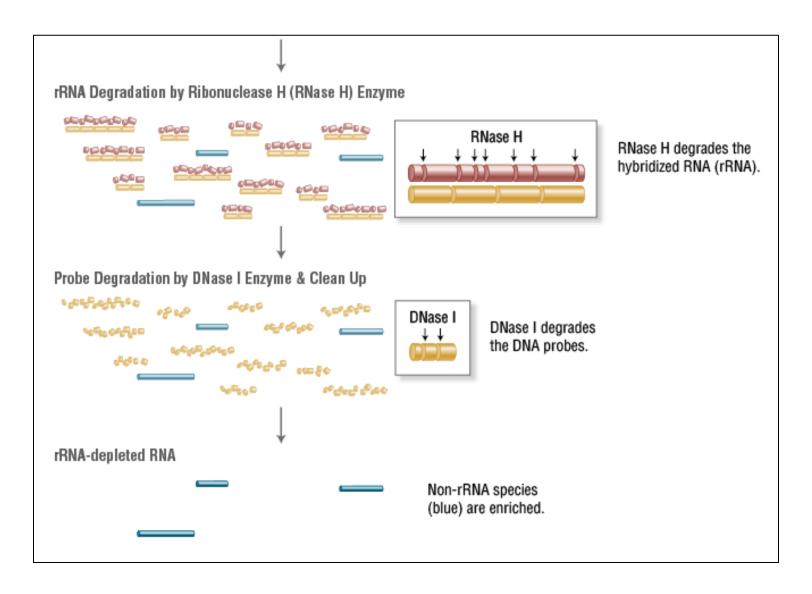


Ribosomal depletion preps

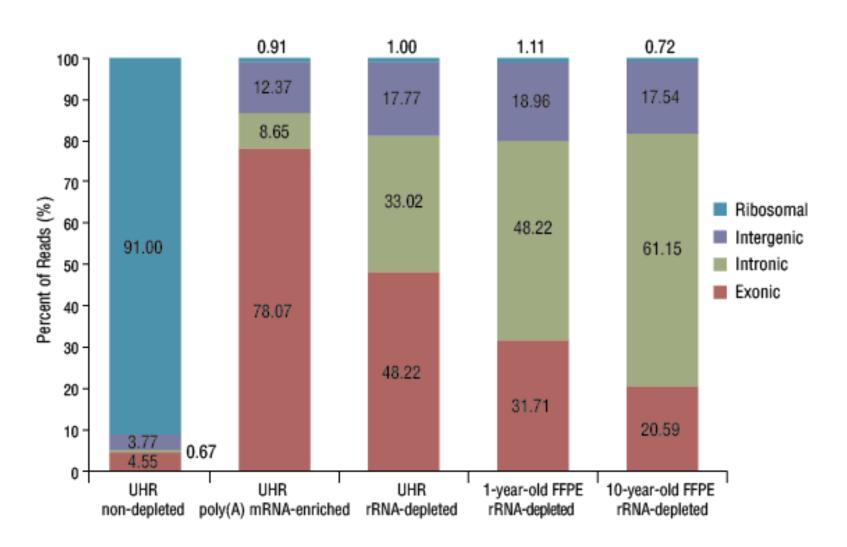
- Majority of cellular RNA is non-coding, particularly ribosomal RNA
- RNA polymerase I (28S, 18S, 5.8S rRNA) and Pol III (5S rRNA)
- Not polyadenylated



Ribosomal depletion preps

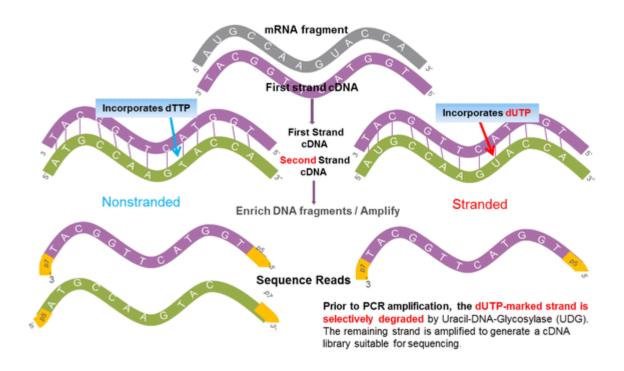


Ribosomal depletion preps

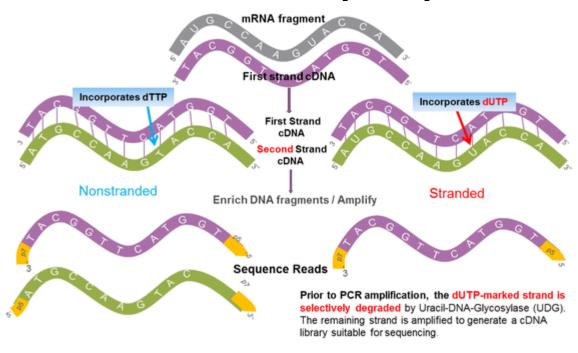


Stranded preps

- Standard cDNA → library prep retains no information about transcript strand.
- Some loci have antisense transcripts



Stranded preps



- Normal first strand synthesis. 2nd strand incorporates uracil
- Uracil-DNA glycosylase excises U-base from DNA
- Endonuclease VIII breaks backbone at those sites

Sequencing choices

Short reads

- Illumina
 - Single-end vs. paired-end
 - Paired-end superior. Estimates insert size empirically
 - Read length
 - Greater cycle number preferred
 - 2x75 good compromise
 - Depth / coverage
 - Very different from DNA seq
 - Variable gene length and expression level
 - Several tools to estimate

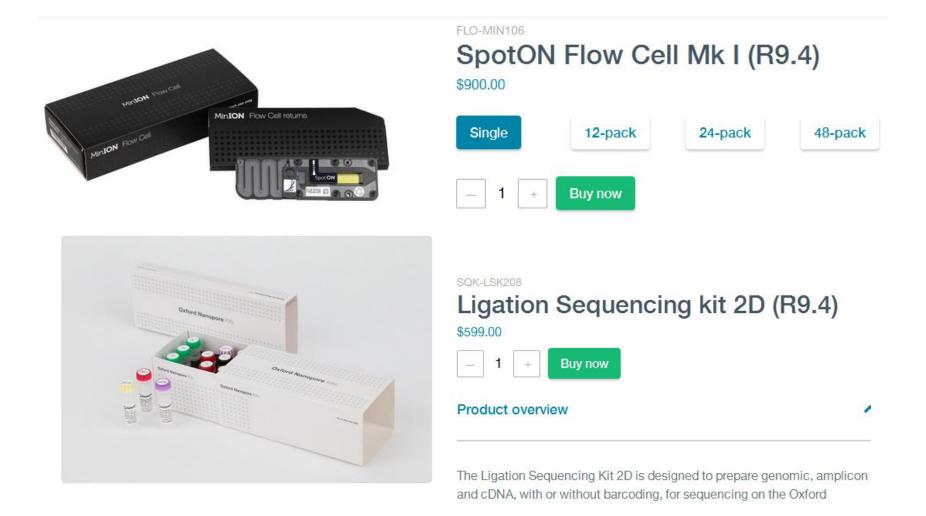
Long reads – Pac Bio SMRT

• Full-length isoform sequencing (\$\$\$\$)

Pacbio Library Construction and Sequencing	Cost Per Sample	
Sequencing SMRT Cell	\$257	Min \$1,100 / sample for whole transcriptome.
Standard Library Prep	\$560	
Low_input Library Prep	\$603	
Iso-Seq Whole Transcriptome Lib_Prep	\$875	
Iso-Seq Targeted Lib_Prep	\$664	

Long reads – Oxford nanopore

MinION / PromethION (also \$\$\$\$\$)



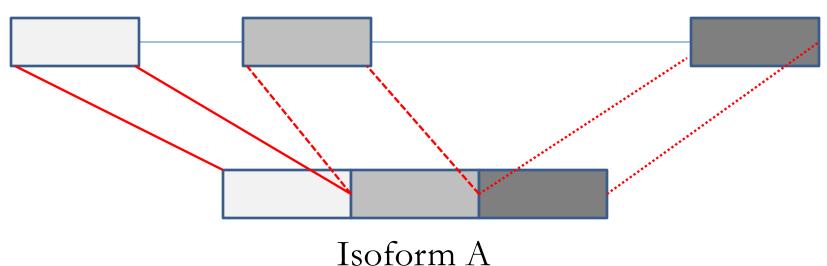
Alignment and counting

Genome locus

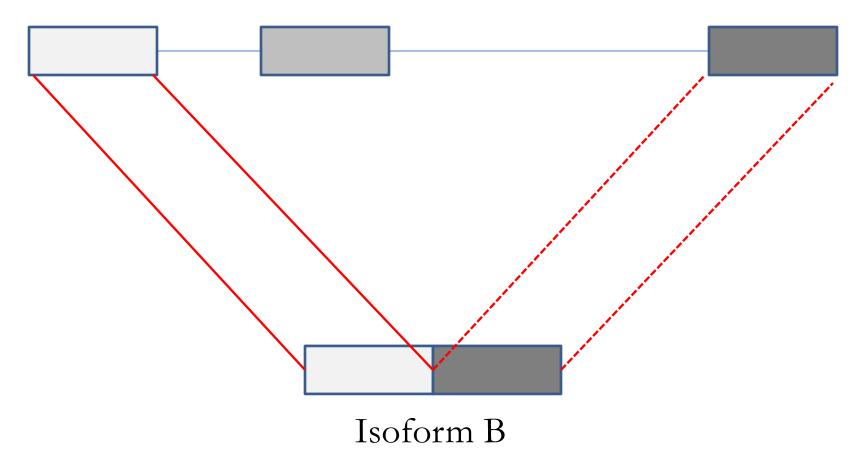
Isoform A

Isoform B

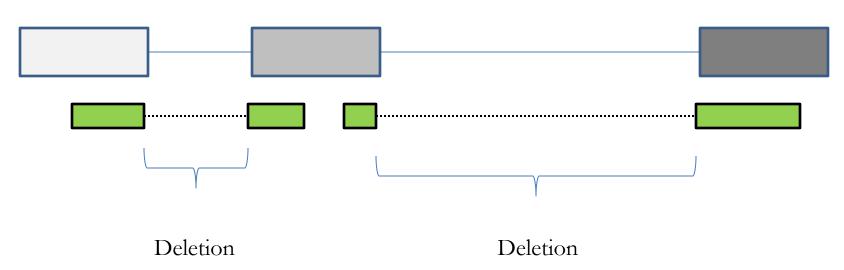
Genome locus



Genome locus



Genome locus



Genomic aligners expect the library to reflect **genome** architecture. Intron splicing looks like large deletions, and can confuse aligner.

One alternative is to align to transcript FASTA rather than whole-genome.

Splice-aware genome aligners

- Tophat2 (bowtie derived)
- Spliced Transcript Alignment to Reference (STAR)

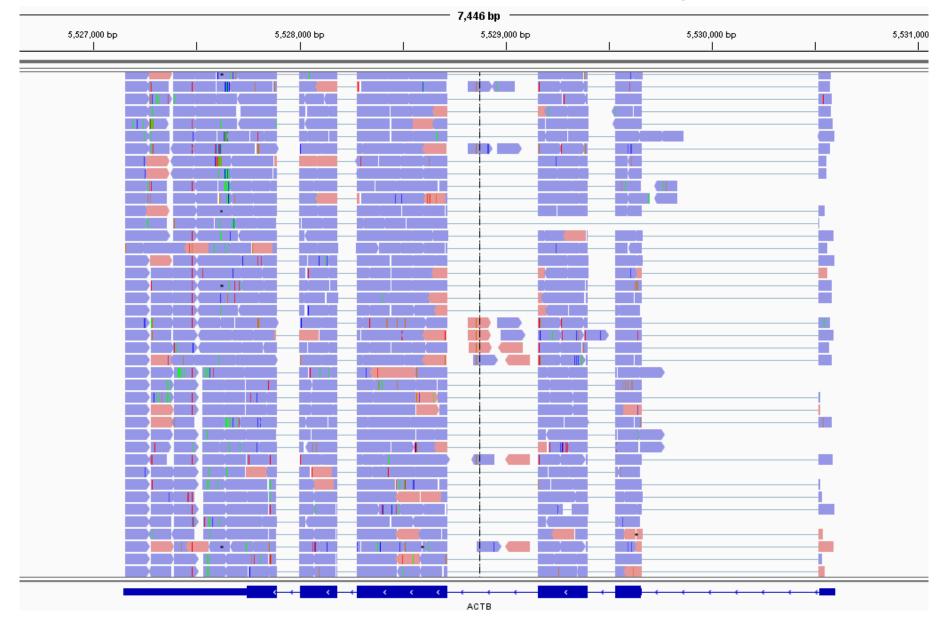
Transcript alignment

- Kallisto
- Sailfish / Salmon

De novo assembly

- ABySS / TransABySS
- Trinity
- SOAPdenovo-Trans

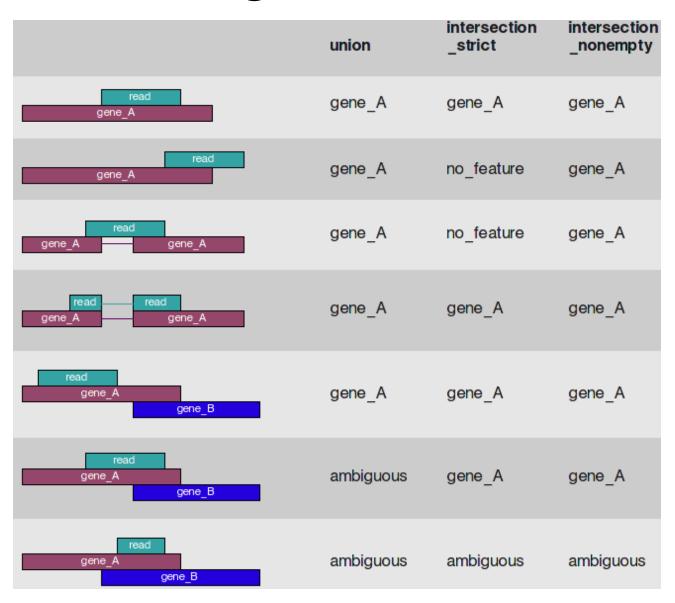
Ribosomal depletion alignment



Counting reads

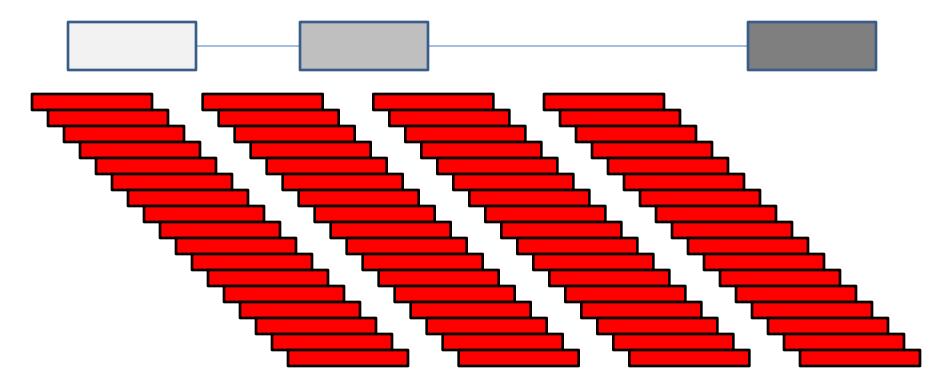
Counting tools

HTSeq count subread featureCount RSEM



RNA-Seq complications - duplicates

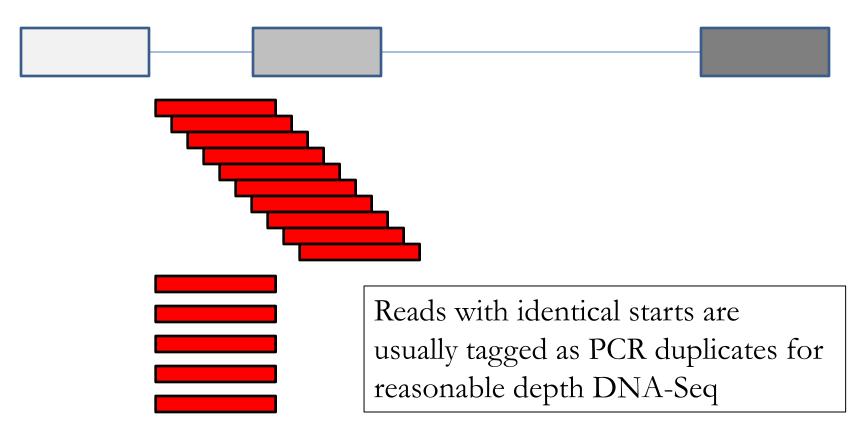


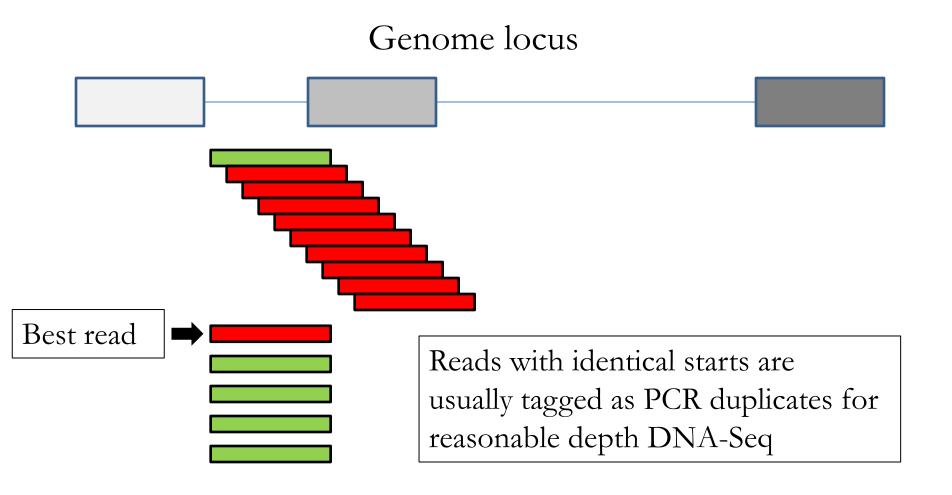


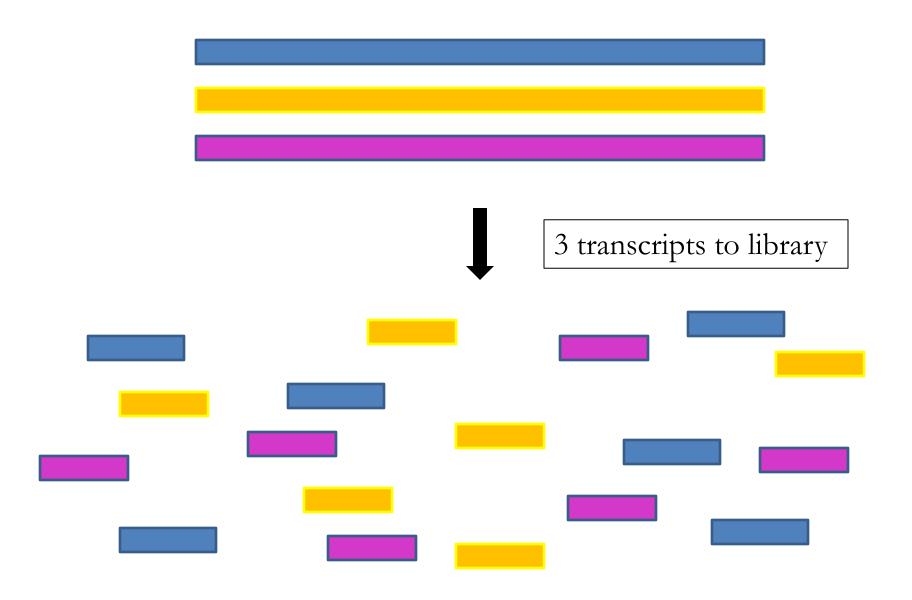
Highly covered genomic locus for DNA-Seq

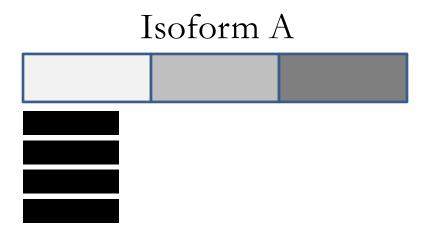
RNA-Seq complications - duplicates

Genome locus

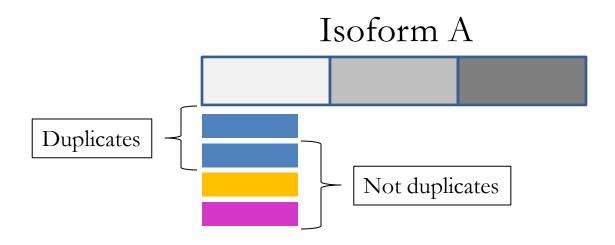








- For DNA-Seq the target is equimolar, but RNA-Seq is more complicated
- Important considerations:
 - Sequencing depth
 - Gene relative expression level
 - Gene size



- But this is a solvable problem
- Adding a short barcode to each fragment during PCR, called a unique molecular identifier (UMI) we know whether reads are truly unique, even with identical 5' mapping

Estimating gene abundance & differential expression

Abundance metrics

• Fragments per Kb of exon per million mapped reads (FPKM)

• Transcripts per million (TPM)

- TPM preferable
 - Different total reads between experiments skew FPKM
 - TPM consistent, i.e. 1 TPM sample A and sample B really means similar abundance

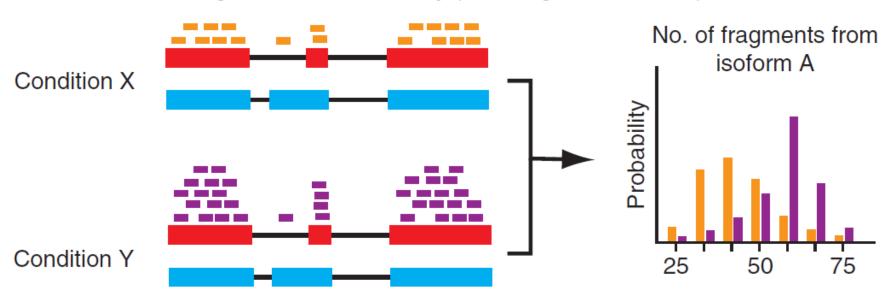
Modeling counts – edgeR, DESeq2

Counts are not normally distributed

- What models counts?
 - Poisson distribution
 - But assumes variance & mean equal
 - Negative binomial
 - Mean ≠ variance
 - edgeR, DESeq2 R packages

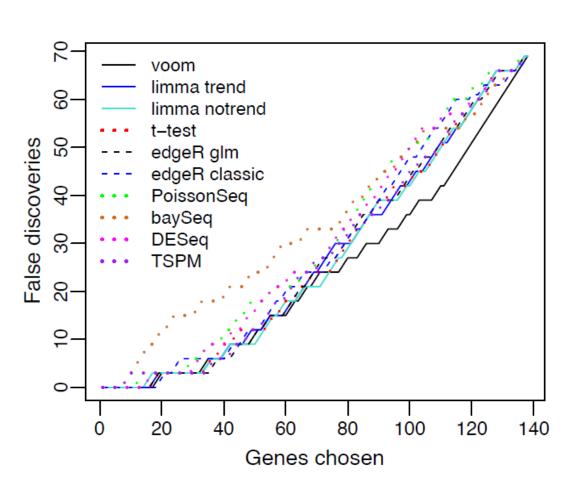
Modeling counts – cuffdiff2

 Combine uncertainty and overdispersion into a single model of fragment count variability (beta negative binomial)



- 5) Test for signficance of changes between conditions in transcript-level counts
- edgeR & DESeq2 model <u>gene-level</u> differential expression
- cuffdiff2 tests for significant isoform-level DE

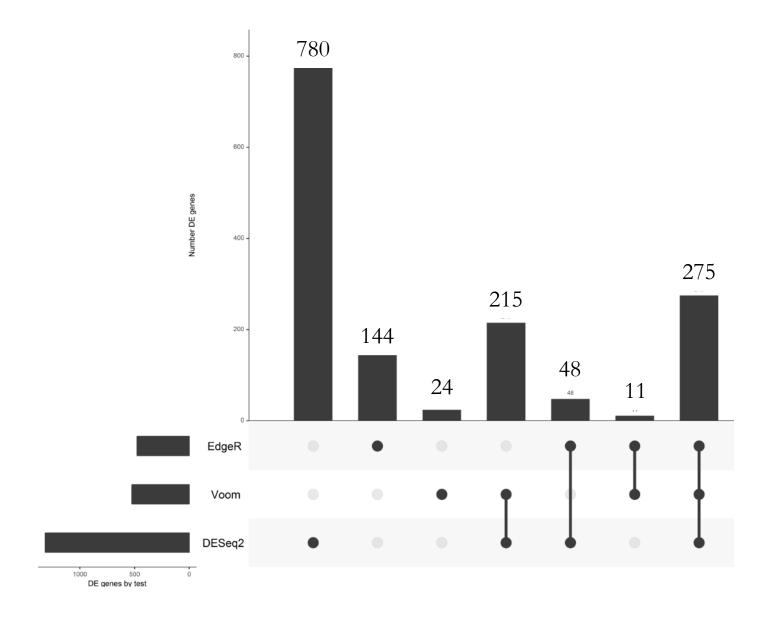
Modeling counts - VOOM



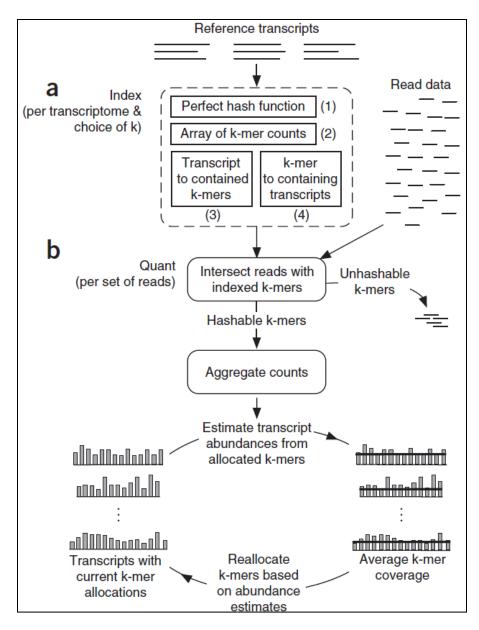
 VOOM uses log2 of counts per million normalization factor

Differential
 expression using the
 empirical Bayes
 limma pipeline

How similar are gene DE algorithms?

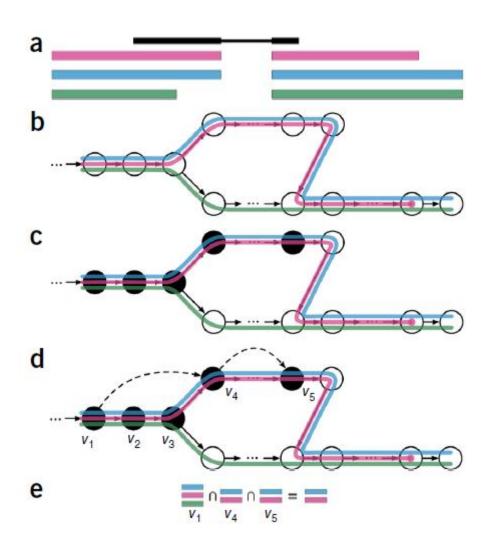


Transcript k-mer modeling - sailfish



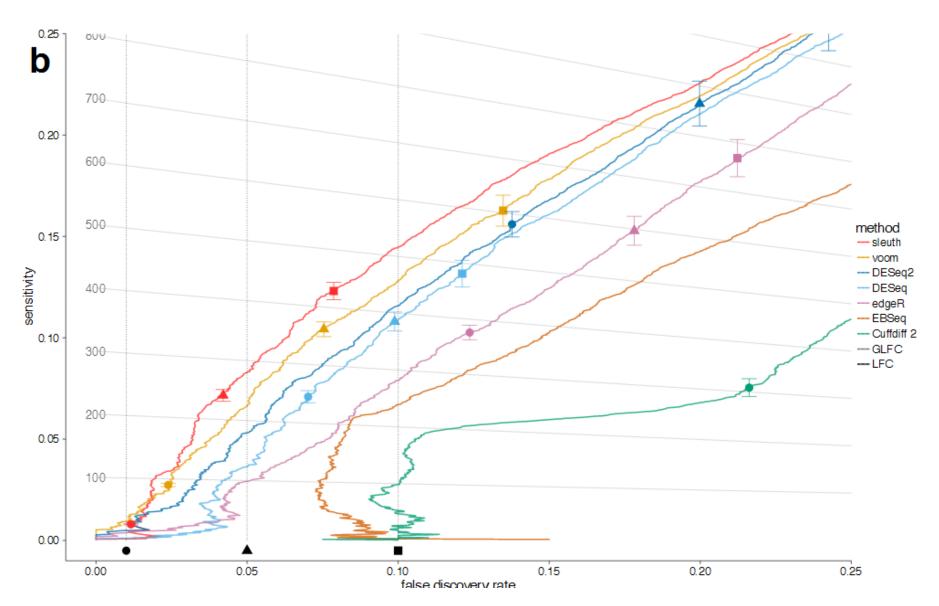
- Hashing uses a large amount of memory
- But lookups are blazingly fast
- Calculates TPMs

Transcript pseudoalignment - kallisto

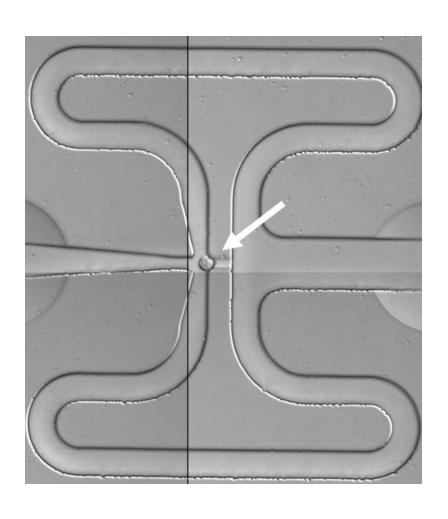


- Builds a de Bruijn graph of transcript sequences
- Pseudoalignment –
 compatible transcripts,
 not where in transcript
- Very fast and efficient
- Allows for bootstrapping

Transcript DE from kallisto - Sleuth

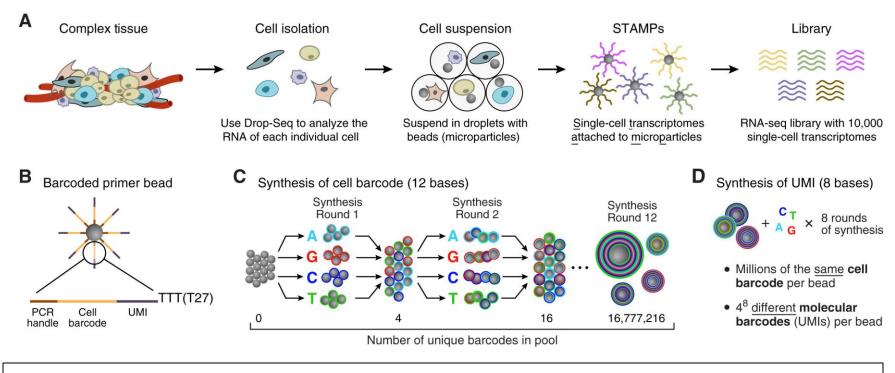


Single cell RNASeq – Fluidigm C1



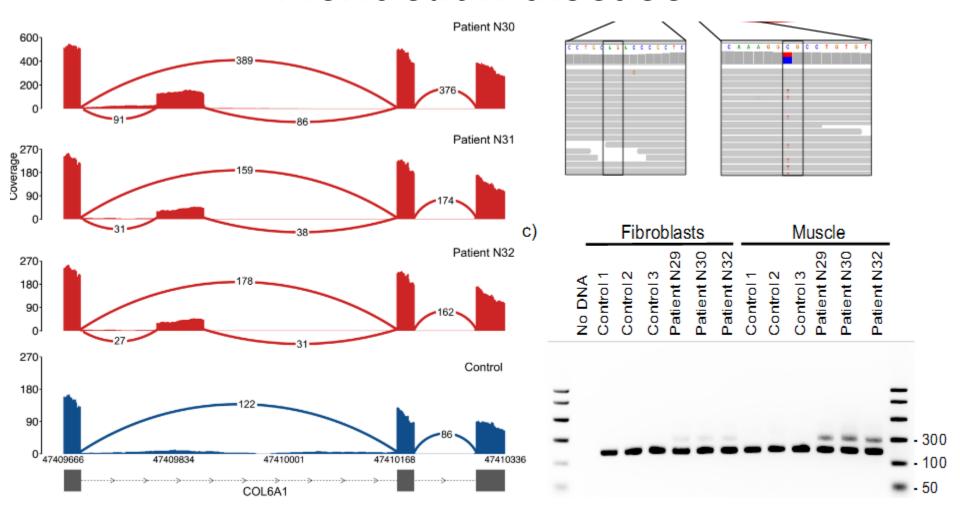
- Microfluidics capture single cells
- Lyse cells and generate cDNA
- Requires live cells
- 96-800 cells / chip

Single cell RNASeq - DropSeq



- Flows beads in a droplet.
- Cells, usu singles, merge into a droplet for library prep.
- 10s of thousands of cells.

RNA-Seq to diagnose Mendelian disease



^{*} Cummings et al. 2015. http://biorxiv.org/content/biorxiv/early/2016/09/08/074153.full.pdf

- RNA-Seq vs. microarrays
 - Microarray requires knowing target sequence
 - Poor dynamic range (hard to detect low-level expressors, saturate at high-levels)

- Sample collection & storage
 - Blood should be spun down to PBMCs
 - Can be directly lysed
 - Tissue should have RNAlater applied soon as possible, followed by disruption.

- cDNA / library prep
 - Polyadenylation library methods selectively capture polyA transcripts
 - Ribosomal depletion methods degrade ribosomal RNA, but leave non-polyadenylated
 - Strandedness
 - There are an appreciable number of genes with antisense transcripts.
 - Also useful for identifying genes in species without a reference genome

- Sequencing technologies
 - Long reads (expensive), but sequence full isoform
 - Short read. Reasonable price.
- Aligner
 - Must use an aligner that is aware of introns
 - May align to either genes (STAR, etc) or transcriptome (Tophat 2 and kallisto)
- Counting
 - Subread featureCounts, HTSeq count, RSEM

- Differential expression
 - Gene, negative binomial: edgeR, DESeq2
 - Gene, log2 counts per million: VOOM
 - Transcript, TPM, kallisto.
 - Transcript, TPM, Sailfish
- Single-cell
 - Isolate a single cell and make a libraries (low-output)
 - Spike-ins help
 - UMIs help also