RNA-Seq: Sequencing the Transcriptome

Kasper Daniel Hansen

Division of Biostatistics, UC Berkeley Using Bioconductor for ChIP-Seq experiments FHCRC, Seattle 12th-14th of November 2008 Potential for surveying the entire transcriptome, including novel, un-annotated regions.

Potential for determining gene structure and isoform level expression using reads mapping to splice junctions.

Potential for making better presence/absence calls on regions.

Con: the assay is dependent on sequencing effort, low expressed regions will be missed.

The current standard protocol for RNA-Seq is

Extraction of RNA, polyA purification Fragmentation of RNA RT of RNA to cDNA Ligation of adapters Size selection ~ 200bp (perhaps ~300bp) PCR amplification (15 rounds) Injection into flowcell

This produces reads from polyadenylated RNA without strand information.

Attempts are being made to make the assay strand specific and to assay total RNA as well.

Data from D. melanogaster



Image from Brenton Gravely

Base effect - single sample



position

Base effect - multiple samples





Base effect - different study (and prep)



position

Base effect - different prep



position

MAQ and ELAND, Human data



position

Reproducible base effect - like probe affinities in microarrays.

Seems to be prep dependent.

Creates issues for comparing different regions in the genome.

Less of an issue for comparing the same region across samples?

Mapping reads to the transcriptome



Illustration from Lior Patcher

Mapping transcripts



Junction reads



Image from Brenton Gravely

Junction reads, zoom



Image from Brenton Gravely

Map to known junctions.

Map to combination of known exons.

Map completely de-novo using canonical acceptor and donor sites. The combinatorics makes this an intimidating approach.

Map de-novo, but constrain the search to canonical acceptor and donor sites between and in transcribed region: transcript assembly. This is the approach taken by TopHat.

Paired-end data will make de-novo mapping a real possibility.

Mapping to transcript space is not easy.

But essential for really understanding alternative splicing.

Constructing all novel splice junctions based on canonical splice sites but only accepting splicing within genes (and small regions upstream/downstream of the gene) in *D. Melanogaster* yields 605,000,000 splice junctions.



Detection in Cerevisiae



<u>Background</u>: outside any transcribed feature, subtracted a boundary, subtracted any region detected as transcribed in recent studies

Detection in Drosophila



percent of intergenic regions detected

Sources of variation

Lane variation

Flowcell variation

Library prep variation

Biological variation

good fit

Poisson model

less good fit

Systematic differences

?: Is absolute quantification possible

We have developed two **R** packages to help us

GenomeGraphs (Durinck, Bullard) : plots annotation and experimental data along a genome. Makes it easy to construct high quality images as well as to do data exploration. Available from Bioconductor.

Genominator (Bullard, Hansen) : provides support for managing, accessing and analyzing data oriented along a chromosome, together with annotation. Uses a SQLite backend. Works very well for unpaired reads mapped to the genome. Available from our home page.

R> summarizeByAnnotation(expData, annoData, fx)

Genome Graphs, example



mcl.9, chr: 5

Acknowledgements

Statistics Jim Bullard Sandrine Dudoit Elizabeth Purdom Margaret Taub Steffen Durinck Terry Speed

RNA assembly Cole Trapnell <u>S. Cerevisiae</u> Gavin Sherlock Albert Lee

D. Melanogaster Brenton Gravely Mike Duff Li Yang Steven Brenner Angela Brooks