Yanagi: Transcript Segment Library Construction for RNA-Seq Quantification

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Motivation

- Challenges with Transcript-level Quantification, e.g.
 - 95% of human genes with multiple exons undergo Alternative Splicing (AS)^[1]



- Ambiguity due to multi-mapped reads
 - Usually resolved using probabilistic models like EM





Motivation

- Challenges with Transcript-level Quantification, e.g.
 - Considering local splicing variations leads to a combinatorial number of transcripts.^[2,3]
 - Standard Annotations list only a minimal subset

 Short-read sequencing does not provide information for correlation between distant splicing events.^[4]





Motivation

- Our Vision:
 - Eliminate multi-mapping trivially caused by the significant share of genomic regions.
 - Building sufficient statistics describing individual events.
 - Independently from the estimation of transcript abundances.
 - Utilize the graph representation of the transcriptome.
 - Without building a special graph-based aligners.





Our Approach Transcriptome Segmentation





- Idea Overview:
 - Segment the transcriptome into a set of disjoint regions.
 - Without losing any possible transcriptome sub-sequences.
 - I.e. Linearizing the splice graph
 - Then use the generated segments as a reference instead of the transcriptome.





• A Segment:

seg(Exs, loc, w)

- Segments are *L*-Disjoint width[overlap(seg_i, seg_j)] < L; $i \neq j$
- L corresponds to the read length
- No read of length at least *L* can map to both segments
 - Ignoring sequencing errors and paralogs for now!





- Generating Segments:
 - Naïve Approach



~30% of exons in UCSC hg38 are shorter than 100bp





- Generating Segments:
 - Yanagi's Approach



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• Segments Graph





Yanagi-based Quantification Workflow



Experiments Segments Analysis



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Segments Analysis

Segments Length







Segments Analysis

• Number of Segments







Use Case: Alt. Splicing Quantification

Differential Exon Skipping



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Differential Exon Skipping

- Synthetic Data:^[5]
 - 2 conditions, 3 replicas each.
 - Simulated reads are based on real RNA-Seq data.
 - For 1000 genes with at least two transcripts.
 - Transcription levels of the most abundant two transcripts are switched across conditions.
- Differential Analysis:
 - Exon Skipping events.
 - Linear Model based on the segment counts.
 - Using Limma-Voom.





Differential Exon Skipping

ROC plots:



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Use Case: Linearized Population Reference Graph

Aligning Over Genomic Variants for WGS



Background

- Several databases of Genomic Variants
 - Rapidly-growing, public archives.
 - E.g. IPD-IMGT/HLA Database currently has 17,344 allele sequences.



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Background

• In principle, Graphs are reasonable representation.





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Yanagi + Genomic Variants

- Project Variants Graphs into Splice Graphs.
 - Start from Multi-Sequence Alignment (MSA) of gene alleles





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Yanagi + Genomic Variants

- Preliminary Results: Number of Aligned Reads •
 - Simulated reads from 6 HLA genes

Dataset	Total Reads	HISAT- genotype	HISAT- genotype bwa-mem		RapMap		
		(WG Graph)	Ref	Ref+Segs	Ref	Ref+Segs	
ClassIEasy	6,000	5,900	6,000	6,000	4,163	5,990	
ClassIHard	6,000	5,966	5,797	6,000	3,553	5,990	
ClassIIHard	14,000	13,844	12,232	13,997	7,628	13,975	
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Summary

• Yanagi perform a transcriptome segmentation into *L*-disjoint segments.

• Introduces fine-grained counts as statistics for DE analysis.

- Flexible approach that can be used in different use cases:
 - Alternative Splicing Quantification
 - Variant Calling





Future Extensions

- Further Analysis, e.g.
 - Experiments on Real Data
 - Comparison with other tools, e.g. SUPPA2 for AS
 - Detailed analysis on multi-mapped reads
- Discovering unannotated transcripts
- Handling paralogs and intersecting genes
- Handling complex repeats and structural variants







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Thank you!

Questions?

