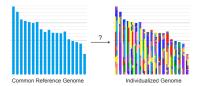
Quantifying gene and allele-specific expression simultaneously using personalized human genomes



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Large-scale genome sequencing efforts have characterized millions of common genetic variants across human populations. However development of tools that can effectively utilize this individual-specific variation to inform quantitation of gene expression abundance have lagged behind. We present tools to utilize individual genetic variation in RNA-seq and quantify gene expression and allele-specific expression (ASE)



- Seqnature: Builds individualized genomes using known genetic variations: SNPs and Indels (Munger, Raghupathy et. al. submitted to Genetics)
- **EMASE: Simultaneous quantitation of gene expression and allele-specific expression (Raghupathy, Choi et. al. to be submitted to **Genetics**)
- We use Seqnature to build personalized diploid human genome using 1000 Genomes variation data and apply EMASE to quantify gene expression and allele specific expression simultaneously.

Genetic Variations Matter in RNA-seq **Analysis**

If one is working with RNA-seq data from a genetically diverse population, a large number of genes can have one or more SNPs and indel

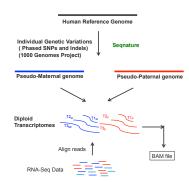
Genomic SNPs Genomic Indels	~ 3.6 Million ~2.5*10 ⁵
Coding SNPs	~25000
Coding Indels	~200

SNP/Indel Statistics: An YRI Individual from 1000

Aligning short sequencing reads to a common reference genome is the first step in RNA-seq analysis. Genetic variations present in the sample, but not in the reference genome can lead to misalignments and incorrect expression estimates and biased allele-specific expression.

SEQNATURE

Sequature incorporates known SNPs and indels in to reference genomes to construct individualized diploid or haploid genomes, for human, genetically diverse heteroxygous model organisms and inbred strains. Sequature also updates the annotation file and it is readily usable for read alignment by common aligners.



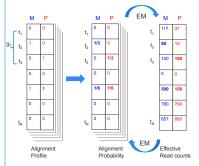
Segnature is an open access tool written in Python and a beta version is available at

EMASE: EM for Allele Specific Expression

EMASE is a model-based quantitation approach that employs an Expectation Maximization (EM) algorithm to apportion

- · gene multireads
- isoform multireads
- · haplotype multireads

EMASE takes alignment results from diploid transcriptome from a common aligner and uses an EM algorithm to estimate gene expression and ASE simultaneously.



Alignment profile and probability are sparse matrices of size NxRx2, where N is number of transcripts, R is number of reads.

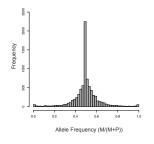
EMASE Offers Tools to Mine RNA-seg Alignments

In addition to estimating expression and ASE simultaneously, EMASE offers functions to mine RNA-seq alignments, like information content in terms of uniquely aligned reads, alignment probabilities, mappability, and analyzing alignments from simulation.

EMASE is implemented in Python. It is available upon request now and will be available soon from http://ord.iav.org/

Allele Specific Gene Expression

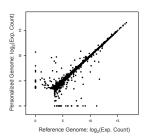
A common problem in quantifying ASE from RNA-seq data is the alignment bias due to common reference genome. Our diploid model accounting for known genetic variations removes the alignment bias. The allele frequency histogram below is symmetric without any reference bias and it shows allele specific expression quantitative.



Gene Expression: Common Reference Genome vs. Personalized Genome

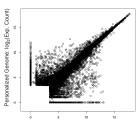
Comparing alignment and gene expression results from using common reference genome and personalized of alignments and expression estimates onalized diploid genome show improvement in

In a single sample, typically about 500-1000 genes show improved expression abundance estimate from aligning to diploid transcriptome (slimulation results not shown). This is mainly due to gained alignments from reads that did not align when a single reference transcriptome is used and the ability for sealve alignments to pseudo-genes and homologous gene families correctly. For example, SNPs/Indels between a gene and its pseudo-gene, and a gene and its homologus family members help differentiate the alignments.



Population-level gene expression analysis

Using personalized diploid genome/transcriptome for expression quantitation is highly relevant for population level RNA-seq analysis. Comparing over 50 RNA-seq sangles from Happhag1000 Genomes forculan population show that about 4,000 genes improve in expression estimate and lead to a better understanding of CIs regulation.



Reference Genome: log₂(Exp. Count)

References

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Conclusions

Current RNA-seq approaches employ two steps to quantify gene expression abundance and ASE from RNA-seq data

- · ASE is assessed separately by analyzing only reads that overlap known SNP locations. Often multimapping reads are ignored in ASE quantitation

Both the gene expression and ASE estimates suffer from alignment biases if one is working with genetically diverse population. We developed complimentary software (open access) tools Segnature and EMASE that can account for individual genetic variations in a diploid model and quantify gene expression & ASE simultaneously by using gene, isoform, and haplobye level multireads.

Our results show that using diploid transcriptome at the alignment step and dealing with multireads using an EM approach results in better estimates of gene expression and ASE. Our results also show the potential use of diploid model in understanding cis-regulation using population data.