# Accurate identification of A-to-I RNA-editing in human by transcriptome sequencing

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# Outline

### Introduction

### Methods

- Reads mapping
- Identification of (putative) RNA editing sites
- Evaluation of mapping bias for single-nucleotide differences

### Validation of predicted A-to-I editing events

### Other results (selected)

- Characterization of predicted A-to-I editing events
- A structural motif in ADAR editing
- Other types of DNA-RNA differences

### Discussion

# Introduction

# • Use transcriptome sequencing data (RNA-seq) for global identification of RNA editing.

- The RNA-seq data:
  - a human glioblastoma cell line: U87MG.
    - Samples are transfected with either a siRNA that targets the ADAR gene or a control siRNA.



# Introduction

- Use transcriptome sequencing data (RNA-seq) for global identification of RNA editing.
- The RNA-seq data:
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# Introduction (contd.)

- 9,636 DNA-RNA differences (RDDs) were identified, and 62% (5,965) are putative A-to-I editing sites.
- Estimation editing levels from RNA-seq correlated well with those based on traditional clonal sequencing.
- Genes with predicted A-to-I editing were significantly enriched with those known to be involved in cancer.
- Similar results are obtained from primary breast cancer samples despite their difference in cell type, cancer type, and genomic backgrounds.



# Restrictions of previous bioinformatic methods

Identify disparities between DNA and RNA sequences by analyzing cDNA, EST, and gDNA.

- Require priori knowledge of editing patterns to restrain the search.
  - The feature of clustering of putative editing sites;
  - The presence of dsRNA structure;
  - . . .
  - \* However, incorporation of such constraints often *limits* the results to editing sites with the corresponding characteristics.
- The estimation of RNA editing levels is usually not afforded.



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# Identification of RNA-editing sites





Identification of RNA editing by RNA-seq Methods

Reads mapping

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Identification of RNA editing by RNA-seq Methods Reads mapping

# Reads mapping

- Map each end of the paired-end reads to hg19 genome using a combination of tools (:. they could differ significantly for some reads):
  - Nowtie, BLAT, TopHat.
  - Exon-exon junction allowed: BLAT and TopHat.
  - The mapping parameters are given in the paper (p. 149).



Identification of RNA editing by RNA-seq Methods Reads mapping

# Reads mapping (contd.)

- Initial mapping:  $\leq 12$  mismatches in each 60-nt read.
- All mappings of each pair of reads were examined to determine **if they pair correctly** (with the expected orientation & the distance between the pair being < 500,000 bp in the genome).
- Require that the pair of reads:
  - map uniquely (as a pair, not necessarily individually) with ≤ 5 mismatches on each reads,
  - o NOT map to anywhere else in the genome as a pair with ≤ 12 mismatches.



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# Identification of RNA editing sites (I)

- For homozygous sites derived from the U87MG genome sequencing data,
  - pile up reads overlapping these sites;
  - examine whether mismatches to the genome sequence existd in the RNA reads;
  - Remove all duplicate reads within each RNA-seq library.
    - $\therefore$  amplication bias in the RT-PCR process  $\Rightarrow$  for the accuracy of the estimated editing ratio.



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# Identification of RNA editing sites (II)

- Infer the strand of the reads based on the strand of genes they were mapped to.
  - Reads mapped to regions with bidirectional transcription (sense & antisense gene pairs) were discarded.
  - for comprehensive gene annotation: Ensembl, RefSeq, UCSC KnownGenes, Gencode genes, and VegaGenes.
  - Extend teh gene boundaries by 1kb each beyond the two ends.



# Identification of RNA editing sites (III)

- A statistical approach to see whether RDDs are likely authentic.
- Calculate the prob. of observing the specific nucleotide (n) for A-to-I editing assuming that
  - the site is *edited* with the true editing ratio r;
  - the quality score of the observed *n* is *q*;
  - the position of *n* in the read is *i*.

$$\mathbf{Pr}[n \mid r, q, i] = \mathbf{Pr}[n \mid \mathsf{freq}(A) = 1 - r, \mathsf{freq}(G) = r, q, i].$$

- Assume that q and i affect the likelihood of a base-call being a sequencing error (similar to the approach used by SNP calling algorithm by Li & Durbin 2009; Li et al. 2009).
- The optimal r: the one maximizing the above function.



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# Identification of RNA editing sites (III contd.)

• LLR to evaluate the significance of a predicted event:

$$\mathsf{LLR} = \mathsf{log}_{10} \left( \frac{\max \left\{ \mathsf{Pr}[n \mid r, q, i] \right\}}{\mathsf{Pr}[n \mid r = 0, q, i]} \right).$$

\* r = 0: not editing.

• Use LLR  $\geq$  2.

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- Indicating that the site is 100 times more likely being a true locus with RDD than a result of sequencing error.
- Require ≥ 2 edited reads and ≥ 5 reads in total for each considered site.
- Mismatches within the first and last five bases of a read were discarded.



Identification of RNA editing by RNA-seq

#### Methods

Evaluation of mapping bias for single-nucleotide differences

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# Evaluation of mapping bias

### Relatve ratio:

$$\frac{\frac{N_{\text{mapped\_ref}}}{N_{\text{simulated\_ref}}}}{\frac{N_{\text{mapped\_ref}}}{N_{\text{simulated\_ref}}} + \frac{N_{\text{mapped\_edit}}}{N_{\text{simulated\_edit}}} := \frac{\alpha}{\alpha + \beta}.$$
Hence,  

$$\frac{\alpha}{\alpha + \beta} = \frac{1}{2} \Rightarrow \alpha : \beta = 1 : 1$$
That is,  

$$\frac{N_{\text{mapped\_ref}}}{N_{\text{mapped\_ref}}} = \frac{N_{\text{simulated\_ref}}}{N_{\text{simulated\_edit}}}.$$

$$\frac{1.0}{\alpha + \beta} = \frac{1}{2} \Rightarrow \alpha : \beta = 1 : 1$$

$$\frac{N_{\text{mapped\_ref}}}{N_{\text{mapped\_edit}}} = \frac{N_{\text{simulated\_edit}}}{N_{\text{simulated\_edit}}}.$$



Identification of RNA editing by RNA-seq Methods Evaluation of mapping bias for single-nucleotide differences

# Evaluation of mapping bias (contd.)

- Simulate 870,280 reads (60nt in length) covering 21,757 heterozygous genomic sites assumed to have alternative alleles (1:1 ratio).
- 40 pairs of reads were generated to overlap each genomic site with a random (uniformly) insert size in the range of [60, 240] bp and random start position relative to the site.
- The base at the heterozygous site was chosen as one of the alternative alleles with *equal probability*.



Identification of RNA editing by RNA-seq Validation of predicted A-to-I editing events

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# RDD identified via RNA-seq





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# Sanger sequencing of gDNA and cDNA & PCR

- $\star$  gDNA sequencing: confirm that it's not a heterozygous SNP.
- \* cDNA sequences: enable detection of edited nucleotides.
- However, cDNA is not sensitive and quantitive enough to detect low-level editing or to provide accurate estimates of editing ratios (?)
- Instead, the traditional clonal sequencing approach is used to analyze the cDNA sequences and PCR sequencing is only used to confirm the gDNA sequences only.
- Four genes were randomly picked where a number of A-to-I editing sites are located within 400 bases.
  - Their cDNA sequences were amplified and cloned into a TOPO vector.
  - 20 clones for each gene were randomly picked and analyzed by Sanger sequencing.



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# Sanger sequencing of gDNA and cDNA & PCR (contd.)



FDR (false-discovery rate):  $4/(93 - 4) \approx 4.5\%$ .



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Identification of RNA editing by RNA-seq Validation of predicted A-to-I editing events

### Sanger sequencing of gDNA and cDNA & PCR (contd.)





Identification of RNA editing by RNA-seq Other results (selected) Characterization of predicted A-to-I editing events

Characterization of predicted A-to-I editing events

• Consider 4,141 A-to-I editing sites with  $\geq$  20% editing level identified from the control siRNA samles.



#### Identification of RNA editing by RNA-seq

Other results (selected)

Characterization of predicted A-to-I editing events

Туре	Total	Coding transcripts				Noncoding	Intergenic
		Coding	Introns	5' UTR	3' UTR		8
A→G	4,141	45	2,015	45	1,293	485	258
		1.1%	48.7%	1.1%	31.2%	11.7%	6.2%
A→C	94	31	9	4	38	5	7
		33.0%	9.6%	4.3%	40.4%	5.3%	7.4%
A→U	48	4	16	0	22	4	2
		8.3%	33.3%	0.0%	45.8%	8.3%	4.2%
C→A	57	6	16	2	24	1	8
		10.5%	28.1%	3.5%	42.1%	1.8%	14.0%
C→G	50	9	11	12	13	2	3
		18.0%	22.0%	24.0%	26.0%	4.0%	6.0%
C→U	173	26	45	5	64	18	15
		15.0%	26.0%	2.9%	37.0%	10.4%	8.7%
G→A	149	18	46	8	46	20	11
		12.1%	30.9%	5.4%	30.9%	13.4%	7.4%
G→C	51	9	14	7	11	8	2
		17.6%	27.5%	13.7%	21.6%	15.7%	3.9%
G→U	73	9	24	3	31	2	4
		12.3%	32.9%	4.1%	42.5%	2.7%	5.5%
T→A	54	6	16	2	23	4	3
		11.1%	29.6%	3.7%	42.6%	7.4%	5.6%
T→C	506	42	239	9	48	123	45
		8.3%	47.2%	1.8%	9.5%	24.3%	8.9%
T→G	109	28	19	10	39	10	3
		25.7%	17.4%	9.2%	35.8%	9.2%	2.8%

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Characterization of predicted A-to-I editing events

### Characterization of predicted A-to-I editing events (contd.)



In Alu elements vs. outside of Alu elements.



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### Characterization of predicted A-to-I editing events (contd.)





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# Motifs near editing sites far away from Alus

Supplemental Table 8. Motif enrichment near predicted A-to-I editing sites in non-Alu regions.

Motif score (ms) cutoff	Number of editing sites in non-Alu regions with motif	Mean of number of motifs in the random sets	P-value
ms > 6.6	51	56.25	0.755
ms > 16.8	21	7.71	2.047x10 <sup>-7</sup>
ms > 21.4	15	5.09	3.082x10 <sup>-6</sup>
ma > 24.4	6	2.71	0.02



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Characterization of predicted A-to-I editing events

### Conservation of neighborhood of predicted A-to-I editing sites





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# A structural motif in ADAR editing





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# Other types of DNA-RNA differences

**Supplemental Table 10.** Co-occurrence of other types of DNA-RNA differences with the predicted A-to-G events in the same gene (1,167 genes with predicted A-to-G events)

	#	# genes also with A-to-G	
Туре	genes	events	P- value
A→C	91	19	1.79x10 <sup>-5</sup>
A→U	45	13	4.69 x10 <sup>-6</sup>
C→A	56	17	1.19 x10 <sup>-7</sup>
C→G	47	13	8.28 x10 <sup>-6</sup>
C→U	155	53	$< 10^{-17}$
G→A	123	39	$1.55 \text{ x} 10^{-15}$
G→C	49	17	1.10 x10 <sup>-8</sup>
G→U	66	20	1.31 x10 <sup>-8</sup>
T→A	50	11	$3.54 \text{ x} 10^{-4}$
T→C	105	20	4.99 x10 <sup>-5</sup>
T→G	258	62	1.11 x10 <sup>-16</sup>



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# Other types of DNA-RNA differences (contd.)

- Regions with unknown sense-antisense transcription may lead to confusion of an actual A-to-G events as T-to-C events, vice versa.
- Indeed, if most T-to-C events were resulted from A-to-I editing on the opposite strand, then they are expected to be as highly enriched in *Alus* as the A-to-G events.
- Yet, 63% of T-to-C events occur in *Alus*, significantly lower than the 88% among A-to-G events ( $p < 1 \times 10^{-10}$ ).



Identification of RNA editing by RNA-seq Discussion

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# Discussion

- It is still possible to have false-positive prediction due to sequencing or mapping errors.
  - Mapping errors arise due to highly homologous regions in mammalian genomes.
- Increased read coverage at putative editing sites enable better accuracy in the esitmation of editing ratios.



Identification of RNA editing by RNA-seq Discussion

# Discussion (contd.)

- The predicted A-to-I editing sites are often associated with lower genomic conservation compared with their flanking regions.
- However, changing the A to I (G) via editing *increases sequence* conservation in primates.
- G-to-A genomic mutationss may be corrected by RNA editing.



# Discussion (contd.)

- Editing levels of the A-to-I editing sites tend to be relatively low (mean, 0.35; median, 0.33).
- Among all 5,965 A-to-G sites in U87MG cells,
  - 31%: editing level  $\leq$  0.2;
  - 5%: editing level  $\geq$  0.8.
  - ▷ Consistent with the continuous probing (COP) hypothesis (Gommans *et al.* 2009).
    - Low-level editing is prevalent due to COP of the transient and dynamic RNA secondary structures by the editing machinary.



Identification of RNA editing by RNA-seq

# Thank you.



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