Computational approaches for detection and quantification of A-to-I RNA-editing



















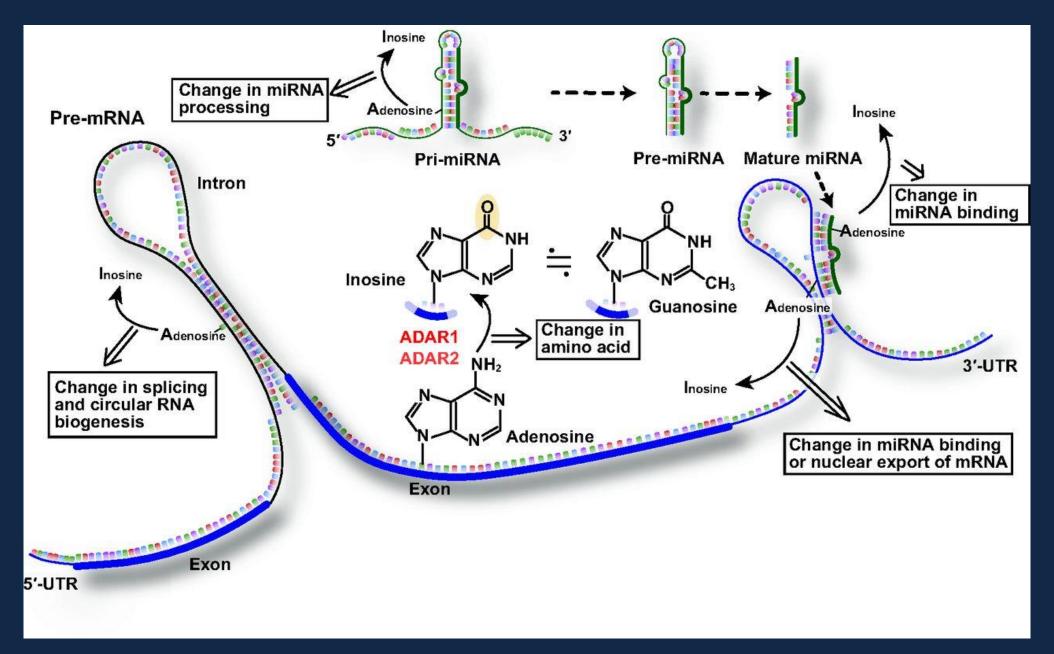
TRAINING COURSE IN Computational Methods for Epitranscriptomics

Bari, 11th-13th September 2024



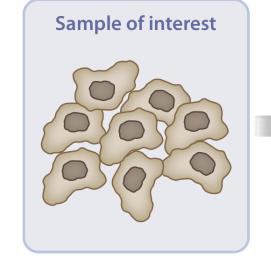
Functional significance of A-to-I RNA editing

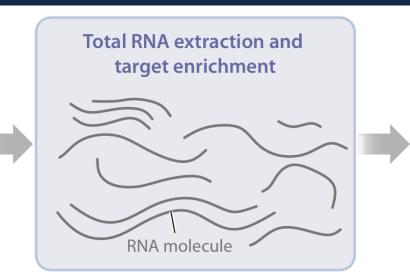


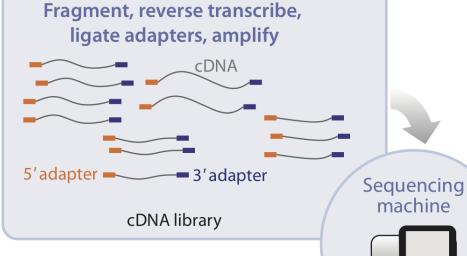


From sample to data analysis



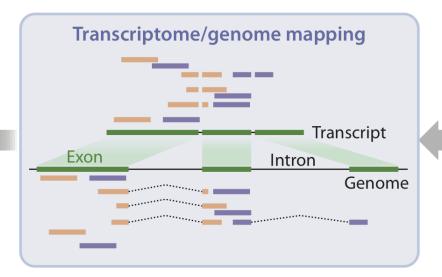


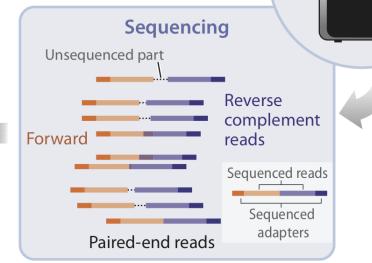




Data analysis

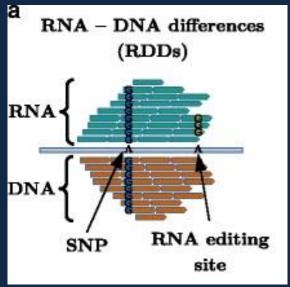
- Differential expression
- Variant calling
- Annotation
- Novel transcript discovery
- RNA editing

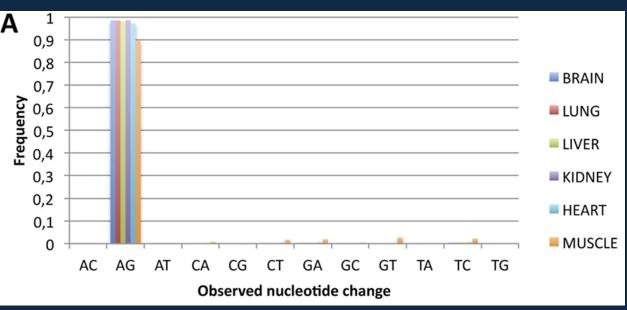


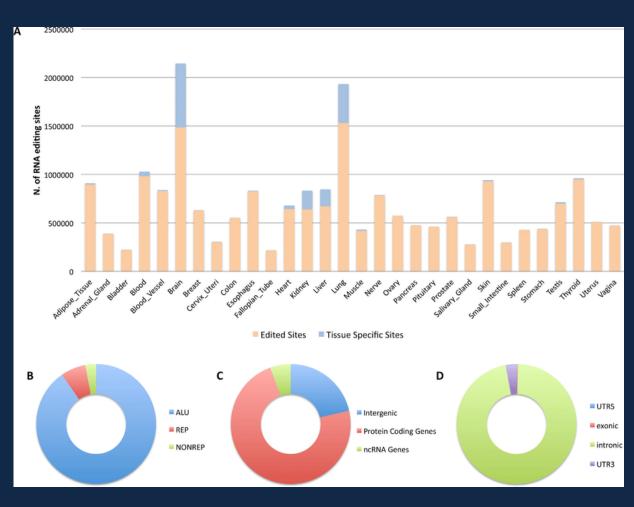


Design of RNA sequencing experiments for identifying RNA editing





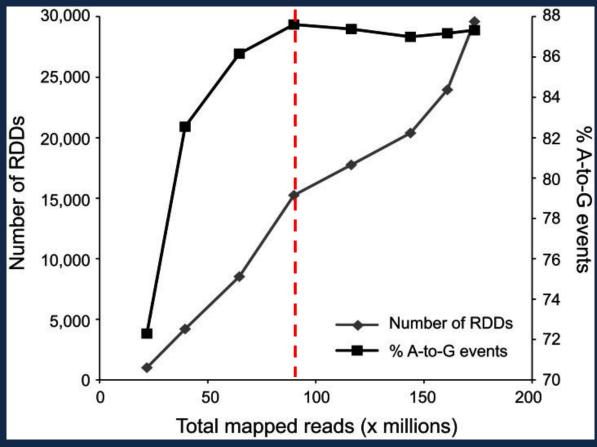




RNA editing distribution along human tissues and a graphical overview of sites stored in REDIportal. A-to-I events collected in REDIportal derive from RNAseq data encompassing 55 human body sites grouped in 30 different tissues.

Design of RNA sequencing experiments for identifying RNA editing





Number of RDDs and % of A-to-G events identified in RNA-Seq closely depend on the sequencing depth.

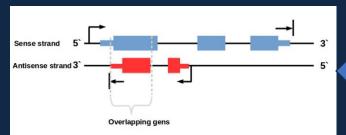


TABLE 1. Recommended variables for consideration in the design of RNA-Seq experiments for identifying RNA-editing events

Variables	Rationale and consideration
Sequencing depth	Number of RDDs and % of A-to-G events increase with sequencing depth; accuracy of estimated editing levels increases with read coverage of putative RDDs.
Biological replicates	Recommended in order to ensure high total coverage of candidate RDD sites after removal of duplicate reads.
Paired or single-end sequencing	Paired-end sequencing and read pairing during data analysis can significantly improve RDD accuracy.
Quality of sequencing library	High fidelity enzymes for RT and PCR should be adopted. Rate of duplicate reads should be evaluated and minimized. Base quality of reads should be inspected and optimized by sequencing chemistry.
Type of sequencing library	Strand-specific libraries are advantageous for pinpointing specific types of RDDs.

Lee JH, Ang JK, Xiao X. Analysis and design of RNA sequencing experiments for identifying RNA editing and other single-nucleotide variants. RNA. 2013 Jun;19(6):725-32.



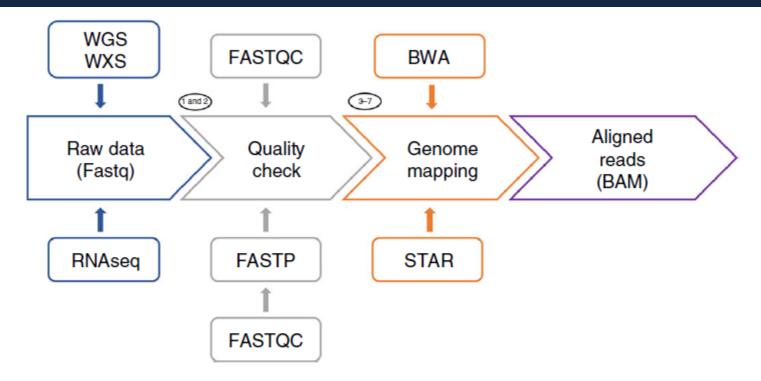
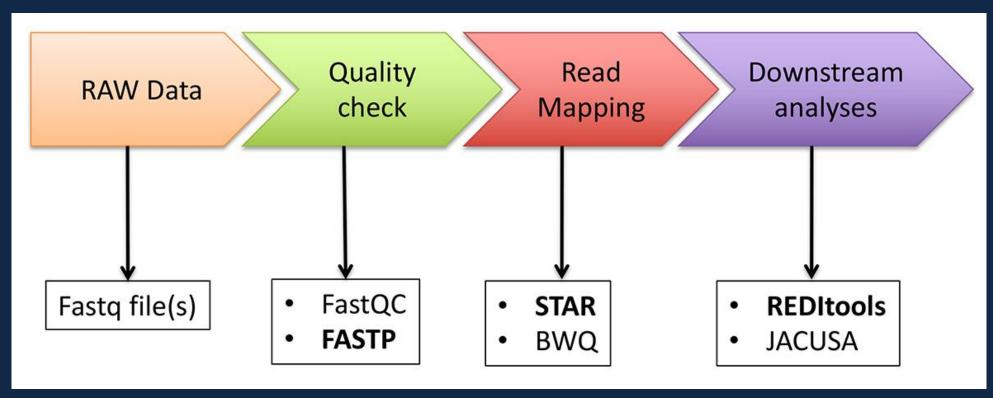


Fig. 1 | Overview of the bioinformatics workflow to preprocess data. Reliable RNA editing calls require good quality WGS and RNAseq reads. Once obtained from public databases, raw reads in fastq format are quality checked using FASTQC and cleaned using FASTP (Steps 1 and 2, Procedures 1 and 2). Then, RNAseq reads are aligned to the reference genome using a splice-aware software like STAR, while WGS reads are mapped using BWA (Steps 3-7, Procedures 1 and 2). Finally, aligned reads are converted into the standard BAM format for the downstream detection of RNA editing.





RNA editing detection and quantification

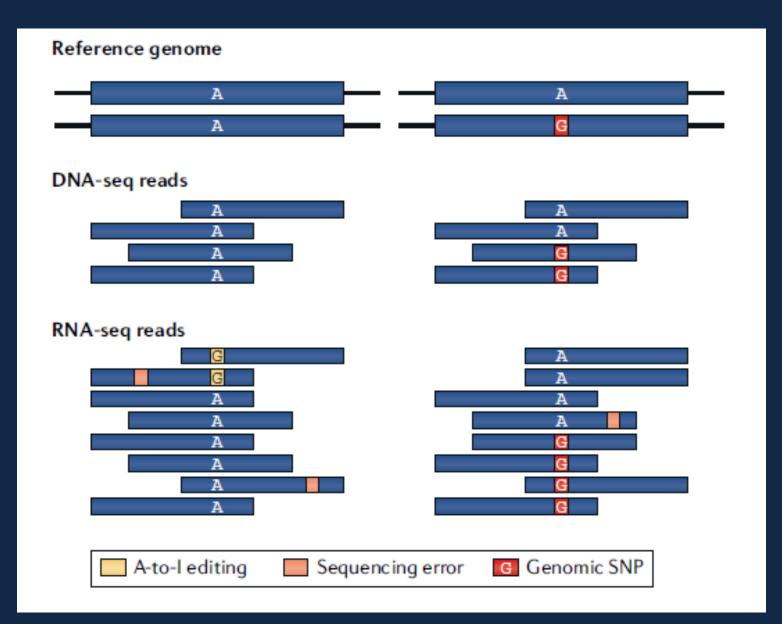
De novo approach

RNA editing candidates can be detected using REDItools. There are two current versions: 1) <u>REDItools 1.3</u> or 2) <u>REDItools 2.0</u>. REDItools2 is a faster re-implementation of REDItools1 for HPC clusters. Its serial version is about ten times faster than REDItools1.

"Known" approach

While the *de novo* approach provides a list of most likely editing candidates, the "known" approach focuses on a limited pool of known events in order to better investigate RNA editing dynamics in different experimental contexts. The "known" approach can be carried out using the REDItools package and a list of events from own data or from public databases such as <u>DARNED</u>, <u>RADAR</u> and <u>REDIportal</u>.





Reverse transcription replaces inosines in mRNA with guanosines in the cDNA. Thus, the hallmark of RNA editing is a consistent A \rightarrow G mismatch between RNA sequencing (RNA-seq) data and the reference genomic sequence to which it is aligned. However, most of these mismatches arise from sequencing errors and genomic polymorphisms, including somatic mutations and incorrect alignment. Matched DNA sequencing (DNA-seq) data may be utilized to distinguish between editing events and genomic polymorphisms. At an editing site, the DNA reads agree with the genome reference (left), while a genomically polymorphic site exhibits mismatches in both DNA-seq and RNA-seq data.



RNA editing quantification by RNA-Seq

r1 r2 r3 r4

r5

r6

r7

r8 r9 r10 r11 r12 r13 r14 r15

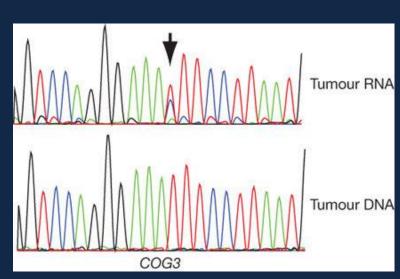
r16

r17

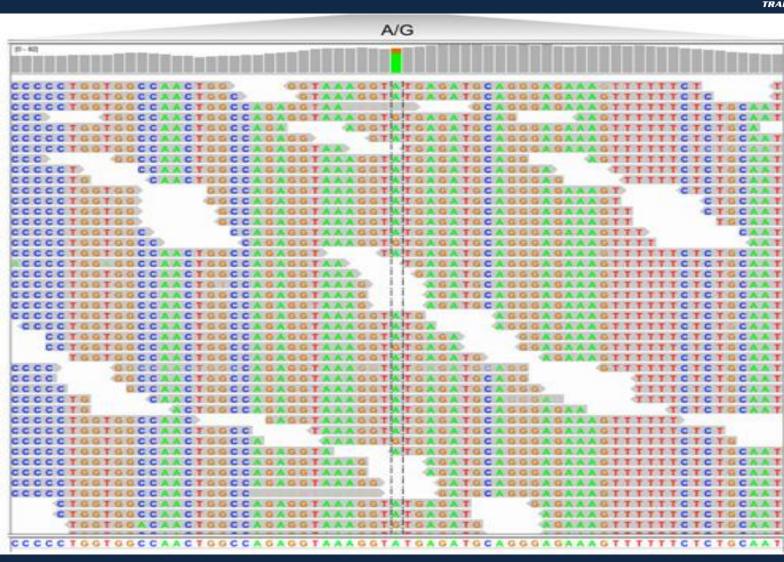
r18

r19 r20 r21

r22 r23 r24



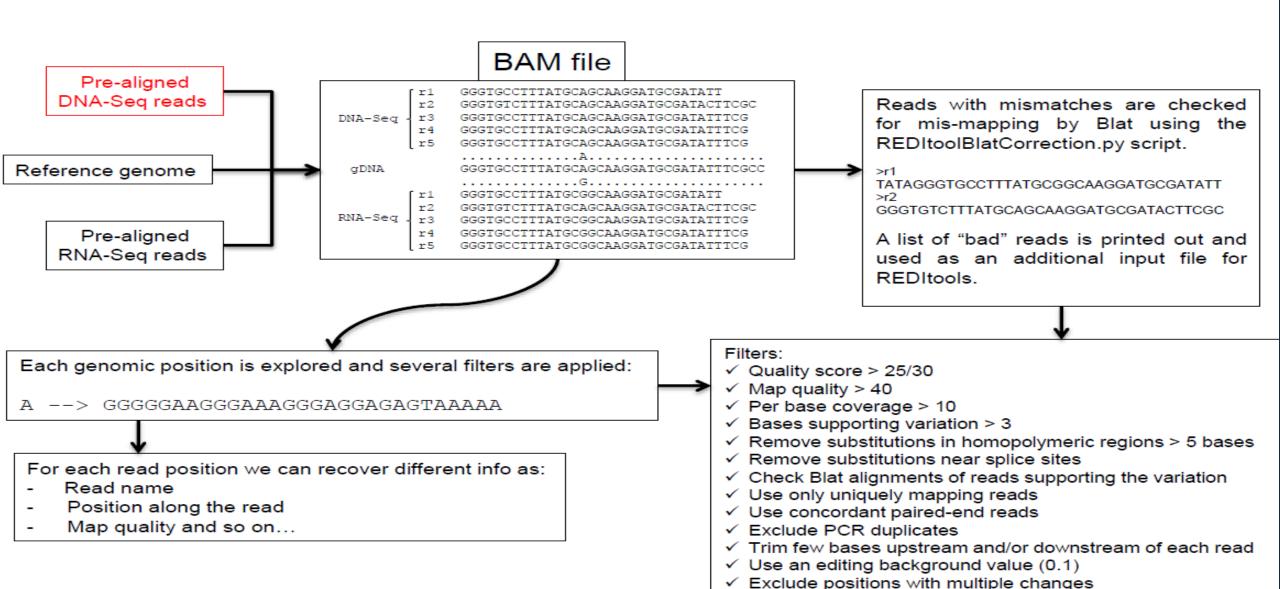
Sanger validation of editing candidates



coverage 32: n.A 27 n.G 5 -> editing frequency 15,6%

RNA editing and NGS

Workflow to call RNA editing by REDItools.



Coding Sequences

ALU Elements



d.

errors

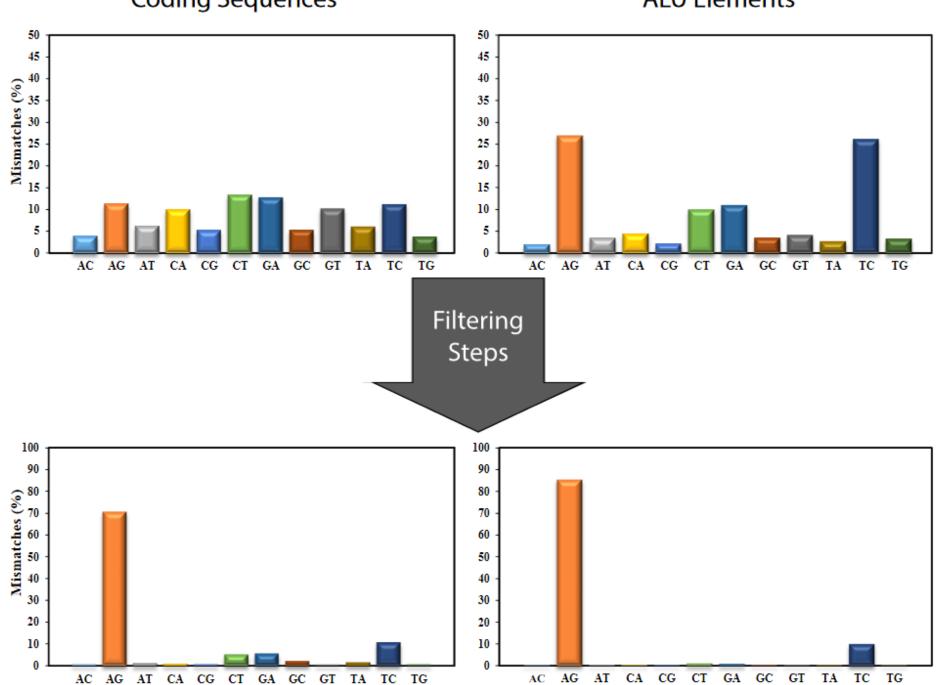
alogous regions

Suggested filteri

Filter

Min. read covera Min. quality sco Min. mapping sc Min. number of Min. editing free Exclude multi-m Exclude duplicat Exclude genomic Trim bases in re Sites near splice Exclude position Clip overlapping Remove sites in Remove sites in Remove sites wi Filter DNA varia

* Filtering ste ** Require ma

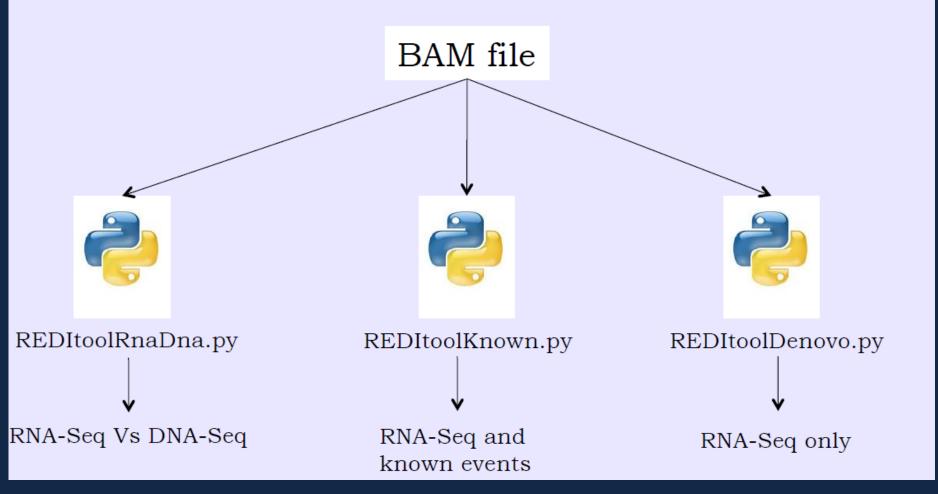


Using REDItools to Detect RNA Editing Events in NGS Datasets



"REDItools" are a suite of python scripts to investigate RNA editing at large-scale employing RNA-Seq as well as DNA-Seq (WGS/WES) massive data.

Starting point is a BAM file of aligned reads onto the reference genome.



REDItoolDnaRna.py is the main script devoted to the identification of RNA editing events taking into account the combined information from RNA-Seq and DNA-Seq data in BAM format. To look at potential RNA editing candidates, RNA-Seq data alone can be used.

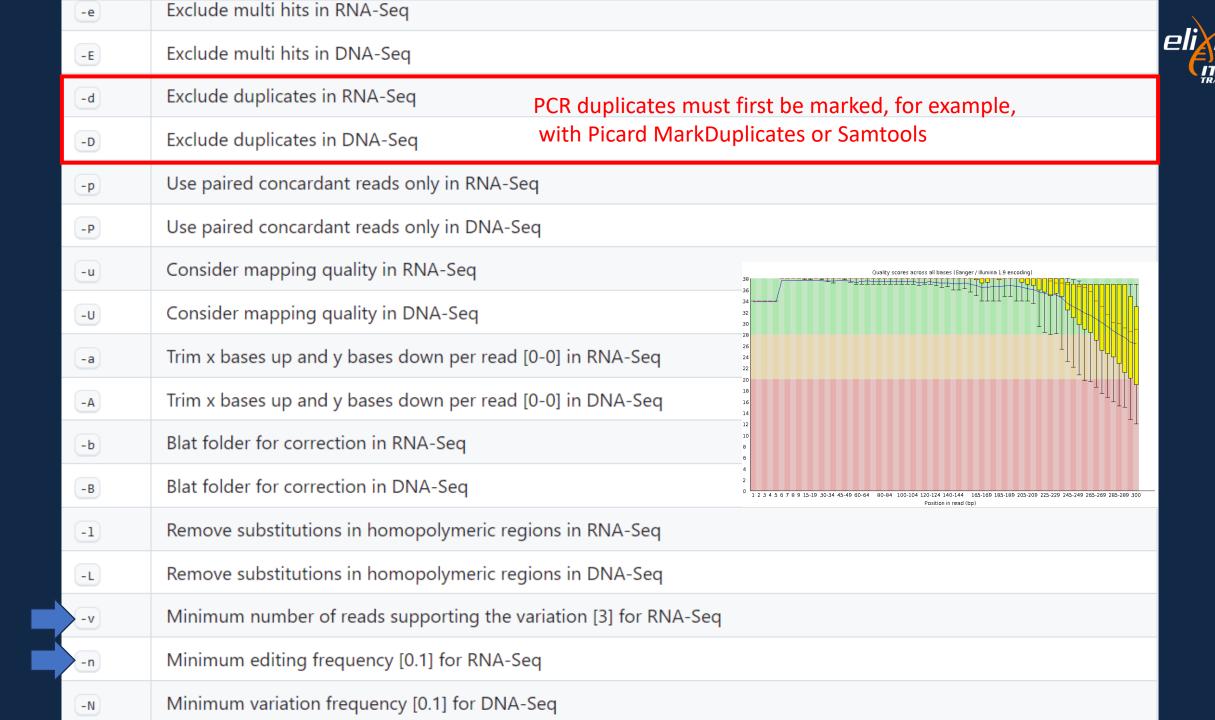


Options:

-i	RNA-Seq BAM file
- j	DNA-Seq BAM files separated by comma or folder containing BAM files. Note that each chromosome/region must be present in a single BAM file only.
-I	Sort input RNA-Seq BAM file
-J	Sort input DNA-Seq BAM file
-f	Reference file in fasta format. Note that chromosome/region names in the reference must match chromosome/region names in BAMs files.
-c	Base interval to explore [100000]. It indicates how many bases have to be loaded during the run.
-k	List of chromosomes to skip separated by comma or file (each line must contain a chromosome/region name).
-t	Number of threads [1]. It indicates how many processes should be launched. Each process will work on an individual chromosome/region.
-0	Output folder [rediFolder_XXXX] in which all results will be stored. XXXX is a random number generated at each run.
_F	Internal folder name [null] is the main folder containing output tables.
-М	Save a list of columns with quality scores. It produces at most two files in the pileup-like format.

-с	Minimum read coverage (dna,rna) [10,10]									
- q	Minimum quality score (dna,rna) [25,25]									
- m	Minimum mapping quality score (dna,rna) [25,25] only usable if you are working with a strand-oriented RNA-Seq									
-0	Minimum homoplymeric length (dna,rna) [5,5] https://rseqc.sourceforge.net/#infer-experiment-py									
-s	Infer strand (for strand oriented reads) [1]. It indicates which read is in line with RNA. Available values are: 1:read1 as RNA,read2 not as RNA; 2:read1 not as RNA,read2 as RNA; 12:read1 as RNA,read2 as RNA; 0:read1 not as RNA,read2 not as RNA.									
-g	Strand inference type 1:maxValue 2:useConfidence [1]; maxValue: the most prominent strand count will be used; useConfidence: strand is assigned if over a prefixed frequency confidence (-x option)									
-x	Strand confidence [0.70]									
-5	Strand correction. Once the strand has been inferred, only bases according to this strand will be selected.									
-G	Infer strand by GFF annotation (must be GFF and sorted, otherwise use -X). Sorting requires grep and sort unix executables.									
-к	GFF File with positions to exclude (must be GFF and sorted, otherwise use -X). Sorting requires grep and sort unix executables.									
-т	Work only on given GFF positions (must be GFF and sorted, otherwise use -X). Sorting requires grep and sort unix executables.									
- X	Sort annotation files. It requires grep and sort unix executables.									

ITALY TRAINING



-z	Exclude positions with multiple changes in RNA-Seq
-Z	Exclude positions with multiple changes in DNA-Seq
-W	Select RNA-Seq positions with defined changes (separated by comma ex: AG,TC) [default all]
-R	Exclude invariant RNA-Seq positions
-V	Exclude sites not supported by DNA-Seq
- W	File containing splice sites annotations (SpliceSite file format see above for details)
-r	Num. of bases near splice sites to explore [4]
gzip	Gzip output files
-h, help	Print the help

Example:

REDItoolDnaRna.py -i rnaseq.bam -j dnaseq.bam -f myreference.fa -o myoutputfolder

Mapping quality value may change according to the aligner:

- For Bowtie use 255
- For Bowtie2 use 40
- For BWA use 30
- For RNA-STAR use 255
- For HiSAT2 use 60
- For Tophat1 use 255
- For Tophat2 use 50
- For GSNAP use 30

Output strand oriented RNA-Seq



	-bash-4.2\$ head -n 20 outTable_378204159																			
Region	Position	Refere	ence	Strand	d Coverage-q30	MeanQ	BaseC	Count[A,C,G	3, T]	AllSubs	s Frequency	_y	gCove?	erage-q30	gMeanQ	gBaseCc	ount[A,C,G,	1, T] C	gAllSubs	gFreq
uency																				
chr13	19659651	A	1	10	70.30 [7, 0,	3, 0]	AG		13		[13, 0,	0, 0]		0.00						
chr12	9079672 A	0	38	65.24	[27, 0, 11, 0]	AG	0.29	30	64.03	[30, 0,	, 0, 0]		0.00							
chr12	26978013	A	0	11	50.36 [8, 0,		AG	0.27	10	62.80	[10, 0,	0, 0]		0.00						
chr12	26978020	A	0	11	51.64 [8, 0,	3, 0]	AG	0.27	11	64.73	[11, 0,	0, 0]		0.00						
chr12	26978033	A	0	11	50.45 [4, 0,		AG	0.64	13	62.92	[13, 0,	0, 0]		0.00						
chr12	26978058	A	0	11	56.09 [4, 0,	, 7, 0]	AG	0.64	20	66.90	[20, 0,	0, 0]		0.00						
chr12	53178370	A	0	23	67.57 [20, 0,	0, 3, 0]	AG	0.13	29	59.69	[29, 0,	0, 0]		0.00						
chr12	53178406	A	0	13			AG	0.31	27	59.07	[27, 0,	0, 0]		0.00						
chr12	53178424	A	0	12	55.50 [8, 0,	, 4, 0]	AG	0.33	16	60.56	[16, 0,	0, 0]		0.00						
chr12	57543105	A	0	22	60.59 [17, 0,	0, 5, 0]	AG	0.23	10	34.00	[10, 0,	0, 0]		0.00						
chr12	57543106	A	0	22				0.14	10	34.50	[10, 0,	0, 0]		0.00						
chr12	68766820	A	1	11			AG	0.36	62	60.71	[62, 0,	0, 0]		0.00						
chr12	110644850	A	1	13			AG	0.31	11	48.36	[11, 0,	0, 0]		0.00						
chr11	117836389	A	2	10	59.20 [7, 0,	, 3, 0]	AG	0.30	28	54.96	[28, 0,	0, 0]		0.00						
chr10	4997838 A	0	10	62.90	[6, 0, 4, 0]	AG	0.40	24	57.96	[24, 0	, 0, 0]		0.00							
chr10	73248973	A	0	12	66.83 [8, 0,	4, 0]	AG	0.33	12	57.33	[12, 0,	0, 0]		0.00						
chr10	77637540	A	0	13			AG	0.23	20	52.20	[20, 0,	0, 0]		0.00						
chr17	3619478 A	0	12	74.00	[9, 0, 3, 0]	AG	0.25	23	56.43		, 0, 0]		0.00							
chr17	18271818	A	0	10	66.60 [7, 0,		AG	0.30	43					0.00						

Output unstranded RNA-Seq

strand (+ or -). You can also indicate strand by 0 (strand -), 1 (strand +) or 2 (+ and - or unknown)

chr1	14542	Α	2	14	37.64	[4, 0, 10, 0]	AG	0.71	29	29.86	[28, 0, 1, 0]	AG	0.03
chr1	14574	Α	2	11	38.09	[7, 0, 4, 0]	AG	0.36	37	30.11	[37, 0, 0, 0]	-	0.00
chr1	14907	Α	2	22	37.32	[11, 0, 11, 0]	AG	0.50	115	30.13	[61, 0, 54, 0]	AG	0.47
chr1	14925	Α	2	24	37 . 75	[24, 0, 0, 0]	-	0.00	122	30.20	[121, 0, 1, 0]	AG	0.01
chr1	14930	Α	2	24	38.29	[12, 0, 12, 0]	AG	0.50	97	29.93	[67, 0, 30, 0]	AG	0.31
chr1	15180	Α	2	14	38.21	[14, 0, 0, 0]	-	0.00	79	29.78	[78, 0, 1, 0]	AG	0.01
chr1	15274	Α	2	6	40.50	[0, 0, 0, 6]	ΑT	1.00	48	30.38	[0, 0, 10, 38]	AT AG	1.00
chr1	15717	Α	2	2	35.50	[2, 0, 0, 0]	-	0.00	28	28.71	[27, 0, 1, 0]	AG	0.04
chr1	16186	Α	2	27	37.07	[25, 0, 2, 0]	AG	0.07	73	30.42	[73, 0, 0, 0]	_	0.00
chr1	16497	Α	2	24	39.83	[20, 0, 4, 0]	AG	0.17	127	30.36	[108, 0, 19, 0]	AG	0.15
chr1	136573	Τ	2	5	37.00	[0, 3, 0, 2]	TC	0.60	44	29.43	[0, 0, 0, 44]	_	0.00
chr1	136586	Τ	2	3	38.33	[0, 2, 0, 1]	TC	0.67	32	29.78	[0, 0, 0, 32]	_	0.00
chr1	136671	Τ	2	3	34.33	[0, 2, 0, 1]	TC	0.67	31	29.94	[0, 0, 0, 31]	_	0.00
chr1	136817	T	2	3	39.67	[0, 0, 0, 3]	_	0.00	51	28.88	[0, 3, 0, 48]	TC	0.06

Accurate identification of human *Alu* and non-*Alu* RNA editing sites



Even though the editing level of *Alu* sequences is typically low (0.6% on average) and varies considerably, almost any one of the adenosines within these sequences will potentially be targeted by ADARs.

<u>STRATEGY</u>: apply different filtering criteria according to the type of region (repeated or not) in which the putative edited site falls

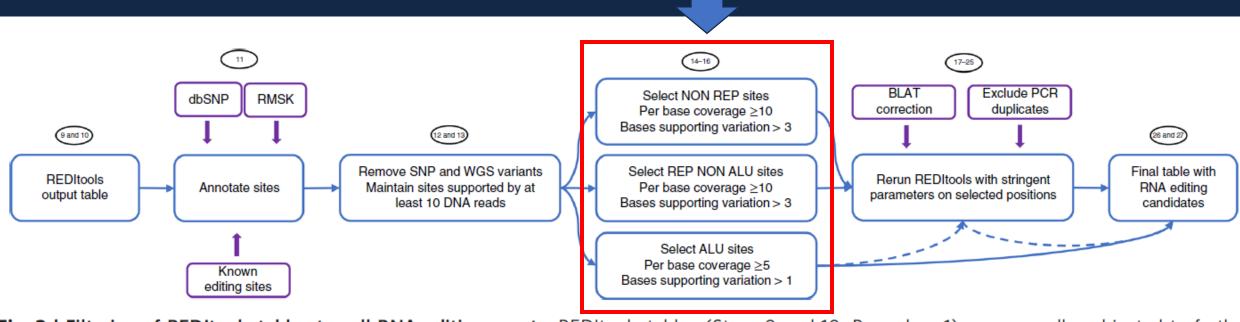
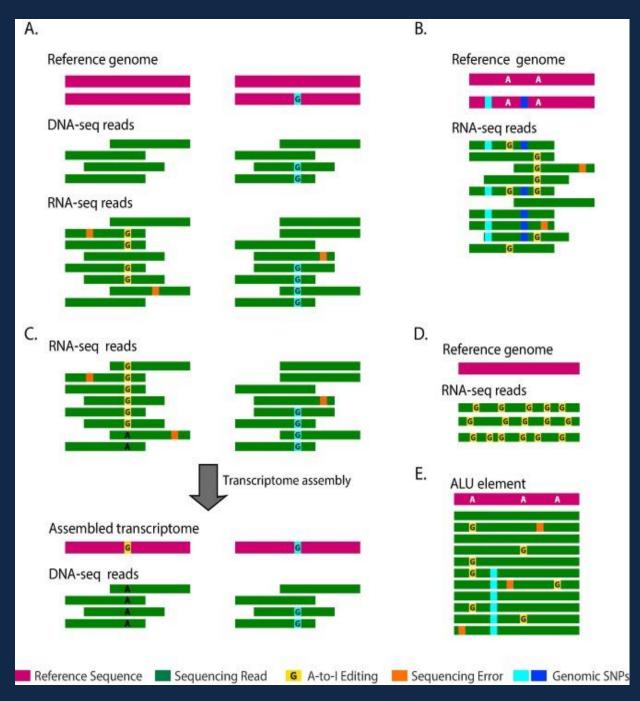


Fig. 3 | Filtering of REDItools tables to call RNA editing events. REDItools tables (Steps 9 and 10, Procedure 1) are generally subjected to further filters to remove potential artifacts and errors. The procedure begins with the annotation of all individual positions using known SNP sites, repeated elements in RepeatMasker and known editing events stored in the REDIportal database (Step 11, Procedure 1). Then, SNPs and sites not supported by ≥10 WGS reads are removed (Steps 12 and 13, Procedure 1) and divided into three groups: ALU, REP NON ALU and NON REP (Steps 14-16, Procedure 1). NON REP and REP NON ALU sites undergo more stringent call criteria than ALU sites that take into account mis-mapping reads and PCR duplicates (Steps 17-25, Procedure 1). Optionally, stringent filters can also be applied to ALU sites (Steps 20 and 21, Procedure 1). Finally, filtered positions are collected in the final list of RNA editing candidates (Steps 26 and 27, Procedure 1).





Quantifying RNA editing in deep transcriptome datasets

The quantification of RNA editing is important to compare values across samples and study its potential role in different experimental conditions or in human disorders.

Determine the fraction of edited transcripts of a site (editing levels) by dividing the number of the 'G'-containing transcripts that map to the site, by the total number of transcripts mapped to the position.

For example, the editing levels of the leftmost editing site in Fig. E is 4/11 as we found evidence for editing in 4 reads out of 11 reads.

The accuracy of measuring the editing levels of a site depends on the site coverage in the RNA-seq dataset, which in turn is determined mainly by the sequencing depth and the expression levels of the transcript of interest. Unfortunately, sufficient coverage for each editing site is often not available in a typical RNA-seq. In order to overcome this limitation have been developed metrics for unbiased RNA editing quantification in a sample:

Overall editing level

The overall editing is defined as the total number of reads with G at all known editing positions over the number of all reads covering the positions without imposing specific sequencing coverage criteria. It can be calculated using REDItools tables obtained imposing loosing parameters.

ALU editing index

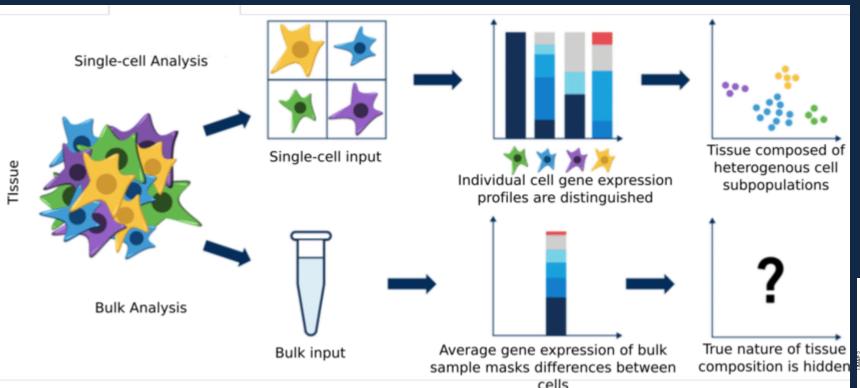
The Alu editing index (AEI) is a metric to quantify the global RNA editing activity of sample and is defined as the weighted average of editing events occurring in all Alu elements. The pipeline to calculate AEI is described in Roth et al. (2019) and available here.

Recoding index

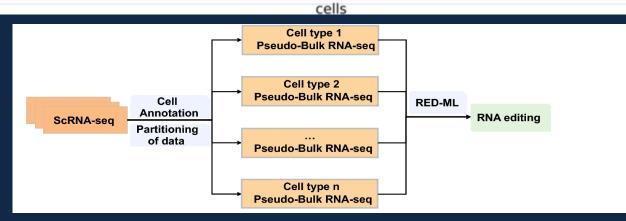
The overall editing calculated at recoding positions residing in coding protein genes is named recoding index (REI). It has been initially described in <u>Silvestris et al. (2019)</u>. This metric, used to investigate the activity of ADAR2, can be calculated using REDItools tables obtained imposing loosing parameters and a list of recoding sites from <u>REDIportal</u>.

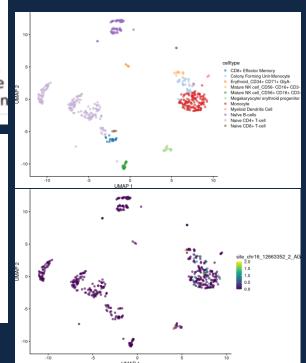
RNA editing at single-cell resolution (with 10x data)





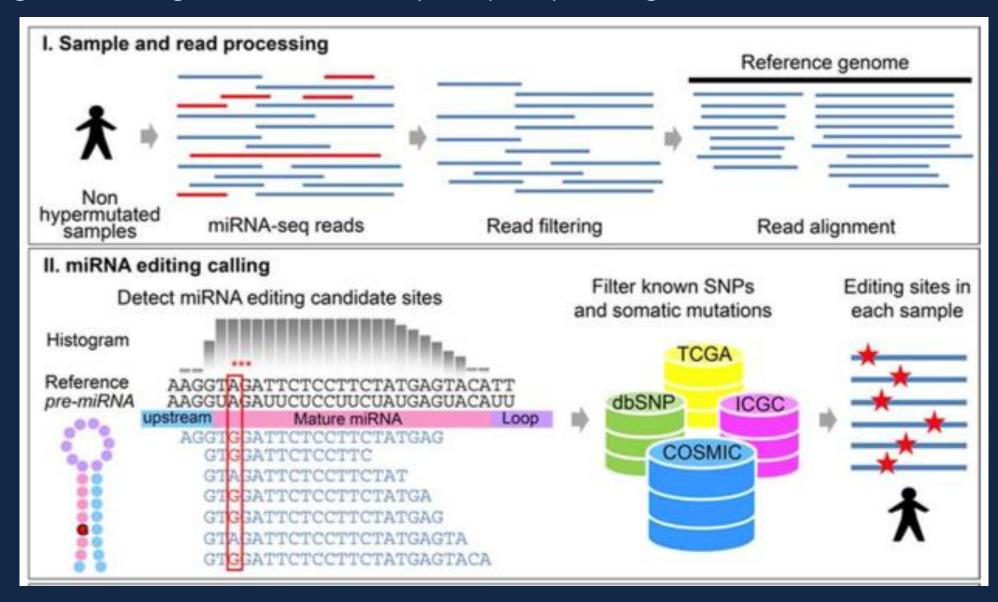
The cell type annotation information was used to combine the mapped reads of the same cell type in scRNA-seq to obtain pseudo-Bulk RNA-seq for each cell type.







Identifying RNA Editing Sites in miRNAs by Deep Sequencing



https://www.tau.ac.il/~elieis/miR_editing/



<u>Pipeline</u>

step1. Filtering Low- Quality Reads and Trimming Sequence Adapters

perl Process_reads.pl Input_fastq_file The_filtered_fastq_file

step2. Aligning the reads against the genome

bowtie -n 1 -e 50 -a -m 1 --best --strata --trim3 2 The_bowtie_folder/The_genome_indexes The_filtered_fastq_file > The_output_file

step3. Mapping the mismatches against the pre-miRNA sequences

perl Analyze_mutation.pl The_output_file main_output.txt

step4. Using binomial statistics to remove sequencing errors

perl Binomial_analysis.pl main_output.txt >binomial_output.txt

