SeqMap: mapping massive amount of oligonucleotides to the genome

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- high-throughput sequencing
- oligo mapping tools
- mismatches and ins/del

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SeqMap: introduction

- <u>http://biogibbs.stanford.edu/~jiangh/SeqMap/</u>
- Command-line tool for mapping large amount of short oligonucleotides to the reference genome
- Data generation:

 ✓ Illumina-Solexa and ABI-SOLiD = 50-100M reads (30-50 nt)
 ✓ Roche 454 = 400,000 reads (200-300 nt)
- Alternatives:
 - ✓ ELAND only for short reads, max 2 substitutions
 - ✓ SOAP slow with larger genomes, max 2 substitutions or one gap (1-3 nt.)
 - ✓ RMAP ungapped mapping, accounts read quality

SeqMap: method

- Flexibility in the mapping:
 - ✓ Allows 5 mixed substitutions and ins/dels
 - ✓ Various command-line options and output formats
 - ✓ Sequences can contain 'N'
 - ✓ Sequences can have different lengths
 - ✓ Operates with FASTA files
- Alternatives:
 - ✓ ELAND only for short reads, max 2 substitutions
 - ✓SOAP slow with larger genomes, max 2 substitutions or one gap (1-3 nt.)
 - ✓ RMAP ungapped mapping, accounts read quality

SOAP: algorithm

- Store the genomic sequence in RAM. Two bits for each base, so one byte can store 4 bps
- Split reads into 4 parts a,b,c,d, two mismatches will be distributed on at most two of the 4 parts at the same time
- Use look up table to judge how many mismatches between reference and read. To have best efficiency, the table used 3 bytes to check a fragment of 12-bp on a time. The table occupied 2^24=16Mb RAM
- Search for identical hits first, if no hits, then 1-mismatch hits will be picked up, then 2-mismatch hits, then gapped hits.

SeqMap: algorithm

- Split query sequence into several fragments
 - ✓ Using 2 mismatches and splitting into 4 fragments, two of them have always perfect matches



- Size of the fragment is dynamic
- Creating hash table from the query not the genome

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SeqMap: speed

Software	Running time	Memory used	Mapped reads	
SeqMap	2213 s	3.0 GB	455 384	
ELAND	345 s	721 MB	455 384	
SOAP	5464 s	979 MB	452 005	
RMAP	14 h	3.1 GB	321 651	

11 530 816 Solexa reads (25 nt) are mapped to mouse chrX (166 650 296 bp) using SeqMap, ELAND, SOAP and RMAP, respectively. The running time, memory usage, and number of mapped reads for each program are reported. For each program, up to two substitutions are allowed and no gap is allowed. The experiments are done on a machine with 3 GHz Intel Xeon CPU and 32 GB memory, running 64-bit Linux.

SeqMap: sensitivity

Table 2. Mapping 100000 randomly perturbed reads with SeqMap, ELAND, SOAP and RMAP

Software	Running time (s)	Memory used (MB)	Mapped reads	
SeqMap	82	923	78 211	
ELAND	3	261	27 561	
SOAP	2	142	38 256	
RMAP	4	232	31 891	

 1Mbp random sequence, 100Kbp random substitutions, 'N's and ins/dels against 100K random 25-mers

SeqMap: PCR primer mapping

- 1000 primer pairs against human genome (2MM) 3896.77s (> 1 h)
- RAM = 6MB (query hash tables)

trans_id	trans_	coord	target_seq	probe_id	probe_se	eq num	_mismatch	strand
chr1	1166	CAAG	AGGGCCCTGCAGTGCC	90560_L	CAAGAGGGCCC	FGCACTTCC	2	-
chr1	16663	ATTA	CAGGCGTGAGCCGCTG	90224_L	ATTACAGGCGT	GAGCCACCG	2	-
chrl	16670	TGCT	GGGATTACAGGCGTGA	90588_L	TGCTGGGATTA	CAGGCGTGA	0	-
chr1	16672	AAAG	TGCTGGGATTACAGGCG	r 90514_L	AAAGTGCTGGGA	ATTACAGGCG	G 1	-
chrl	16681	ATCC	CAGCACTTTGGGAGGC	90377_L	ATCCCAGCAAT	FTGGGACGC	2	+

Substantial biases in ultra-short read data sets from high-throughput DNA sequencing

Juliane C. Dohm, Claudio Lottaz, Tatiana Borodina and Heinz Himmelbauer

Nucl. Acids Res. 2008 36:e105

- Illumina-Solexa technology
- short read error rates
- substitution types
- GC content correlation with the read coverage

Solexa sequencing biases: overview

- Solexa read lengths up to 36 bases
- >40 M reads with 3 days
- Main drawbacks of the high-throughput DNA sequencing using short reads:
 - ✓ Wrong base calls
 - ✓ Coverage of low-complexity regions

Solexa sequencing biases: research topics

- Detect biases of error positions, rates and erroneous base calls (neighboring bases and ins/dels)
- Determine the compensation of erroneous base calls by correct ones with higher coverage
- Analyze read start positions, coverage along target sequence and coverage dependencies of the local sequence characteristics
- Assess the reliability of quality values for wrong and correct base calls

Solexa sequencing biases: data

- 27mer reads from Beta vulgaris clone ZR-47B15 (2 788 286 in total)
- 32mer reads from Helicobacter acinonychis (12 288 791 in total)
- ELAND software for mapping reads
- Perl scripts for ins/del detection

Solexa sequencing biases: ELAND results



ELAND categories are:

QC - no matching done because of low quality of the read (more than two positions with quality score=5),

NM - no match found;

- U0 unique exact match found;
- **U1** unique match with one error;
- **U2** unique match with two errors;

 \square NM

UU 🛛

🛛 U1

U2

B012

- R0 multiple exact matches found;
- **R1** multiple matches with one error;
- R2 multiple matches with two errors.

Solexa sequencing biases: GC content



Each data point corresponds to the number of reads recorded for a 1-kbp window (shift of 100 bp in Beta and 1 kbp in Helicobacter)

Solexa sequencing biases: overall coverage

Table 1. Proportion of reference sequence and coverage ranges (based on ELAND U0, U1, U2, R0 matched reads and reads with single indels)

Beta		Helicobacter		
Coverage	BAC (%)	Coverage	Genome (%)	
200-300	4.27	<100	3.53	
300-400	23.93	100-150	26.06	
400-500	25.64	150-200	42.28	
500-600	23.93	200-250	21.49	
600-700	12.82	250-300	4.44	
700-800	4.27	300-350	1.29	
800–900	5.13	>350	0.90	

- Read distribution along the Beta vulgaris BAC sequence (with cloning vector pBeloBACII). 2 166 892 27mer reads were matched against the finished sequence (enclosed by the cloning vector,117 kbp in total). The read coverage was calculated in 200 consecutive 0.58 kbp windows.
- Read distribution along the 1.55Mbp Helicobacter genome, based on 8 700 113 32mer reads. The local coverage is shown in 200 consecutive windows of 7.77 kbp.

Solexa sequencing biases: wrong base calls (1)



(a) Error frequency per position calculated from considering wrong base calls only. The highest error frequency is observed at the read 30 end. (b) Per-base error rates (overall error frequency per position considering all base calls).

Solexa sequencing biases: wrong base calls (2)



Distance between two errors on a read in the Helicobacter and Beta vulgaris data sets. 'O' indicates that the erroneous base calls are next to each other.

Solexa sequencing biases: error context



Dohm JC, Lottaz C, Borodina T and Himmelbauer H. "Substantial biases in ultra-short read data sets from high-throughput DNA sequencing." Nucleic Acids Research 2008 36:e105.

Solexa sequencing biases: score quality (1)

Table 3. Observed and expected error rates for base calls of different quality values in the *Beta* and *Helicobacter* data sets

Score	Beta (%)	Helicobacter (%)	Expected (%)	
Q = 40	1.39	0.43	0.01	
$\tilde{Q} = 30$	3.55	1.06	0.10	
$\tilde{Q} = 20$	5.21	1.70	0.99	
$\tilde{Q} = 10$	9.68	4.40	9.09	
$\tilde{Q} = 0$	39.65	28.68	50.00	

The Solexa base caller Bustard reports the quality of each base call by estimating a quality score similar to the phred score based on the image output without considering the reference sequence.

Q=40 -> expected error probability of P=0.01% Q=0 -> expected error probability of P=50%

Solexa sequencing biases: score quality (2)



Histograms of base quality values for all correct base calls (a) and all wrong base calls (b) in the Beta and Helicobacter data sets.

Six percent of all wrong base calls in Helicobacter and 19% of all wrong base calls in Beta have Solexa quality scores Q=40.

Solexa sequencing biases: conclusions

- General sequence bias around read starting positions were not detected
- Strong correlation between GC richness and read coverage
- Base call errors occur preferentially at the 3' end of the reads
- Sequence tuples before error position are mainly G-rich
- Solexa software underestimates true-error rate up to 100fold for high quality values and overestimates for low quality values

Eddy J, Maizels N. "Conserved elements with potential to form polymorphic G-quadruplex structures in the first intron of human genes." *Nucl. Acids Res.* 2008 36: 1321-1333.