

MAQ: Mapping and Assembly with Qualities

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Abstract

The vast numbers of very short reads produced by new sequencing technologies such as Illumina GA and AB SOLiD pose a great challenge to data analysis. MAQ, which stands for Mapping and Assembly with Qualities, is a software for mapping short reads to diploid mammalian-sized genomes and calling variants.

MAQ differs from most existing alignment software in several aspects. First, it calculates a mapping quality for each alignment, measuring the probability of the alignment being wrong. This greatly helps accurate variant calling. Second, it maps every read that has a match, placing repetitive reads randomly amongst equally good alternatives, but with a low mapping score, instead of discarding them. This avoids any ambiguity in defining "unique", and provides more data for the subsequent analysis. Third, MAQ aligns mate-pair reads in a sliding window, effectively examining proper paired positions with little computational effort, and allowing reads in repeats to be aligned with high confidence if their mates are in unique regions. Furthermore, MAQ calls the diploid genotype at each base position in the reference with a Phred-like quality, allowing the user to control the sensitivity/specificity trade-off, and facilitating the use of MAQ calls in downstream analysis.

MAQ has been used at multiple sites for human chromosome resequencing, human structural variation analysis, and multiple smaller genome variation studies. In addition to its key functionalities, MAQ can also call short indels, work with SOLiD data, simulate reads and comes with a fast graphic visualizer for the read alignments. Most of its functions come with user friendly interfaces and are well documented. It is available from <http://maq.sourceforge.net>.

Methods

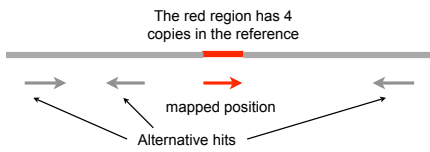
Indexing reads and scanning the reference



Scoring hits and random mapping

A hit is scored as 2^{24-q+h} , where q is sum of qualities of mismatched bases and h is a 24-bit integer from hashing the coordinate of the hit and read identifier. As a result, MAQ *randomly* maps a read if there are multiple equally best hits due to genomic repeats.

The number of hits a read has implies the reference copy number of the region where the read is mapped:



Mapping quality

Mapping quality is the phred-scaled probability (Ewing and Green, 1998) that a read alignment may be wrong. Given L -long reference x and l -long read z , the probability that z is mapped at u is:

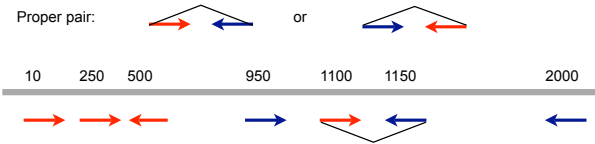
$$p_s(u|x, z) = \frac{p(z|x, u)}{\sum_{v=1}^{L-l+1} p(z|x, v)}$$

where $p(z|x, u)$ equals the product of the error probabilities of mismatched bases. The mapping quality is:

$$Q_s(u|x, z) = -10 \log_{10} [1 - p_s(u|x, z)]$$

Paired-end (PE) alignment

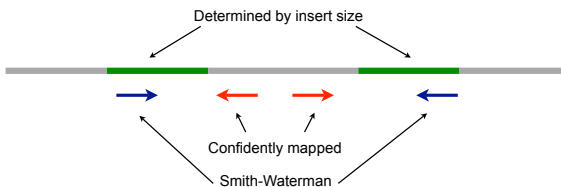
Hit to a read found on the *forward* strand: keep the position in a 2-element queue
 Hit to a read found on the *reverse* strand: check the positions in the queue of its mate



Scan at 1150bp, *read1 queue*: (250,1100); *read2 queue*: (950). Find *read2 hit* and check distance: (250,1150) and (1100,1150).

Short indels

For read pairs where only one end maps, use gapped Smith-Waterman alignment for the other end in a restricted region.



SOLiD alignment

Map in color space:

- The complement of a colour is itself, and therefore a colour read only needs to be reversed.
- The correct orientation for a read pair is:



Decoding colour sequence

Given the reference sequence $b_1 \dots b_{i+1}$ and the colour read sequence $\hat{c}_1 \dots \hat{c}_i$, let $f_i(\hat{b}_i)$ be the best decoding up to position i . Then:

$$f_1(\hat{b}_1) = q_0 \cdot (1 - \delta_{\hat{b}_1, b_1})$$

$$f_{i+1}(\hat{b}_{i+1}) = \min_{b_i} \left\{ f_i(\hat{b}_i) + q_0 \cdot (1 - \delta_{b_{i+1}, \hat{b}_{i+1}}) + q_i \cdot [1 - \delta_{c_i, g(b_i, \hat{b}_{i+1})}] \right\}$$

where $g(\cdot, \cdot)$ translates adjacent nucleotides to the corresponding colour.

Consensus calling

Given data D at an aligned position, calculate $P(D|(b_1, b_2))$, $P(D|(b_2, b_2))$ and $P(D|(b_1, b_2))$, assuming the sample is diploid. Knowing the prior of a heterozygote (b_1, b_2) , we can calculate the posterior probability of each genotype. The consensus \hat{g} is the genotype that maximizes the posterior probability and its quality is $-10 \log_{10}[1 - P(\hat{g}|D)]$.

Results

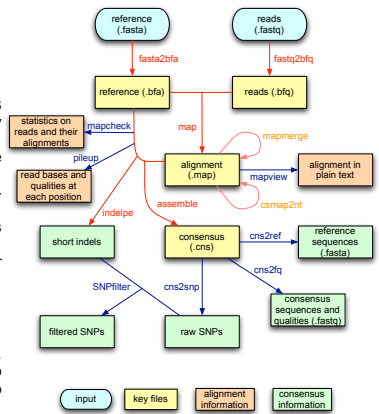
Designed for human genome resequencing

- In alignment, 6 CPU hours and 800MB memory per 1 million read pairs. Easy parallelization on clusters.
- Compressed binary alignment file: 1 byte per nucleotide on reads.
- Compressed binary consensus file: 4 bytes per nucleotide on the reference.
- Quickly retrieval of reads in any regions (via *maqview*).
- Quickly and compact alignment viewer (*maqview*)

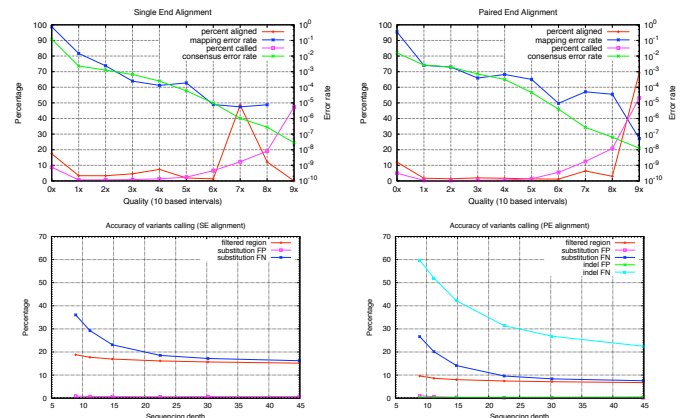
Simulation

100 million 35bp read pairs on human chrX. Added 0.1% substitutions and 0.01% 1bp indels, assuming diploid sample. Mapped to the human reference genome.

MAQ Work Flow



Accuracy of alignments, consensus and variants calls



Maq has been used for several projects including 1000genome and Cancer Genome Project at this meeting.

Acknowledgements

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References

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