

# Genome-scale technologies 2 / Algorithmic and statistical aspects of DNA sequencing

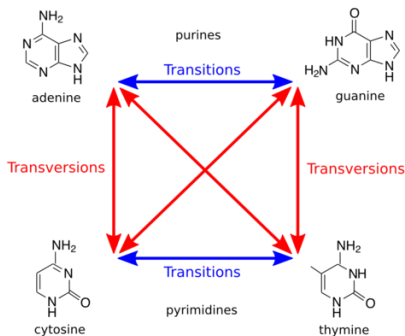
Single nucleotide polymorphism discovery

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# Short biological prerequisites

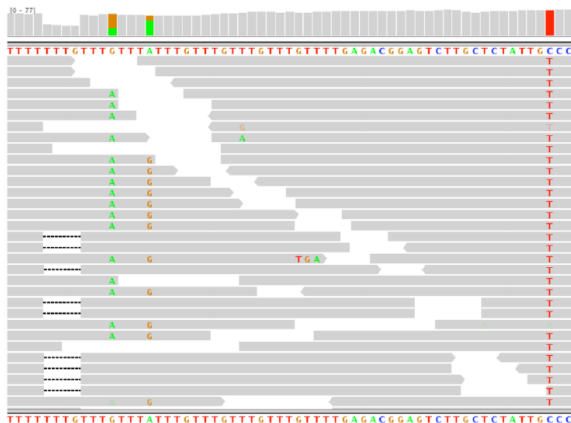
- ▶ Haploid genome: each chromosome has a single copy in the nucleus
- ▶ Diploid genome: each chromosome has two copies in the nucleus (one from father, one from mother)
- ▶ Mutation types: transitions and transversions



# Short biological prerequisites

- ▶ Allele: one of a number of alternative forms of the same genetic locus.
- ▶ Haplotype:
  1. a specific group of genes that a progeny inherits from one parent
  2. a collection of specific alleles (that is, specific DNA sequences) in a cluster of tightly-linked genes on a chromosome that are likely to be inherited together
  3. a set of (several) single-nucleotide polymorphisms (SNPs, "snips")—also known as DNA sequence variations at specific nucleotide sites, or as polymorphic sites—on a single chromosome that are associated statistically.

## Single nucleotide polymorphism calling



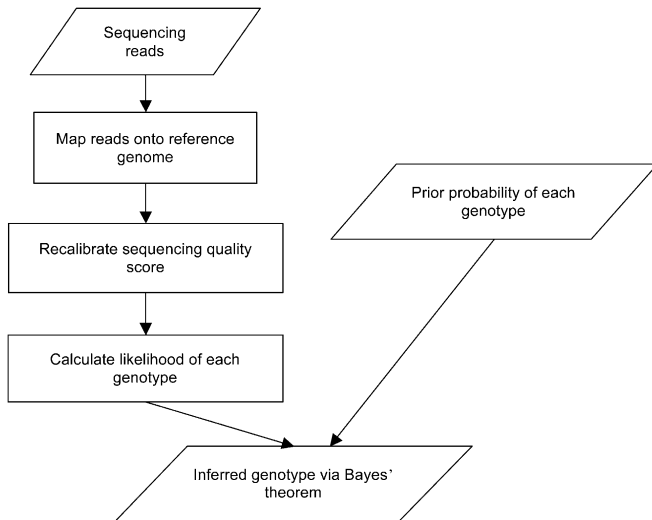
## IGV visualization

- ▶ Reads: arrows oriented by increasing machine cycle;
- ▶ Highlighted bases: mismatches to the reference
- ▶ coverage histogram per base above the reads.

# Single nucleotide polymorphism calling: challenges

- ▶ Reads need to be accurately mapped to the reference
- ▶ Each read aligned independently  $\Rightarrow$  many reads spanning indels will be misaligned
- ▶ Per base Phred scores are inaccurate, co-vary with machine cycle, or sequence context.
- ▶ Separating true variation from machine artifacts due to the high rate and context-specific nature of sequencing errors requires sensitive and specific statistical models.

# SNP-calling probabilistic model in SOAPsnp



# SNP-calling probabilistic model in SOAPsnp

- ▶ The probability of genotype  $T_i$  given observed data  $D$  (reads)s at a locus  $i$  is given by



$$P(T_i|D) = \frac{P(T_i)P(D|T_i)}{\sum_{x=1}^S P(T_x)P(D|T_x)}, \quad (1)$$

where  $S$  is the total number of genotypes.

- ▶ For a haploid genome  $H_m$  there are four types of genotypes:
  - ▶  $T_i = H_m \in \{A, C, G, T\}$ ,  $S = 4$ .
- ▶ For a diploid genome  $H_m H_n$  there are four types of genotypes:
  - ▶  $T_i = H_m H_n \in \{AA, CC, GG, TT, AC, AG, AT, CG, CT, GT\}$ ,  $S = 10$ .

The genotype with the highest posterior  $P(T_i|D)$  is chosen as the consensus, with a Phred-like score  $-10 \log_{10}[1 - P(T_i|D)]$ .

# SNP-calling model in SOAPsnp:

## Prior probability of genotypes

- ▶  $P(T_i)$  for a haploid genotype
  - ▶ Assumptions: SNP rate is 0s.001, transitions are  $4\times$  more frequent than transversions.
  - ▶ Given the reference allele G, the prior probabilities for relevant allele in reads are:  $6.67 \times 10^{-4}$  for A,  $1.67 \times 10^{-4}$  for C and T and 0.999 for G.



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- ▶  $P(T_i)$  for a diploid genotype

	A	C	G	T
A	$3.33 \times 10^{-4}$	$1.11 \times 10^{-7}$	$6.67 \times 10^{-4}$	$1.11 \times 10^{-7}$
C		$8.33 \times 10^{-5}$	$1.67 \times 10^{-4}$	$2.78 \times 10^{-8}$
G			0.9985	$1.67 \times 10^{-4}$
T				$8.33 \times 10^{-5}$

Assuming that the reference allele is G, the homozygous SNP rate is 0.0005, the heterozygous SNP rate is 0.001, and the ratio of transitions versus transversions is 4.

# SNP-calling model in SOAPsnp:

## Observation likelihood

- ▶  $P(D|T)$  calculated from observed allele types in the sequencing reads.
- ▶ Let  $P(d_k|H)$  be the likelihood of observing allele  $d_k$  for a possible haploid genotype  $H$ .
- ▶ For a diploid genome with the assumption that the two copies are independent



$$P(d_k|T) = \frac{P(d_k|H_m) + P(d_k|H_n)}{2},$$

- ▶ For a set of  $n$  observed alleles at a locus  $i$ ,  $D = \{d_1, \dots, d_n\}$ ,
  - ▶  $P(D|T) = \prod_{k=1}^n P(d_k|T)$ .

# SNP-calling model in SOAPsnp:

## Recalibration of base calling quality

- ▶ For each allele  $d_k$  observed for an assumed genotype  $T$  we define
  1.  $o_k$ , observed allele type
  2.  $q_k$ , quality score
  3.  $c_k$ , sequencing cycle (coordinate on read).
- ▶ Then the likelihood  $P(d_k|T)$  becomes
  - ▶

$$P(d_k|T) = P((o_k, q_k, c_k)|T) = P((o_k, c_k)|(T, q_k))P(q_k|T).$$

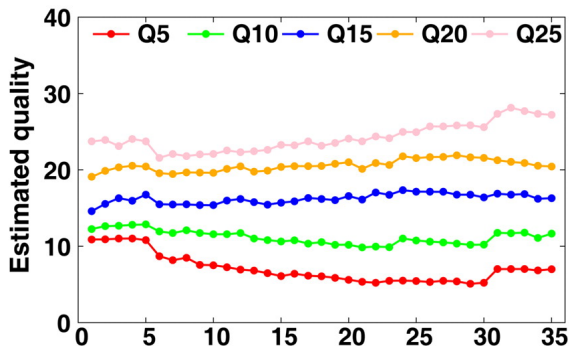
# SNP-calling model in SOAPsnp:

## Recalibration of base calling quality

- ▶ Four-dimensional matrix built to store  $P((o_k, c_k)|(T, q_k))$ .
- ▶ Precalculated from unique alignments, counting the number of substitutions, and estimating the mismatch rate for each combination of  $q_k$ ,  $c_k$  and substitution type.
- ▶ Effectively, each quality score rescaled by each sequencing cycle and for each substitution combination.
- ▶  $P(q_k|T)$  is the probability of a genotype  $T$  to have an observation with quality  $q_k$ 
  - ▶ Assumed that for  $T = A, C, G, T$  these distributions are the same
  - ▶  $P(q_k|T)$  becomes a function of  $q_k$  only,  $P(q_k|T) = f(q_k)$ ,
  - ▶  $f(q_k)$  reduces in the Bayesian formula (1).

# SNP-calling model in SOAPsnp:

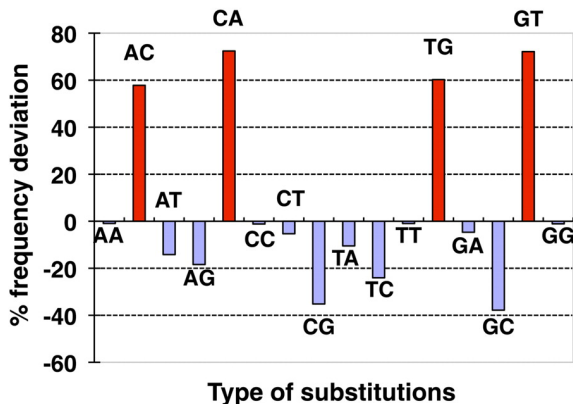
## Recalibration of base calling quality



- ▶ Extracted bases with each quality value from raw aligned reads
- ▶ Estimated quality =  $-10 \log_{10}(\text{mismatchrate})$ .

# SNP-calling model in SOAPsnp:

## Estimated vs quality-based mismatch rate



- ▶ % frequency deviation =  $\frac{[(\text{Error rate by alignment mismatch rate}) - (\text{error rate by quality value})]}{(\text{Error rate by quality value})}$ .

# SNP-calling model in SOAPsnp:

## Dealing with dependent errors

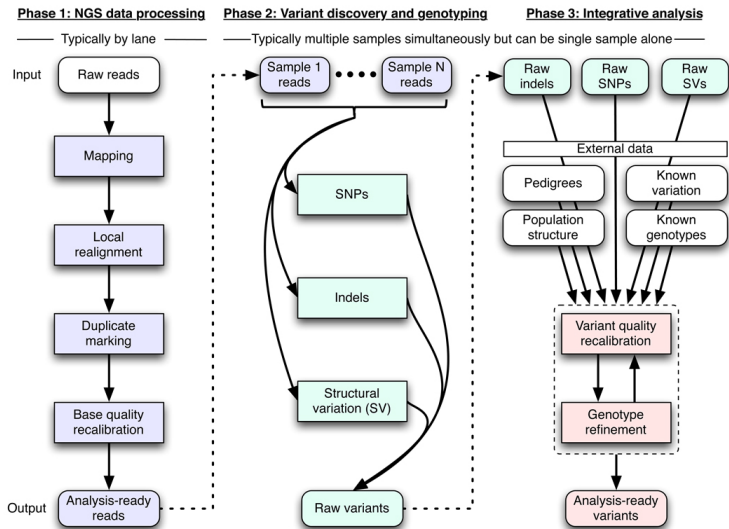
- ▶ The same alleles from reads mapped at the same location ranked by sequencing quality (low to high).
- ▶ Reduction of quality based on dependence for  $t_k$ -th observation

$$q'_k = \theta^{t_k} q_k,$$

where the dependency coefficient  $0 < \theta < 1$ .  $\theta = 0$  means the completely dependent model,  $\theta = 1$  means completely independent model.

- ▶ The reduced qualities  $q'_k$  used instead of the original  $q_k$  in the likelihood matrix.

# Genome Analysis Toolkit (GATK)





# Genome Analysis Toolkit (GATK):

## Steps of local realignment of reads spanning an indel.

1. Identify regions for realignment where
  - ▶ at least one read contains an indel,
  - ▶ there exists a cluster of mismatching bases or
  - ▶ an already known indel segregates at the site (e.g., dbSNP).
2. At each region, construct haplotypes by incorporating
  - ▶ any known indels at the site,
  - ▶ indels in reads spanning the site or
  - ▶ Smith-Waterman alignment of all reads that do not perfectly match the reference sequence.
3. For each haplotype  $H_i$ , each read  $R_j$  is aligned without gaps to  $H_i$  and assigned the **likelihood**  $L(R_j|H_i)$ .
4. Realign the reads to  $i$  if the log likelihood ratio satisfies

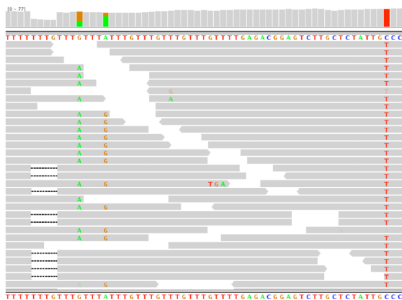
$$\log \frac{\prod_j \max[L(R_j|H_i), L(R_j|H_0)]}{L(R_j|H_0)} > 5$$

## Genome Analysis Toolkit (GATK):

The likelihood  $L(R_j|H_i)$  in the local realignment of reads spanning an indel.

$$\begin{aligned} L(R_j|H_i) &= \prod_k L(R_{j,k}|H_{i_k}) \\ L(R_j|H_i) &= \begin{cases} 1 - \epsilon_{j,k} & \text{if } R_{j,k} = H_{j,k}, \\ \epsilon_{j,k} & \text{otherwise.} \end{cases} \end{aligned} \quad (2)$$

# Read realignment in GATK points at common misalignment errors.



HiSeq data, raw BWA alignments



HiSeq data, after MSA

# Bibliography

- ▶ R. Li et al., *SNP detection for massively parallel whole-genome resequencing*. Genome Res. 2009.
- ▶ H. Li and N. Homer, *A survey of sequence alignment algorithms for next-generation sequencing*. Briefings in Bioinformatics 2010.
- ▶ M. A. DePristo et al., *A framework for variation discovery and genotyping using next-generation DNA sequencing data*. Nature Genetics 2011.