Variant calling Detecting variants in NGS data

Samtools and the Genome Analysis ToolKit (GATK)

University of Cambridge

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Courtesy of Marta Bleda

The pipeline

Objective

Assign a genotype to each position

Problems

Some variation observed in BAM files is caused by mapping and sequencing artifacts:

- **PCR artifacts**:
	- Mismatches due to errors in early PCR rounds
	- PCR duplicates
- **Sequencing errors**: erroneous call, either for physical reasons or to properties of the sequenced DNA
- Mapping errors: often happens around repeats or other low-complexity regions

Separate **true variation** from machine artifacts

Variant calling process pipeline

1. Mark duplicates

Duplicates should not be counted as additional evidence

2. Local realignment around INDELS

Reads mapping on the edges of INDELS often get mapped with mismatching bases introducing false positives

3. Base quality score recalibration (BQSR)

Quality scores provided by sequencing machines are generally inaccurate and biased

4. Variant calling

Discover variants and their genotypes

1. Mark duplicates

- The same DNA molecule can be **sequenced several times during PCR**
- **Not informative**
- **Not** to be counted as **additional evidence** for or against a putative variant
- Can result in **false variant calls**

Tools

- **Samtools**: samtools rmdup
- **Picard**: MarkDuplicates

1. Mark duplicates

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- **Not informative**
- **Not** to be counted as **additional evidence** for or against a putative variant
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Tools

- **Samtools**: samtools rmdup
- **Picard**: MarkDuplicates

1. Mark duplicates The reason why duplicates are bad

 $* =$ sequencing error propagated in duplicates

... and thus be more likely to make the right call

1. Mark duplicates Duplicate identification

Duplicates have the **same starting position** and the **same CIGAR** string

2. Local realignment around INDELS

- Reads **near INDELS** are mapped with mismatches
- **Realignment** can identify the most consistent placement for these reads
	- **1. Identify** problematic regions
	- **2. Determine the optimal** consensus sequence
- **Minimizes mismatches** with the reference sequence
- **Refines** location of **INDELS**

DePristo MA, et al. A framework for variation discovery and genotyping using next-generation DNA sequencing data. Nat Genet. 2011 May;43(5):491-8. PMID: 21478889

2. Local realignment around INDELS

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3. Base quality score recalibration

- **Calling algorithms rely** heavily on the **quality scores** assigned to the individual base calls in each sequence read
- Unfortunately, the scores produced by the machines are subject to various sources of **systematic error**, leading to over- or under-estimated base quality scores in the data

How?

- **1. Analyze covariation** among several features of a base:
	- Reported quality score
	- Position within the read
	- Preceding and current nucleotide
- 2. Use a set of **known variants** (i.e.: dbSNP) to model error properties of real polymorphism and determine the **probability that novel sites are real**
- **3. Adjust** the quality scores of all reads in a BAM file

3. Base quality score recalibration

Before After

Phred Quality score:

$$
Q = -10 \log_{10} P(\text{error})
$$

Steps

- **1. Variant calling:** Identify the positions that differ from the reference
- **2. Genotype calling**: calculate the genotypes for each sample at these sites

Initial approach

Independent base assumption

Counting the number of times each allele is observed

Evolved approach

Bayesian inference → Compute genotype likelihood Advantages:

Provide statistical measure of **uncertainty** Lead to **higher accuracy** of genotype calling

Variant discovery process

Reference = A

Variant discovery process

Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAAA *N*=30, *X*=0

Variant discovery process

Reference = A

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Variant discovery process

Reference = A

 N = nucleotides $G =$ true genotype R = reference base $V =$ variant base $X =$ variant nucleotides Outcomes: RR RV VV

Variant discovery process

Reference = A

- AAAAAAAAAAAAAAAAAAAAAAAAAAAAAA *N*=30, *X*=0
- GGGGGGGGGGGGGGGGGGGGGGGGGGGGGG *N*=30, *X*=30
- AAAAAAAAAAAAAAAGGGGGGGGGGGGGGG *N*=30, *X*=15
- AAAAAAAAAAAAAAAAGGGGGGGGGGGGCT *N*=30, *X*=12

AAAGGGCCTT *N*=10, *X*=3

-
-

 N = nucleotides $G =$ true genotype $R =$ reference base $V =$ variant base $X =$ variant nucleotides Outcomes: RR RV VV

Variant discovery process

Reference = A

 AAAAAAAAAAAAAAAAAAAAAAAAAAAAAA *N*=30, *X*=0 GGGGGGGGGGGGGGGGGGGGGGGGGGGGGG *N*=30, *X*=30 AAAAAAAAAAAAAAAGGGGGGGGGGGGGGG *N*=30, *X*=15 AAAAAAAAAAAAAAAAGGGGGGGGGGGGCT *N*=30, *X*=12 AAAGGGCCTT *N*=10, *X*=3

Cutoff for $X \rightarrow$ value or proportion

•
$$
c_1 = 10\%
$$
, $c_2 = 30\%$ $X \le c_1$ \rightarrow RR
 $c_1 < X < c_2$ \rightarrow RV
 $X \ge c_2$ \rightarrow VV

Variant discovery process

Reference = A

 AAAAAAAAAAAAAAAAAAAAAAAAAAAAAA *N*=30, *X*=0 → **RR** GGGGGGGGGGGGGGGGGGGGGGGGGGGGGG *N*=30, *X*=30 → **VV** AAAAAAAAAAAAAAAGGGGGGGGGGGGGGG *N*=30, *X*=15 → **RV** AAAAAAAAAAAAAAAAGGGGGGGGGGGGCT *N*=30, *X*=12 → **RV** AAAGGGCCTT *N*=10, *X*=3 → **RV?**

Cutoff for $X \rightarrow$ value or proportion

•
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c_1 = 10\%
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, $c_2 = 30\%$ $X \le c_1$ \rightarrow RR
 $c_1 < X < c_2$ \rightarrow RV
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-
-
-
-

```
N = nucleotides
G =true genotype
R = reference base
V =variant base
X = variant nucleotides
Outcomes:
    RR RV VV
```
Variant discovery process

 α = nucleotide-base error rate

 $G =$ true genotype $R =$ reference base $V =$ variant base $X =$ variant nucleotides Outcomes: RR RV VV

$$
P(G=RR, X|N, \alpha) = \frac{P}{\alpha}
$$

P of all R calls being correct and \mathbf{H} ι ∪ι all ι\ calls bellig co
all V calls being wrong

$$
P(G=VV, X|N, \alpha)
$$
 = P of all V calls being correct and
all R calls being wrong

 $P\!\left(G\!=\!RV\,,X|N\,,\alpha\right)$ = F $\overline{}$ \overline{a} f a $\frac{1}{11}$ \mathbf{u} P of all R and V calls being correct

Variant discovery process

Bayesian approximation

 α = nucleotide-base error rate

$$
P(G=RR, X|N, \alpha) = {N \choose X} \alpha^{X} (1-\alpha)^{N-X}
$$

\n
$$
P(G=VV, X|N, \alpha) = {N \choose X} (1-\alpha)^{X} \alpha^{N-X}
$$

\n
$$
P(G=RV, X|N, \alpha) = {N \choose X} (\frac{1}{2})^{N}
$$

Variant discovery process

Bayesian approximation

 α = nucleotide-base error rate

$$
\begin{array}{cc}\n\rho_{VV} \\
\rho_{VR}\n\end{array}
$$
 Prior probabilities

 N = nucleotides $G =$ true genotype $R =$ reference base $V =$ variant base $X =$ variant nucleotides Outcomes: RR RV VV

$$
P(G=RR, X|N, \alpha) = {N \choose X} \alpha^{X} (1-\alpha)^{N-X} (1-p_{VV}-p_{RV})
$$

\n
$$
P(G=VV, X|N, \alpha) = {N \choose X} (1-\alpha)^{X} \alpha^{N-X} p_{VV}
$$

\n
$$
P(G=RV, X|N, \alpha) = {N \choose X} \left(\frac{1}{2}\right)^{N} p_{RV}
$$

VCF file format

- Specification defined by the 1000 genomes (current version **4.2**): http://www.1000genomes.org/wiki/Analysis/Variant%20Call%20Format/vcf-variant-call-format-version-41
- Commonly **compressed and indexed** with bgzip/tabix
- Single-sample or multi-sample VCF

- **CHROM**: chromosome
- **POS**: position
- **ID**: identifier
- **REF:** reference base(s)
- **ALT:** non-reference allele(s)
- **QUAL:** quality score of the calls (phed scale)
- **FILTER**: "PASS" or a filtering tag
- **INFO:** additional information
- **FORMAT:** describes the information given by sample

Software

A more complete list is available from http://seganswers.com/wiki/Software/list. LD, linkage disequilibrium; NGS, next-generation sequencing.

Software

Prerequisites: JAVA and Picard tools

- **Requires Java** (**http://www.oracle.com/technetwork/java/javase/downloads/index.html**)
	- Check your java version

java –version

 $GATK \geq 2.6 \rightarrow$ Requires Java version 1.7

- **Picard (current version 1.130)**
	- Website: **http://broadinstitute.github.io/picard/**
	- For a compiled version click on "Latest Release" and download **picard-tools-1.130.zip**
	- Testing:

java -jar picard.jar -h

– Usage

java –jar picard.jar <ToolName> [options]

Samtools installation

● **Samtools 1.2 download**

– Download $\frac{1}{2}$ samtools-1.2

 \pm bcftools-1.2 \pm htslib-1.2.1

– Uncompress each of the files and inside the uncompressed folder execute:

• Check if Samtools is working

samtools

● **Usage**

samtools <command> [options]

GATK installation

● **Usage**

java –jar GenomeAnalysisTK.jar -T <ToolName> [arguments]

Filtering recommendations

Filtering recommendations for SNPs:

- QD $\langle 2.0$
- $MQ < 40.0$
- $-$ FS > 60.0
- HaplotypeScore > 13.0
- MQRankSum < -12.5
- ReadPosRankSum < -8.0

Filtering recommendations for indels:

- QD $\langle 2.0$
- ReadPosRankSum < -20.0
- InbreedingCoeff < -0.8
- $-$ FS > 200.0