

Variant calling

Detecting variants in NGS data

The Genome Analysis ToolKit (GATK)

Edinburgh Genomics

Edinburgh, UK

23rd October 2015

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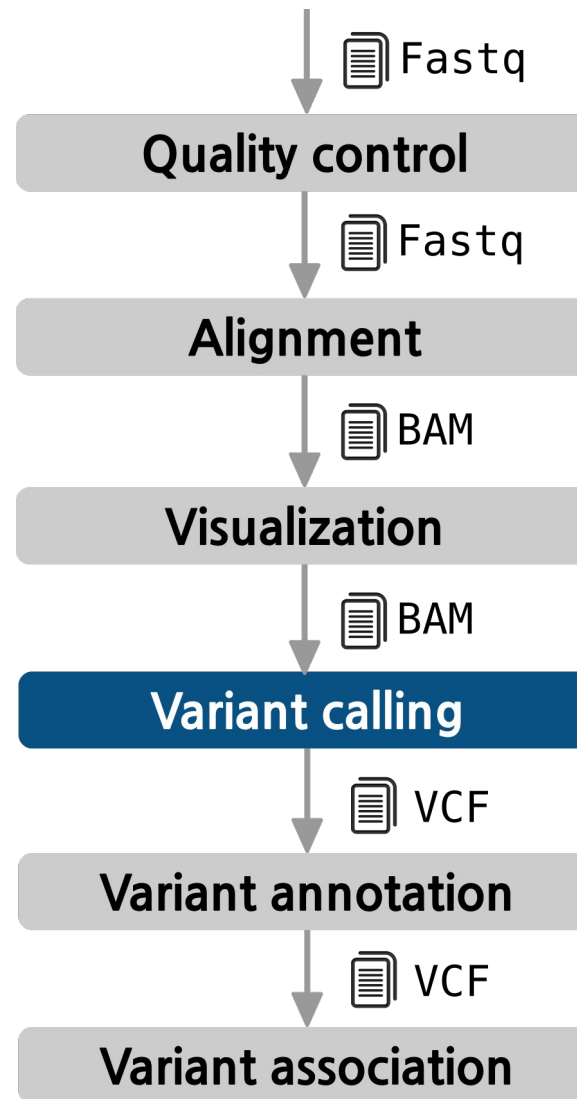
mb2033@cam.ac.uk

Research Assistant at the Department of Medicine

University of Cambridge

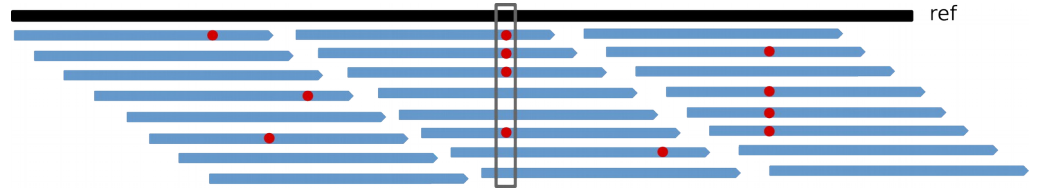
Cambridge, UK

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

- 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**

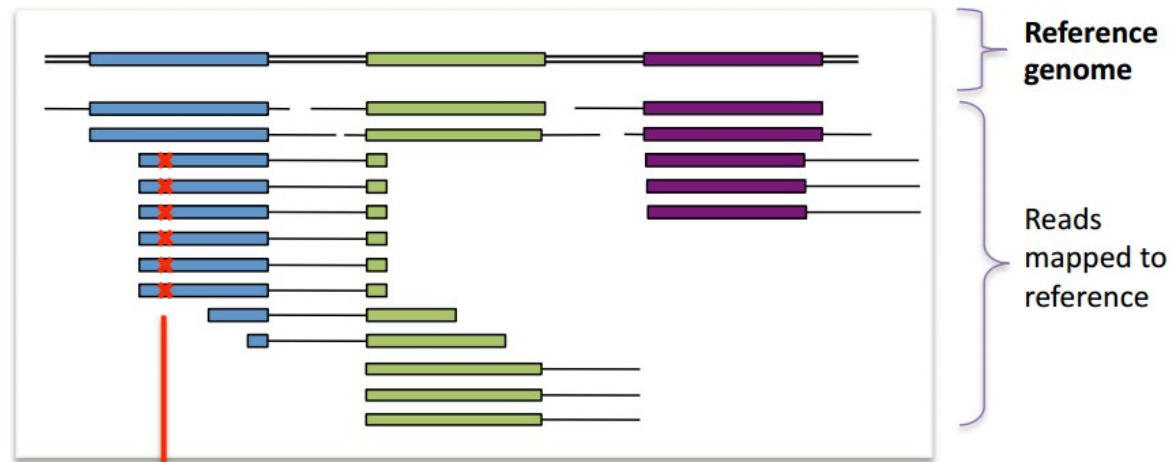
Tools

- Samtools: `samtools rmdup`
- Picard: `MarkDuplicates`

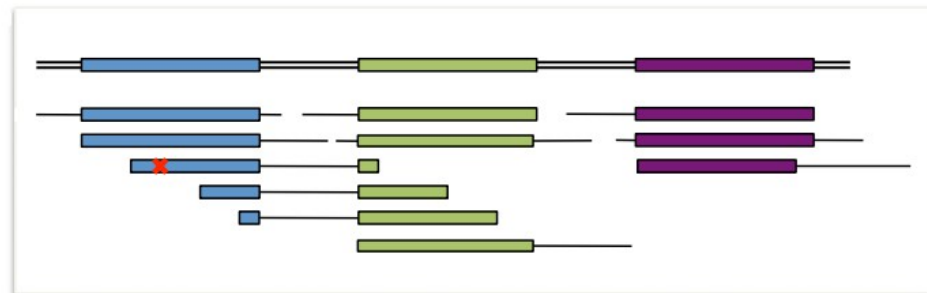
1. Mark duplicates

The reason why duplicates are bad

✗ = sequencing error propagated in duplicates



After marking duplicates, the GATK will only see :

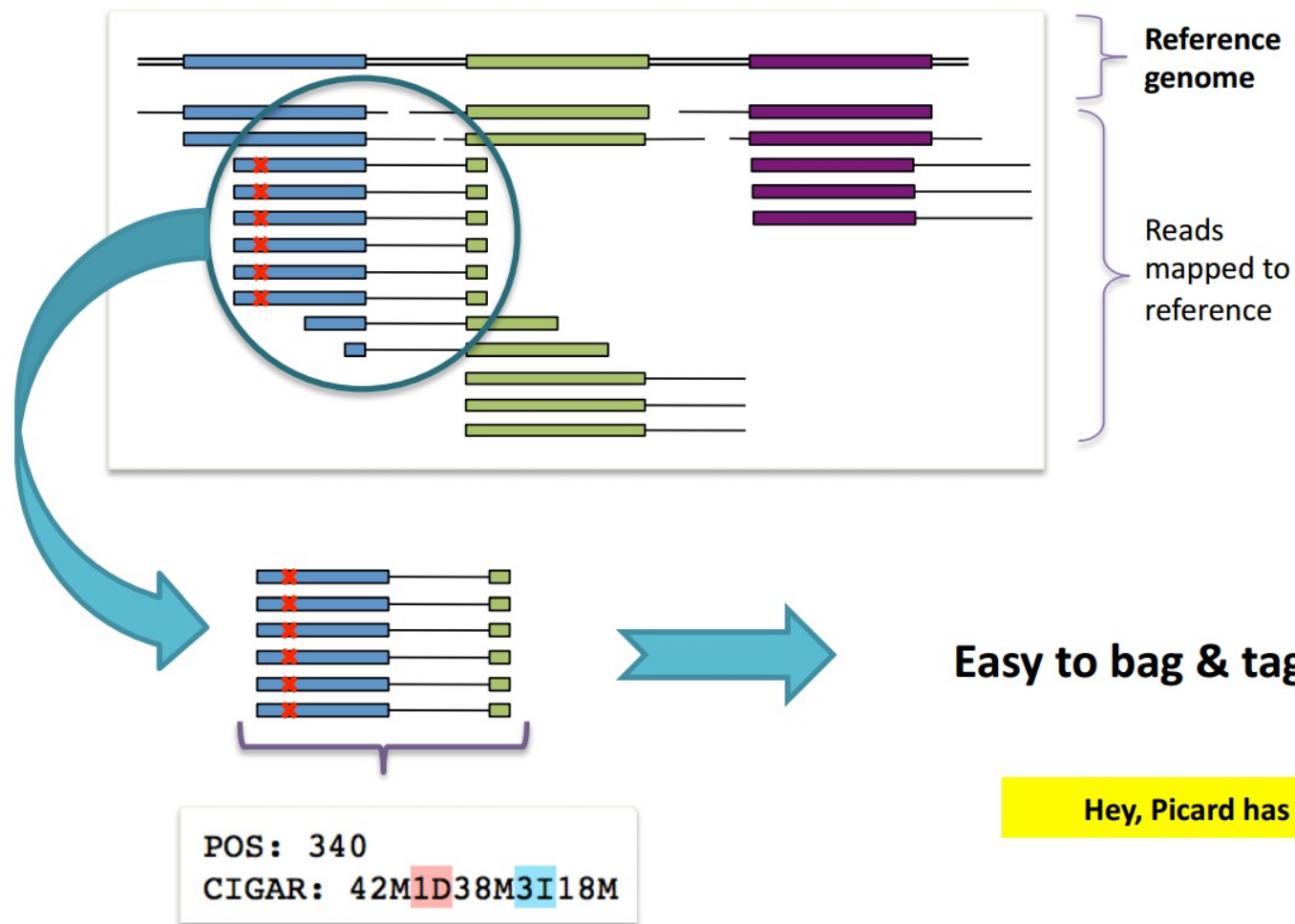


... and thus be more likely to make the right call

1. Mark duplicates

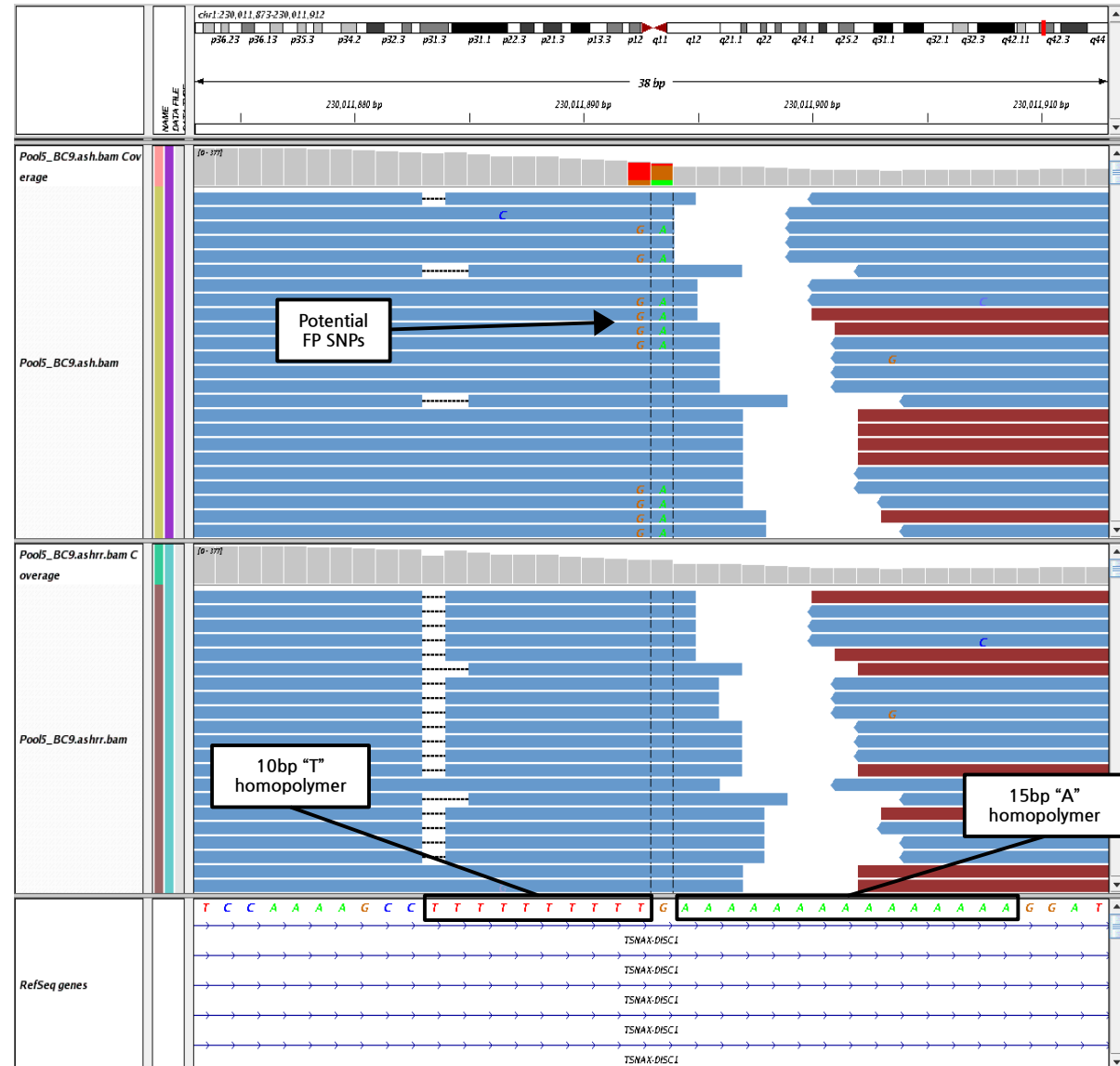
Duplicate identification

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**

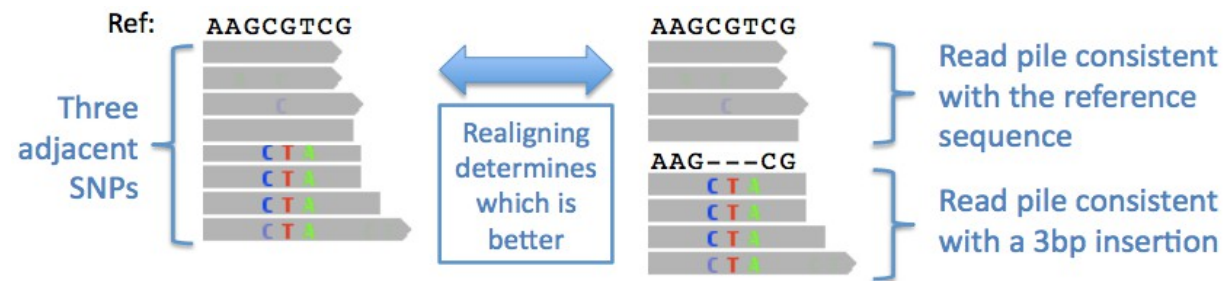


2. Local realignment around INDELS

- 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**



3. Base quality score recalibration

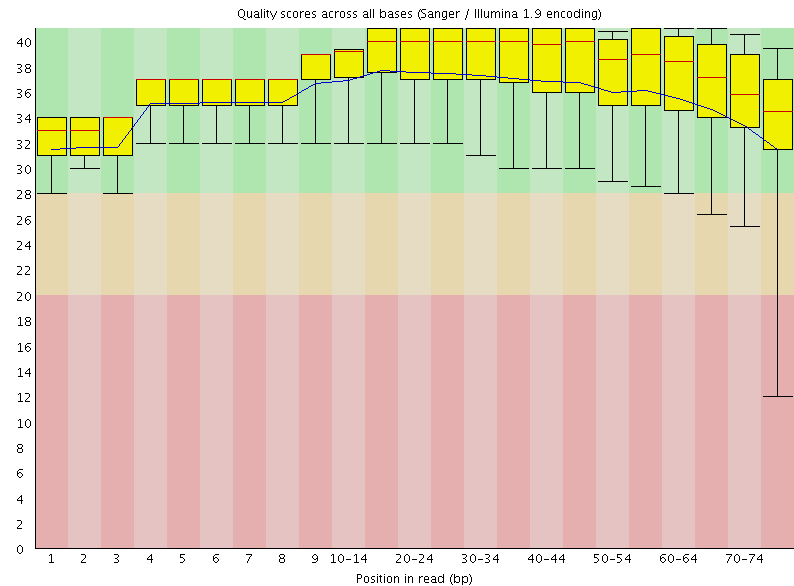
- **Calling algorithms rely** heavily on the **quality scores** assigned to the individual base calls in each sequence read
- Unfortunately, the scores produced by the machines are subject to various sources of **systematic error**, leading to over- or under-estimated base quality scores in the data

How?

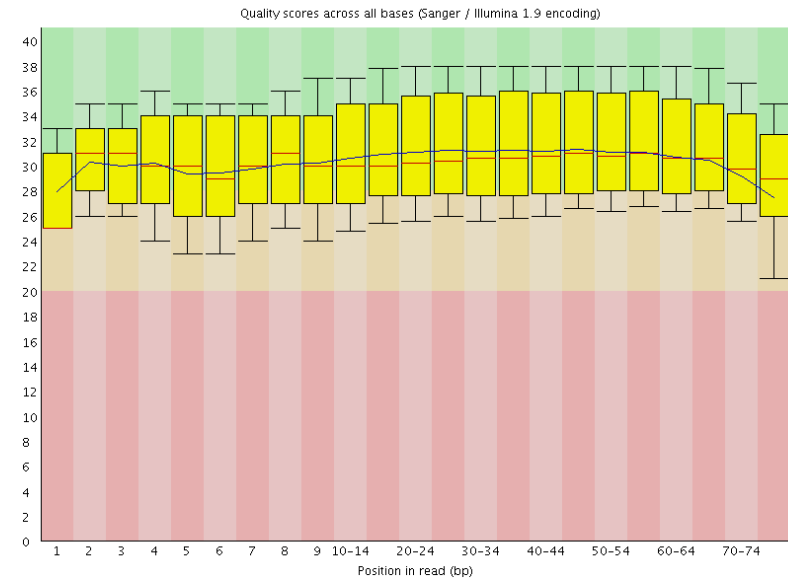
1. **Analyze covariation** among several features of a base:
 - Reported quality score
 - Position within the read
 - Preceding and current nucleotide
2. Use a set of **known variants** (i.e.: dbSNP) to model error properties of real polymorphism and determine the **probability that novel sites are real**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

3. Base quality score recalibration

Before



After



Phred Quality score:

$$Q_{\text{Phred}} = -10 \log_{10} P(\text{error}).$$

A score of 20 corresponds to 1 % error rate in base calling

4. Variant calling

Variant discovery process

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

4. Variant calling

Variant discovery process



Reference = A

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0$

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:

RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0$

GGGGGGGGGGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=30$

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:

RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0$

GGGGGGGGGGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=30$

AAAAAAAAAAAAAAGGGGGGGGGGGGGGGG

$N=30, \quad X=15$

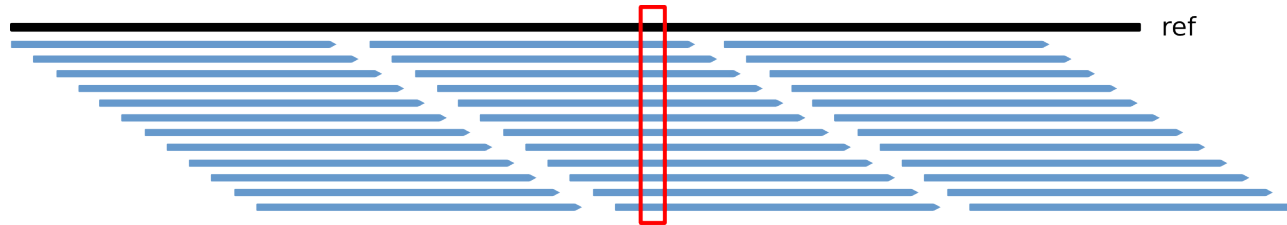
N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:

RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0$

GGGGGGGGGGGGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=30$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=15$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGGGGGCT

$N=30, \quad X=12$

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:
RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0$

GGGGGGGGGGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=30$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGG

$N=30, \quad X=15$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGCT

$N=30, \quad X=12$

AAAGGGCCTT

$N=10, \quad X=3$

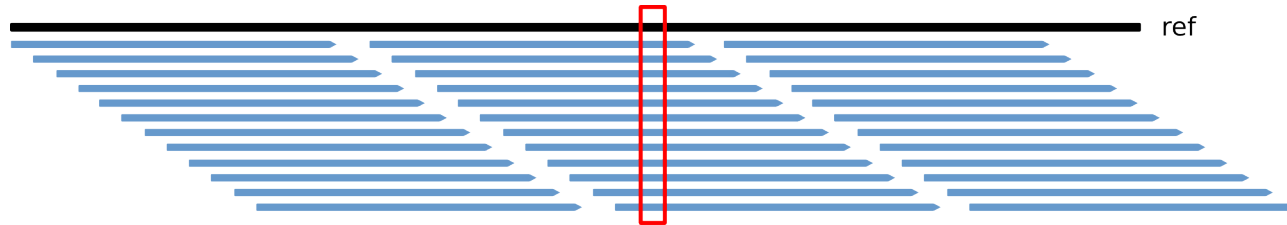
N = nucleotides
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Outcomes:

RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA	$N=30, \quad X=0$
GGGGGGGGGGGGGGGGGGGGGGGGGGGG	$N=30, \quad X=30$
AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGGG	$N=30, \quad X=15$
AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGCT	$N=30, \quad X=12$
AAAGGGCCTT	$N=10, \quad X=3$

Cutoff for $X \rightarrow$ value or proportion

- $c_1 = 10\%, c_2 = 30\%$

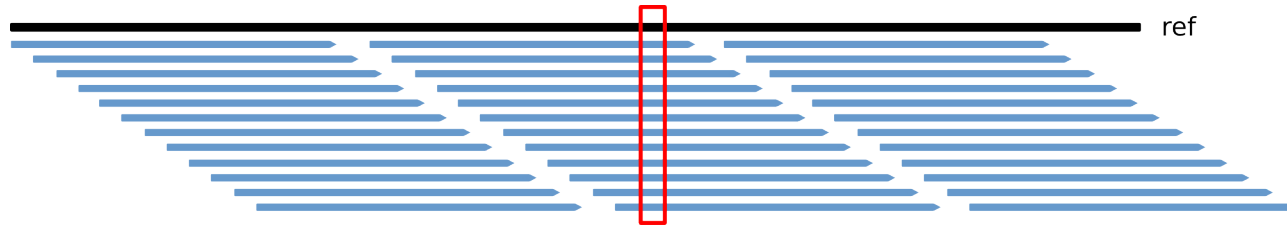
$X \leq c_1$	\rightarrow	RR
$c_1 < X < c_2$	\rightarrow	RV
$X \geq c_2$	\rightarrow	VV

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:
RR RV VV

4. Variant calling

Variant discovery process



Reference = A

AAAAAAAAAAAAAAAAAAAAAAAAAAAAA

$N=30, \quad X=0 \rightarrow RR$

GGGGGGGGGGGGGGGGGGGGGGGGGGGG

$N=30, \quad X=30 \rightarrow VV$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGGG

$N=30, \quad X=15 \rightarrow RV$

AAAAAAAAAAAAAAAAAGGGGGGGGGGGGGCT

$N=30, \quad X=12 \rightarrow RV$

AAAGGGCCTT

$N=10, \quad X=3 \rightarrow ??$

Cutoff for $X \rightarrow$ value or proportion

- $c_1 = 10\%, c_2 = 30\%$

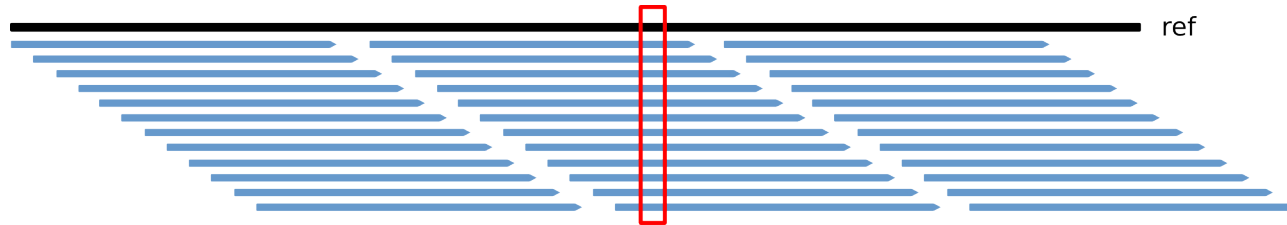
$X \leq c_1$	\rightarrow	RR
$c_1 < X < c_2$	\rightarrow	RV
$X \geq c_2$	\rightarrow	VV

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:
RR RV VV

4. Variant calling

Variant discovery process



Bayesian approximation

α = nucleotide-base error rate

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:

RR RV VV

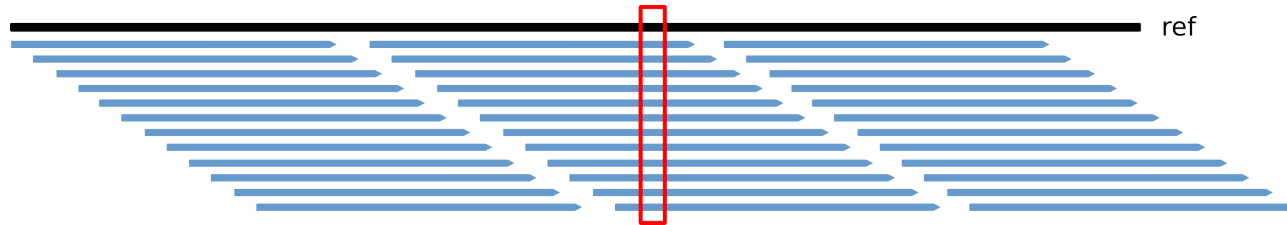
$P(G=RR, X|N, \alpha)$ = P of all R calls being correct and all V calls being wrong

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

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

4. Variant calling

Variant discovery process



Bayesian approximation

α = nucleotide-base error rate

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:

RR RV VV

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

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

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

4. Variant calling

Variant discovery process



Bayesian approximation

α = nucleotide-base error rate

p_{VV}
 p_{VR} } Prior probabilities

N = nucleotides
 G = true genotype
 R = reference base
 V = variant base
 X = variant nucleotides

Outcomes:
RR RV VV

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

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

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

VCF file format

- Specification defined by the 1000 genomes (current version 4.2):
<http://samtools.github.io/hts-specs/VCFv4.2.pdf>
- Commonly **compressed and indexed** with bgzip/tabix
- Single-sample or multi-sample VCF

```
##fileformat=VCFv4.1
##fileDate=20090805
##source=myImputationProgramV3.1
##reference=file:///seq/references/1000GenomesPilot-NCBI36.fasta
##contig=<ID=20,length=62435964,assembly=B36,md5=f126cdf8a6e0c7f379d618ff66beb2da,species="Homo sapiens",taxonomy=x>
##phasing=partial
##INFO=<ID=NS,Number=1,Type=Integer,Description="Number of Samples With Data">
##INFO=<ID=DP,Number=1,Type=Integer,Description="Total Depth">
##INFO=<ID=AF,Number=A,Type=Float,Description="Allele Frequency">
##INFO=<ID=AA,Number=1,Type=String,Description="Ancestral Allele">
##INFO=<ID=DB,Number=0,Type=Flag,Description="dbSNP membership, build 129">
##INFO=<ID=H2,Number=0,Type=Flag,Description="HapMap2 membership">
##FILTER=<ID=q10,Description="Quality below 10">
##FILTER=<ID=s50,Description="Less than 50% of samples have data">
##FORMAT=<ID=GT,Number=1,Type=String,Description="Genotype">
##FORMAT=<ID=GQ,Number=1,Type=Integer,Description="Genotype Quality">
##FORMAT=<ID=DP,Number=1,Type=Integer,Description="Read Depth">
##FORMAT=<ID=HQ,Number=2,Type=Integer,Description="Haplotype Quality">
#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT NA00001 NA00002 NA00003
20 14370 rs6054257 G A 29 PASS NS=3;DP=14;AF=0.5;DB;H2 GT:GQ:DP:HQ 0|0:48:1:51,51 1|0:48:8:51,51 1/1:43:5:.,.
20 17330 . T A 3 q10 NS=3;DP=11;AF=0.017 GT:GQ:DP:HQ 0|0:49:3:58,50 0|1:3:5:65,3 0/0:41:3
20 1110696 rs6040355 A G,T 67 PASS NS=2;DP=10;AF=0.333,0.667;AA=T;DB GT:GQ:DP:HQ 1|2:21:6:23,27 2|1:2:0:18,2 2/2:35:4
20 1230237 . T . 47 PASS NS=3;DP=13;AA=T GT:GQ:DP:HQ 0|0:54:7:56,60 0|0:48:4:51,51 0/0:61:2
20 1234567 microsat1 GTC G,GTCT 50 PASS NS=3;DP=9;AA=G GT:GQ:DP 0/1:35:4 0/2:17:2 1/1:40:3
```

VCF file format

#CHROM	POS	ID	REF	ALT	QUAL	FILTER	INFO
20	14370	rs6054257	G	A	29	PASS	NS=3;DP=14;AF=0.5;DB;H2
FORMAT		NA000001	NA000002		NA000003		
GT:GQ:DP:HQ		0 0:48:1:51,51	1 0:48:8:51,51		1/1:43:5:.,.		

genotype genotype quality read depth haplotype qualities

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

Software

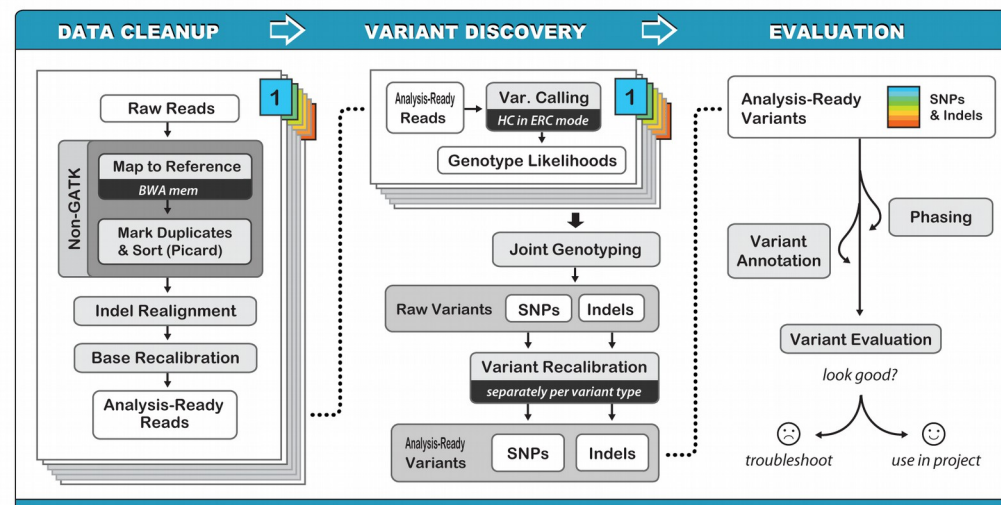
Software	Available from	Calling method	Prerequisites	Comments	Refs
SOAP2	http://soap.genomics.org.cn/index.html	Single-sample	High-quality variant database (for example, dbSNP)	Package for NGS data analysis, which includes a single individual genotype caller (SOAPsnp)	15
realSFS	http://128.32.118.212/thorfinn/realSFS/	Single-sample	Aligned reads	Software for SNP and genotype calling using single individuals and allele frequencies. Site frequency spectrum (SFS) estimation	-
Samtools	http://samtools.sourceforge.net/	Multi-sample	Aligned reads	Package for manipulation of NGS alignments, which includes a computation of genotype likelihoods (samtools) and SNP and genotype calling (bcftools)	53
GATK	http://www.broadinstitute.org/gsa/wiki/index.php/The_Genome_Analysis_Toolkit	Multi-sample	Aligned reads	Package for aligned NGS data analysis, which includes a SNP and genotype caller (Unified Genotyper), SNP filtering (Variant Filtration) and SNP quality recalibration (Variant Recalibrator)	32,33
Beagle	http://faculty.washington.edu/browning/beagle/beagle.html	Multi-sample LD	Candidate SNPs, genotype likelihoods	Software for imputation, phasing and association that includes a mode for genotype calling	42
IMPUTE2	http://mathgen.stats.ox.ac.uk/impute/impute_v2.html	Multi-sample LD	Candidate SNPs, genotype likelihoods	Software for imputation and phasing, including a mode for genotype calling. Requires fine-scale linkage map	44
QCall	ftp://ftp.sanger.ac.uk/pub/rd/QCALL	Multi-sample LD	'Feasible' genealogies at a dense set of loci, genotype likelihoods	Software for SNP and genotype calling, including a method for generating candidate SNPs without LD information (NLDA) and a method for incorporating LD information (LDA). The 'feasible' genealogies can be generated using Margarita (http://www.sanger.ac.uk/resources/software/margarita)	54
MaCH	http://genome.sph.umich.edu/wiki/Thunder	Multi-sample LD	Genotype likelihoods	Software for SNP and genotype calling, including a method (GPT_Freq) for generating candidate SNPs without LD information and a method (thunder_glf_freq) for incorporating LD information	-

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

GATK (Genome Analysis ToolKit)

<http://www.broadinstitute.org/gatk/>

- Probabilistic method: **Bayesian estimation** of the most likely genotype
- Calculates many **parameters** for each position of the genome
- INDEL realignment
- Base quality recalibration
- SNP and INDEL calling
- **Multi-sample** calling
- Uses standard input and output files
- Used in **many NGS projects**, including the 1000 Genomes Project, The Cancer Genome Atlas, etc.



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 → Requires Java version \geq 1.7

- **Picard (current version 1.140)**
 - Website: <http://broadinstitute.github.io/picard/>
 - For a compiled version click on “Latest Release” and download [picard-tools-1.140.zip](#)
 - Testing:



```
java -jar picard.jar -h
```


- Usage

```
java -jar picard.jar <ToolName> [options]
```

GATK installation

- GATK 3.4-46 download

<http://www.broadinstitute.org/gatk/>

- We need to register before download
- Go to Downloads and click 
- Accept the license agreement
- Extract the file

[Download now](#) 

You must be logged into the forums to proceed

You do not seem to be logged into the forums

[Register](#)

[Login Here »](#)

- Check if GATK is working

```
java -jar GenomeAnalysisTK.jar -h
```

Shows GATK help

- Usage

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

THANK YOU.

Filtering recommendations

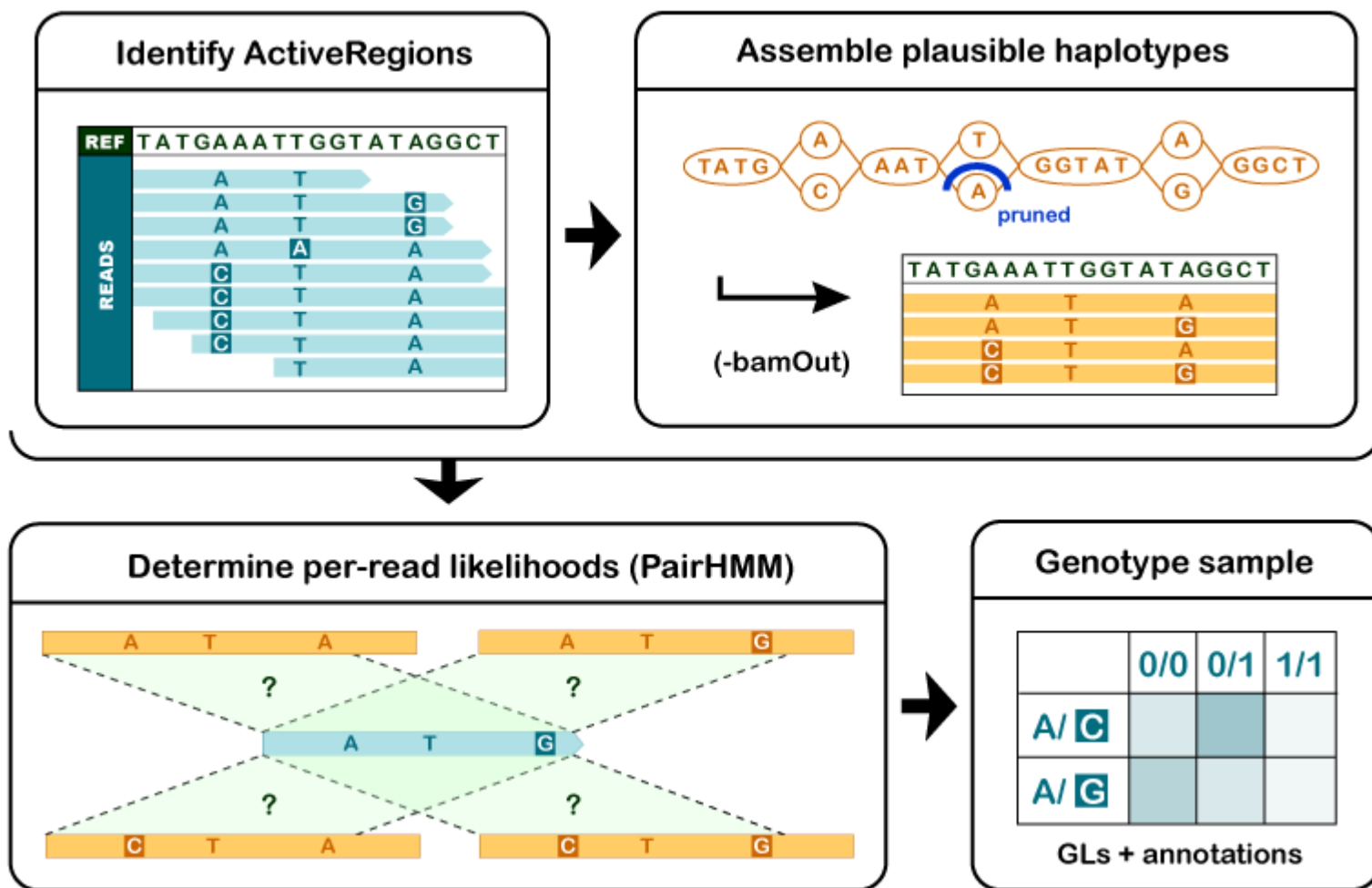
Filtering recommendations for SNPs:

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

Filtering recommendations for indels:

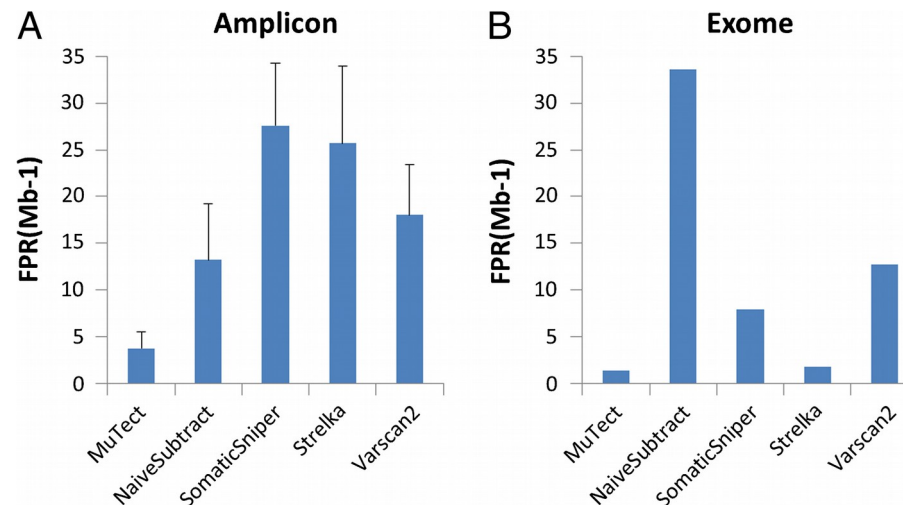
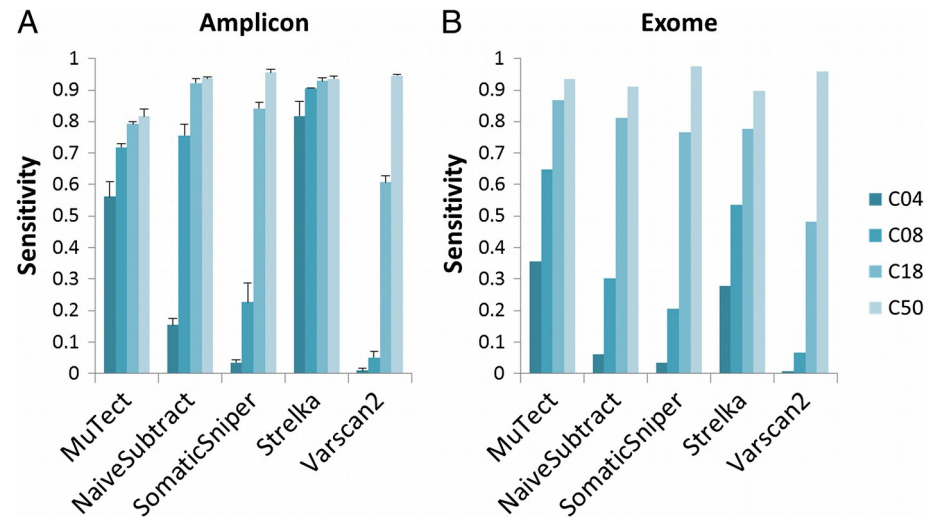
- $QD < 2.0$
- $ReadPosRankSum < -20.0$
- $InbreedingCoeff < -0.8$
- $FS > 200.0$

GATK (Genome Analysis ToolKit)



Somatic calling

Comparison



Xu, Huilei, et al. "Comparison of somatic mutation calling methods in amplicon and whole exome sequence data." *BMC genomics* 15.1 (2014): 244.