

Analysing re-sequencing samples

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What is resequencing?

- You have a reference genome
 - represents one individual
- You generate sequence from other individuals
 - same species
 - closely related species
- You want to identify variation
 - 1) map millions of reads to reference genome
 - 2) SNPs / indels / structural variation

What accuracy is required?

- Is the result of sequencing the final answer or will it be used for something else?
- What is the importance of reducing false positives and false negatives relative to sequencing cost?

Example 1: identification of new mutations

- e.g. comparison of tumour vs. normal tissue or comparison of parents vs offspring
- sensitivity to false positives and false negatives is high
- mutations extremely rare
- FP rate >1 per Mb will swamp signal
- samples may be precious

Example 2: SNP discovery

- Sequencing multiple individuals in order to design a SNP array
- High tolerance to false positives and false negatives (they will be validated by array)
- Does not need to be comprehensive – lower coverage acceptable
- Only interested in identifying markers to (e.g.) analyze population structure

Example 3: selection mapping

- Sequencing multiple individuals in order to scan genetic variation for signals of selection
- Looking for regions with reduced levels of SNP variation
- low false positive rate important
 - or selective sweeps will be obscured by noise

Types of reads

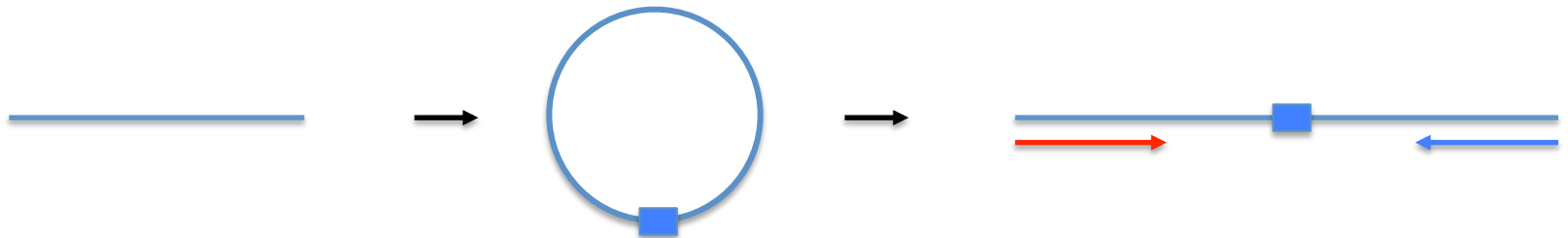
- fragment



- paired-end



- mate pair (jumping libraries)



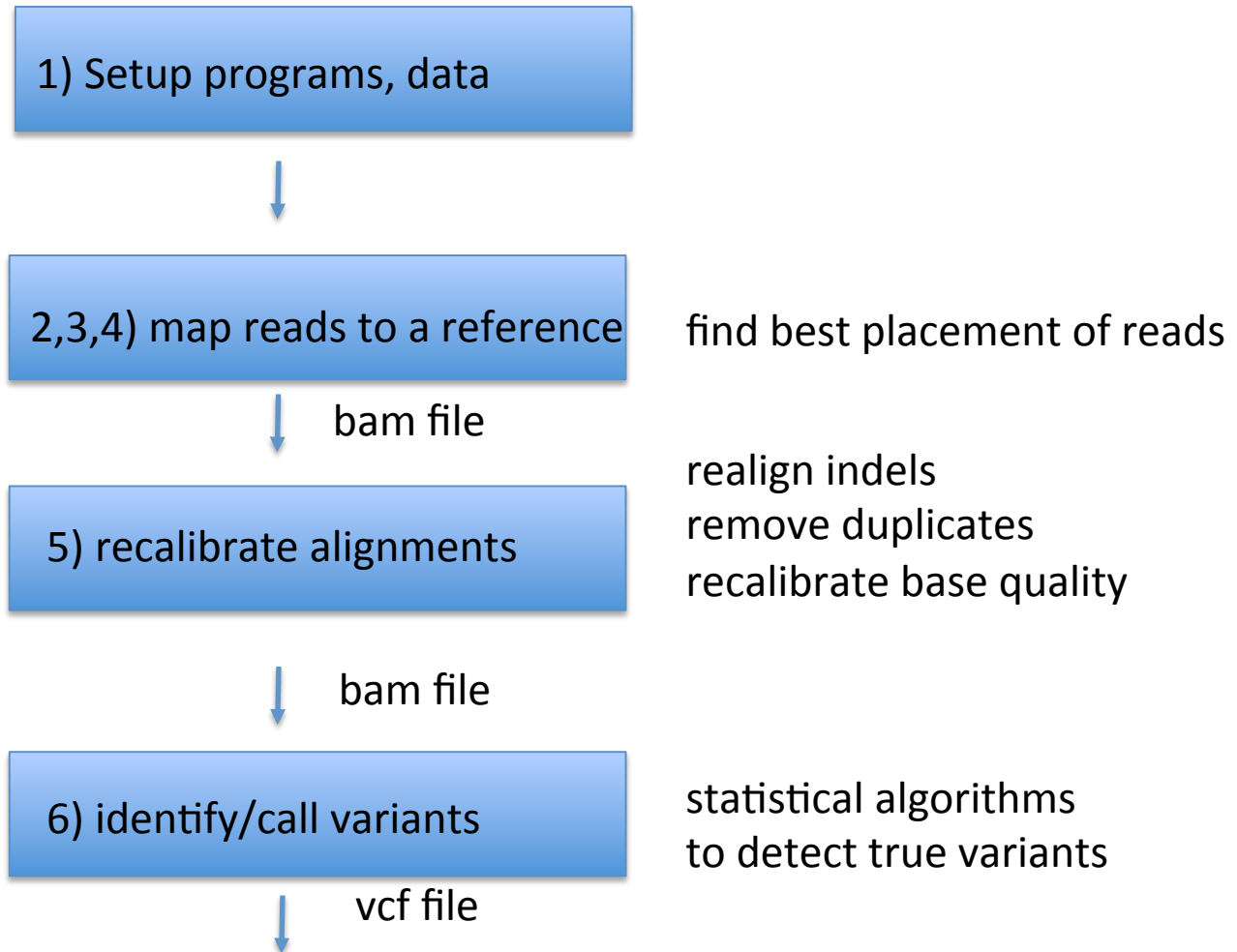
Benefits of each library type

- Fragments
 - fastest runs (one read per fragment)
 - lowest cost
- Paired reads
 - More data per fragment
 - improved mapping and assembly
 - same library steps, more data
 - Insert size limited by fragment length

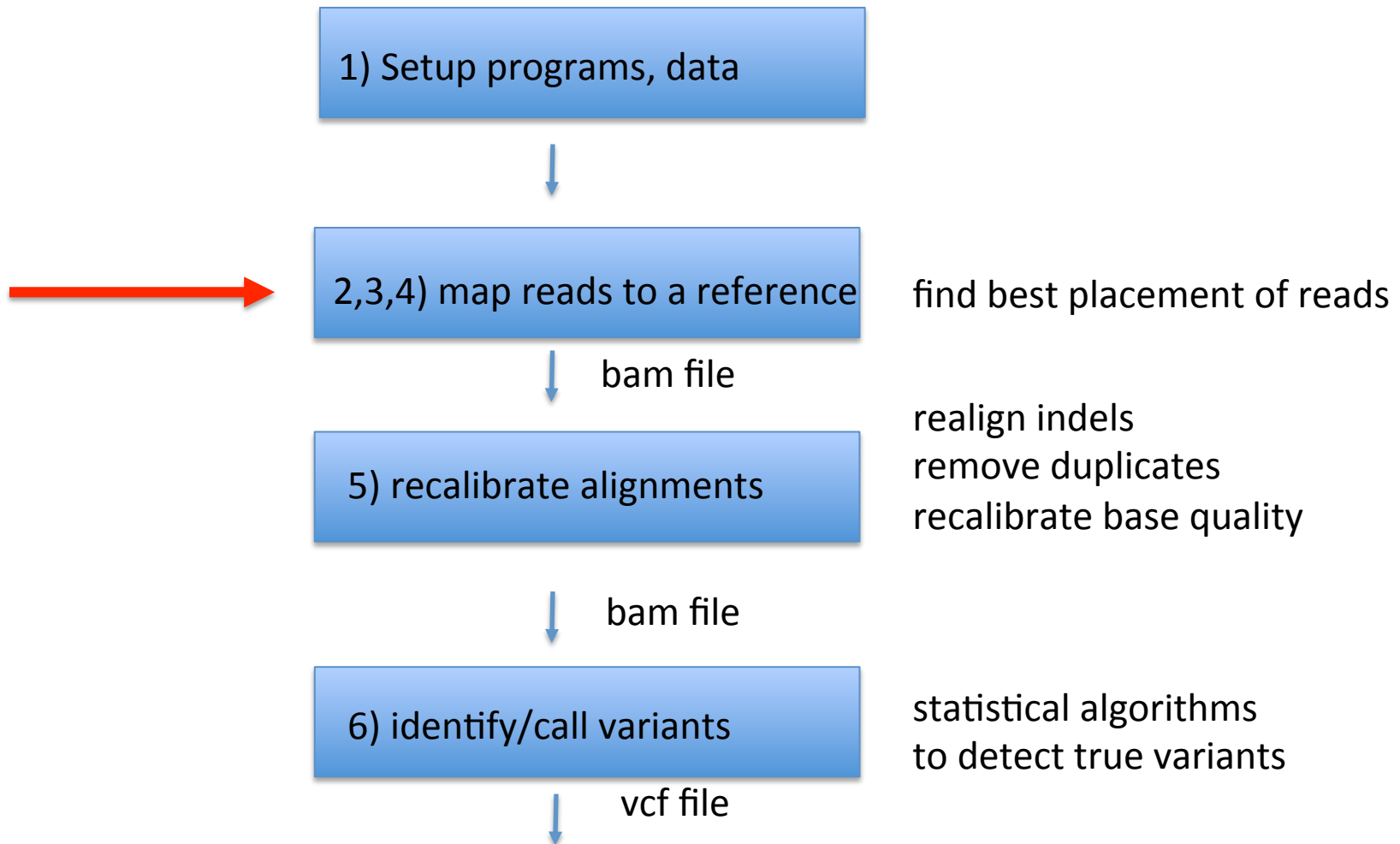
Benefits of each library type

- **Mate pairs**
 - Allows for longer insert sizes
 - Very useful for
 - assembly and alignment across duplications and low-complexity DNA
 - identification of large structural variants
 - phasing of SNPs
 - More DNA and more complex library preparation
 - Not all platforms can read second strand

Steps in resequencing



Steps in resequencing



Step 2: Map reads

- **Maq** (<http://maq.sourceforge.net/>)
 - nongapped
- **BWA** (<http://bio-bwa.sourceforge.net/>)
 - Burroughs-Wheeler aligner
 - gapped
 - successor to Maq
- **bowtie** (<http://bowtie-bio.sourceforge.net/index.shtml>)
 - fast + memory efficient
- **Mosaik** (<http://bioinformatics.bc.edu/marthlab/>)
 - Smith-Waterman

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mapping algorithm tricks

- simple brute force
- hash tables
- suffix trees
- Burroughs-Wheeler transform

brute force

TCGATCC

x

GACCTCATCGATCCCACTG

brute force

TCGATCC

x

GACCTCATCGATCCCACTG

brute force

TCGATCC

x

GACCTCATCGATCCCACTG

brute force

TCGATCC

x

GACCTCATCGATCCCACTG

brute force

TCGATCC
| | x
GACCTCATCGATCCCACTG

brute force

TCGATCC
x
GACCTCATCGATCCCACTG

brute force

TCGATCC

x

GACCTCATCGATCCCACTG

brute force

TCGATCC
| | | | | |
GACCTCA**TCGATCC**CACTG

hash tables

build an index of the reference sequence for fast access

	0	5	10	15		
seed length 7	GACCTCATCGATCCCCTG					
	GACCTCA				→	chromosome 1, pos 0
	ACCTCAT				→	chromosome 1, pos 1
	CCTCATC				→	chromosome 1, pos 2
	CTCATCG				→	chromosome 1, pos 3
	TCATCGA				→	chromosome 1, pos 4
	CATCGAT				→	chromosome 1, pos 5
	ATCGATC				→	chromosome 1, pos 6
	TCGATCC				→	chromosome 1, pos 7
	CGATCCC				→	chromosome 1, pos 8
GATCCCA				→	chromosome 1, pos 9	

hash tables

build an index of the reference sequence for fast access

TCGATCC ?

0 5 10 15

GACCTCATCGATCCCCTG

GACCTCA	→	chromosome 1, pos 0
ACCTCAT	→	chromosome 1, pos 1
CCTCATC	→	chromosome 1, pos 2
CTCATCG	→	chromosome 1, pos 3
TCATCGA	→	chromosome 1, pos 4
CATCGAT	→	chromosome 1, pos 5
ATCGATC	→	chromosome 1, pos 6
TCGATCC	→	chromosome 1, pos 7
CGATCCC	→	chromosome 1, pos 8
GATCCCA	→	chromosome 1, pos 9

hash tables

build an index of the reference sequence for fast access

TCGATCC = chromosome 1, pos 7

0 5 10 15

GACCTCATCGATCCCCTG

GACCTCA → chromosome 1, pos 0

ACCTCAT → chromosome 1, pos 1

CCTCATC → chromosome 1, pos 2

CTCATCG → chromosome 1, pos 3

TCATCGA → chromosome 1, pos 4

CATCGAT → chromosome 1, pos 5

ATCGATC → chromosome 1, pos 6

TCGATCC → chromosome 1, pos 7

CGATCCC → chromosome 1, pos 8

GATCCCA → chromosome 1, pos 9

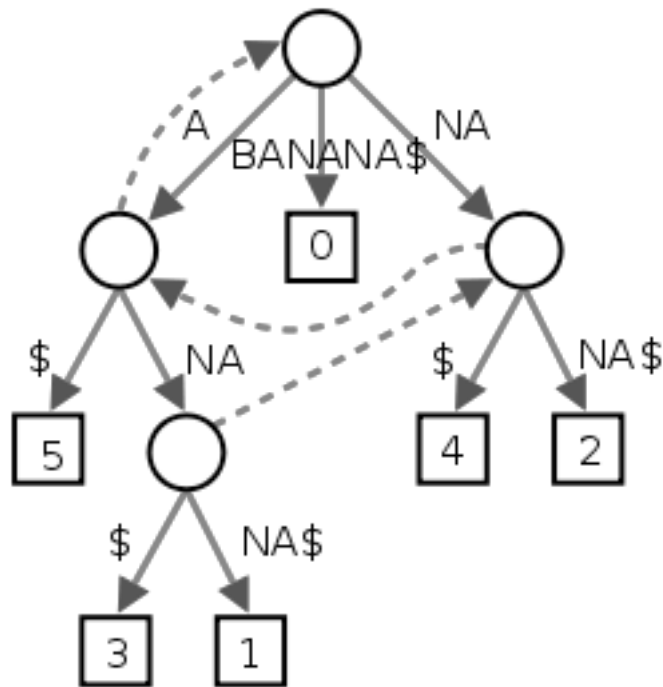
hash tables

Used by MAQ, Eland, SOAP, SHRiMP, ZOOM, partially by Mosaik

Problem: Indexing big genomes/lists of reads requires lots of memory

suffix trees

suffix tree for BANANA



breaks sequence into parts

(e.g. B, A, NA)

allows efficient searching of substrings in a sequence

Advantage: alignment of multiple identical copies of a substring in the reference is only needed to be done once because these identical copies collapse on a single path

Burroughs-Wheeler transform

Transformation				
Input	All Rotations	Sorting All Rows in Alphabetical Order by their first letters	Taking Last Column	Output Last Column
<code>^BANANA </code>	<code>^BANANA </code> <code> ^BANANA</code> <code>A ^BANAN</code> <code>NA ^BANA</code> <code>ANA ^BAN</code> <code>NANA ^BA</code> <code>ANANA ^B</code> <code>BANANA ^</code>	<code>ANANA ^B</code> <code>ANA ^BAN</code> <code>A ^BANAN</code> <code>BANANA ^</code> <code>NANA ^BA</code> <code>NA ^BANA</code> <code>^BANANA </code> <code> ^BANANA</code>	<code>ANANA ^B</code> <code>ANA ^BAN</code> <code>A ^BANAN</code> <code>BANANA ^</code> <code>NANA ^BA</code> <code>NA ^BANA</code> <code>^BANANA </code> <code> ^BANANA</code>	<code>BNN^AA A</code>

algorithm used in computer science for file compression

original sequence can be reconstructed

identical characters more likely to be consecutive → reduces memory required

Mapping algorithms

- Bowtie and BWA exploit suffix tree and BW transform
- Increases speed and decreases memory needed
- Standard output is now SAM or SAM binary (BAM) format

SAM format

HEADER SECTION

```
@HD VN:1.0 SO:coordinate
@SQ SN:1 LN:249250621 AS:NCBI37 UR:file:/data/local/ref/GATK/human_g1k_v37.fasta M5:1b22b98cdeb4a9304cb5d48026a85128
@SQ SN:2 LN:243199373 AS:NCBI37 UR:file:/data/local/ref/GATK/human_g1k_v37.fasta M5:a0d9851da00400dec1098a9255ac712e
@SQ SN:3 LN:198022430 AS:NCBI37 UR:file:/data/local/ref/GATK/human_g1k_v37.fasta M5:fdfd811849cc2fadebc929bb925902e5
@RG ID:UM0098:1 PL:ILLUMINA PU:HWUSI-EAS1707-615LHAAXX-L001 LB:80 DT:2010-05-05T20:00:00-0400 SM:SD37743 CN:UMCORE
@RG ID:UM0098:2 PL:ILLUMINA PU:HWUSI-EAS1707-615LHAAXX-L002 LB:80 DT:2010-05-05T20:00:00-0400 SM:SD37743 CN:UMCORE
@PG ID:bwa VN:0.5.4
```

ALIGNMENT SECTION

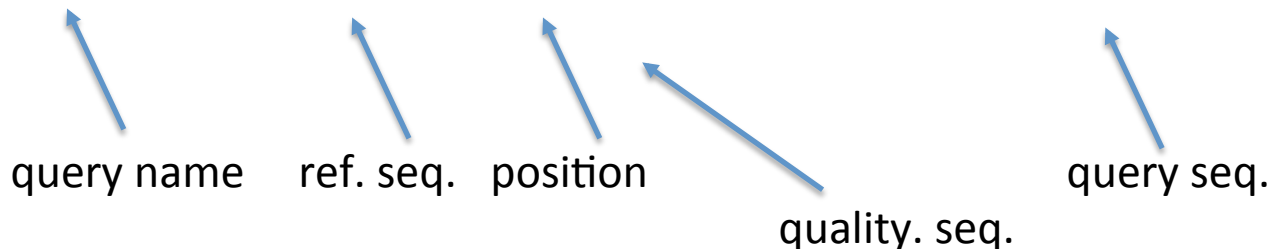
```
8_96_444_1622 73 scaffold00005 155754 255 54M * 0 0 ATGTAAAGTATTTCCATGGTACACAGCTTGGTCGTAATGTGATTGCTGAGCCAG
BC@B5)5CBBCCBCCCBC@@7C>CBCCBCCC;57)8(@B@B>ABBCBC7BCC=> NM:i:0

8_80_1315_464 81 scaffold00005 155760 255 54M = 154948 0 AGTACCTCCCTGGTACACAGCTTGGTAAAAATGTGATTGCTGAGCCAGACCTTC B?@?
BA=>@>>7;ABA?BB@BAA;@BBBBBBAABABBBBCABAB?BABA?BBBAB NM:i:0

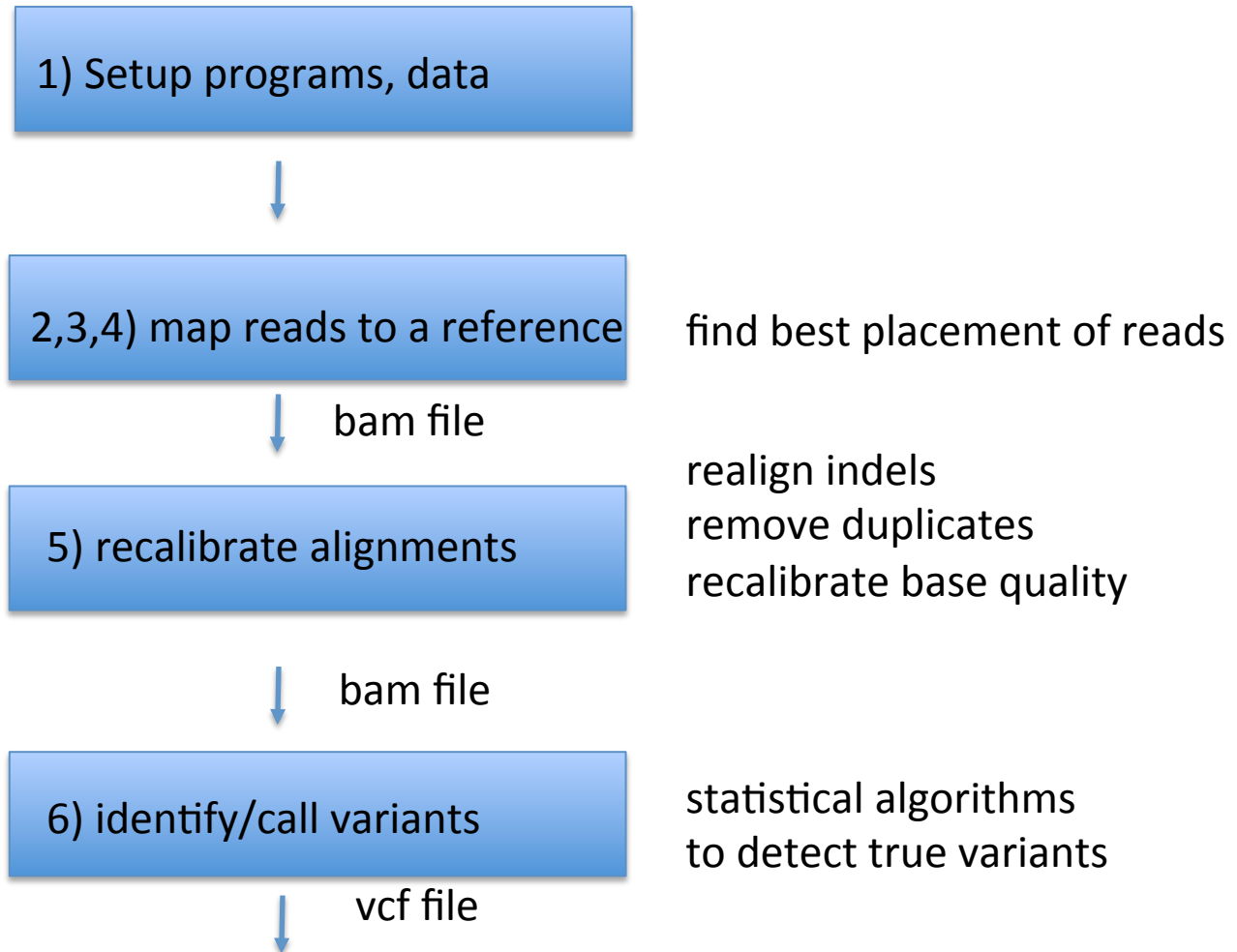
8_17_1222_1577 73 scaffold00005 155783 255 40M1116N10M * 0 0 GGTA AAAATGTGATTGCTGAGCCAGACCTTCATCATGCAGTGAGAGACGC BB@BA??
>CCBA2AAABBBBBBBB8A3@BABA;@A:>B=,;@B=A:BA AAA NM:i:0 XS:A:+ NS:i:0

8_43_1211_347 73 scaffold00005 155800 255 23M1116N27M * 0 0 TGAGCCAGACCTTCATCATGCAGTGAGAGACGCAAACATGCTGGTATTTG
#>8<=<@6/;@9';@7A@BAAA@BABBBABBB@=<A@BBBBBBBCCBB NM:i:2 XS:A:+ NS:i:0

8_32_1091_284 161 scaffold00005 156946 255 54M = 157071 0 CGCAAACATGCTGGTAGCTGTGACACCACATCAACAGCTTGACTATGTTTGTA
BBBBB@AABACBCA8BBBBBABB@BBBBBBA@BBBBBBBBA@:B@AA@=@@ NM:i:0
```



Steps in resequencing



Steps in resequencing

1) Setup programs, data



2,3,4) map reads to a reference

find best placement of reads



bam file

5) recalibrate alignments

realign indels
remove duplicates
recalibrate base quality



bam file

6) identify/call variants

statistical algorithms
to detect true variants



vcf file



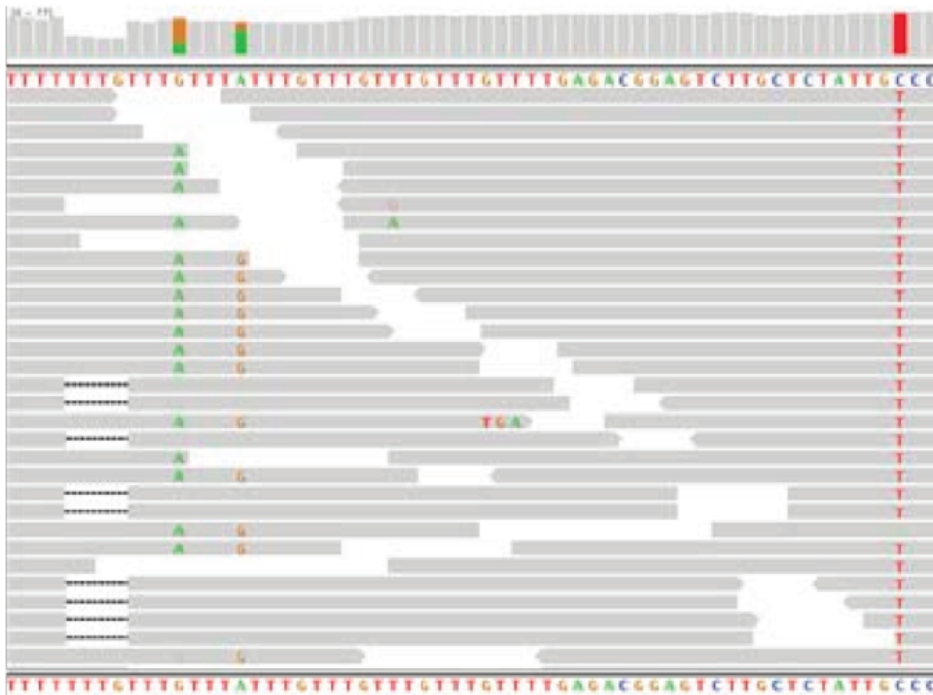
software

- Some very useful programs for manipulation of short reads and alignments:
- SAM Tools (<http://samtools.sourceforge.net/>)
 - provides various utilities for manipulating alignments in the SAM and BAM format, including sorting, merging, indexing and generating alignments in a per-position format.
- Picard (<http://picard.sourceforge.net/>)
 - comprises Java-based command-line utilities that manipulate SAM and BAM files
- Genome Analysis Toolkit (<http://www.broadinstitute.org/gatk/>)
 - GATK offers a wide variety of tools, with a primary focus on variant discovery and genotyping as well as strong emphasis on data quality assurance.
- Integrative Genomics viewer (<http://www.broadinstitute.org/igv/>)
 - IGV is very useful for visualizing mapped reads

step 2: recalibration

- 2.1 realign indels
- 2.2 remove duplicates
- 2.3 recalibrate base quality

2.1: local realignment



HiSeq data, raw BWA alignments



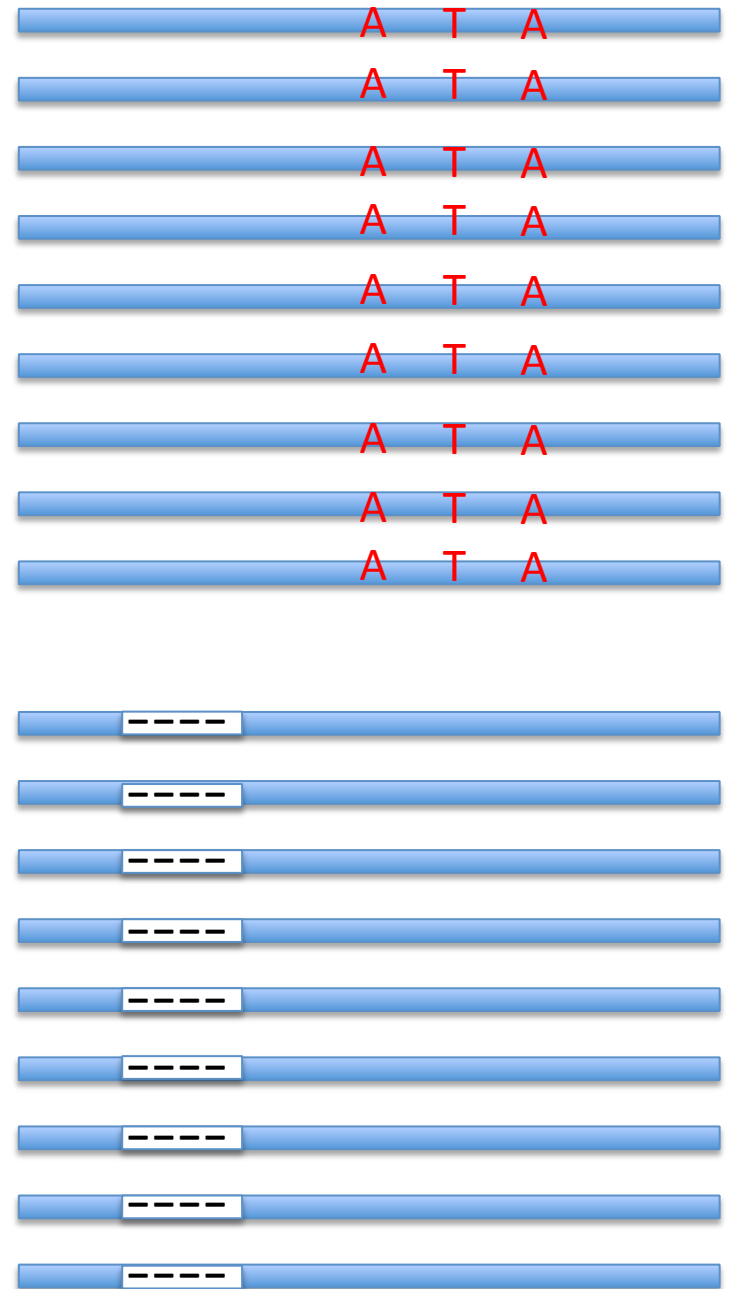
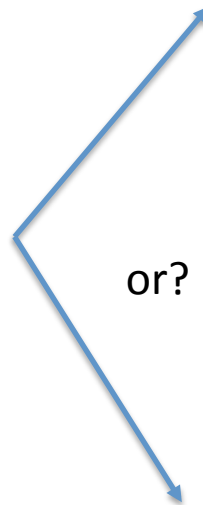
HiSeq data, after MSA

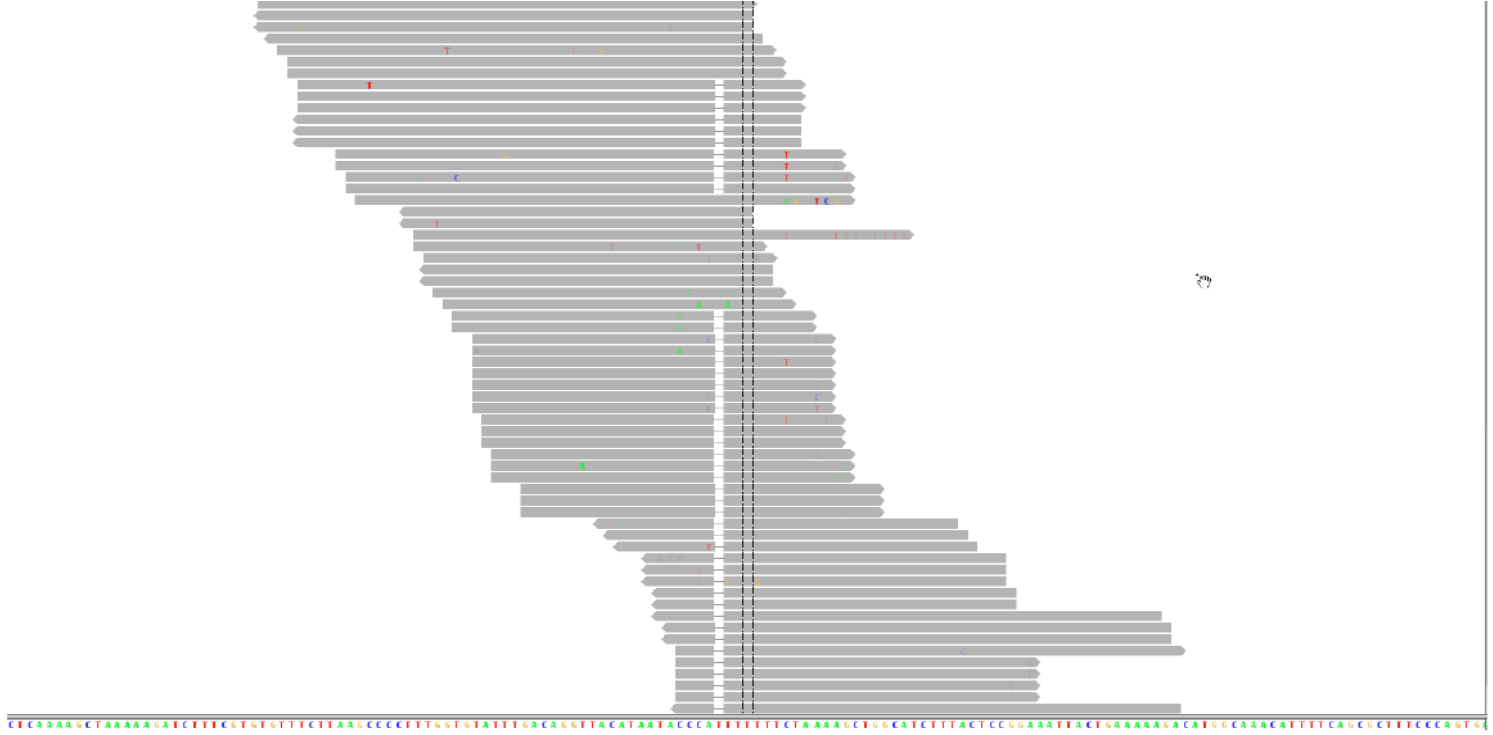
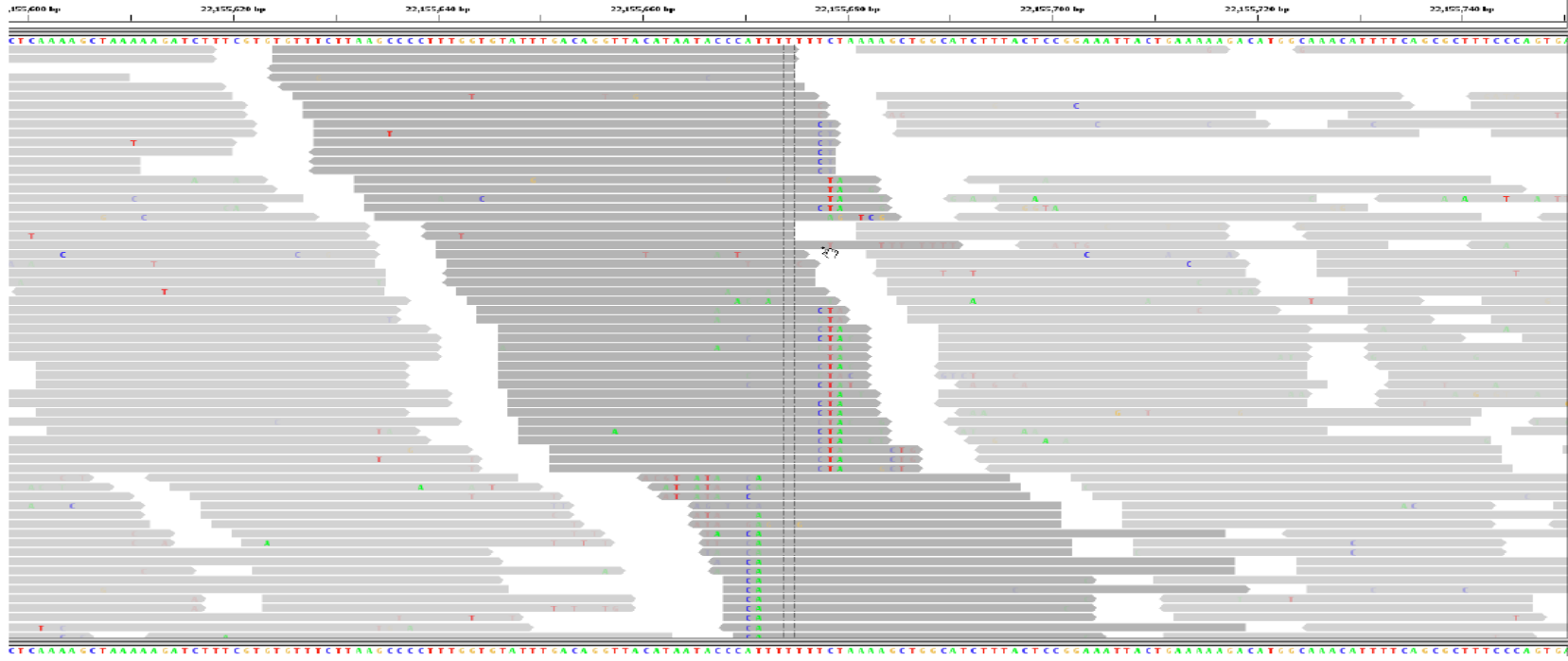
can be performed using GATK commands:
RealignerTargetCreator followed by
IndelRealigner

local realignment

- mapping is done one read at a time
- single variants may be split into multiple variants around indels
- solution: realign these regions taking all reads into account

local realignment





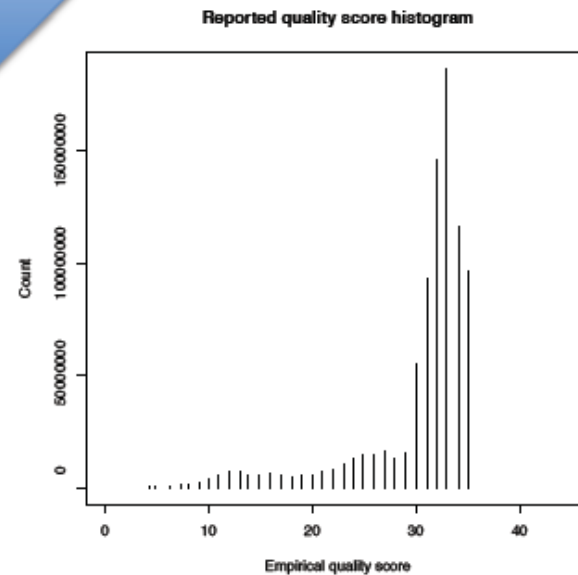
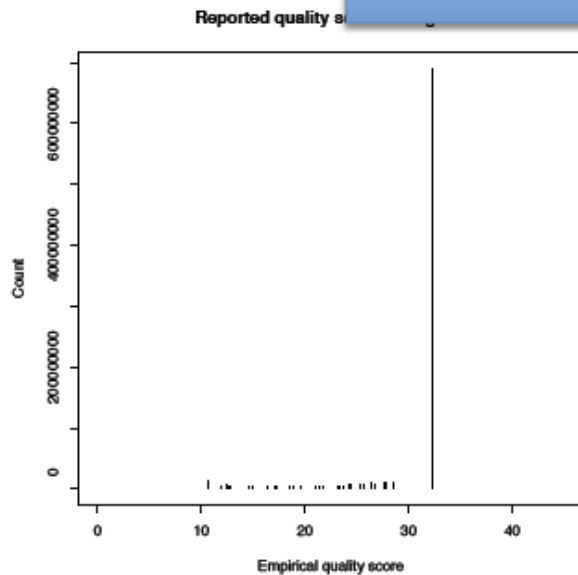
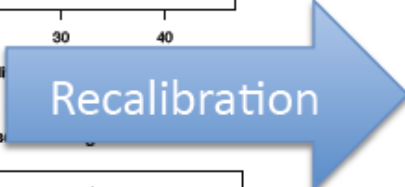
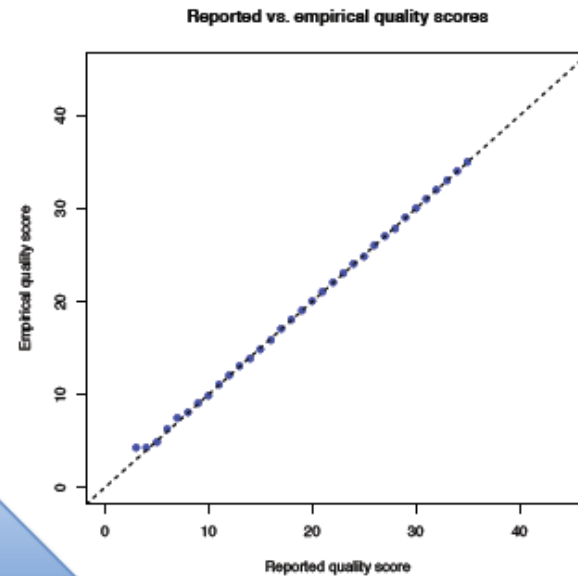
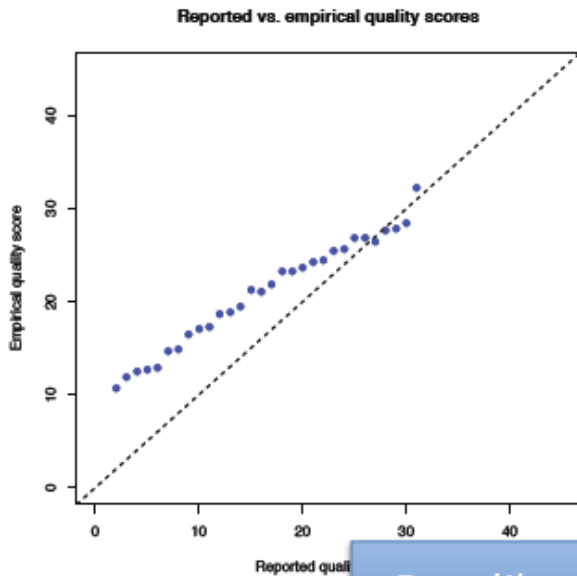
2.2 PCR duplicates

- When two or more reads originate from same molecule (artificial duplicates)
 - not independent observations
 - skew allele frequency and read depth
 - errors double counted
- PCR duplicates occur
 - during library prep, or
 - optical duplicates (one cluster read as two)
- mark or remove

Identify PCR duplicates

- Single or paired reads that map to identical positions
- Picard `MarkDuplicates`
- Optical duplicates occur close to each other on sequencer
- If low coverage, then duplicates are likely artifacts
- If high coverage, then more duplicates are real

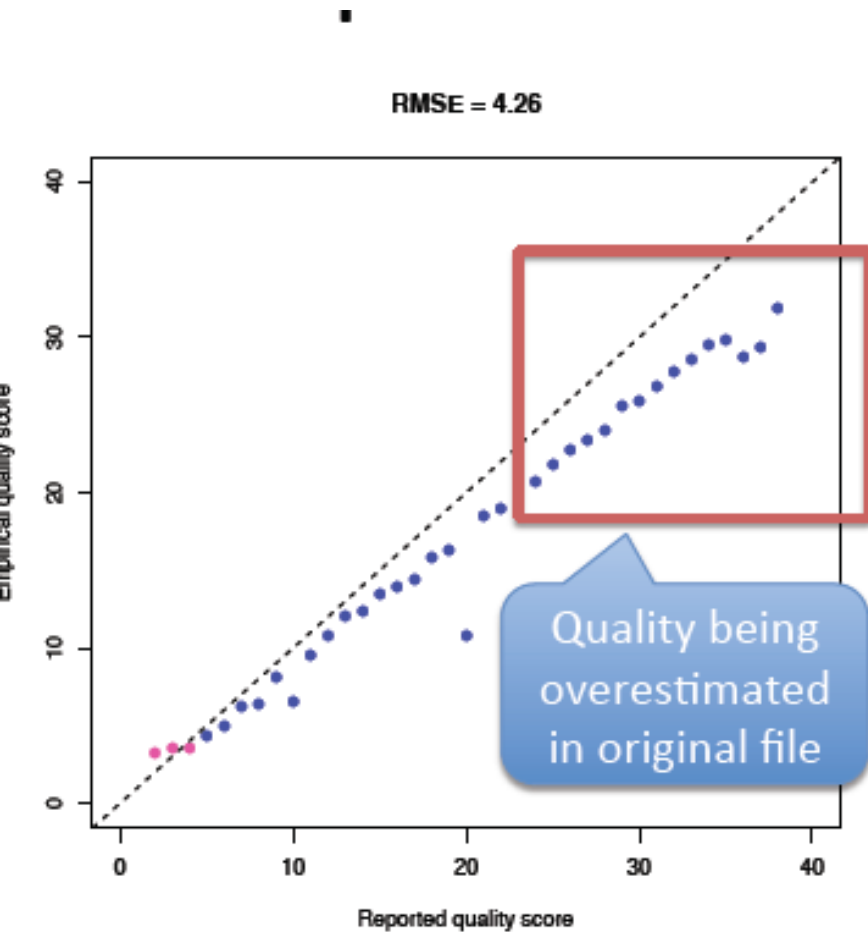
2.3 base quality recalibration



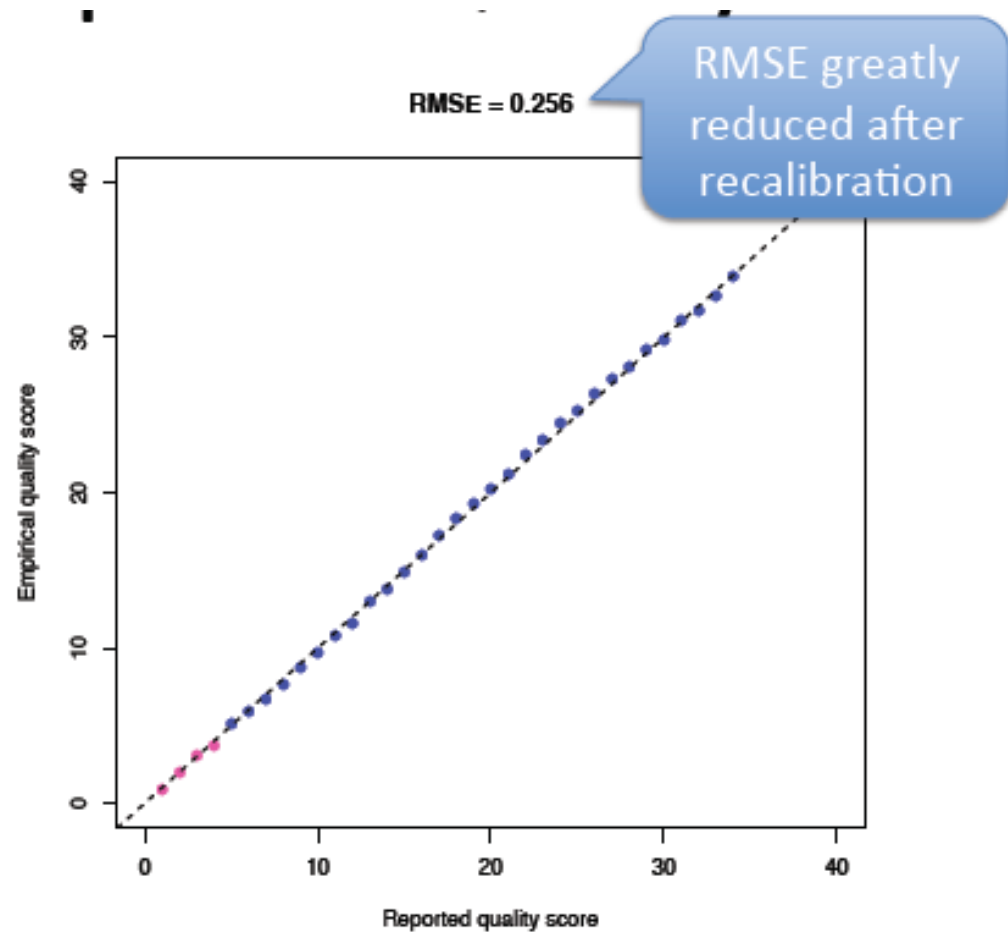
Recalibration Method

- Bin each base by
 - read group
 - called quality
 - position in read
 - local dinucleotide context
- score observed quality per bin
 - # of mismatches +1 / # of observed bases
- scale compared to reported quality

Reported vs empirical quality scores



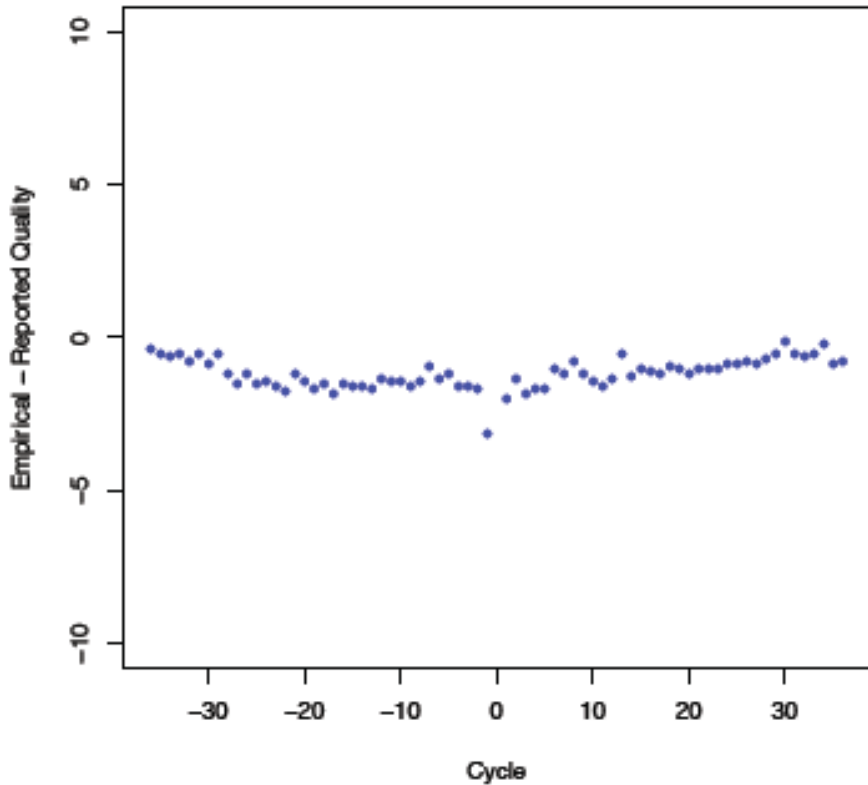
Before Recalibration



After Recalibration

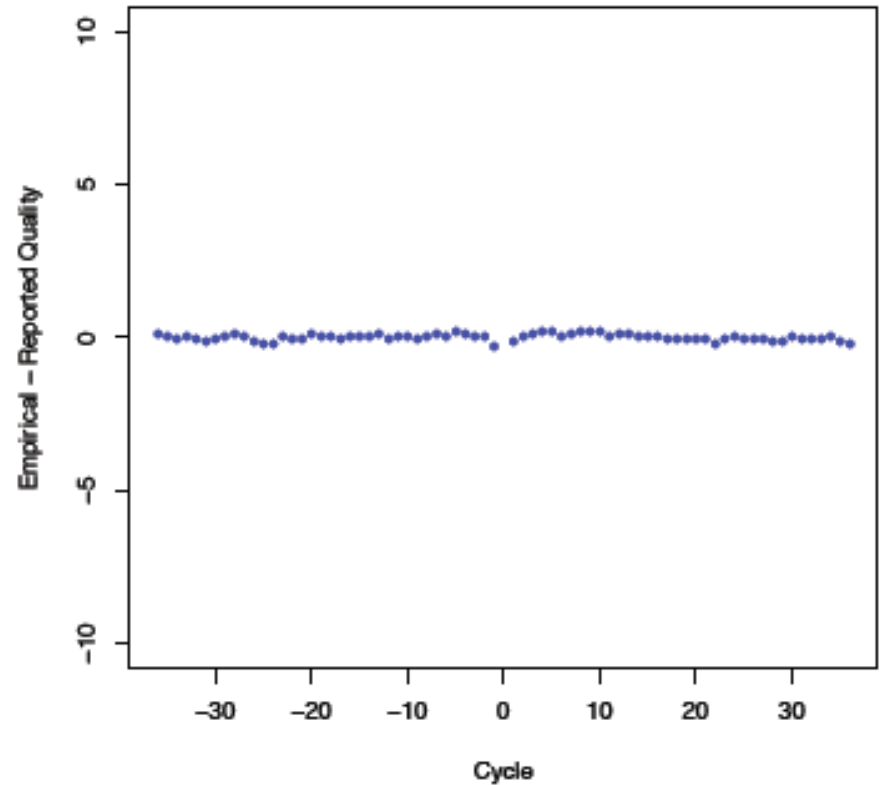
Residual error by machine cycle

RMSE = 1.275



Before Recalibration

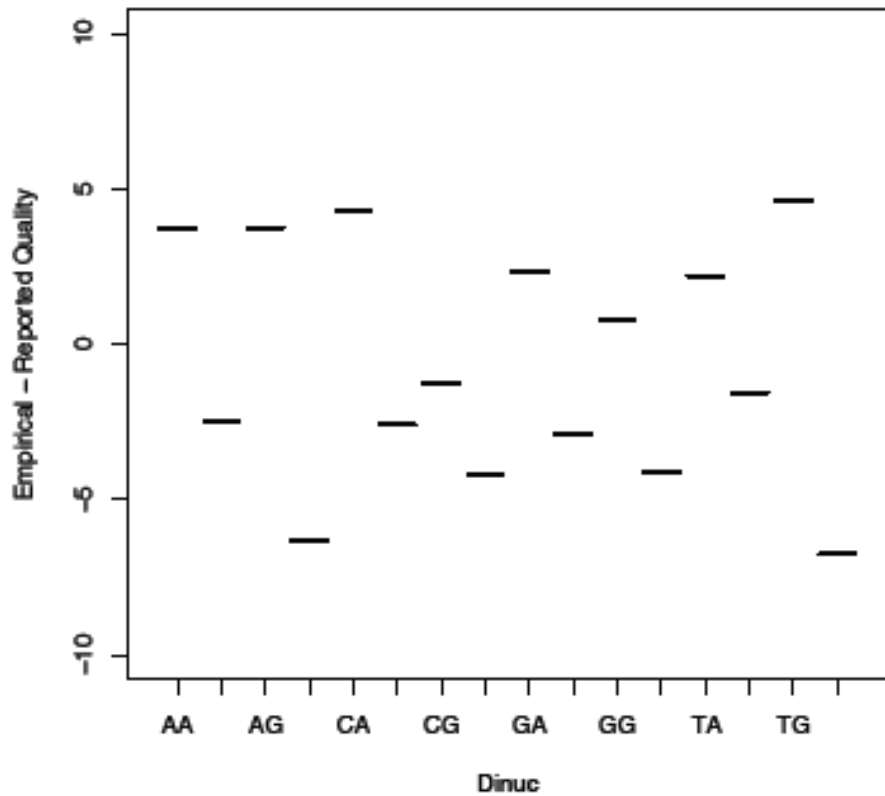
RMSE = 0.105



After Recalibration

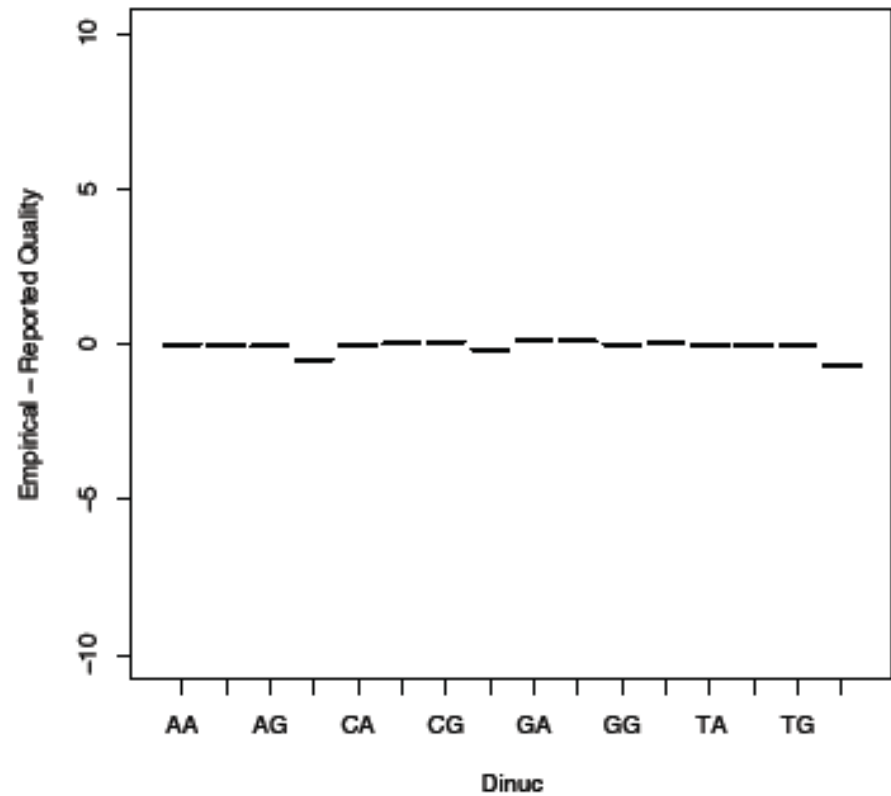
Residual error by dinucleotide

RMSE = 4.188



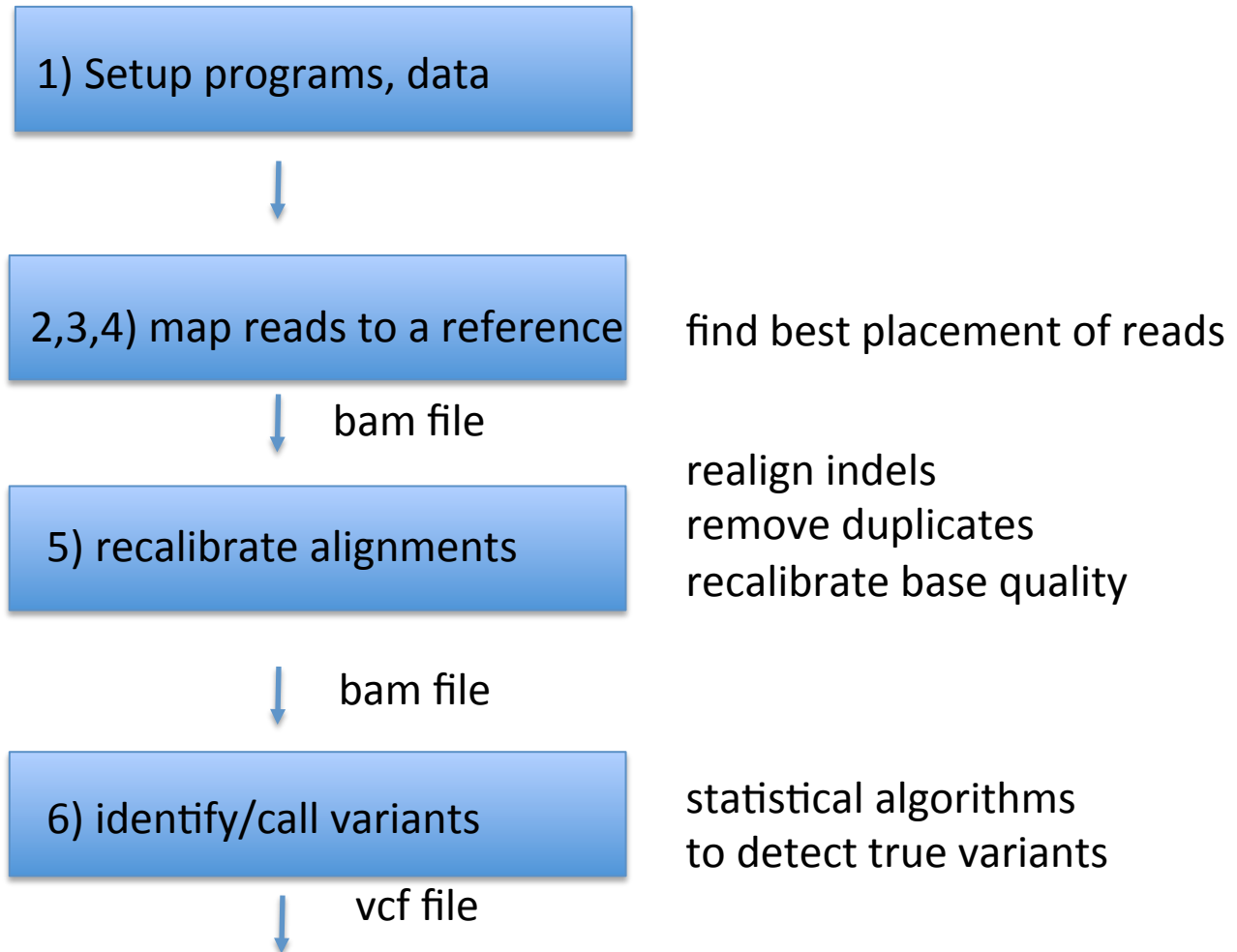
Before Recalibration

RMSE = 0.281



After Recalibration

Steps in resequencing



Steps in resequencing

1) Setup programs, data



2,3,4) map reads to a reference

find best placement of reads



bam file

5) recalibrate alignments

realign indels
remove duplicates
recalibrate base quality



bam file

6) identify/call variants

statistical algorithms
to detect true variants

vcf file



Single Nucleotide Variant calling

- **Genome Analysis Toolkit** (<http://www.broadinstitute.org/gatk/>)
 - Integrated pipeline for SNP discovery (java)
- **FreeBayes** (<http://bioinformatics.bc.edu/marthlab/FreeBayes>)
 - Bayesian SNP calling (C++)

Both programs perform Bayesian population based SNP calling

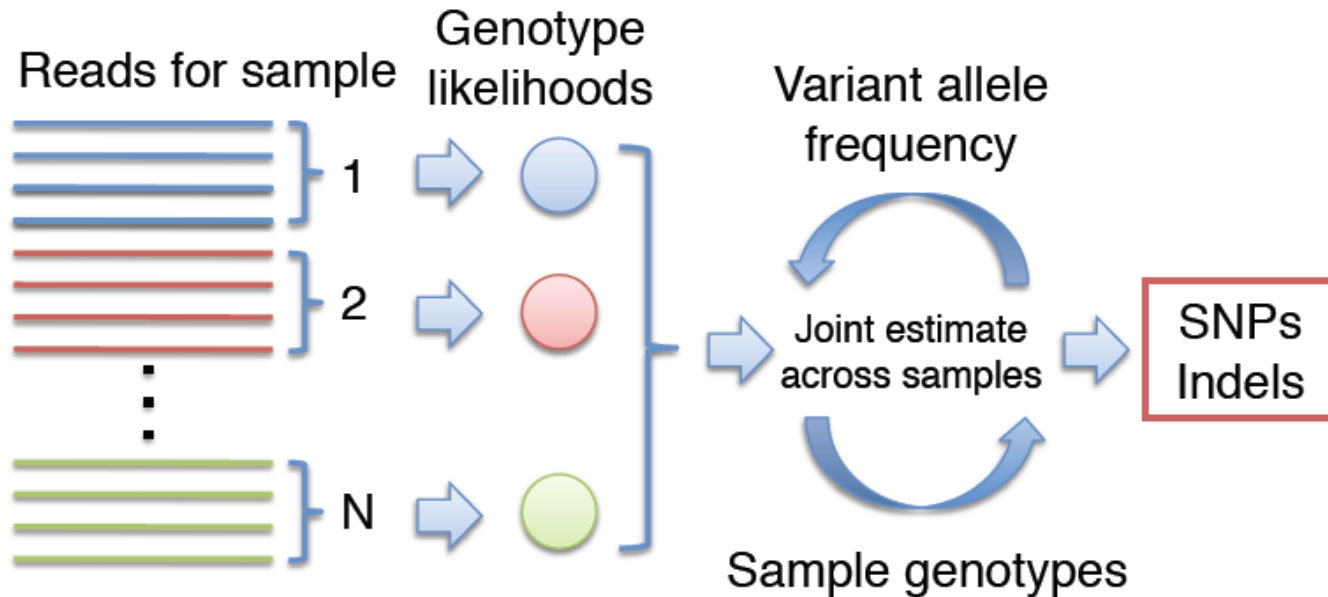
simple pileup methods

acacagatagacatagacatagacagatgag
acacagatagacatagacatagacagatgag
acaca**c**atagacatagacatagacagatgag
acacagatagacatagacatagacagatgag
acacagatagacata**t**acatagacagatgag
acacagatagacata**t**acatagacagatgag
acacagatagacata**t**acatagacag**t**tgag
acacagatagacatagacatagacagatgag
acacagatagacata**t**acatagacagatgag
acacagatagacatagacatagacagatgag

Bayesian population based calling

- Assign calls to specific genotypes
- Probability of genotype given data
- Variants at high frequency are more likely real
- Weak single sample calls are combined to discover variants among samples with high confidence
- "haplotype aware" calling also possible
 - infers haplotypes
 - uses info to impute variants

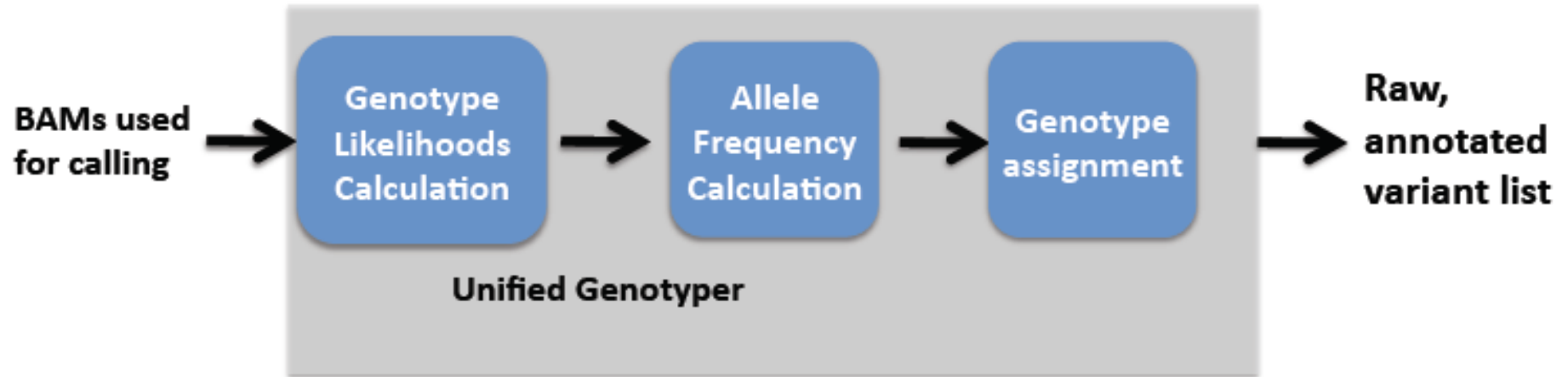
population-based calling



Simultaneous estimation of:

- Allele frequency (AF) spectrum: $\Pr\{AF = i \mid D\}$
- The prob. that a variant exists: $\Pr\{AF > 0 \mid D\}$
- Assignment of genotypes to each sample

GATK unified genotyper - multi sample aware calling



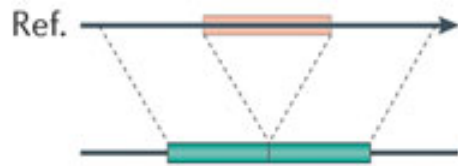
- Computing, for each sample, for each genotype, likelihoods of data given genotypes.
- Computing, the allele frequency distribution to determine most likely allele count, and emit a variant call if determined.
- If a variant is emitted, assign a genotype to each sample.

VCF format

```
##fileformat=VCFv4.0 ##fileDate=20090805
##source=myImputationProgramV3.1
##reference=1000GenomesPilot-NCBI36
##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=.,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 GTCT G,GTACT 50 PASS NS=3;DP=9;AA=G GT:GQ:DP 0/1:35:4 0/2:17:2 1/1:40:3
```

Discovery of structural variants

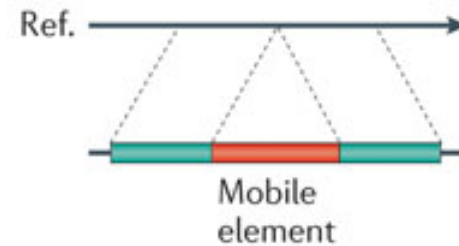
Deletion



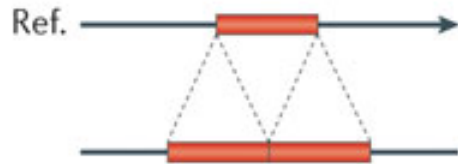
Novel sequence insertion



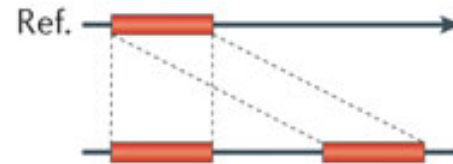
Mobile-element insertion



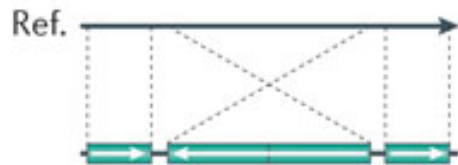
Tandem duplication



Interspersed duplication



Inversion



Translocation



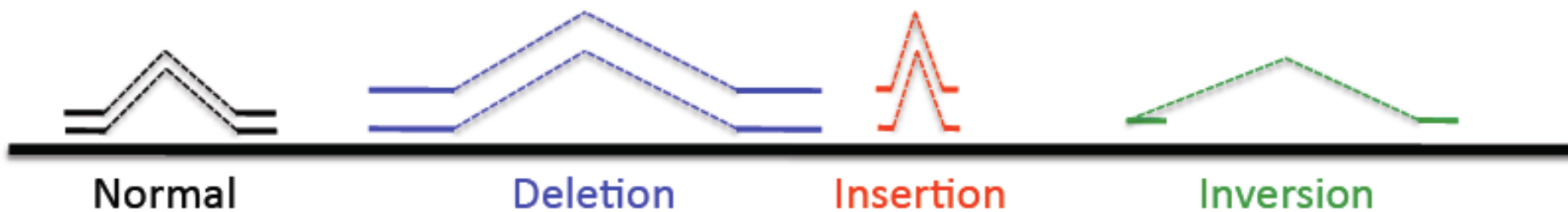
1) Read depth analysis

- Depth of coverage can be used to estimate copy number
- Samples may exhibit variation in depth indicative of polymorphic copy number variants
- How many copies of a duplication in the reference?
- How similar are the copies
- Difficult to distinguish homozygotes and heterozygotes.



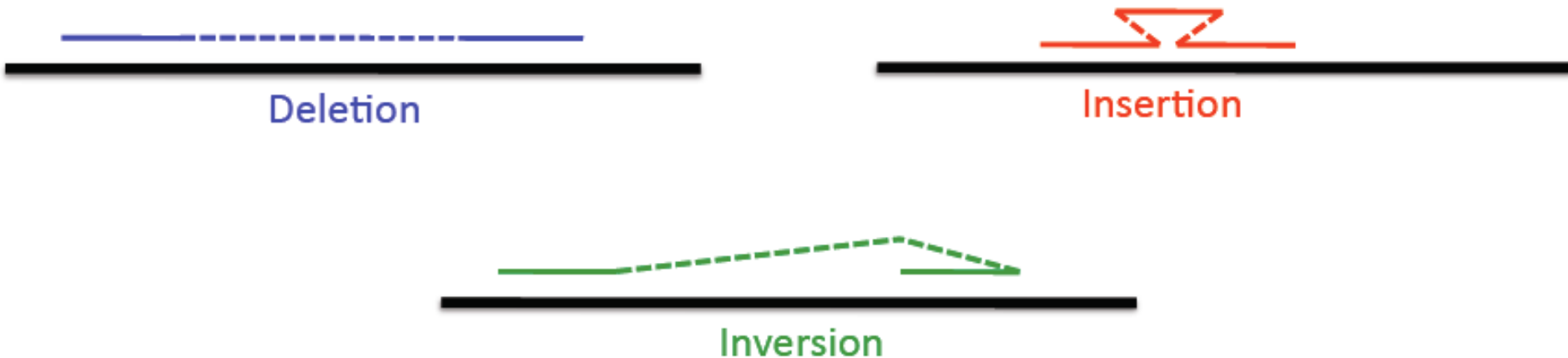
2) Paired end analysis

- Paired ends have a fixed length between them
- Genomic rearrangements cause them to vary
 - Deletion: reads will map too far apart
 - Insertion: reads will map too close
 - Inversion: reads in wrong orientation
- more reliable with long pairs



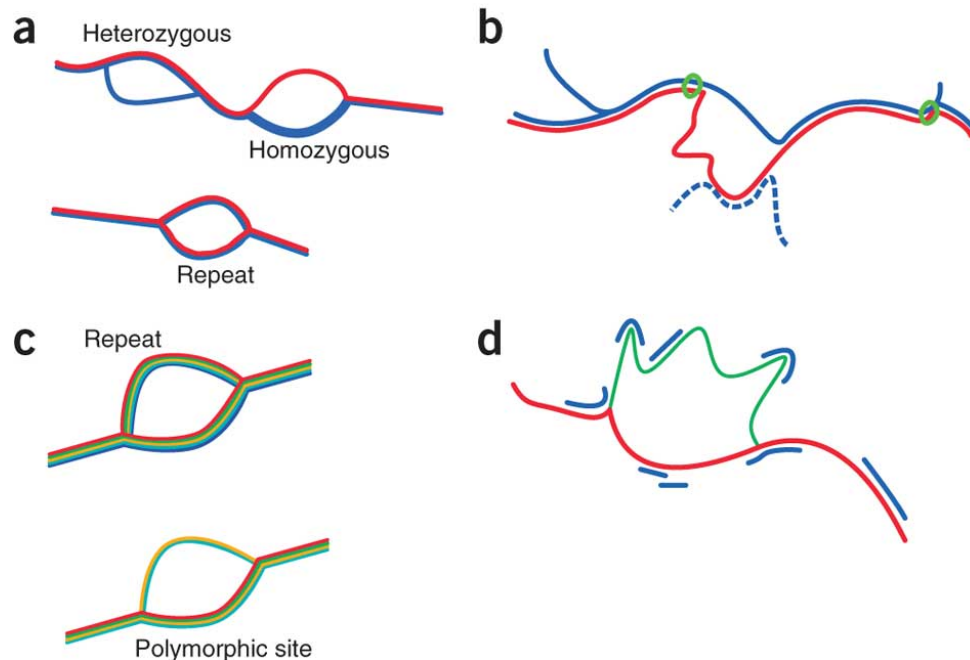
3) Split-read alignments

- Base-level breakpoint resolution
- Only works with long reads
 - short reads have many spurious splits
- Caveat: breakpoints may be duplicated
 - reads won't split if single alignment is good



4) *De novo* assembly to identify structural variants

- Assemble contigs
- Align to reference
- Look for insertions, deletions, rearrangements



Naming conventions

By using standardized names for your files it becomes easier to keep track of the different steps

- Initial file name according to some relevant information about the contents e.g. NA06984.ILLUMINA.low_coverage.17q
- For each step of the pipeline, create a new file with an extension that state what you have been doing, e.g.
 - NA06984.ILLUMINA.low_coverage.17q.merge.bam
 - NA06984.ILLUMINA.low_coverage.17q.merge.realign.bam
 - NA06984.ILLUMINA.low_coverage.17q.merge.realign.dedup.bam
 - NA06984.ILLUMINA.low_coverage.17q.merge.realign.dedup.recal.bam

Downstream analysis

A number of convenient tools exist for working with your bam / vcf / bed files

- BEDTools – enables genome arithmetics
- Vcftools – for manipulations of vcf-files
- Annotations to compare with can be extracted from e.g the UCSC browser, ensemble database, etc
- .. Perl / python / bash / awk

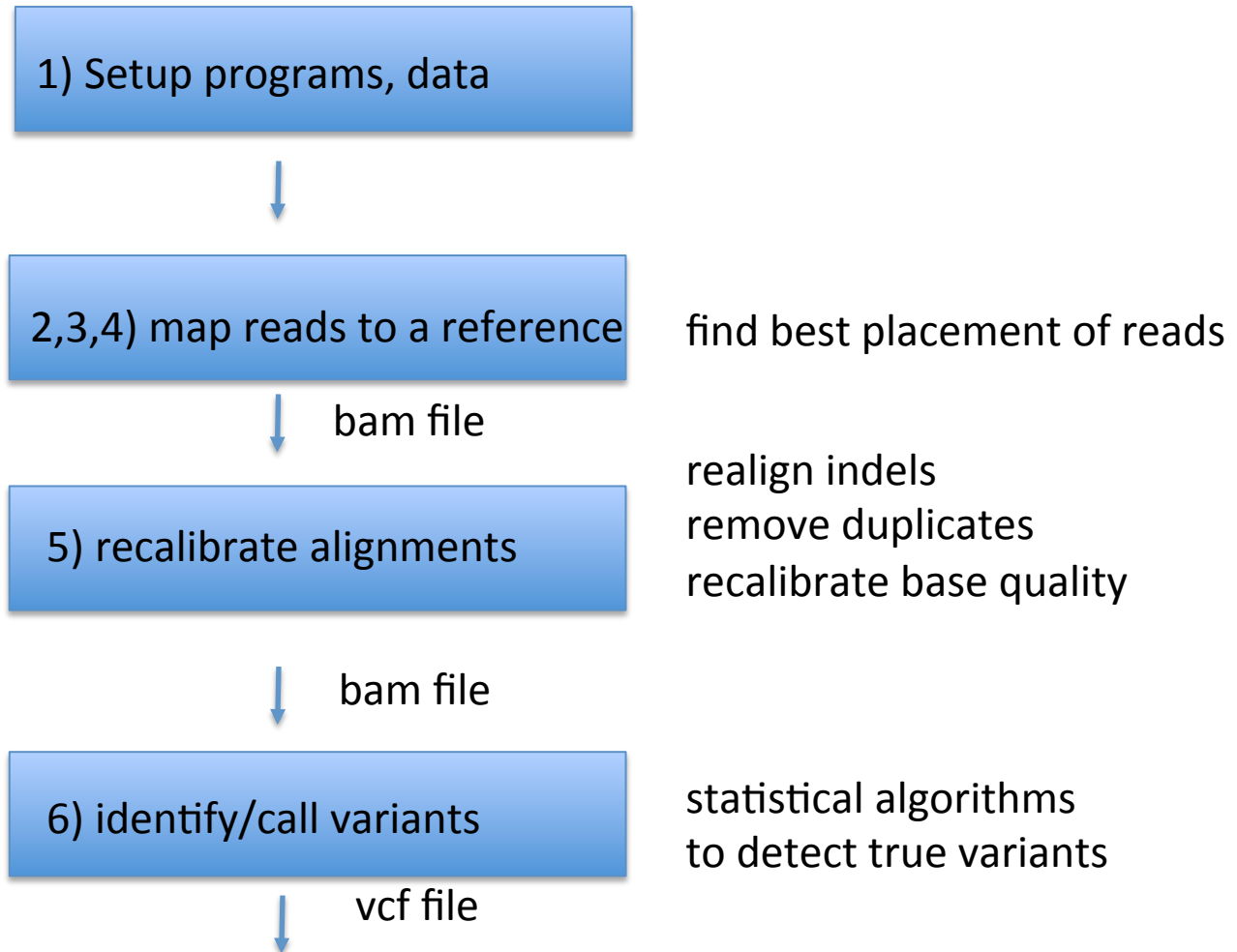
Annotation of variants

- What is the expected effect of a particular variant?
- Which variants in a set are most likely deleterious?
- Most used programs annovar and SnpEff

Overview of exercise

1. Access to data and programs
2. Mapping (BWA)
3. Merging alignments (BWA)
4. Creating BAM files (Picard)
5. Processing files (GATK)
6. Variant calling and filtering (GATK)
7. Viewing data (IGV)

Steps in resequencing



1) Access to data and programs

- Data comes from 1000 genomes pilot project
 - 81 low coverage (2-4 x) Illumina WGS samples
 - 63 Illumina exomes
 - 15 low coverage 454
- ~ 1 Mb from chromosome 17
- Tasks: align one sample to reference, process, recalibration, variant calling and filtering

1) Access to data and programs

- Programs:
- BWA and samtools modules can be loaded:

```
module load bioinfo-tools
```

```
module load bwa
```

```
module load samtools
```

- picard and GATK are set of java programs:

```
/bubo/sw/apps/bioinfo/GATK/1.5.21/
```

```
/bubo/sw/apps/bioinfo/picard/1.69/kalkyl/
```

2) Mapping

- Indexing sequences:

- perform BW transform on reference

```
bwa index -a bwtsv <reference sequence>
```

- make a fasta index for reference

```
samtools faidx <reference sequence>
```

- align each paired end separately

```
bwa aln <reference> <seq 1> >align1.sai
```

```
bwa aln <reference> <seq 2> >align2.sai
```

3) Merging alignments

- command to combine alignments from paired ends into a SAM file

```
bwa sampe <ref> <sai1> <sai2> <fq1> <fq2> >align.sam
```

<ref> = reference sequence
<sai1> = alignment of seq 1 of pair
<sai2> = alignment of seq 2 of pair
<fq1> = fastq reads seq 1 of pair
<fq2> = fastq reads seq 2 of pair

4) Creating and editing BAM files

- create BAM file (samtools)

```
samtools import <reference.fai> <sam file> <bam file>
```

- sort BAM file (samtools)

```
samtools sort <input bam> <output bam rootname>
```

- index BAM file (samtools)

```
samtools index <input bam>
```

- add read groups to BAM (picard)

```
java -Xmx2g -jar /path/AddOrReplaceReadGroups.jar
```

```
INPUT=<sam file>
```

```
OUTPUT=<bam file>
```

```
... more options
```

- index new BAM file (picard)

```
java -Xmx2g -jar /path/BuildBamIndex.jar
```

```
INPUT=<bam file>
```

```
... more options
```

5) Processing files

- mark problematic indels (GATK)

```
java -Xmx2g -jar /path/GenomeAnalysisTK.jar  
-I <bam file>  
-R <ref file>  
-T RealignerTargetCreator  
-o <intervals file>
```

- realign around indels (GATK)

```
java -Xmx2g -jar /path/GenomeAnalysisTK.jar  
-I <bam file>  
-R <ref file>  
-T IndelRealigner  
-o <realigned bam>  
-targetIntervals <intervals file>
```

5) Processing files

- mark duplicates (picard)

```
java -Xmx2g -jar /path/MarkDuplicates.jar
```

```
INPUT=<input bam>
```

```
OUTPUT=<marked bam>
```

```
METRICS_FILE=<metrics file>
```

- quality recalibration - compute covariation (GATK)

```
java -Xmx2g -jar /path/GenomeAnalysisTK.jar
```

```
-T CountCovariates
```

```
-I <input bam>
```

```
-R <ref file>
```

```
-knownSites <vcf file>
```

```
-cov ReadGroupCovariate
```

```
-cov CycleCovariate
```

```
-cov DinucCovariate
```

```
-cov QualityScoreCovariate
```

```
-recalFile <calibration csv>
```

5) Processing files

NEXT:

repeat steps 2-5 for another sample!

5) Processing files

- merge BAM from multiple samples (picard)

```
java -Xmx2g -jar /path/MergeSamFiles.jar
```

```
INPUT=<input bam 1> INPUT=<input bam 2> .. INPUT=<input bam N>
```

```
OUTPUT=<output bam>
```

- unified genotyper (GATK)

```
java -Xmx2g -jar /path/GenomeAnalysisTK.jar
```

```
-T UnifiedGenotyper
```

```
-R <ref file>
```

```
-I <merged bam>
```

```
-o <filename.vcf>
```

```
-glm BOTH
```


6) Variant calling and filtering

- variant filtering

```
java -Xmx2g -jar /path/GenomeAnalysisTK.jar
-T VariantFiltration
-R <reference>
-V <input vcf>
-o <output vcf>
--filterExpression "QD<2.0" --filterName QDfilter
--filterExpression "MQ<40.0" --filterName MQfilter
--filterExpression "FS>60.0" --filterName FSfilter
--filterExpression "HaplotypeScore>13.0" --filterName HSfilter
--filterExpression "MQRankSum<-12.5" --filterName MQRSfilter
--filterExpression "ReadPosRankSum<-8.0" --filterName RPRSfilter
```

7) Viewing data with IGV

Integrative Genomics Viewer



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

pipeline (1)

2. Mapping

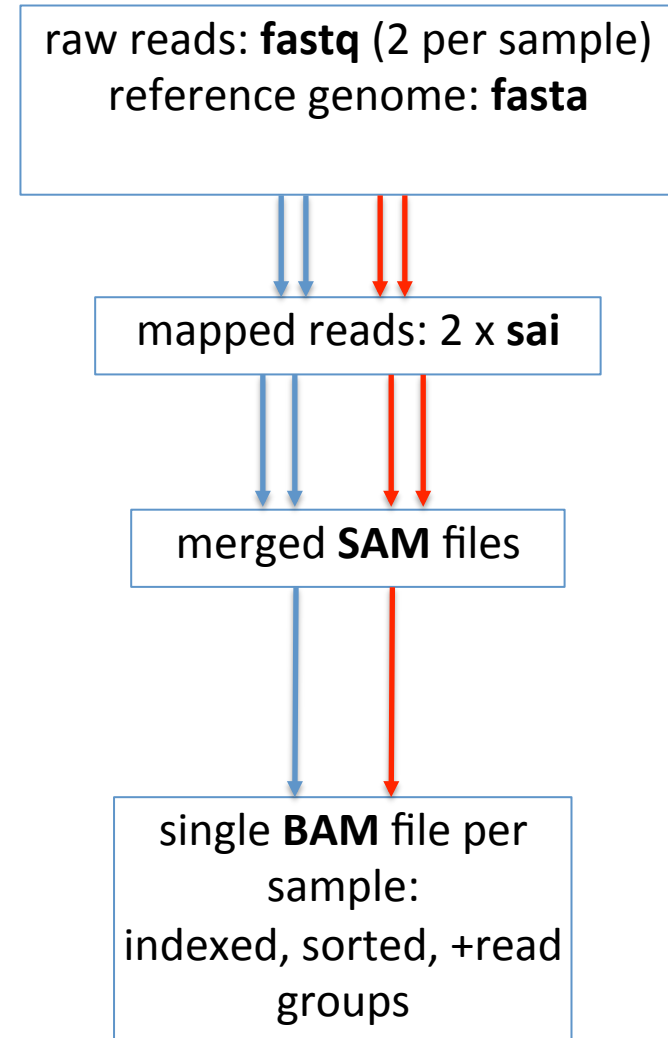
- `bwa index`
- `samtools faidx`
- `bwa aln`

3. Merging alignments

- `bwa sampe`

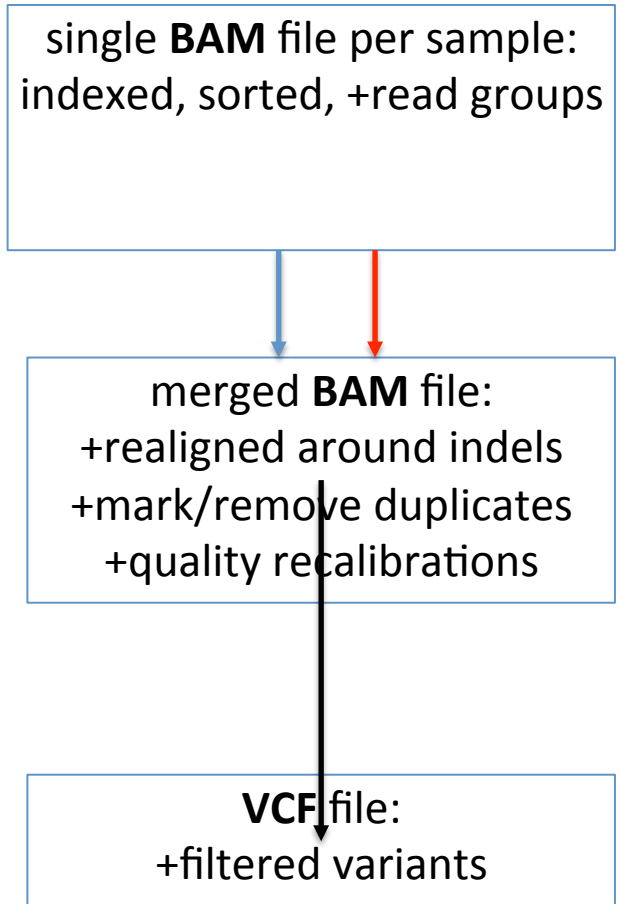
4. Creating BAM files

- `samtools import`
- `samtools index`
- `samtools sort`
- `picard AddOrReplaceReadGroups`
- `picard BuildBamIndex`



pipeline (2)

5. Processing files (GATK)
 - GATK RealignerTargetCreator
 - GATK IndelRealigner
 - picard MarkDuplicates
 - GATK CountCovariates
 - picard MergeSamFiles
6. Variant calling and filtering (GATK)
 - GATK UnifiedGenotyper
 - GATK VariantFiltration
7. Viewing data (IGV)



mapping

2.

raw reads: **fastq** (2 per sample)
reference genome: **fasta**

mapped reads: 2 x **sai** per
sample

3.

merged **SAM** files

4.

single **BAM** file per sample:
indexed, sorted, +read groups

processing

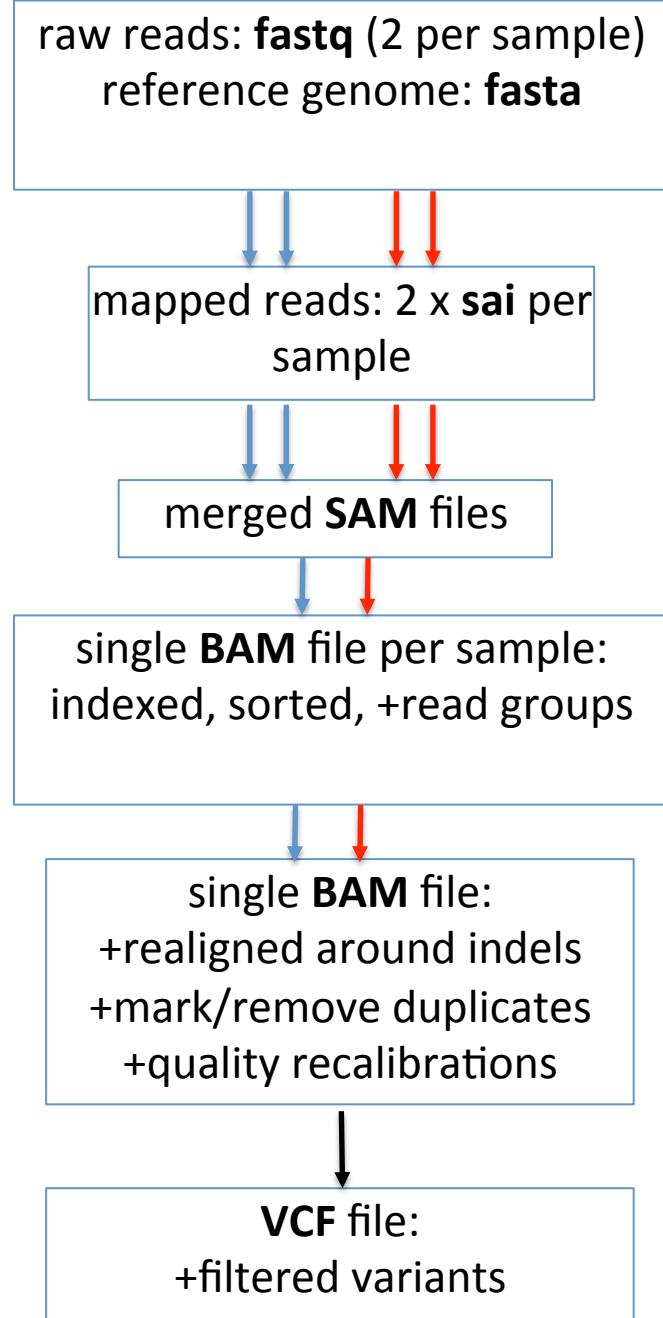
5.

single **BAM** file:
+realigned around indels
+mark/remove duplicates
+quality recalibrations

variant calling

6.

VCF file:
+filtered variants



Thanks!

- + this presentation was made by Matt Webster
- + special thanks to Mike Zody for some slides