

Genetic variation: what, why, how

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EGAG

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What is genetic variation?

- Differences in DNA content or structure among individuals
- Any two individuals have ~99.5% identical DNA.
- But the human genome is big - each haploid set of 23 chromosomes has 3 billion nucleotides.
- The details matter.

~98-99% identical DNA



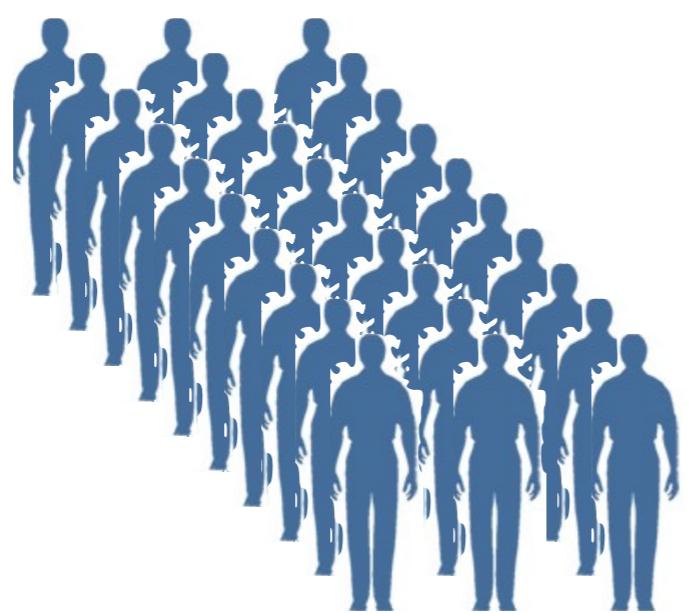
~99.5% identical DNA



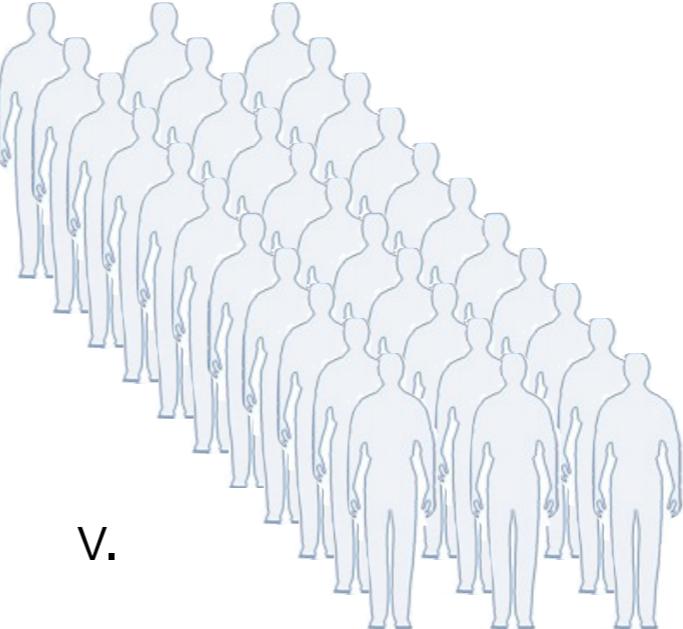
V3073025 [RF] © www.visualphotos.com

Why do we care?

- Understand the relationship between genotype and phenotype.



Cases
(have disease)

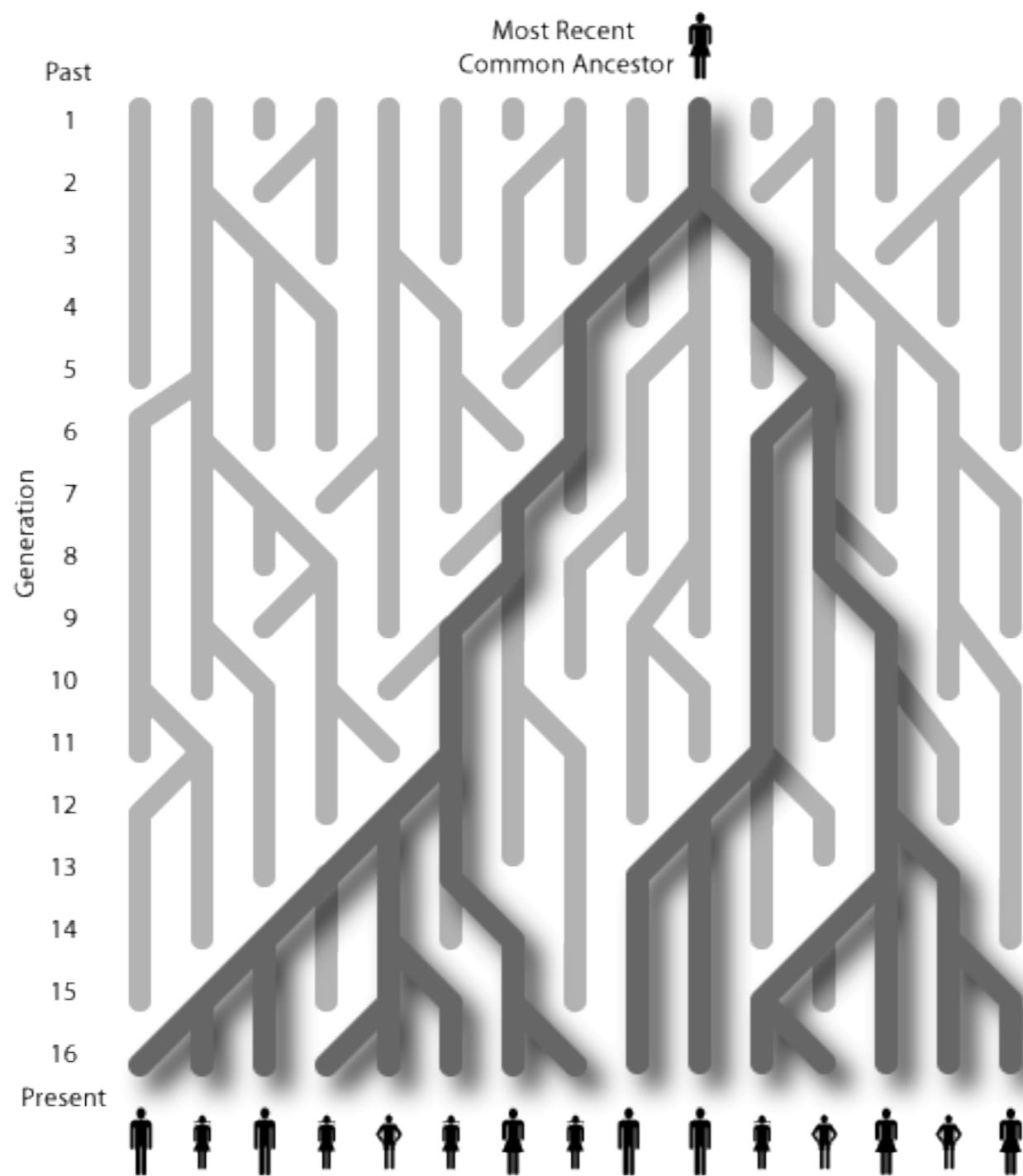


Controls
(no disease)

Complex diseases
(multiple genes contribute to risk)

Why do we care?

- Bread crumbs of evolution



Why do we care?

- How, when, where does our genome evolve?



Types of genetic variation

ctc**c**gag
ctc**t**gag

Single-nucleotide
polymorphisms
(SNPs)

ctc--ag
ctc**tg**ag

Insertion-deletion
polymorphisms
(INDELs)

ctcag
ctc  ag

Structural
variants
(SVs)

“spelling mistakes”

*“extra or missing
letters”*

*“extra, missing
or reordered
chapters”*

Properties of genetic variation

Single-nucleotide (SNPs)	ctc c gag	ctc t gag	ctc ca g
Size	1bp	1-100bp	100bp-1Mb+
Frequency	3 million / genome	300K / genome	3,000 / genome
Detection Difficulty	Easy	Medium	Hard

How different are we?

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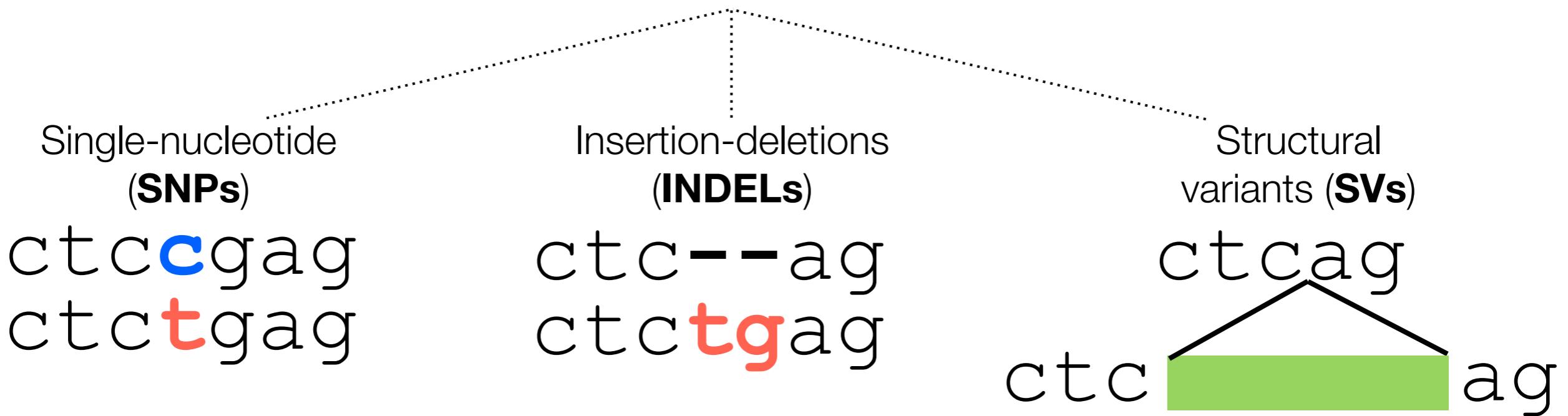
15,000,000 to 21,000,000 different base pairs.

How different are we?

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But, the genome is big.

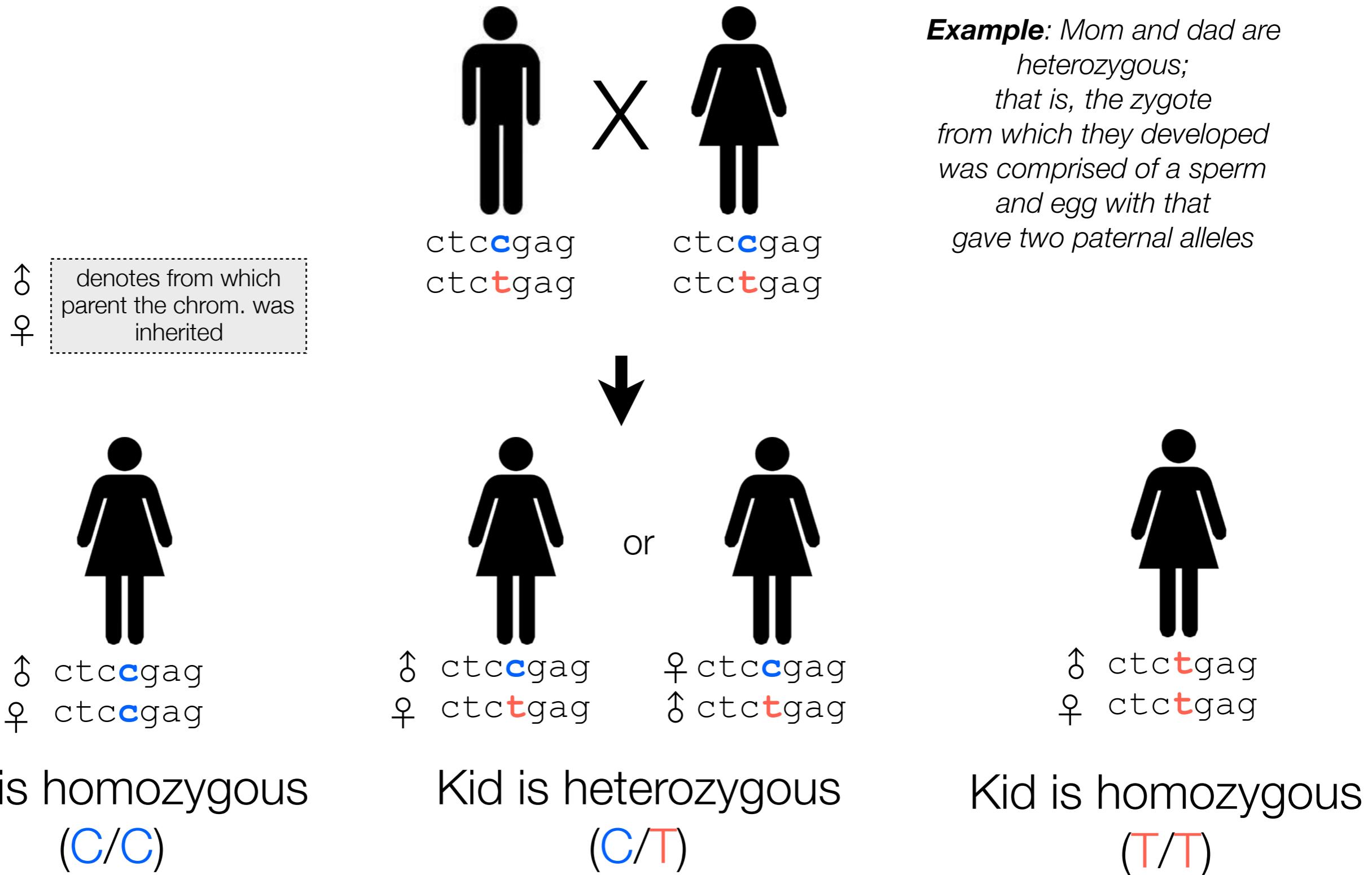
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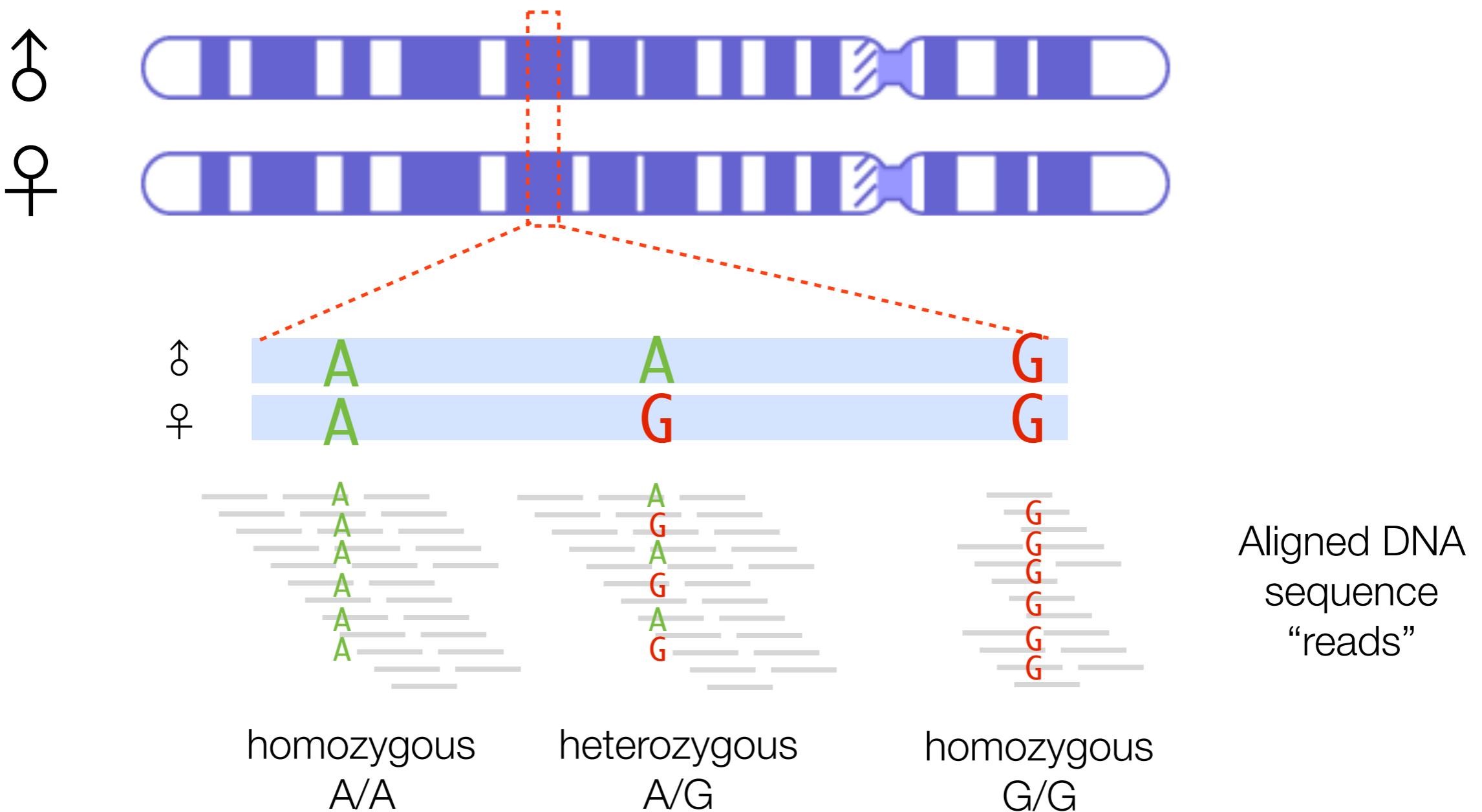
Detecting genetic variation

CGCAAATTGCCGGATTCCTTGTGTTCTGCATGTAGTTAACGAGATTGCCAGCACCGGGTATCATTACCACTTTCTTT
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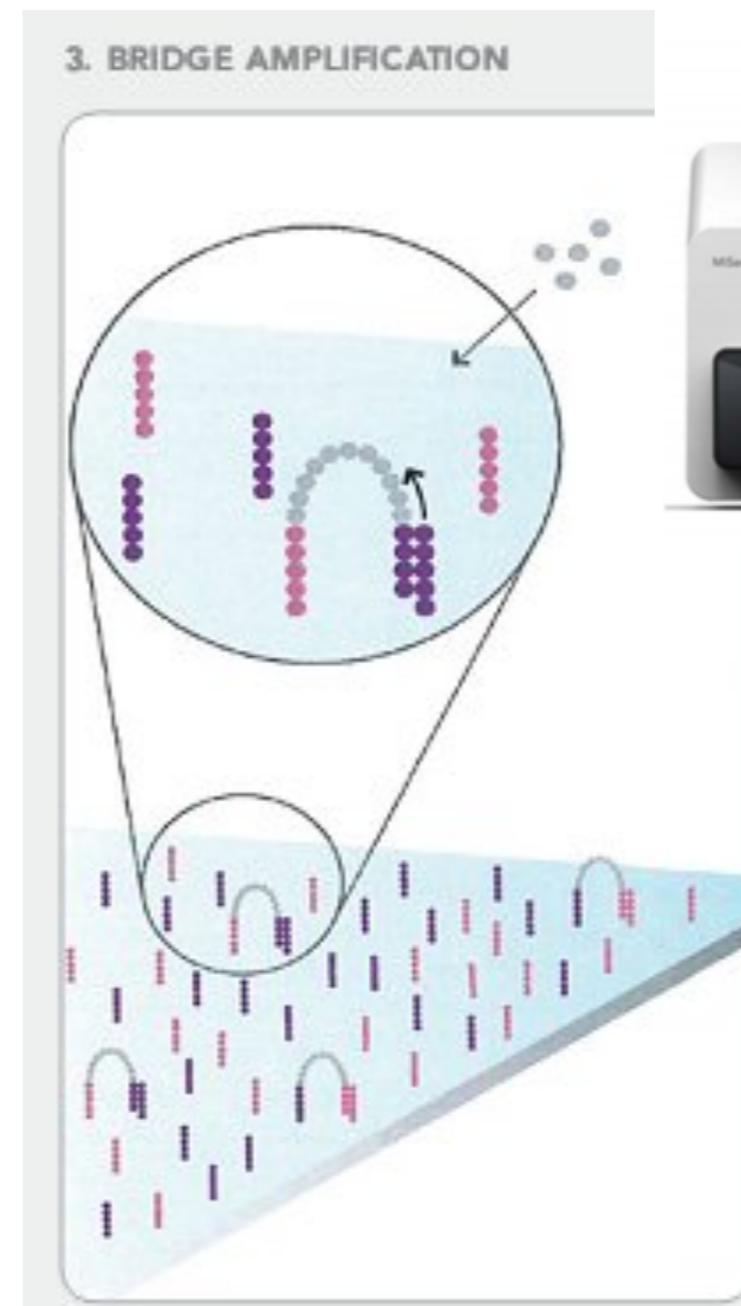
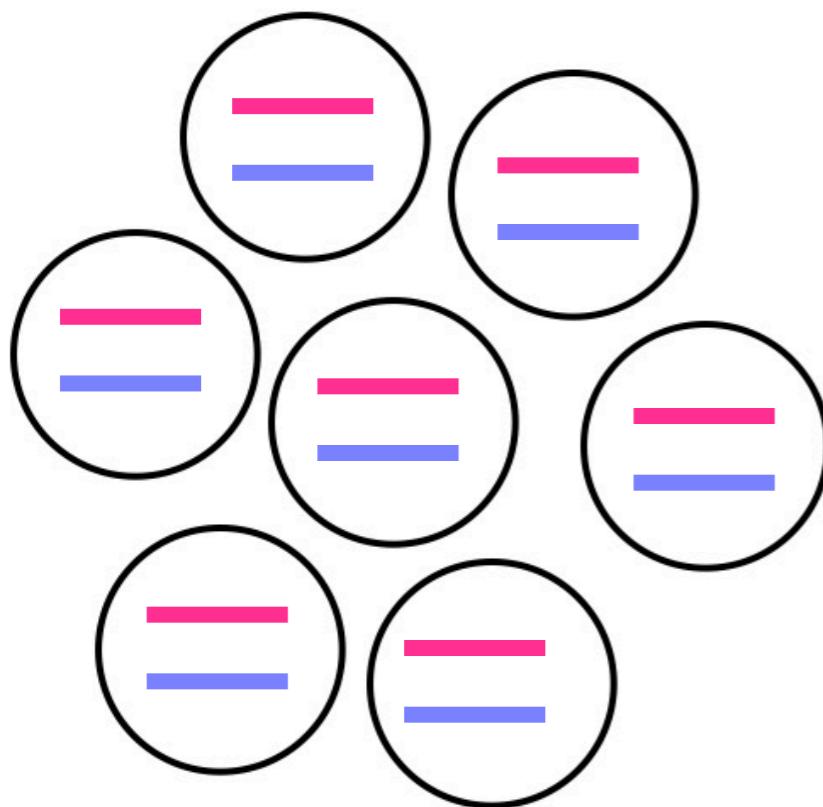
How existing (germline) variation is inherited



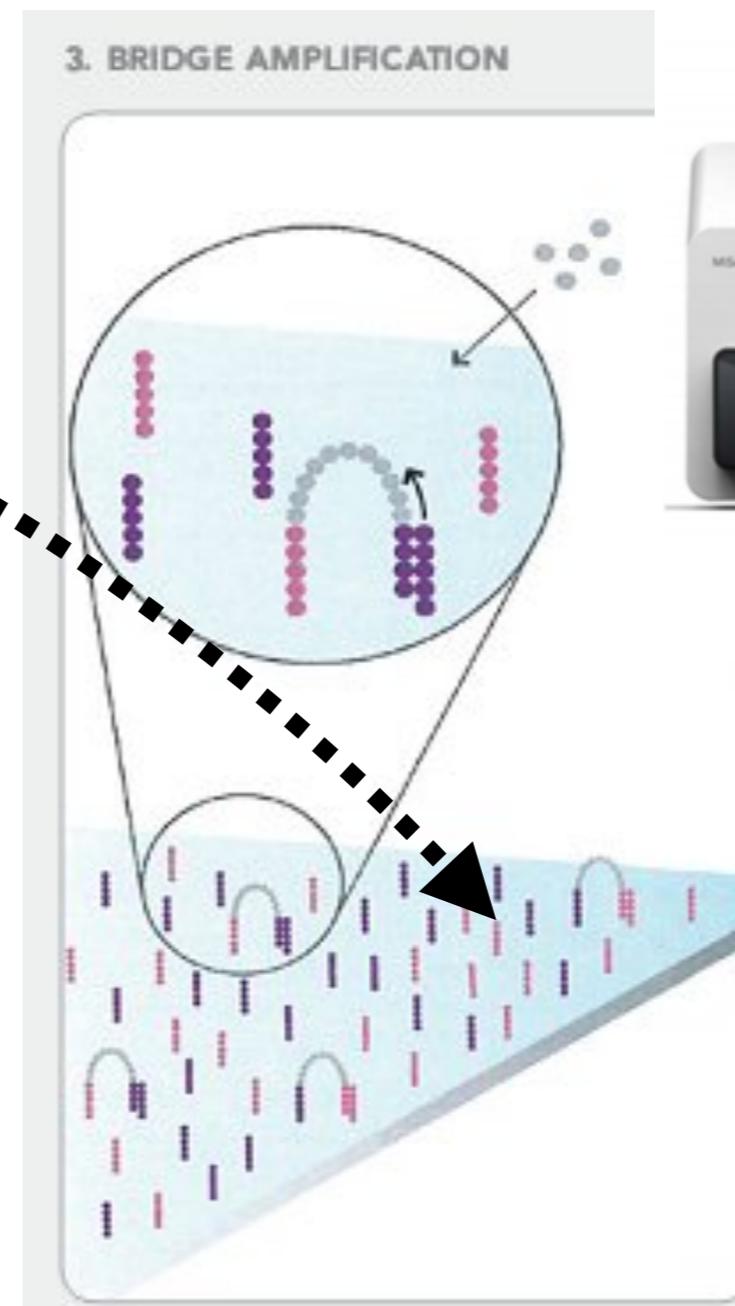
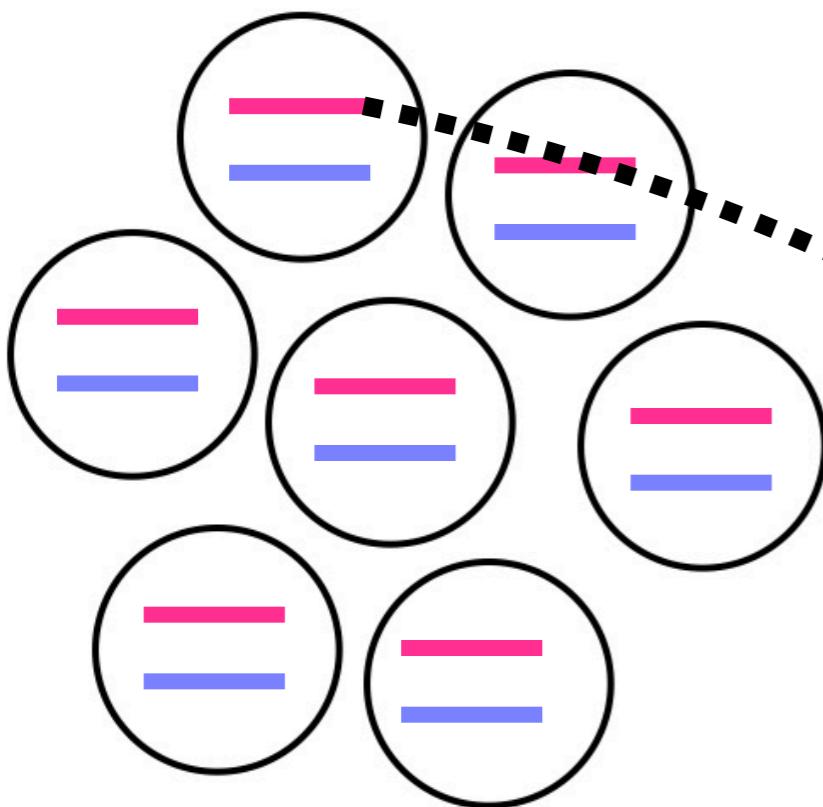
Recall: we are diploid.



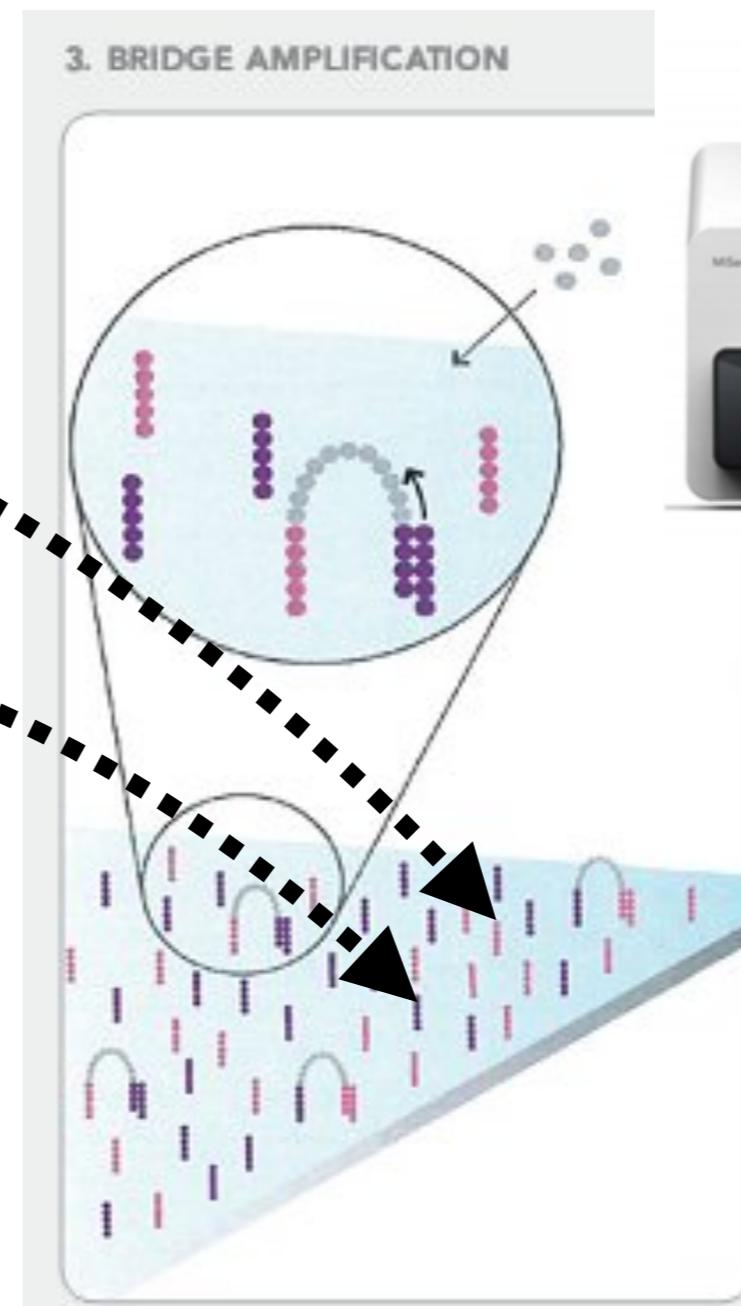
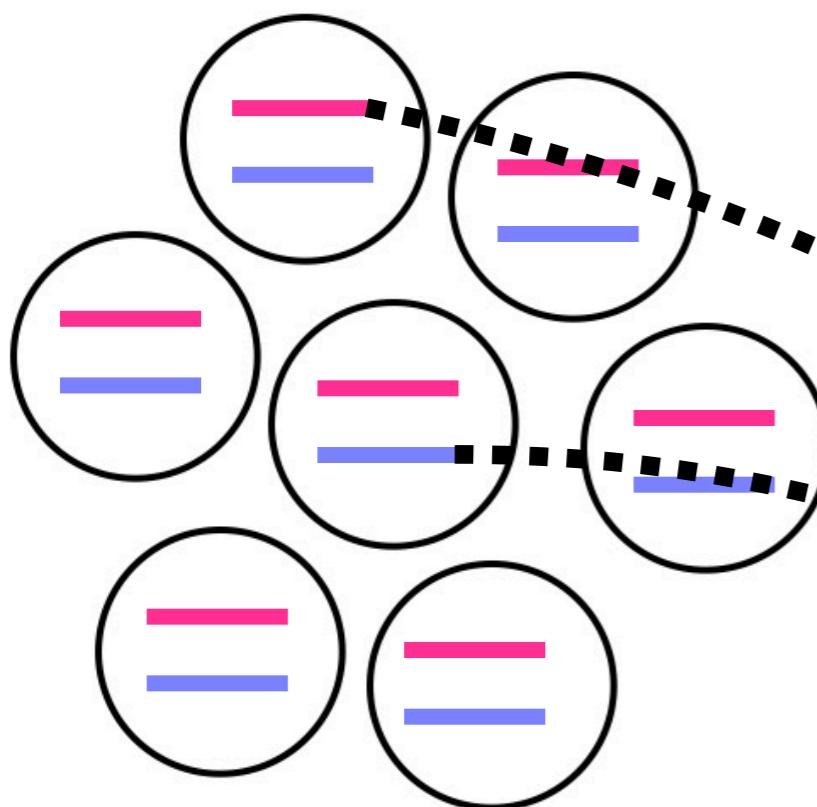
Each sequencing read is a piece of a parental allele from a single cell



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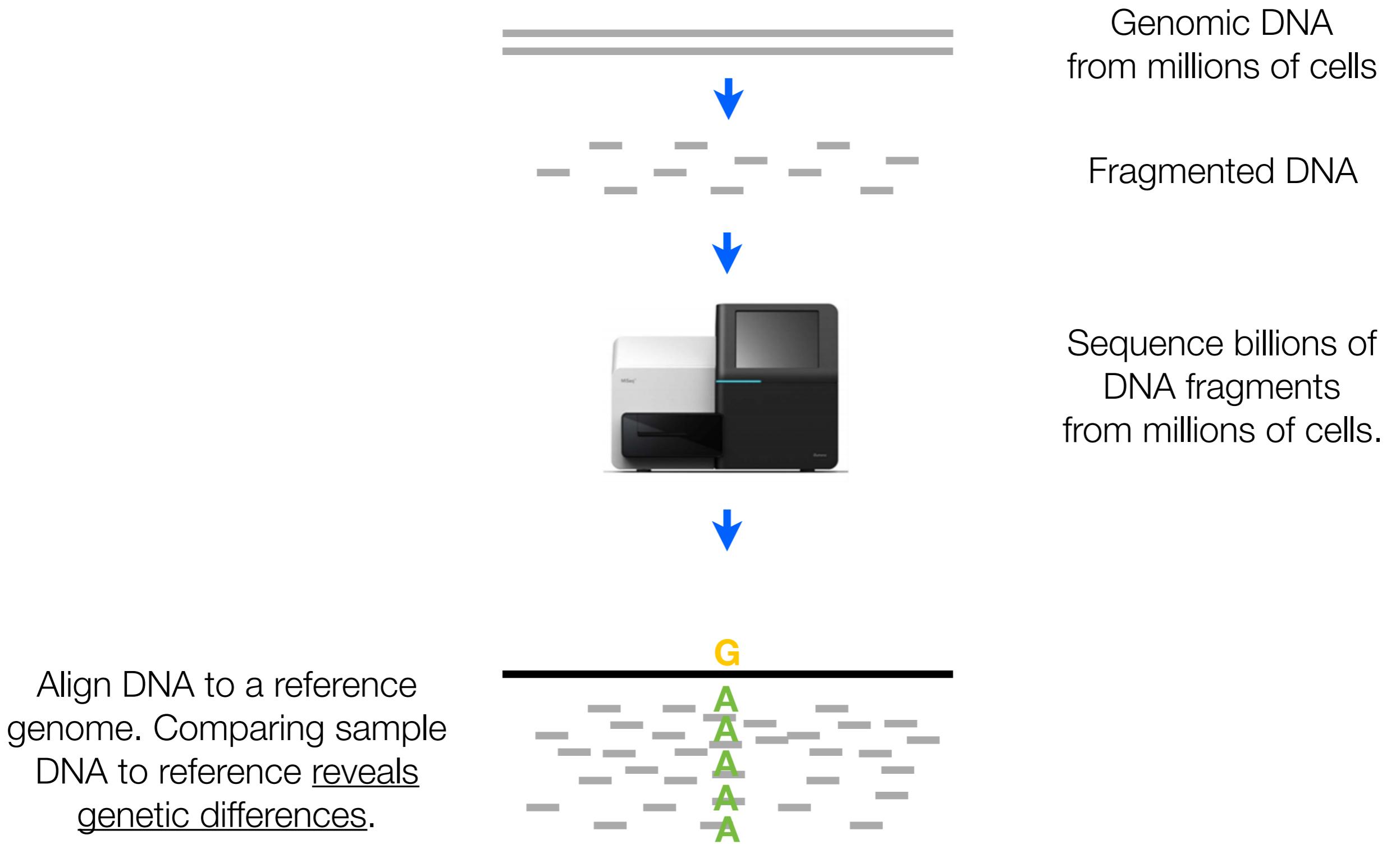


Each sequencing read is a piece of a parental allele from a single cell

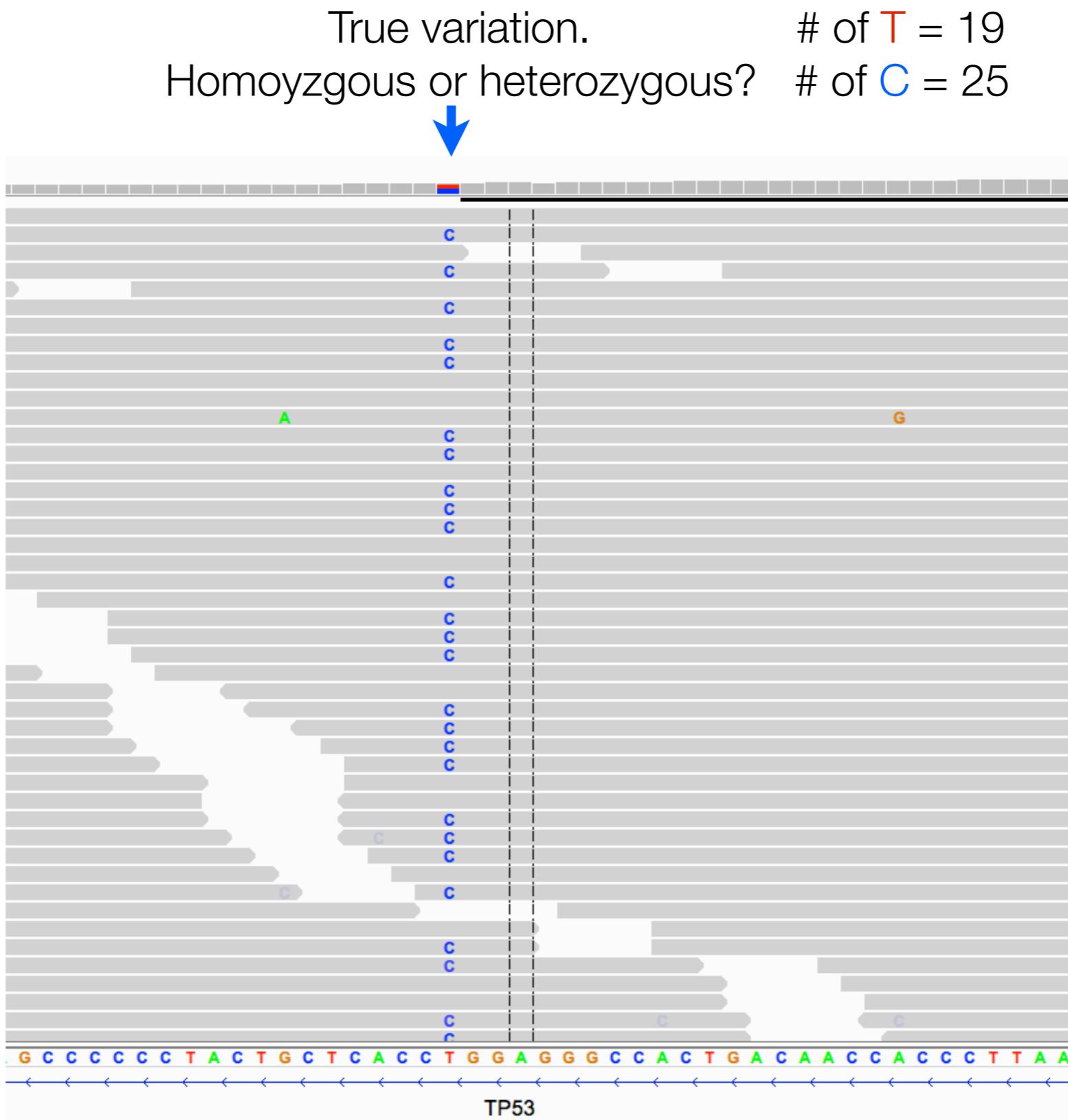


— Paternal
— Maternal

A “pileup” of reads at a given chromosomal position is a sampling of the alleles present in a population of cells.

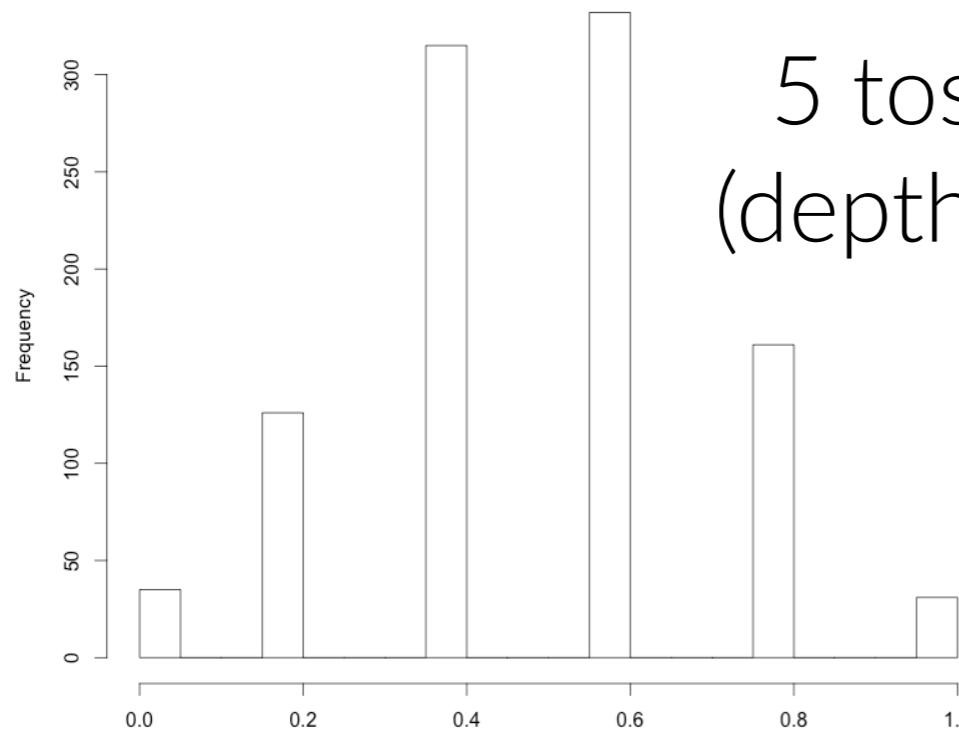


Heterozygotes: expect ~50/50 allele balance (binomial expectation). There are biases...



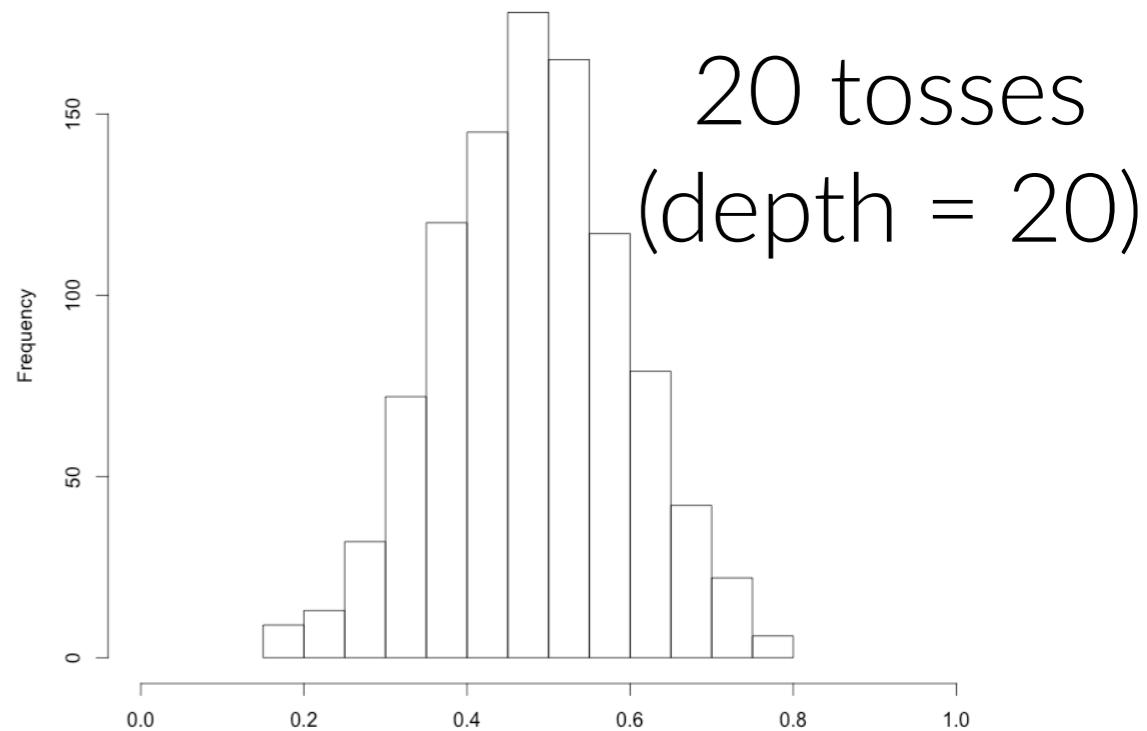
Allele sampling from NGS reads is akin to coin tosses. Deeper sampling “coverage” is better.

Distribution of % heads from 1000 experiments with 5 tosses each



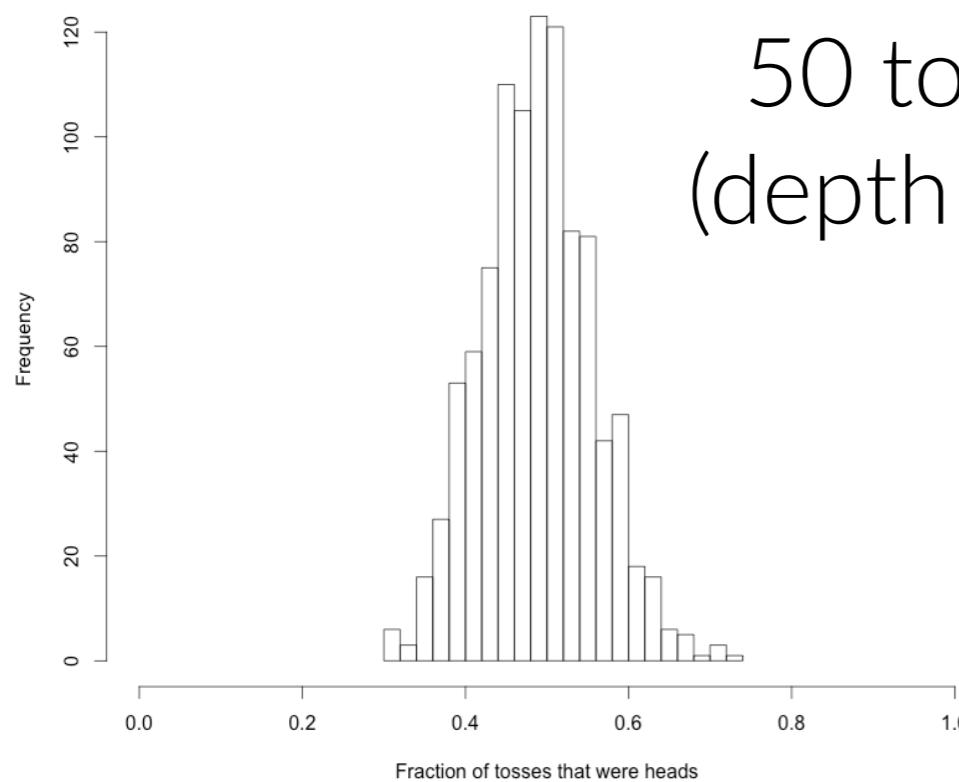
5 tosses
(depth = 5)

Distribution of % heads from 1000 experiments with 20 tosses each



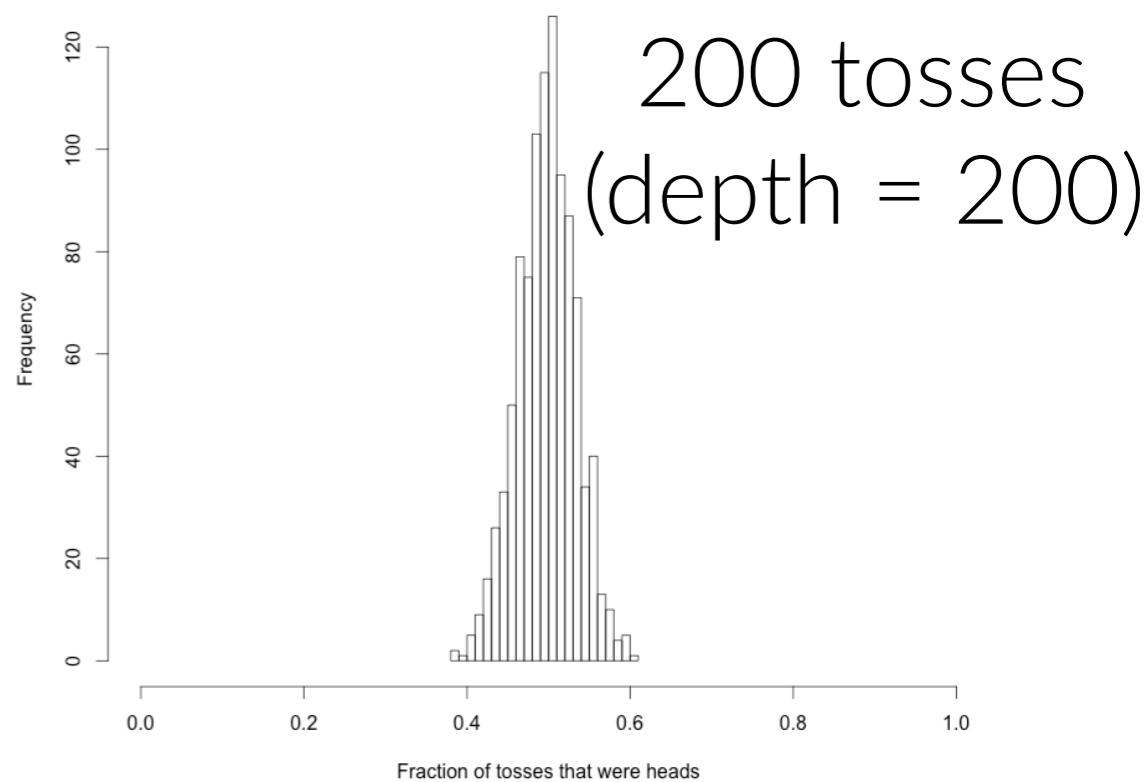
20 tosses
(depth = 20)

Distribution of % heads from 1000 experiments with 50 tosses each



50 tosses
(depth = 50)

Distribution of % heads from 1000 experiments with 200 tosses each

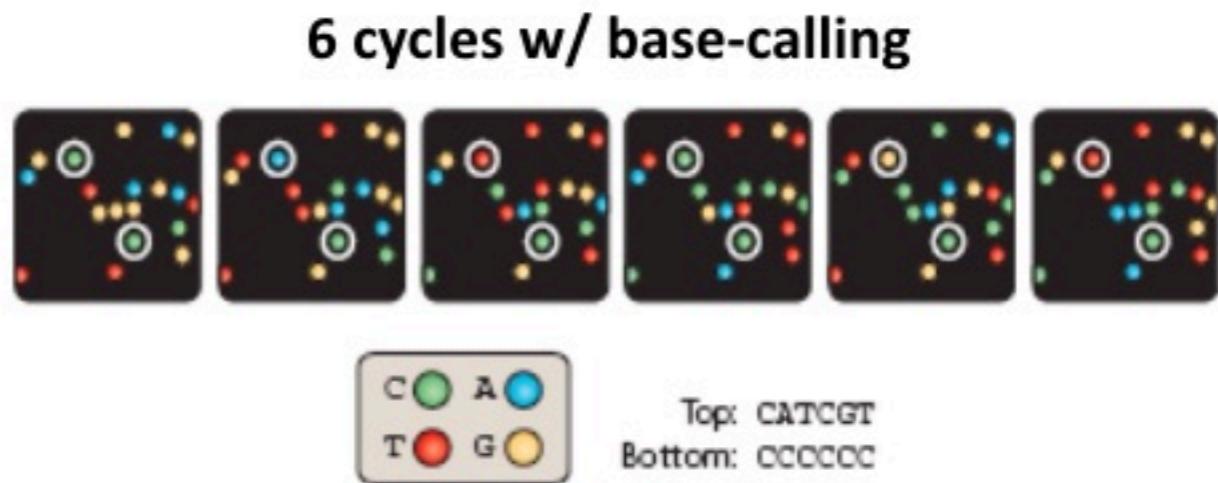
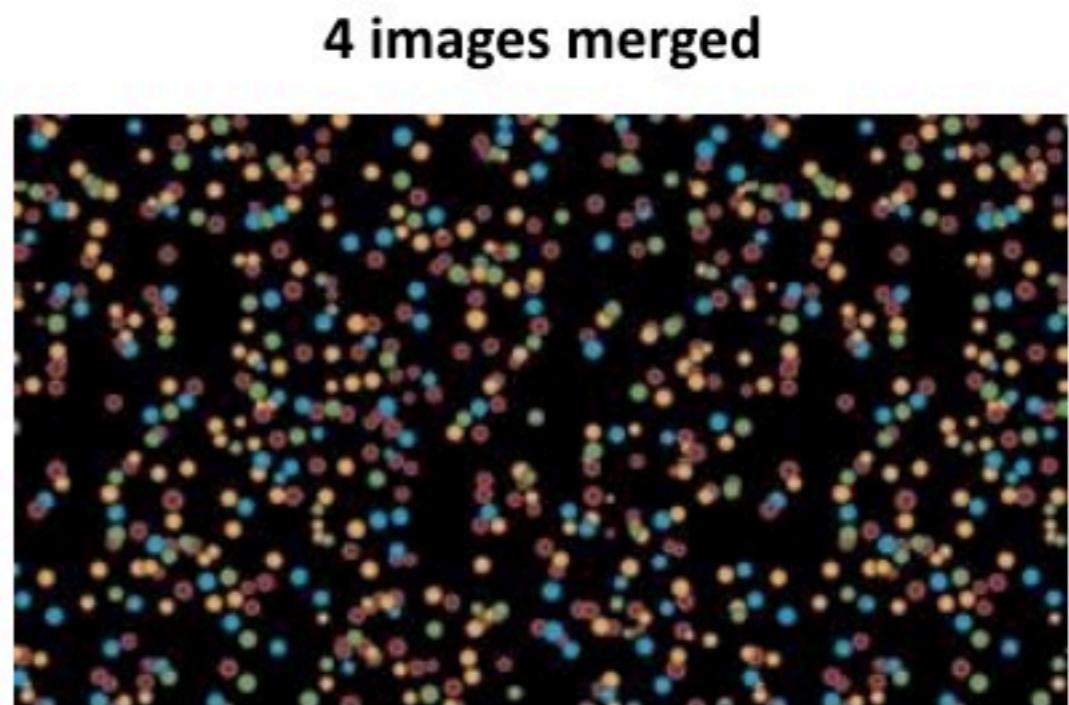
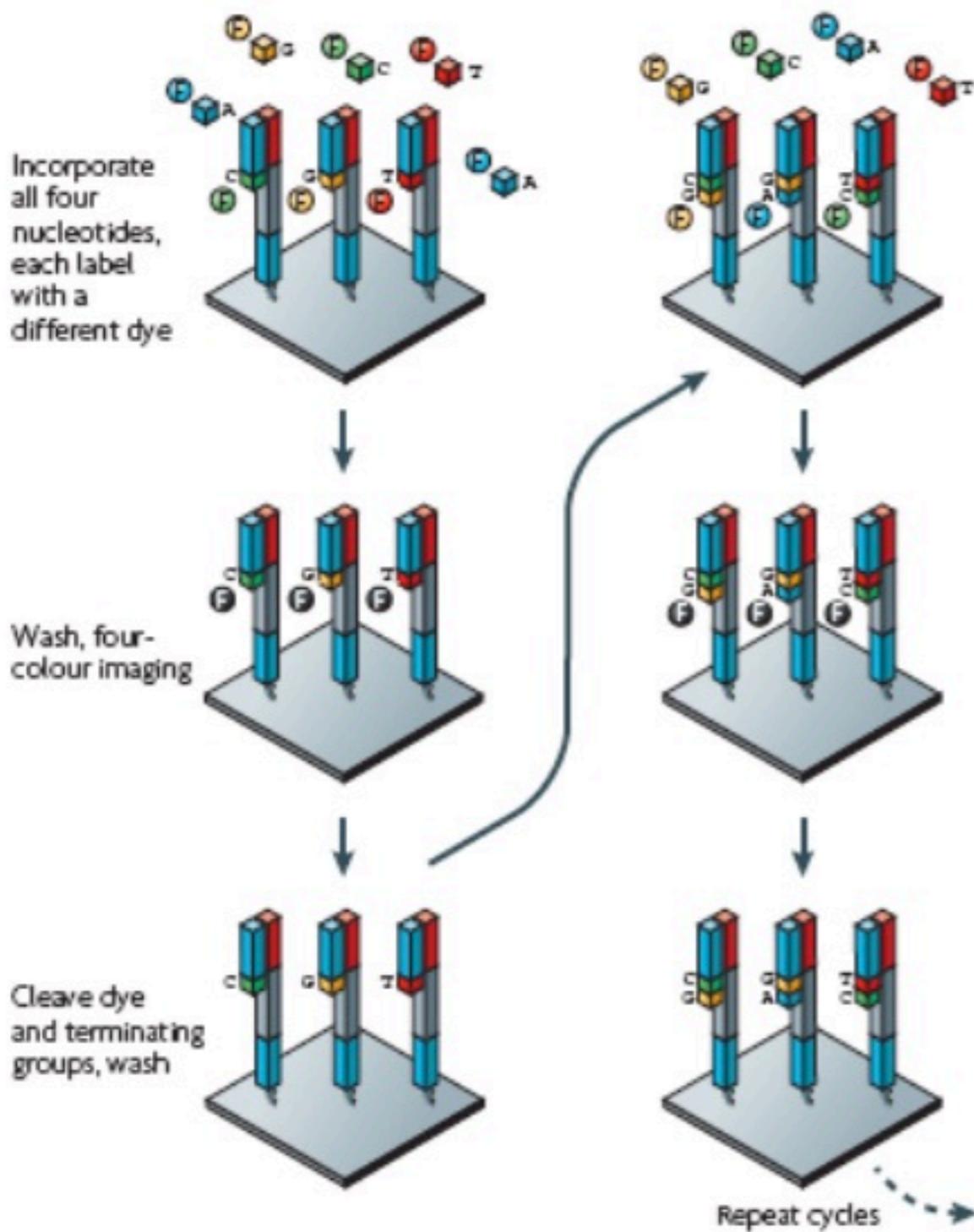


200 tosses
(depth = 200)

Sequencers make mistakes!



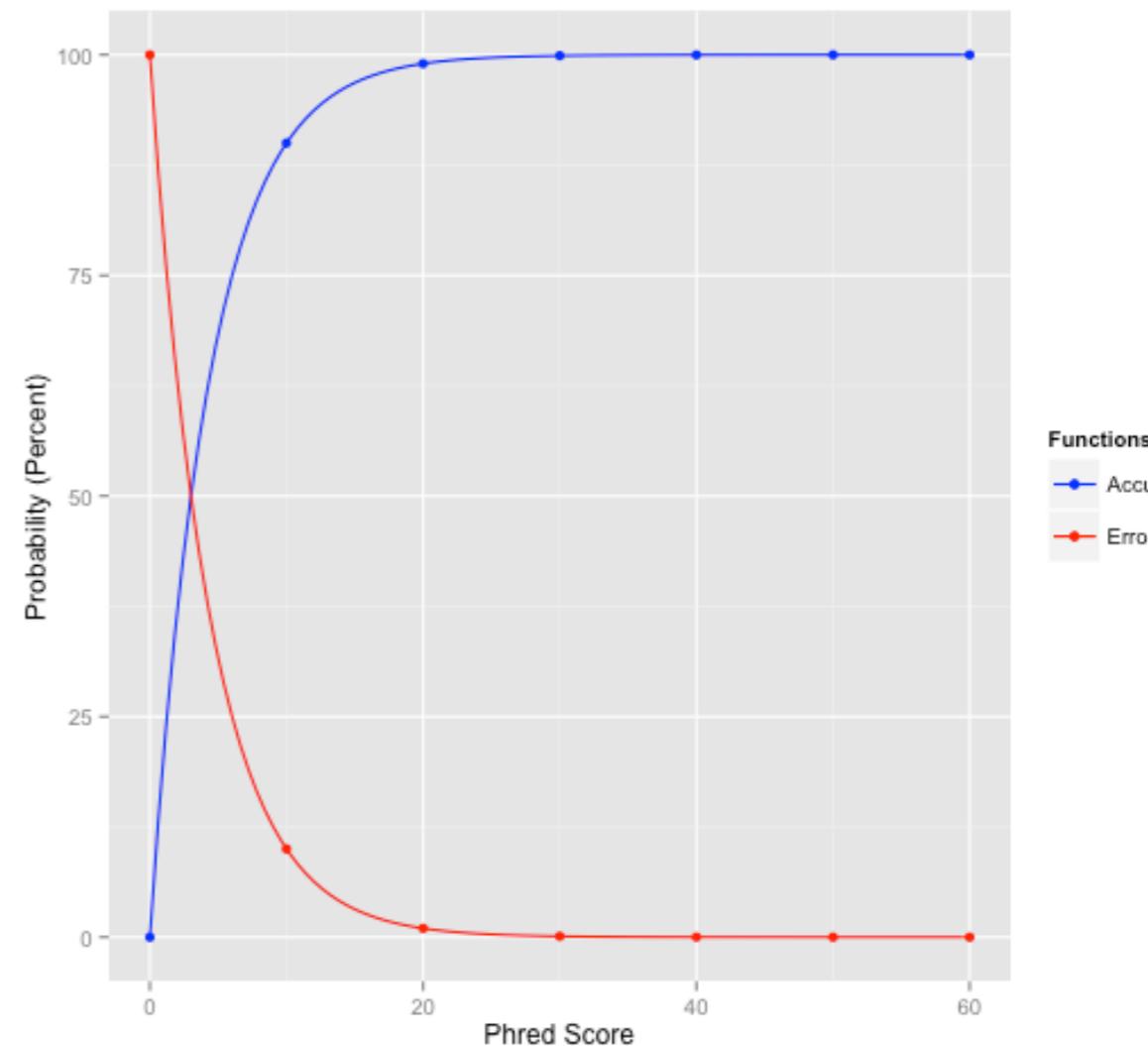
Not surprising: Solexa sequencing is ~stargazing.
Think about this. It is amazing. Error rate is ~0.1%!!!



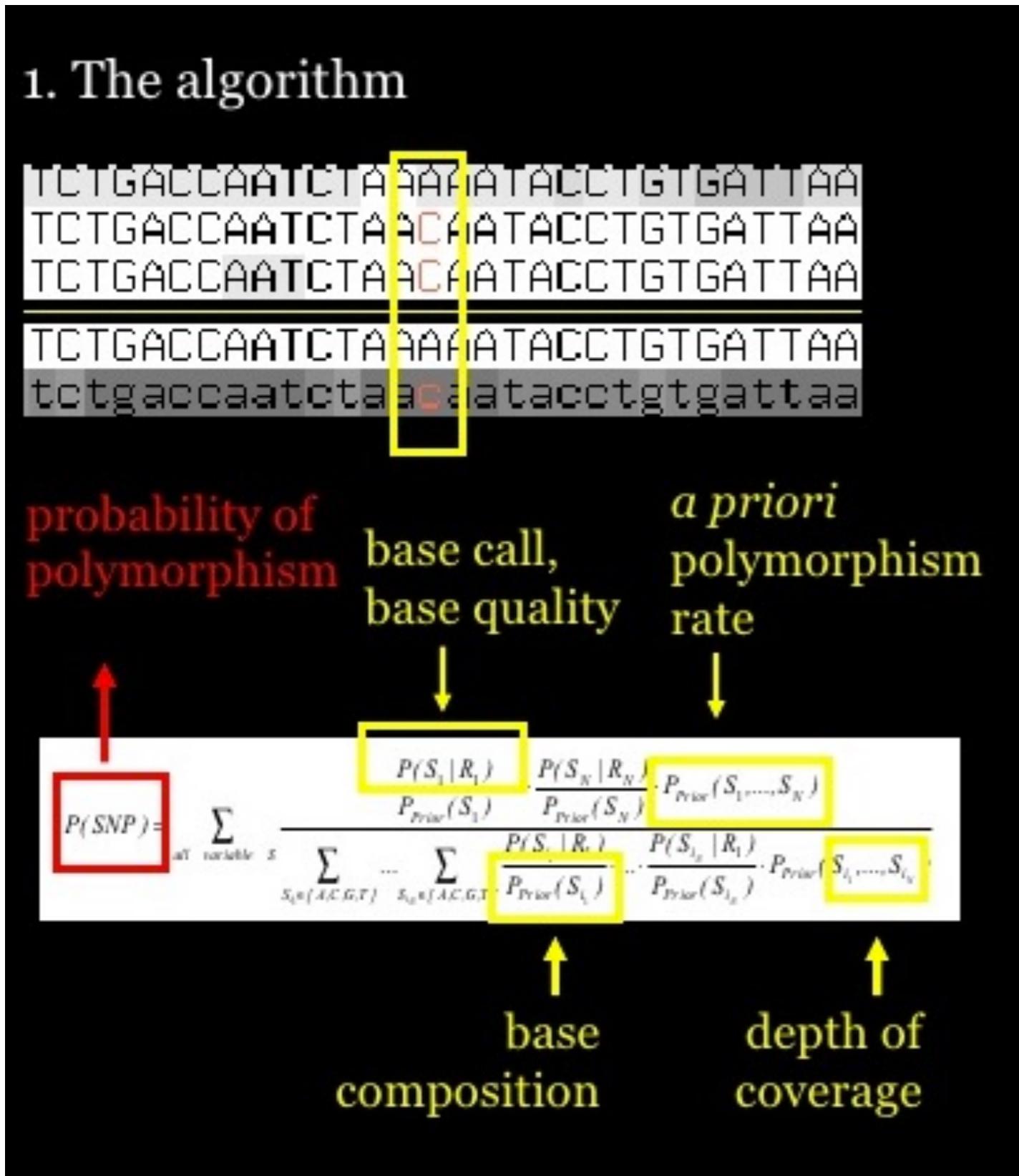
“Phred-scaled” Quality Scores

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

Phred Quality Score	Error	Accuracy (1 - Error)
10	$1/10 = 10\%$	90%
20	$1/100 = 1\%$	99%
30	$1/1000 = 0.1\%$	99.9%
40	$1/10000 = 0.01\%$	99.99%
50	$1/100000 = 0.001\%$	99.999%
60	$1/1000000 = 0.0001\%$	99.9999%



PolyBayes: the first Bayesian approach to SNP discovery. Accounts for base quality. Predecessor to FreeBayes



A general approach to single-nucleotide polymorphism discovery

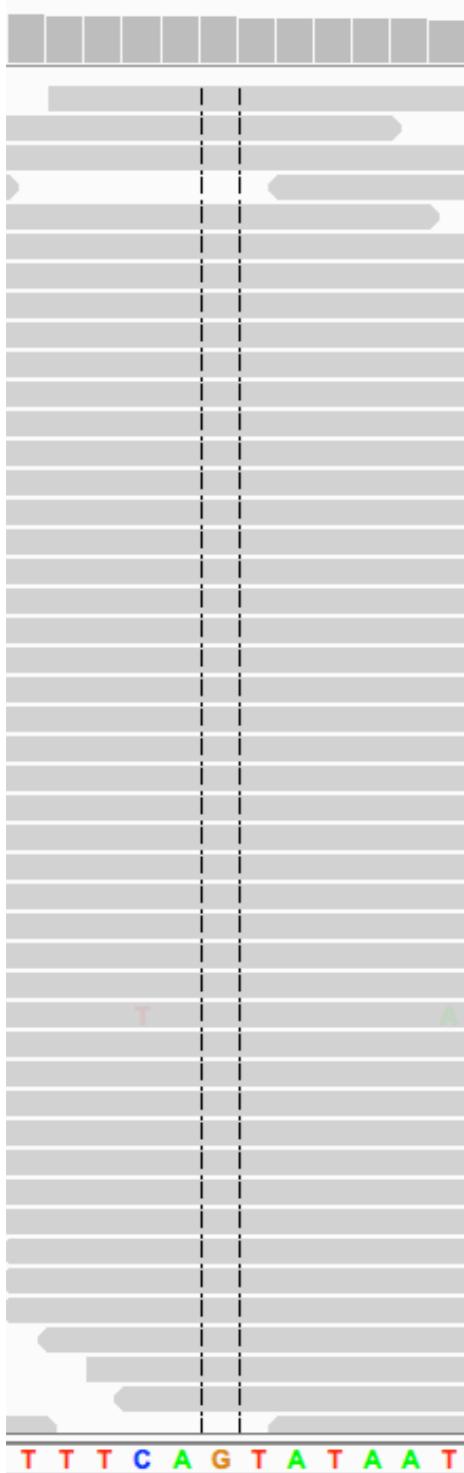
Gabor T. Marth¹, Ian Korf¹, Mark D. Yandell¹, Raymond T. Yeh¹, Zhijie Gu², Hamideh Zakeri², Nathan O. Stitzel¹, LaDeana Hillier¹, Pui-Yan Kwok² & Warren R. Gish¹

identification and multiple alignment. We analyse these sequences with a novel, Bayesian inference engine, POLYBAYES, to calculate the probability that a given site is polymorphic. Rigorous treatment of base quality permits completely automated evaluation of the full length of all sequences, without limitations on alignment depth. We demonstrate this approach by accurate SNP predictions in human ESTs aligned to finished and working-draft quality genomic sequences, a data set representative of the typical challenges of sequence-based SNP discovery.

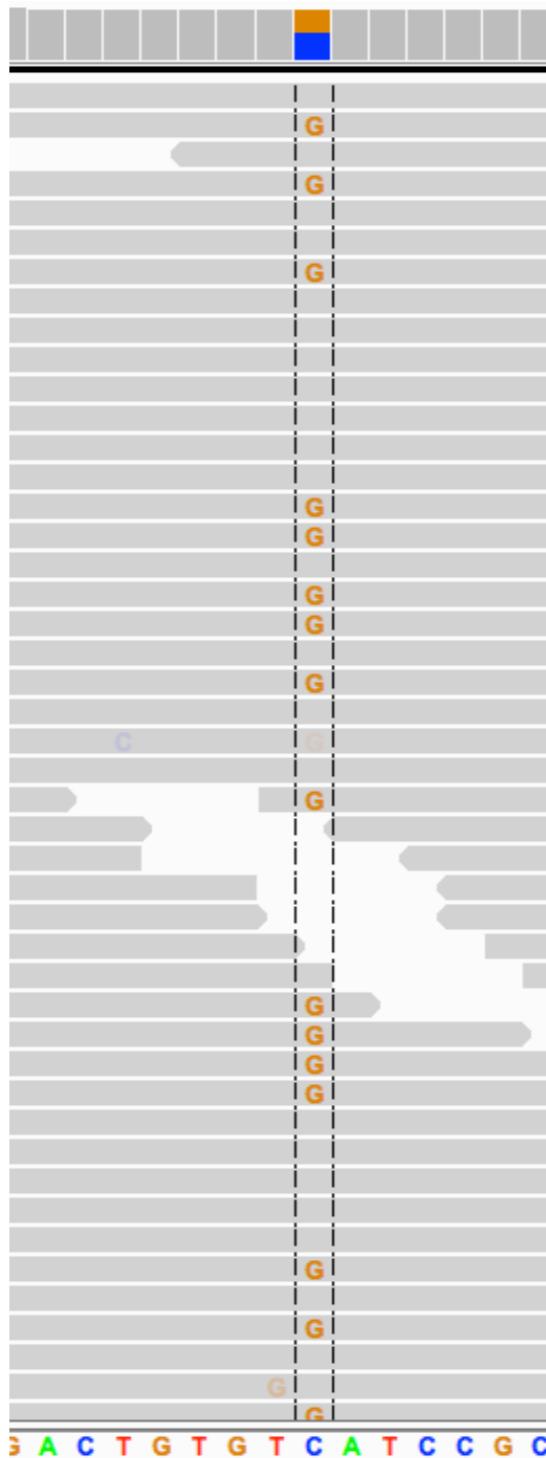
2. Use sequence quality information (base quality values) to distinguish true mismatches from sequencing errors

TGAAA Agg AATT	TTGAT CC CTGT
TGAAA t GAATT	TTGATT T CCTGT
sequencing error	true polymorphism

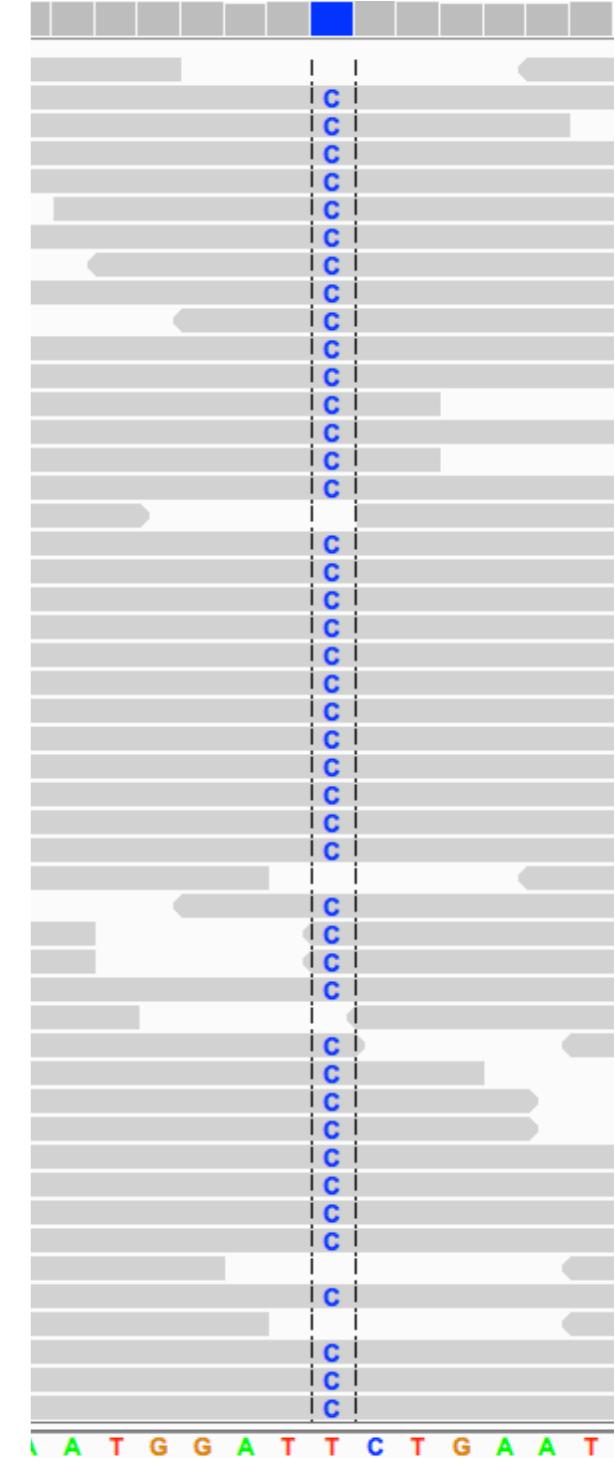
Different (diploid) SNP genotypes



Homozygous for reference
(i.e., both chroms same as ref)

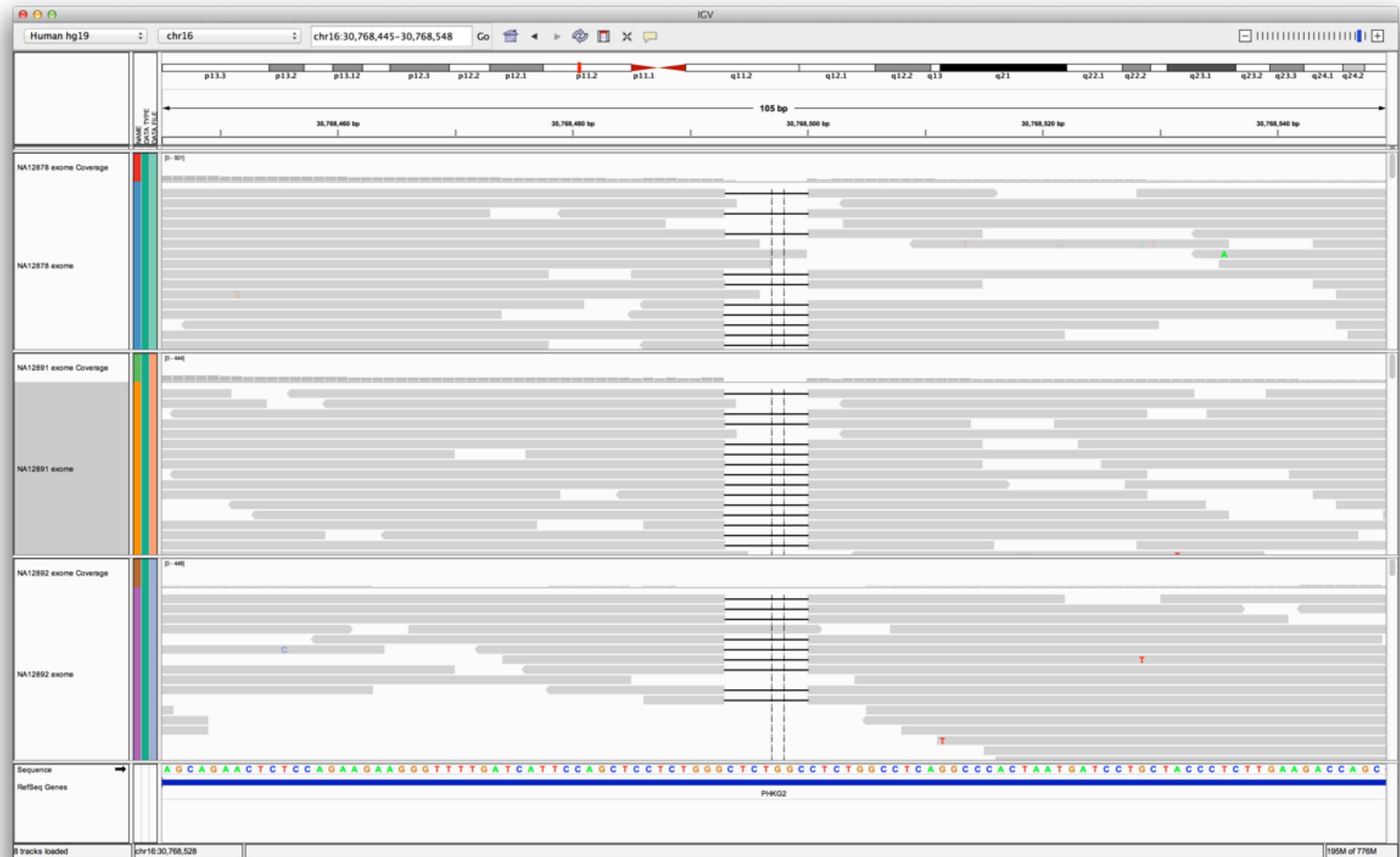


Heterozygous
(i.e., 1 chrom same as ref, 1 diff.)

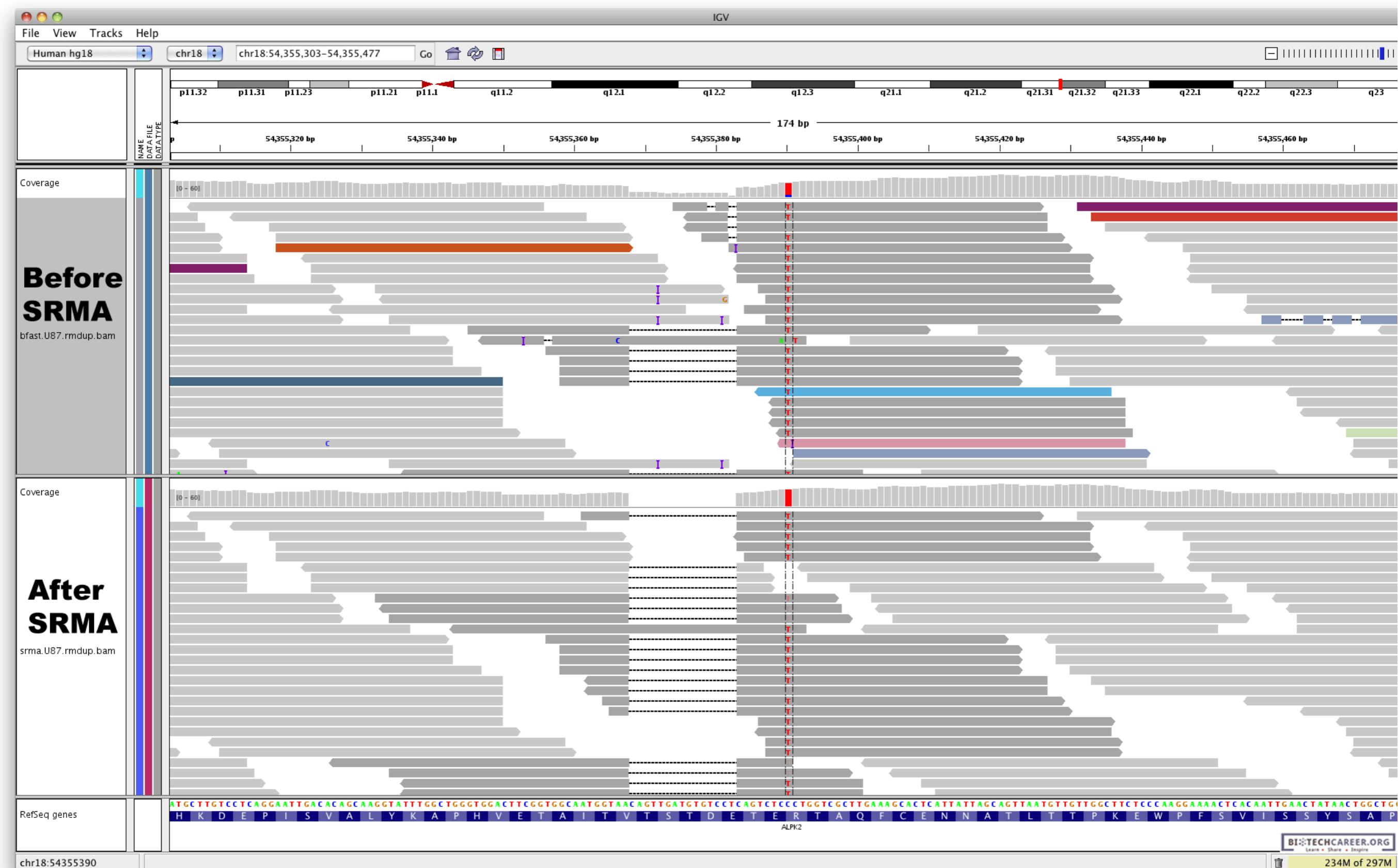


Homozygous for reference
(i.e., both chroms diff than ref)

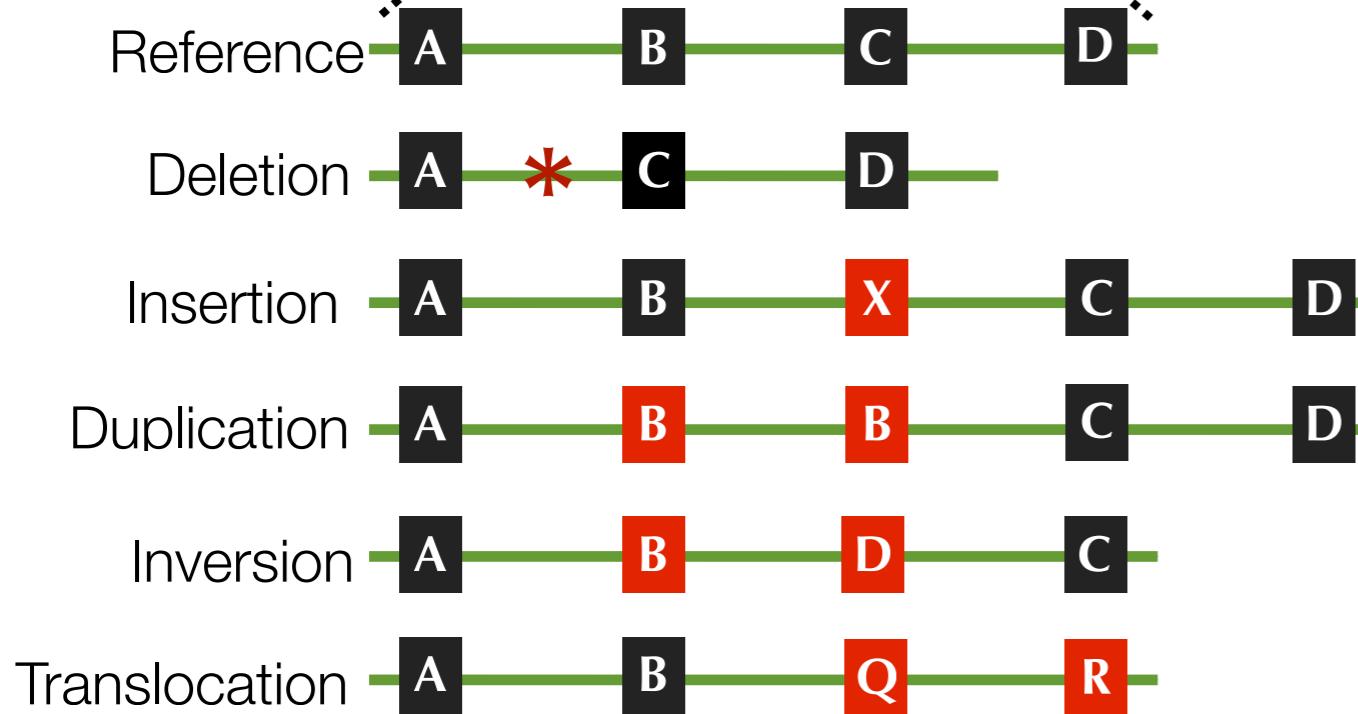
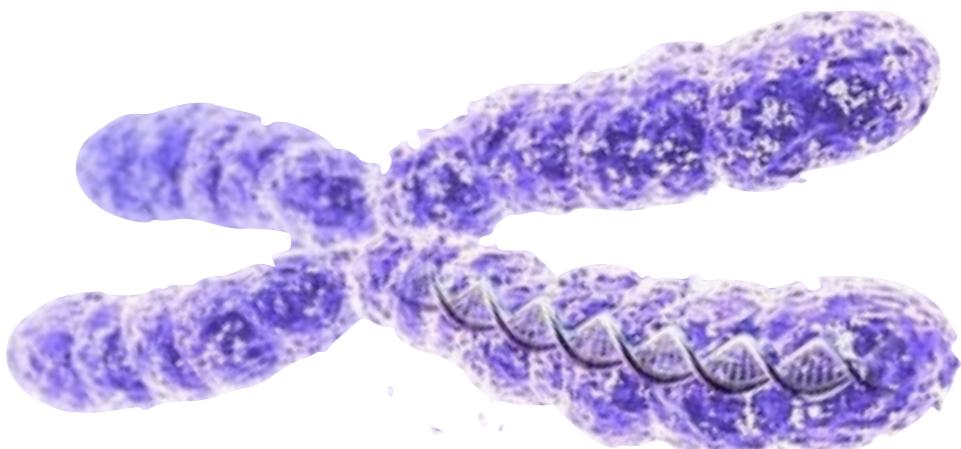
Insertion-deletion polymorphisms (INDELS)



Not always so simple...



Structural Variation



- large (>1kb) differences that affect the copy number, orientation, or location of genomic segments
- Common in mammalian genomes (~3-5 thousand between two people)
- A hallmark of cancer
- A major cause of spontaneous disease
- more are functional than SNPs
- very challenging to identify

SV Breakpoints

Deletion

Test A B C H I J

Reference A B C D E F G H I J

Inversion

Test A B C G F E D H I J

Ref. A B C D E F G H I J

Tandem Duplication

Test A B C D E F G D E F G H I J

Ref. A B C D E F G H I J

Distant Insertion

Test A B C D E X F G H I J

Ref. A B C D E F G H I J W X Y

Reciprocal translocation

Test chr1/2 A B C D E 1 2 3 4 5

1 2 3 4 5 F G H I J Test Chr2/1

Ref. Chr1 A B C D E F G H I J

1 2 3 4 5 6 7 8 9 10 Ref. Chr2

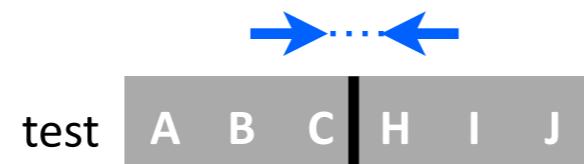
Detecting SVs from alignments

1) Read depth



2) Paired-end mapping

paired-reads (or strobe)



Reference

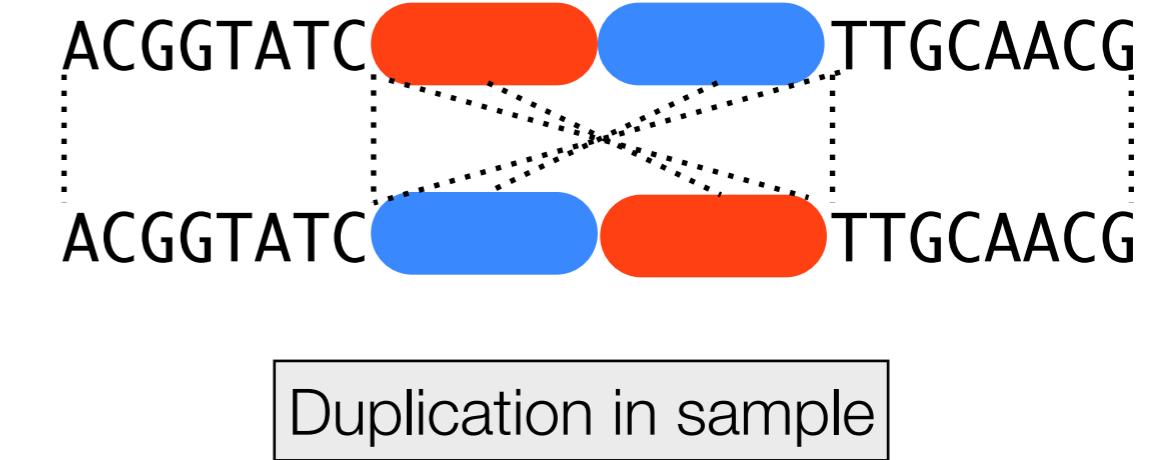
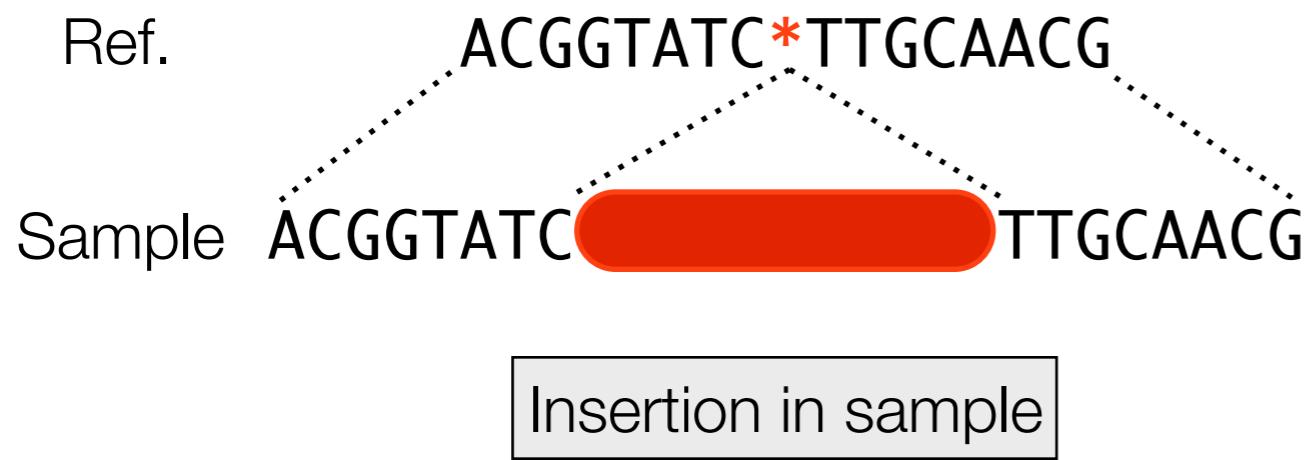
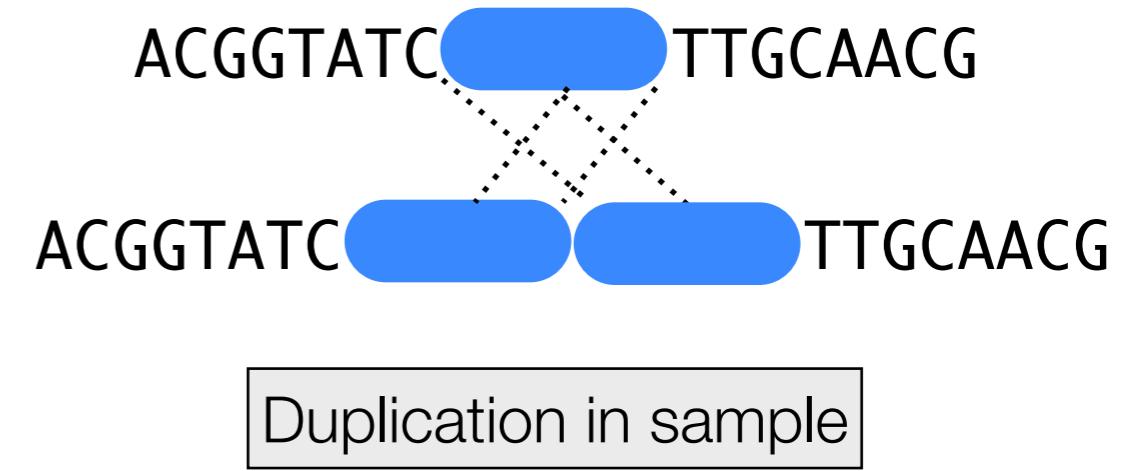
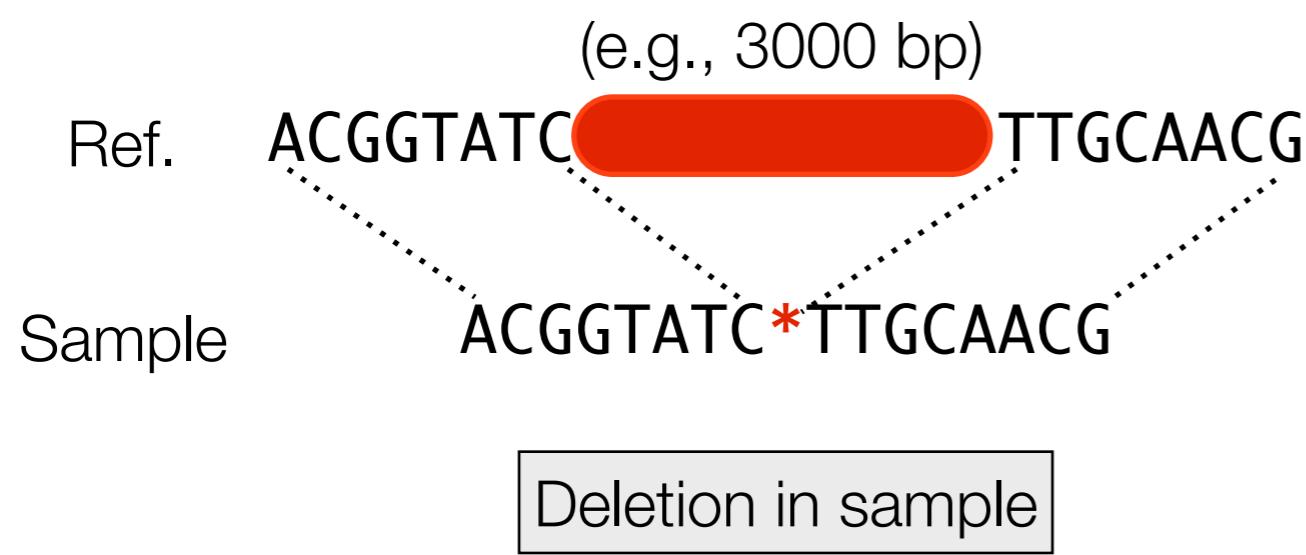


3) Split mapping

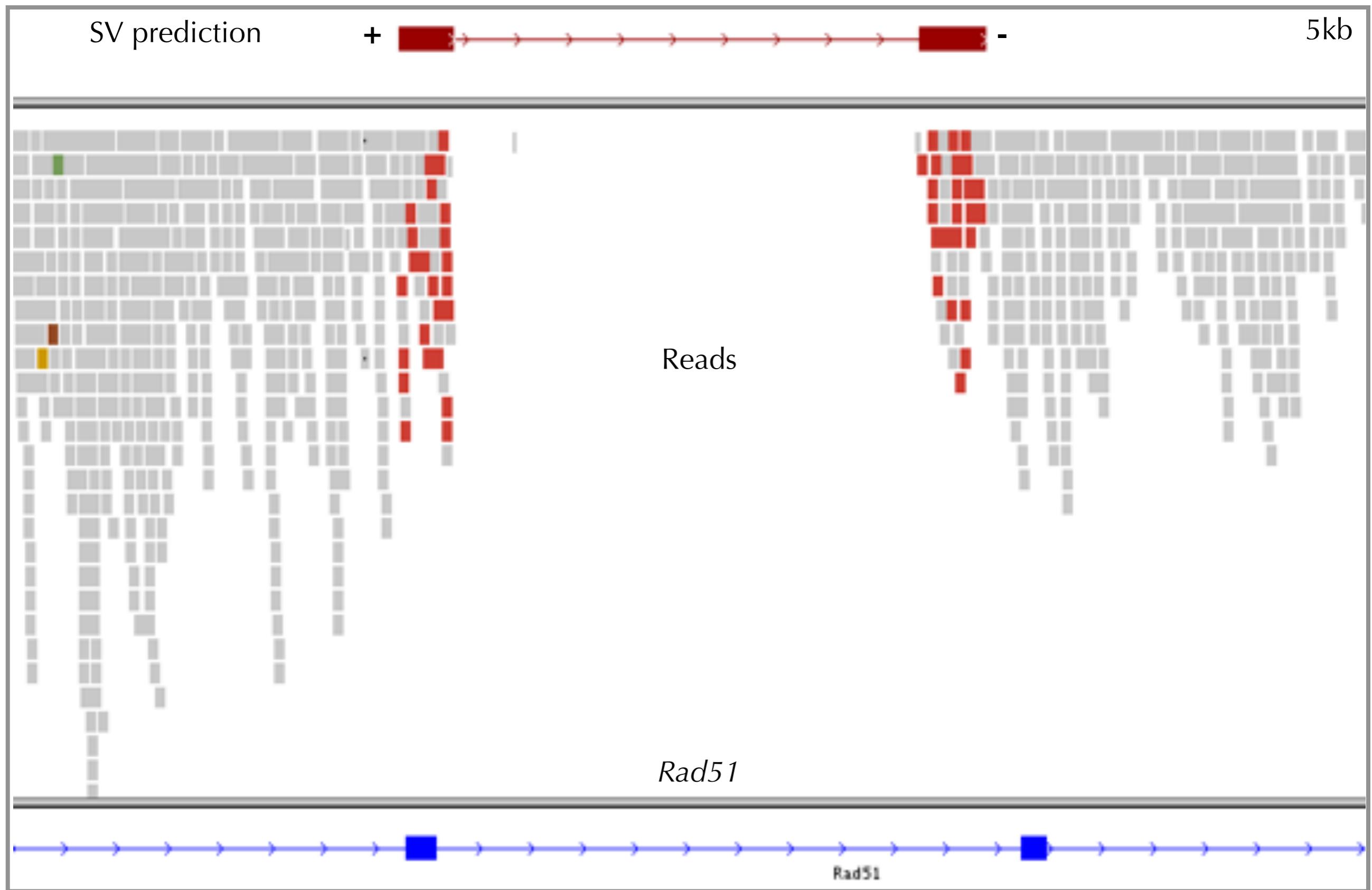
long reads or contigs



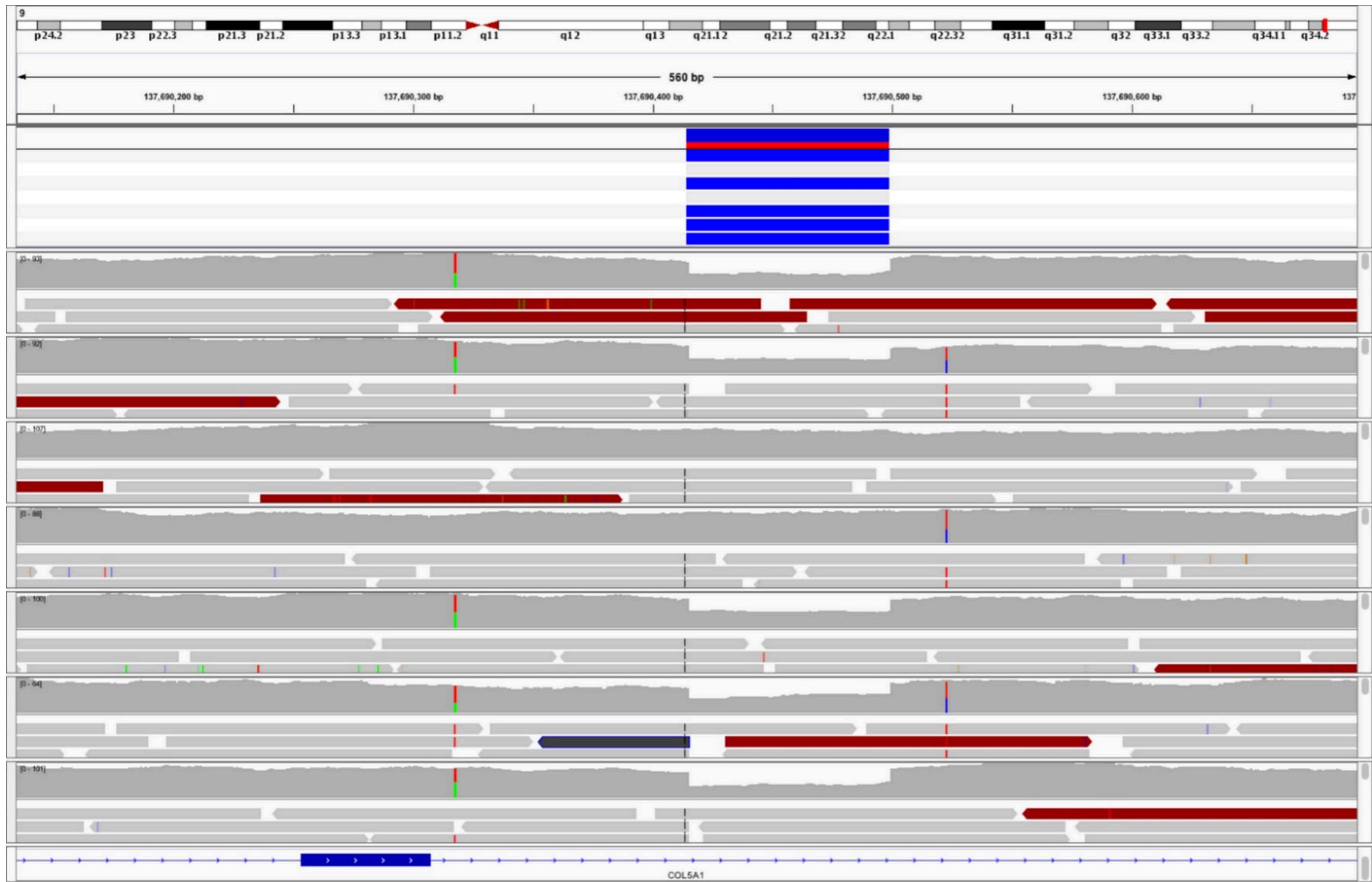
SV Alignment patterns



A real deletion detected with sequencing



A deletion segregating in a family



Summary

- Four major classes of genetic variation: single-nucleotide polymorphisms (SNPs), insertion-deletion polymorphisms (INDELS), structural variants (SVs), and mobile-element insertions (not discussed).
- Modern sequencing technologies provide an excellent substrate for detecting all forms of genetic variations
- However, sufficient sequencing depth and per-base accuracy are necessary for **comprehensive** and **accurate** variant discovery.
- Improved sequencing technologies (e.g. longer, more accurate reads) and better algorithms (e.g., modeling error, phase-aware) are the path forward.