



Genetic variation: what, why, how

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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

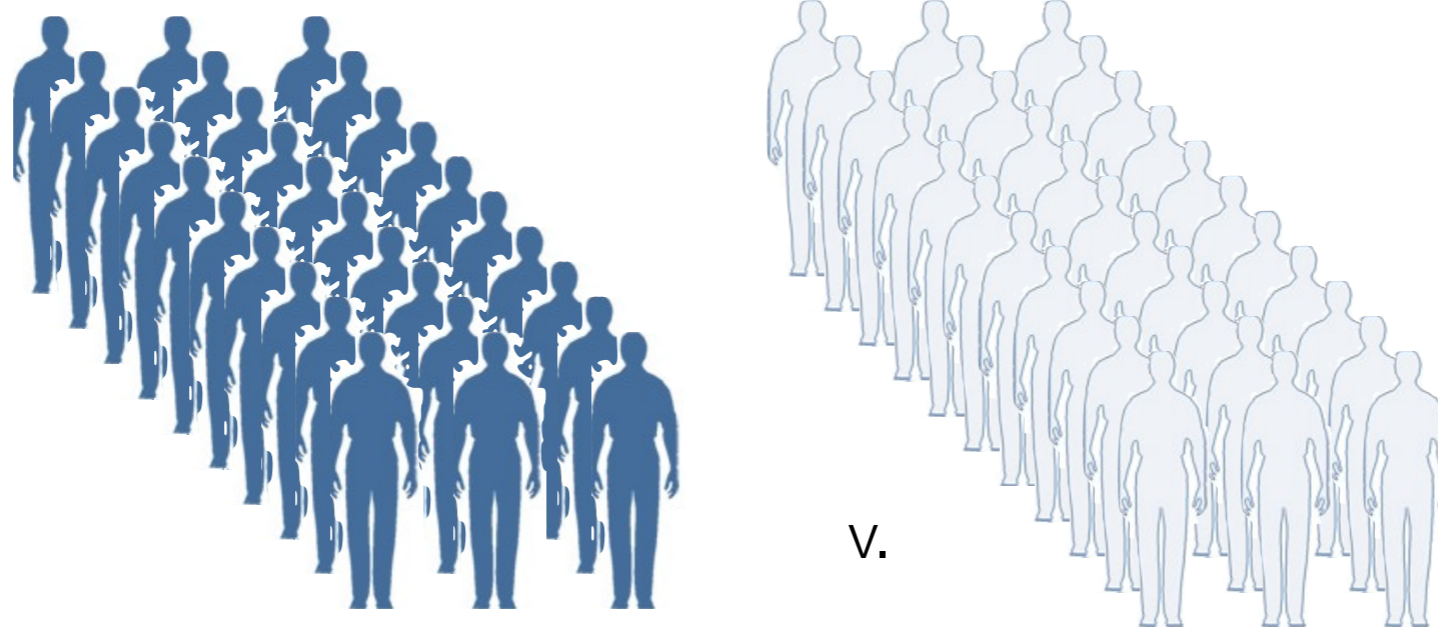


~99.5% identical DNA



Why do we care?

- Understand the relationship between genotype and phenotype.



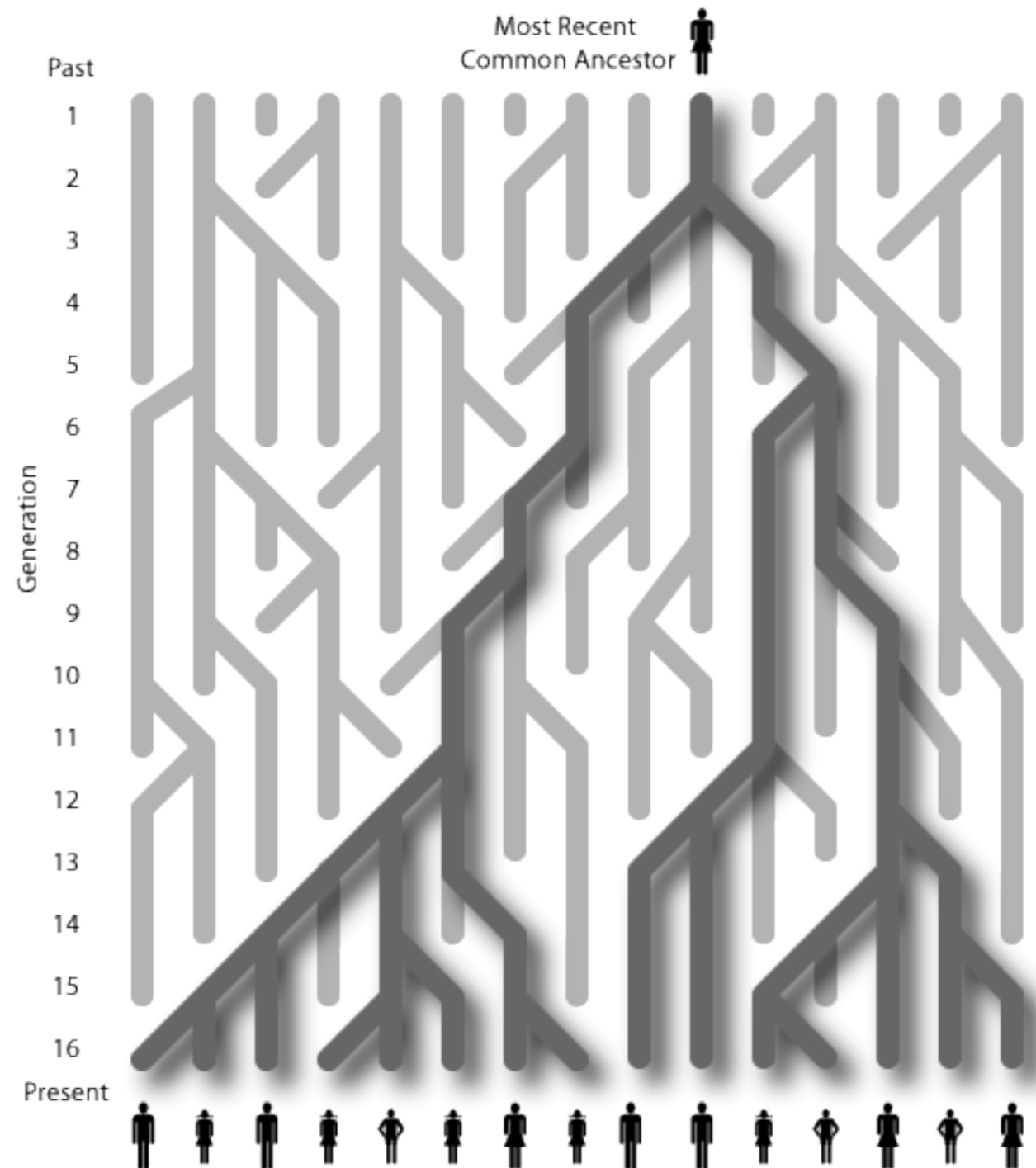
Cases
(have disease)

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**)

“spelling mistakes”

ctc--ag
ctc**tg**ag

Insertion-deletion
polymorphisms
(**INDELs**)

*“extra or missing
letters”*

ctcag
ctc  ag

Structural
variants
(**SVs**)

*“extra, missing
or reordered
chapters”*

Properties of genetic variation

Single-nucleotide
(**SNPs**)

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

Insertion-deletions
(**INDELs**)

ctc--ag
ctc**tg**ag

Structural
variants (**SVs**)

ctcag
ctc  ag

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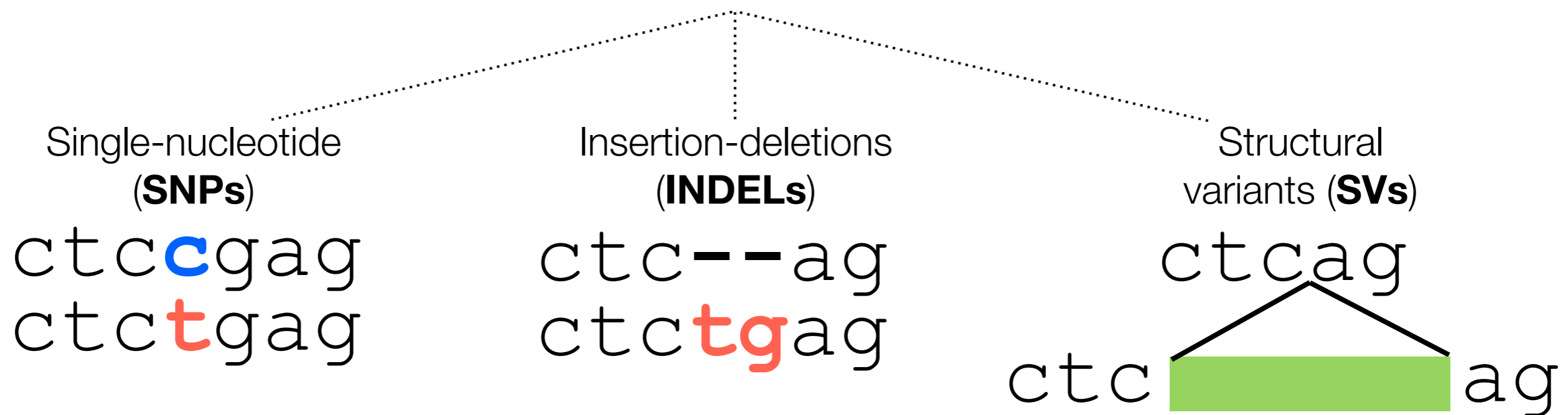
15,000,000 to 21,000,000 different base pairs.

How different are we?

The genomes of any two humans are ~99.5% identical

But, the genome is big.

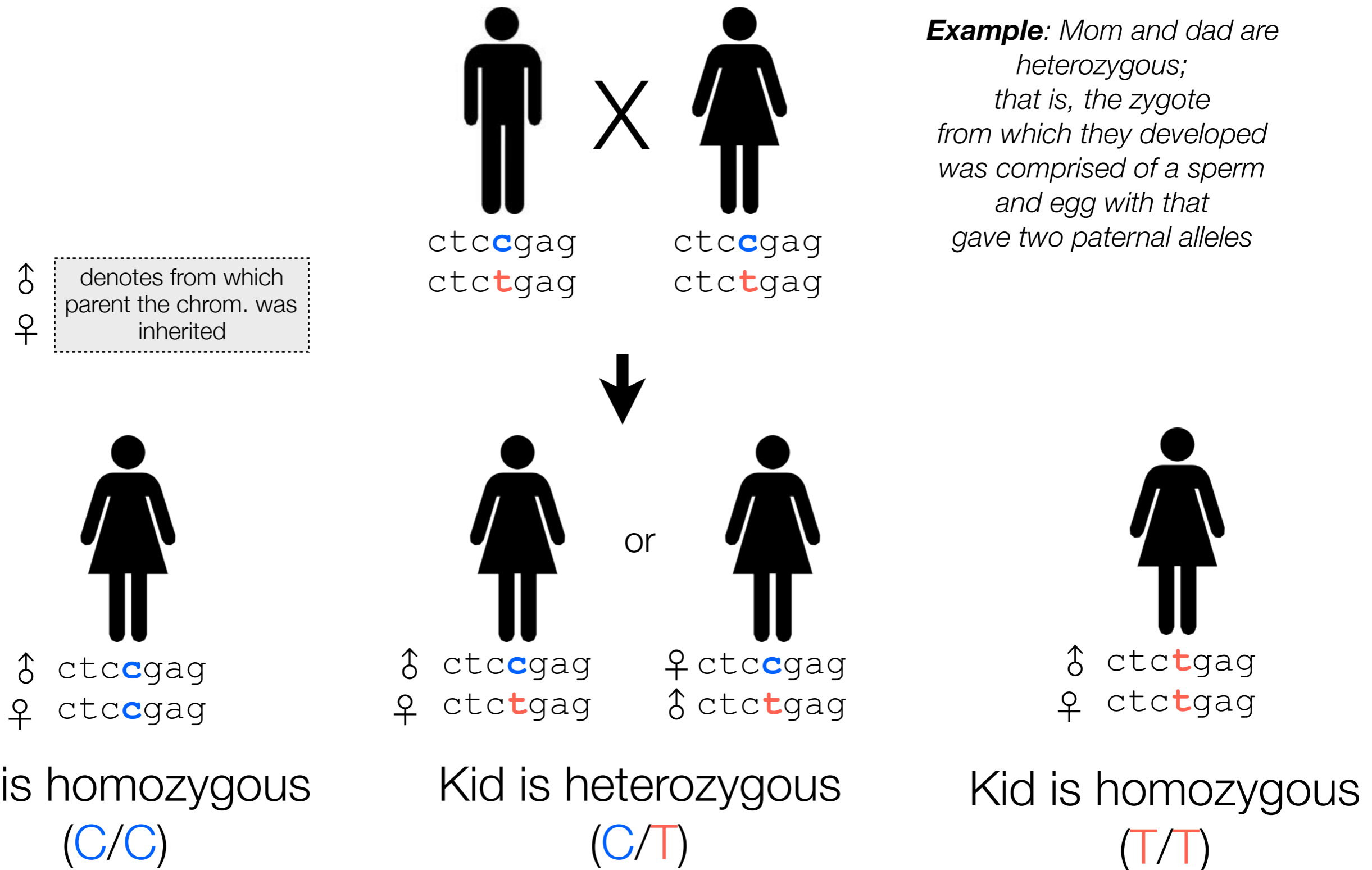
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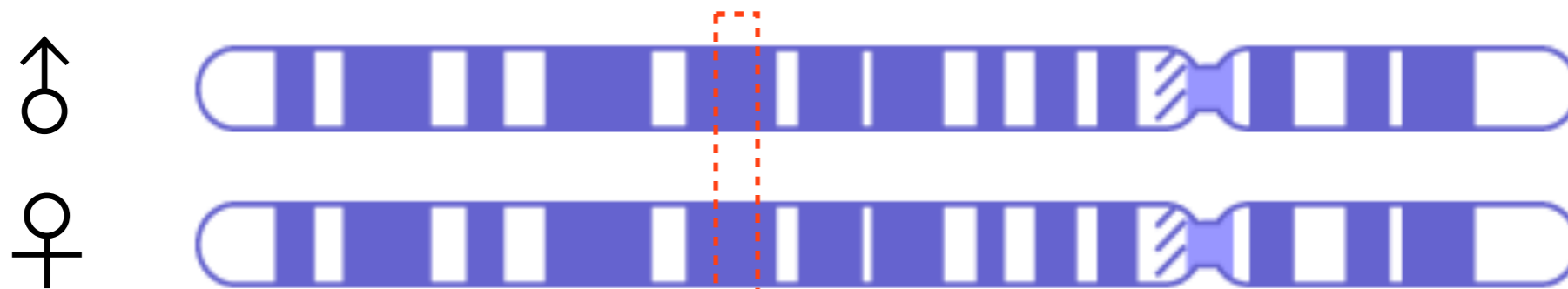
Detecting genetic variation

CGCAAATTTGCCGGATTTCTTTGCTGTTCTGTCATGTAGTTTAAACGAGATTGCCAGCACCGGGTATCATTACCATTTTTCTTTTC
GTTAACTTGCCGTCAGCCTTTTCTTTGACCTCTTCTTTCTGTTTCATGTGTATTTGCTGTCTCTTAGCCAGACTTCCCGTGTCTTTCC
ACCGGGCCTTTGAGAGGTCACAGGGTCTTGATGCTGTGGTCTTCATCTGCAGGTGTCTGACTTCCAGCAACTGCTGGCCTGTGCCAGG
GTGCAAGCTGAGCACTGGAGTGGAGTTTTCTGTGGAGAGGAGCCATGCCTAGAGTGGGATGGGCCATTGTTTCATCTTCTGGCCCCTG
TTGTCTGCATGTAACCTAATAACCAACCAGGCATAGGGGAAAGATTGGAGGAAAGATGAGTGAGAGCATCAACTTCTCTCACAACT
AGGCCAGTAAGTAGTGCTTGTGCTCATCTCCTTGGCTGTGATACGTGGCCGGCCCTCGCTCCAGCAGCTGGACCCCTACCTGCCGTCT
GCTGCCATCGGAGCCCAAAGCCGGGCTGTGACTGCTCAGACCAGCCGGCTGGAGGGAGGGGCTCAGCAGGTCTGGCTTTGGCCCTGGG
AGAGCAGGTGGAAGATCAGGCAGGCCATCGCTGCCACAGAACCAGTGGATTGGCCTAGGTGGGATCTCTGAGCTCAACAAGCCCTCT
CTGGGTGGTAGGTGCAGAGACGGGAGGGGCAGAGCCGCAGGCACAGCCAAGAGGGCTGAAGAAATGGTAGAACGGAGCAGCTGGTGAT
GTGTGGGCCCACCGGCCCCAGGCTCCTGTCTCCCCCAGGTGTGTGGTGTGATGCCAGGCATGCCCTTCCCCAGCATCAGGTCTCCAGAG
CTGCAGAAGACGACGGCCGACTTGGATCACACTCTTGTAGTGTCCCAGTGTTGCAGAGGTGAGAGGAGAGTAGACAGTGAGTGGGA
GTGGCGTCGCCCTAGGGCTCTACGGGGCCGGCGTCTCCTGTCTCCTGGAGAGGCTTCGATGCCCTCCACACCCTCTTGATCTTCCC
TGTGATGTCATCTGGAGCCCTGCTGCTTGCGGTGGCCTATAAAGCCTCCTAGTCTGGCTCCAAGGCCTGGCAGAGTCTTCCCAGGGA
AAGCTACAAGCAGCAAACAGTCTGCATGGGTCATCCCCTTCACTCCCAGCTCAGAGCCCAGGCCAGGGGCCCCCAAGAAAGGCTCTGG
TGGAGAACCTGTGCATGAAGGCTGTCAACCAGTCCATAGGCAAGCCTGGCTGCCTCCAGCTGGGTTCGACAGACAGGGGCTGGAGAAGG
GGAGAAGAGGAAAGTGAGGTTGCCTGCCCTGTCTCCTACCTGAGGCTGAGGAAGGAGAAGGGGATGCACTGTTGGGGAGGCAGCTGTA
ACTCAAAGCCTTAGCCTCTGTTCCCACGAAGGCAGGGCCATCAGGCACCAAAGGGATTCTGCCAGCATAGTGCTCCTGGACCAGTGAT
ACACCCGGCACCCCTGTCCTGGACACGCTGTTGGCCTGGATCTGAGCCCTGGTGGAGGTCAAAGCCACCTTTGGTTCTGCCATTGCTGC
TGTGTGGAAGTTCACCTCCTGCCTTTTCTTTCCCTAGAGCCTCCACCACCCCGAGATCACATTTCTCACTGCCTTTTGTCTGCCCAGT
TTCACCAGAAGTAGGCCTCTTCCCTGACAGGCAGCTGCACCACTGCCTGGCGCTGTGCCCTTCTTTGCTCTGCCCGCTGGAGACGGTG
TTTGTTCATGGGCCTGGTCTGCAGGGATCCTGCTACAAAGGTGAAACCCAGGAGAGTGTGGAGTCCAGAGTGTTGCCAGGACCCAGGCA
CAGGCATTAGTGCCCGTTGGAGAAAACAGGGGAATCCCGAAGAAATGGTGGGTCTGGCCATCCGTGAGATCTTCCCAGGTGTGCCGT
TTTCTCTGGAAGCCTCTTAAGAACACAGTGGCGCAGGCTGGGTGGAGCCGTCCCCCATGGAGCACAGGCAGACAGAAGTCCCCGCC
CAGCTGTGTGGCCTCAAGCCAGCCTTCCGCTCCTTGAAGCTGGTCTCCACACAGTGCTGGTTCCGTCACCCCTCCCAAGGAAGTAGG
TCTGAGCAGCTTGTCTGGCTGTGTCCATGTCAGAGCAACGGCCCAAGTCTGGGTCTGGGGGGGAAGGTGTCATGGAGCCCCCTACGA
TTCCCAGTCGTCCTCGTCCTCCTCTGCCTGTGGCTGCTGCGGTGGCGGCAGAGGAGGGATGGAGTCTGACACGCGGGCAAAGGCTCCT
CCGGGCCCCTCACAGCCCCAGGTCTTTCCCAGAGATGCCTGGAGGGAAAAGGCTGAGTGAGGGTGGTTGGTGGGAAACCCTGGTTC
CCCCAGCCCCCGGAGACTTAAATACAGGAAGAAAAAGGCAGGACAGAATTACAAGGTGCTGGCCCAGGGCGGGCAGCGGCCCTGCCTC
CTACCCTTGCGCCTCATGACCGGAGCCATAGCCCAGGCAGGAGGGCTGAGGACCTCTGGTGGCGGCCAGGGCTTCCAGCATGTGCC
TAGGGGAAGCAGGGGCCAGCTGGCAAGAGCAGGGGGTGGGCAGAAAGCACCCGGTGGACTCAGGGCTGGAGGGGAGGAGGCGATCTTG
CCAAGGCCCTCCGACTGCAAGCTCCAGGGCCCGCTCACCTTGCTCCTGCTCCTTCTGCTGCTGCTTCTCCAGCTTTCGCTCCTTCAT
GCTGCGCAGCTTGGCCTTGCCGATGCCCCAGCTTGGCGGATGGACTCTAGCAGAGTGGCCAGCCACCGGAGGGGTCAACCACTTCCC

How existing (germline) variation is inherited

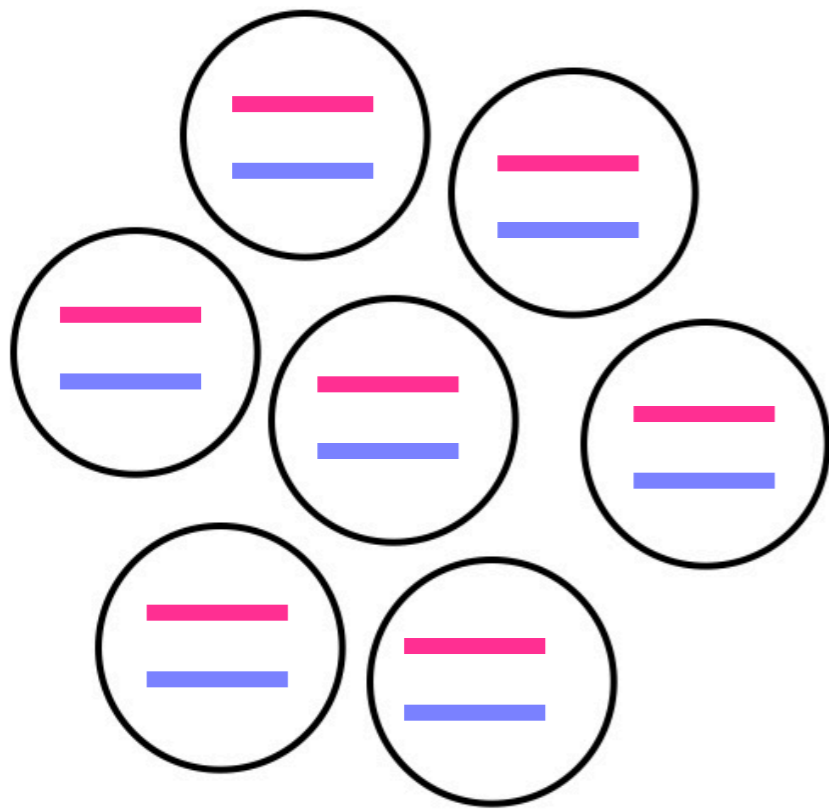


Recall: we are diploid.

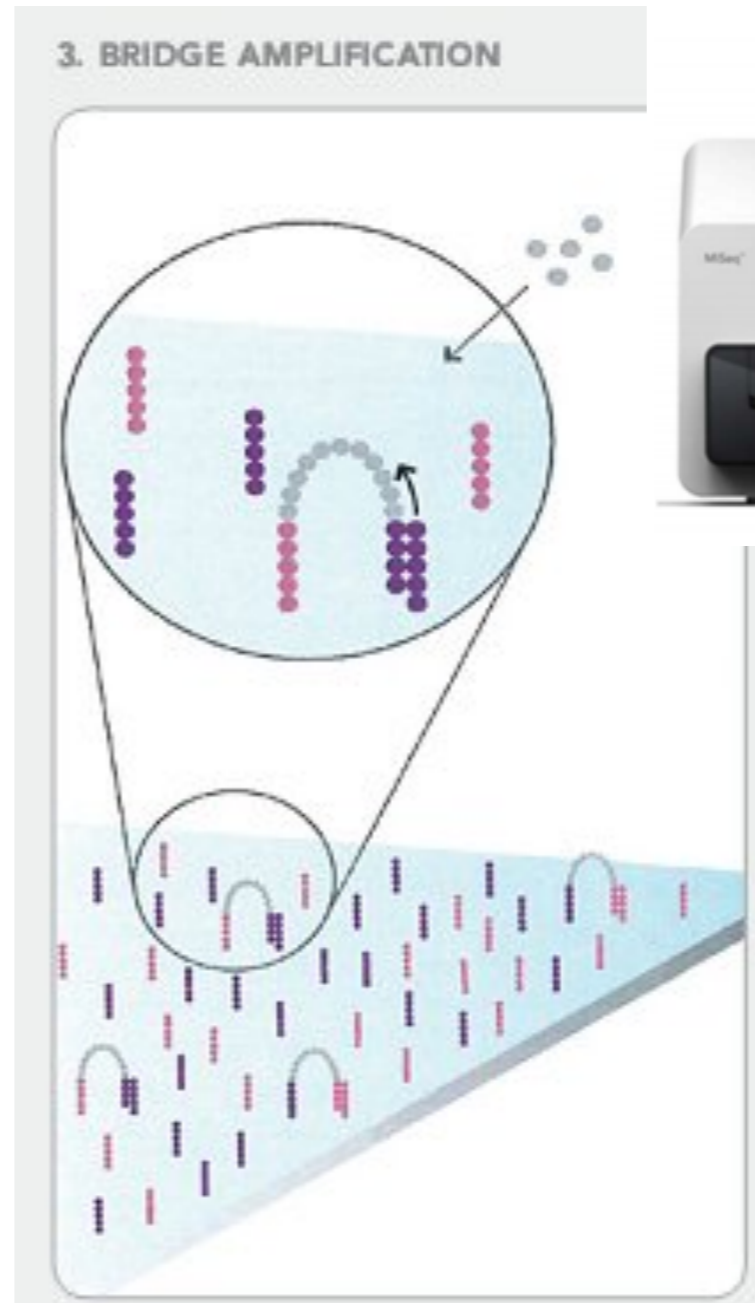


Aligned DNA
sequence
“reads”

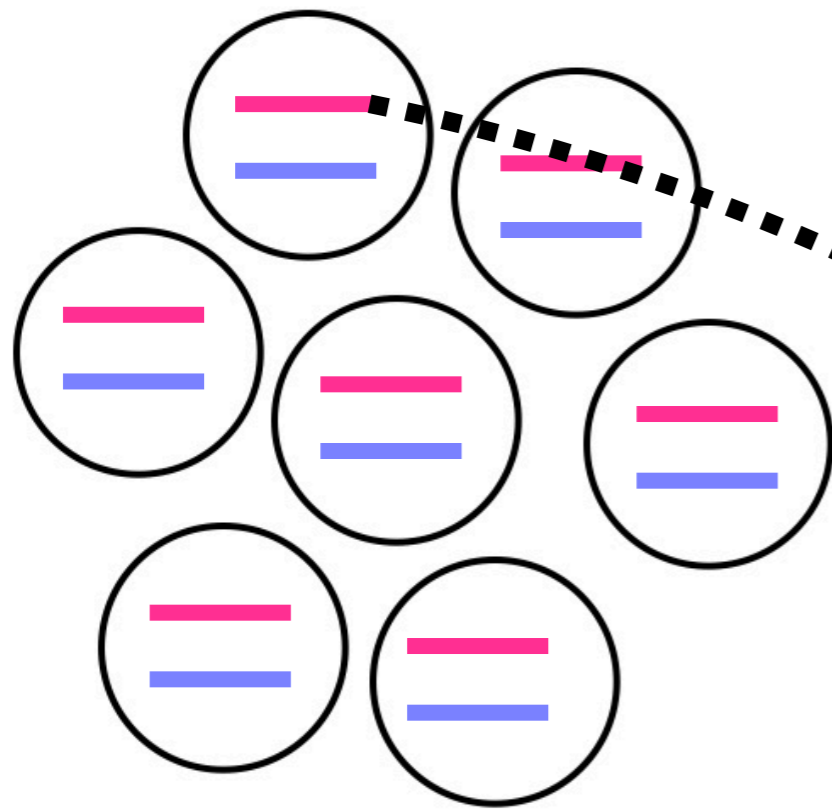
Each sequencing read is a piece of a parental allele from a single cell



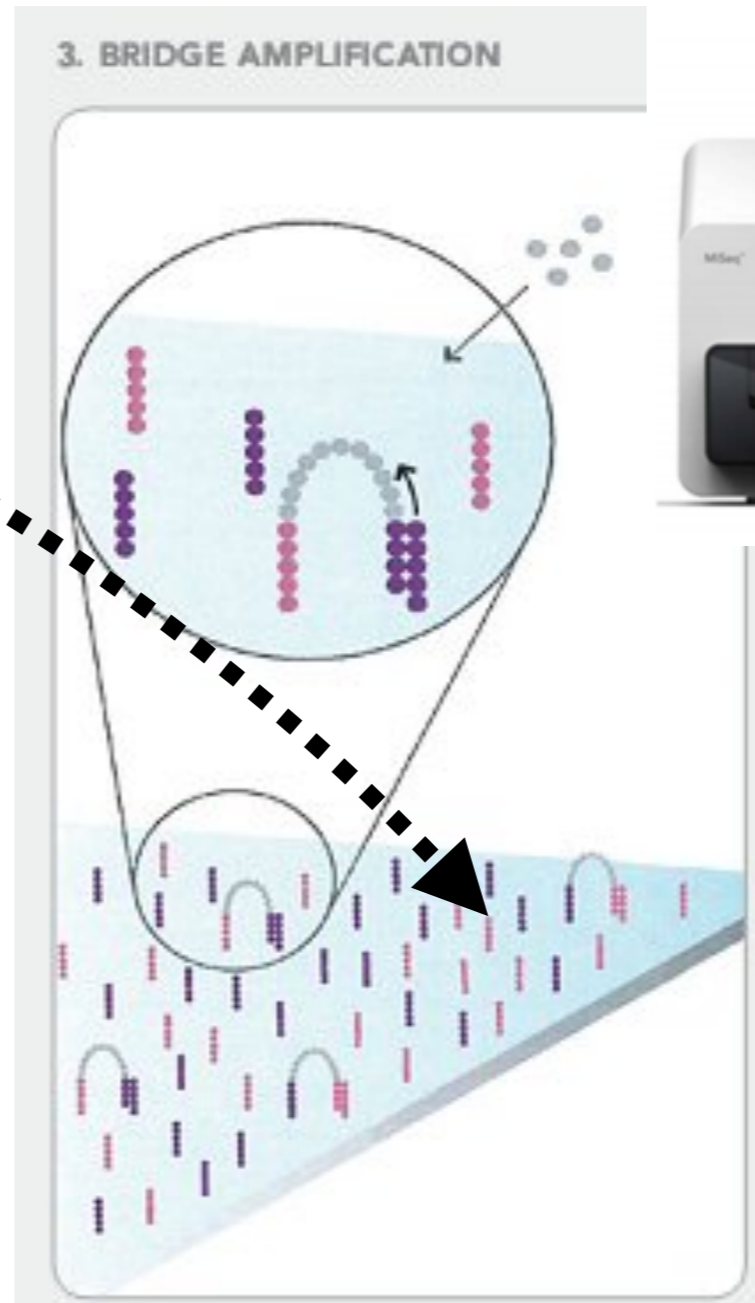
— Paternal
— Maternal



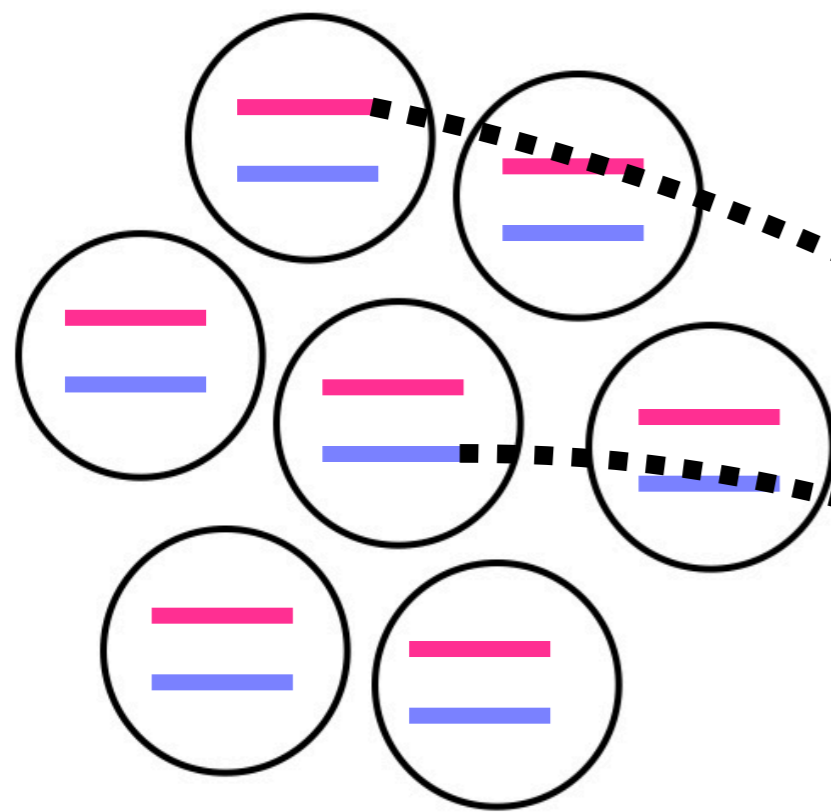
Each sequencing read is a piece of a parental allele from a single cell



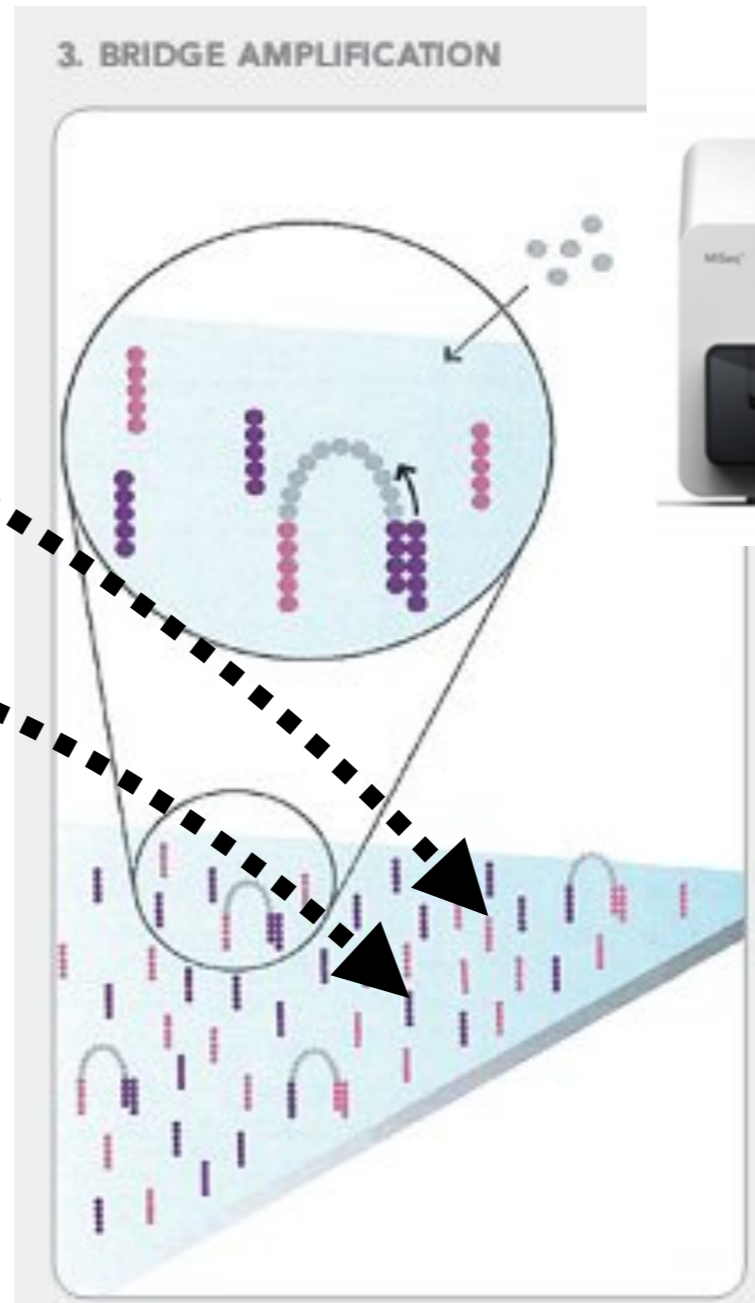
— Paternal
— Maternal



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.



Genomic DNA
from millions of cells



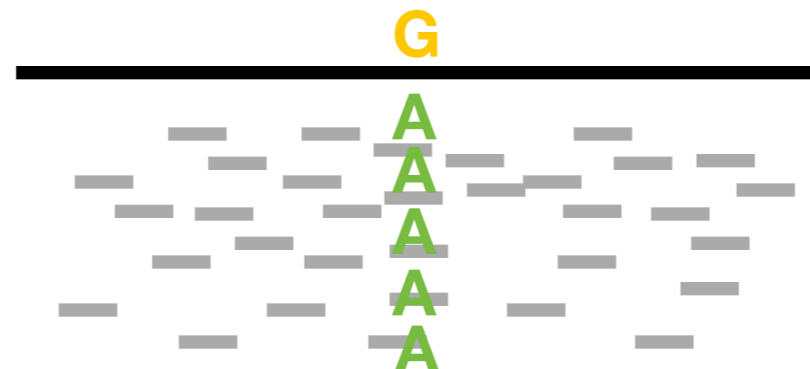
Fragmented DNA



Sequence billions of
DNA fragments
from millions of cells.

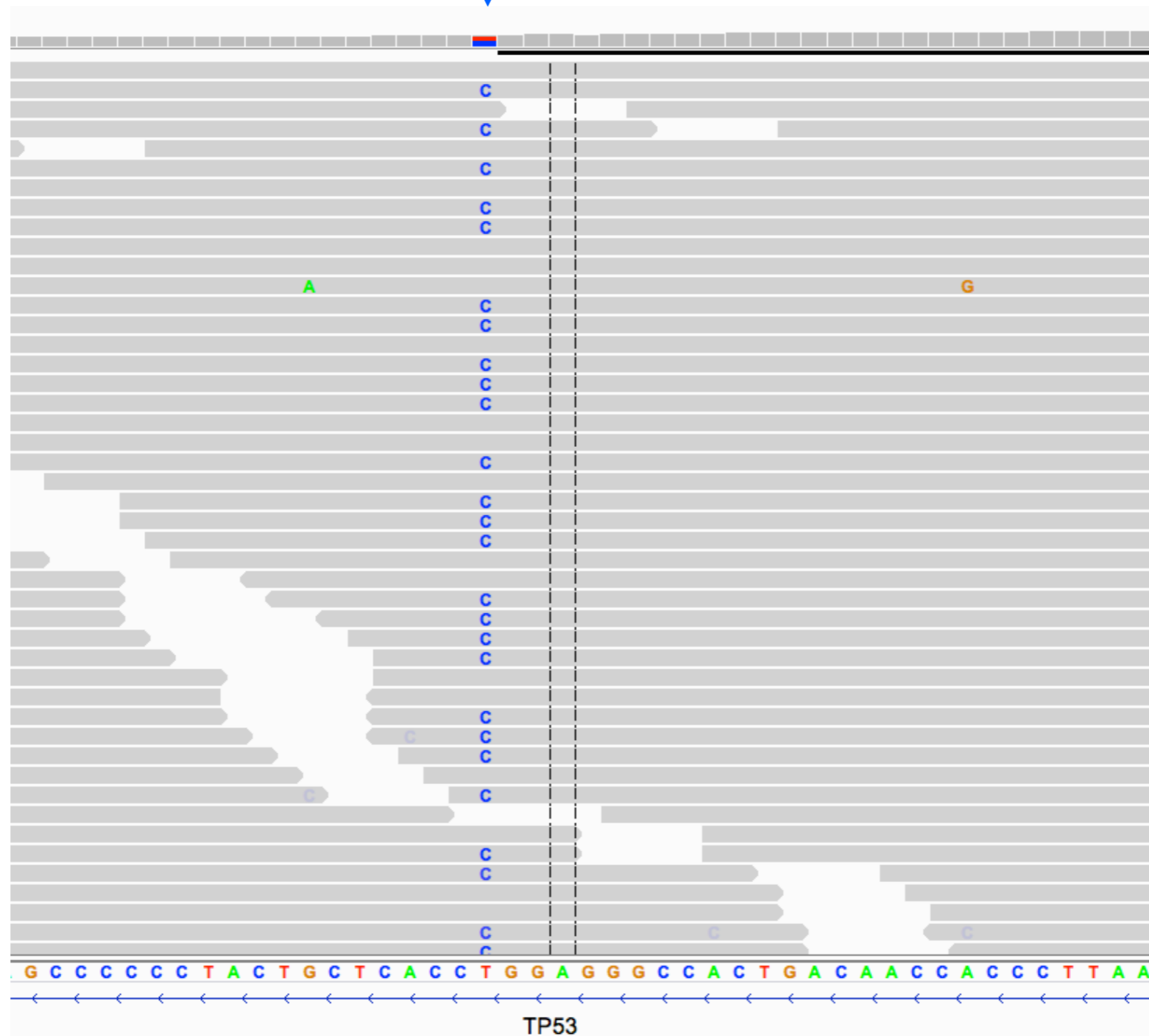


Align DNA to a reference
genome. Comparing sample
DNA to reference reveals
genetic differences.



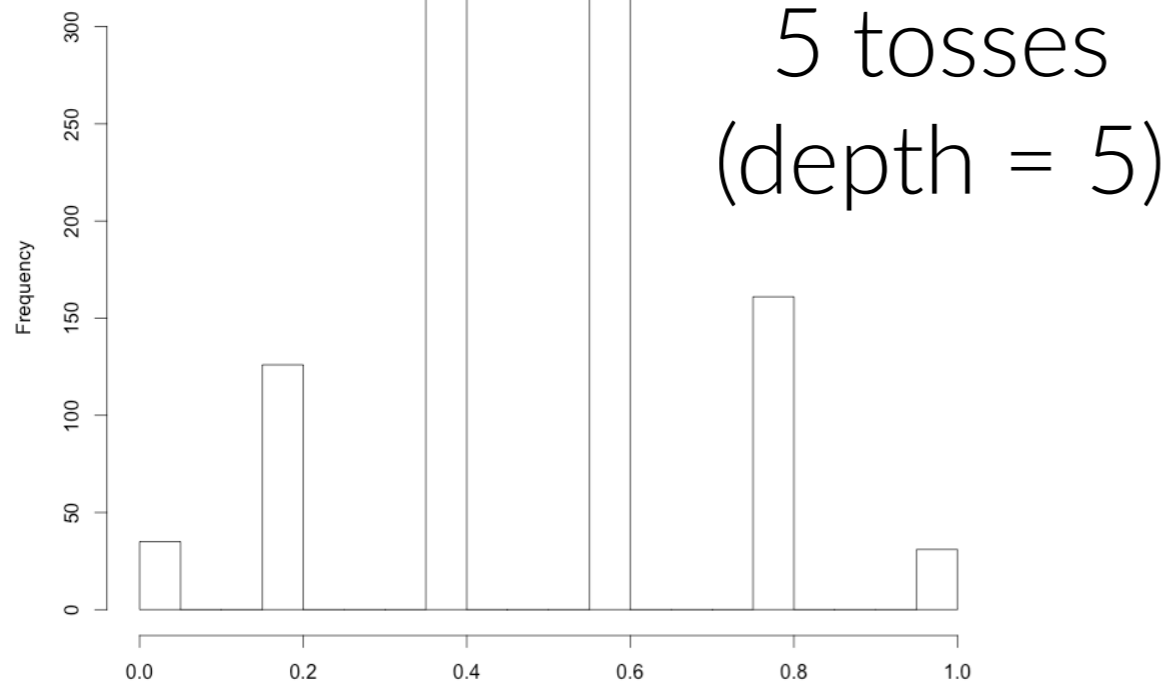
Heterozygotes: expect ~50/50 allele balance (binomial expectation). There are biases...

True variation. # of T = 19
Homoyzgous or heterozygous? # of C = 25

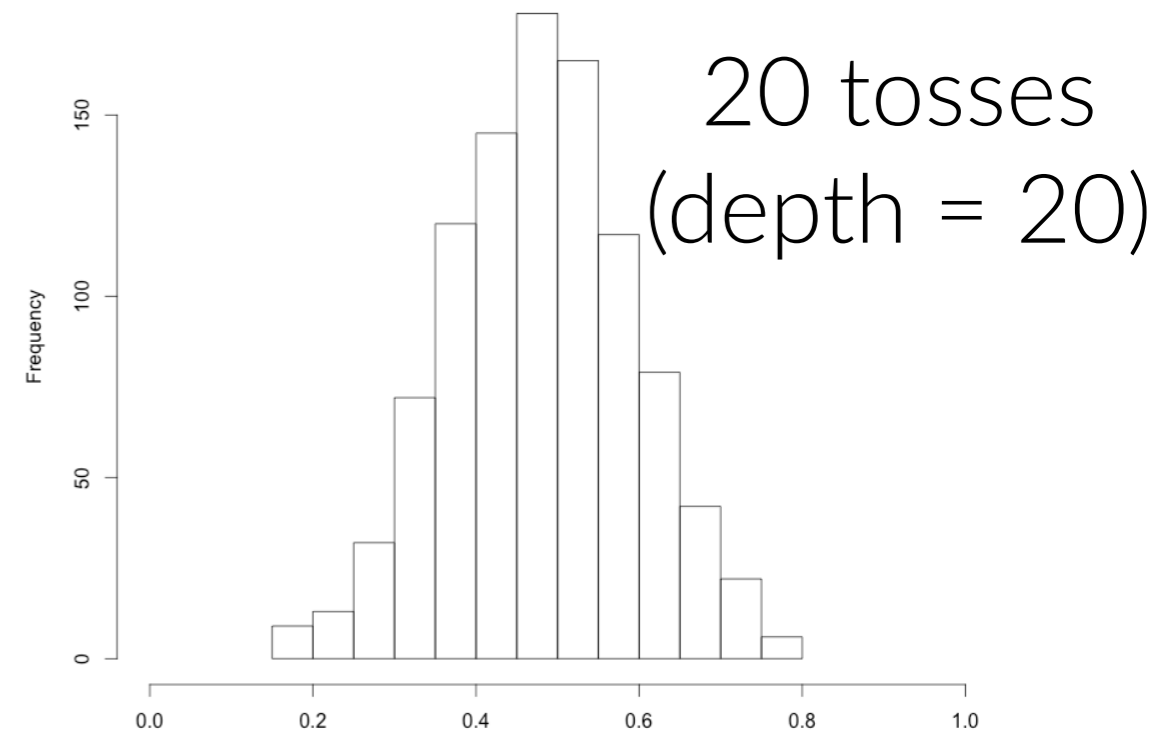


Allele sampling from NGS reads is akin to coin tosses. Deeper sampling “coverage” is better.

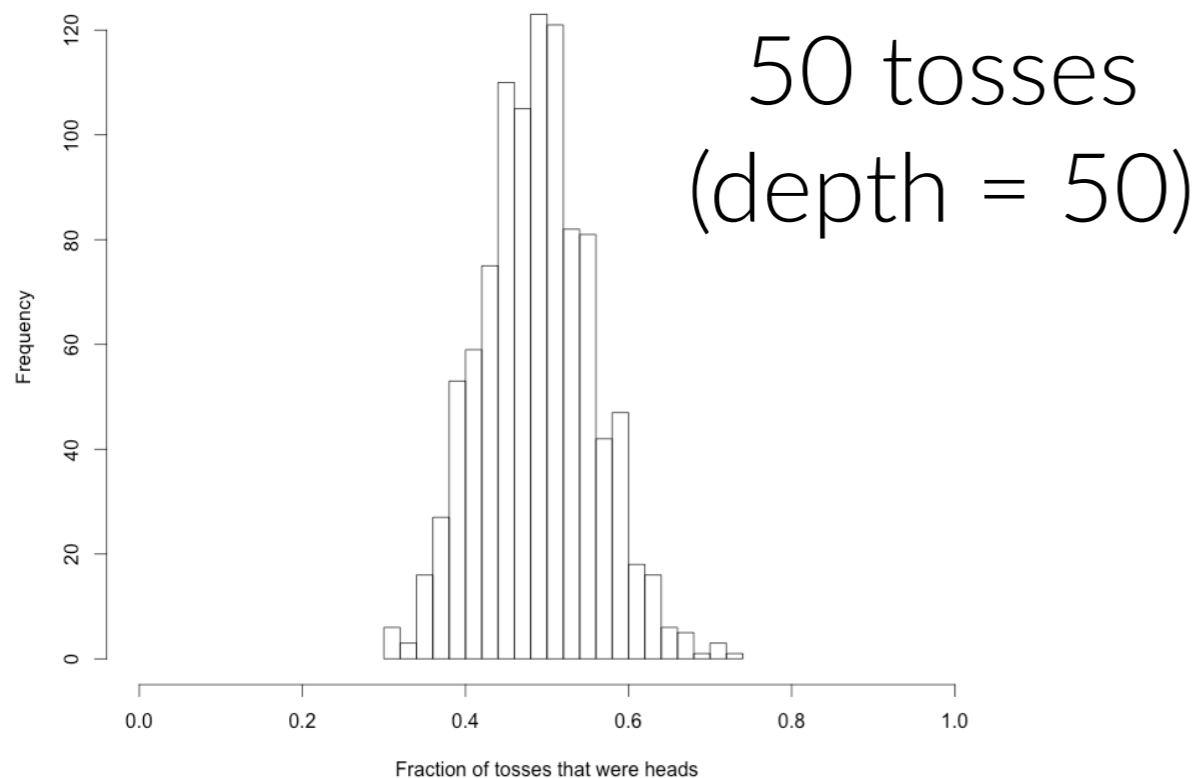
Distribution of % heads from 1000 experiments with 5 tosses each



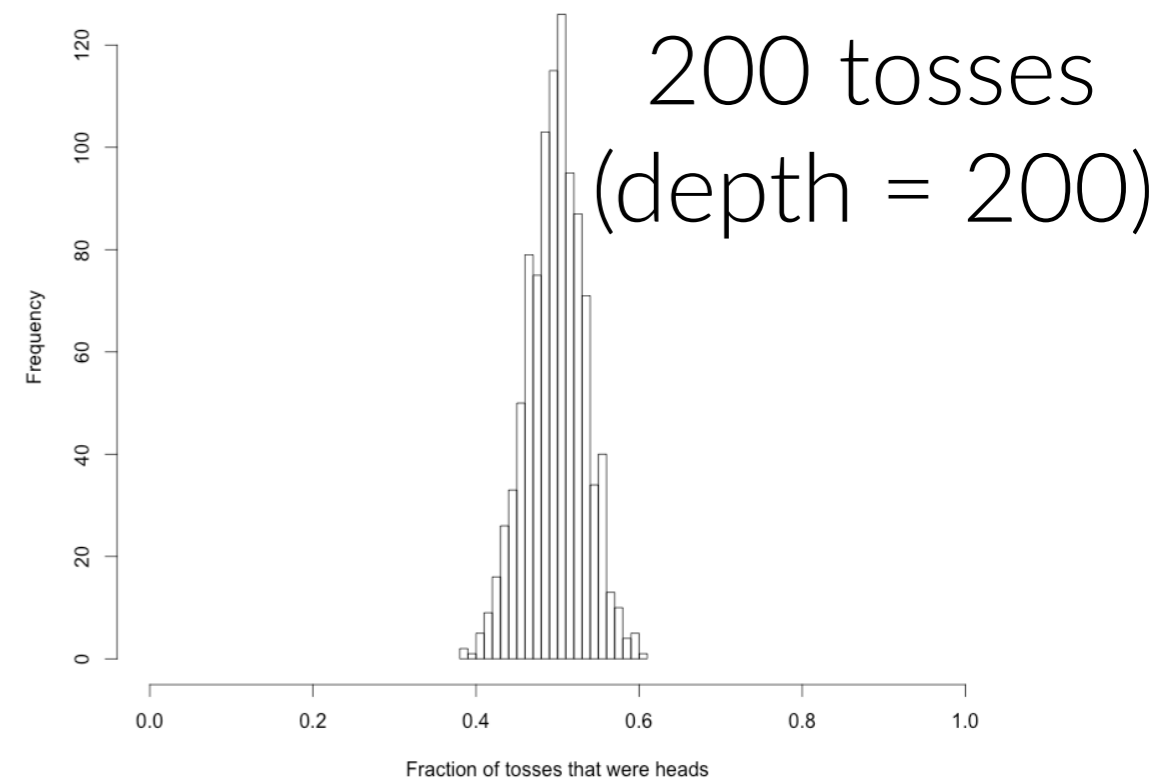
Distribution of % heads from 1000 experiments with 20 tosses each



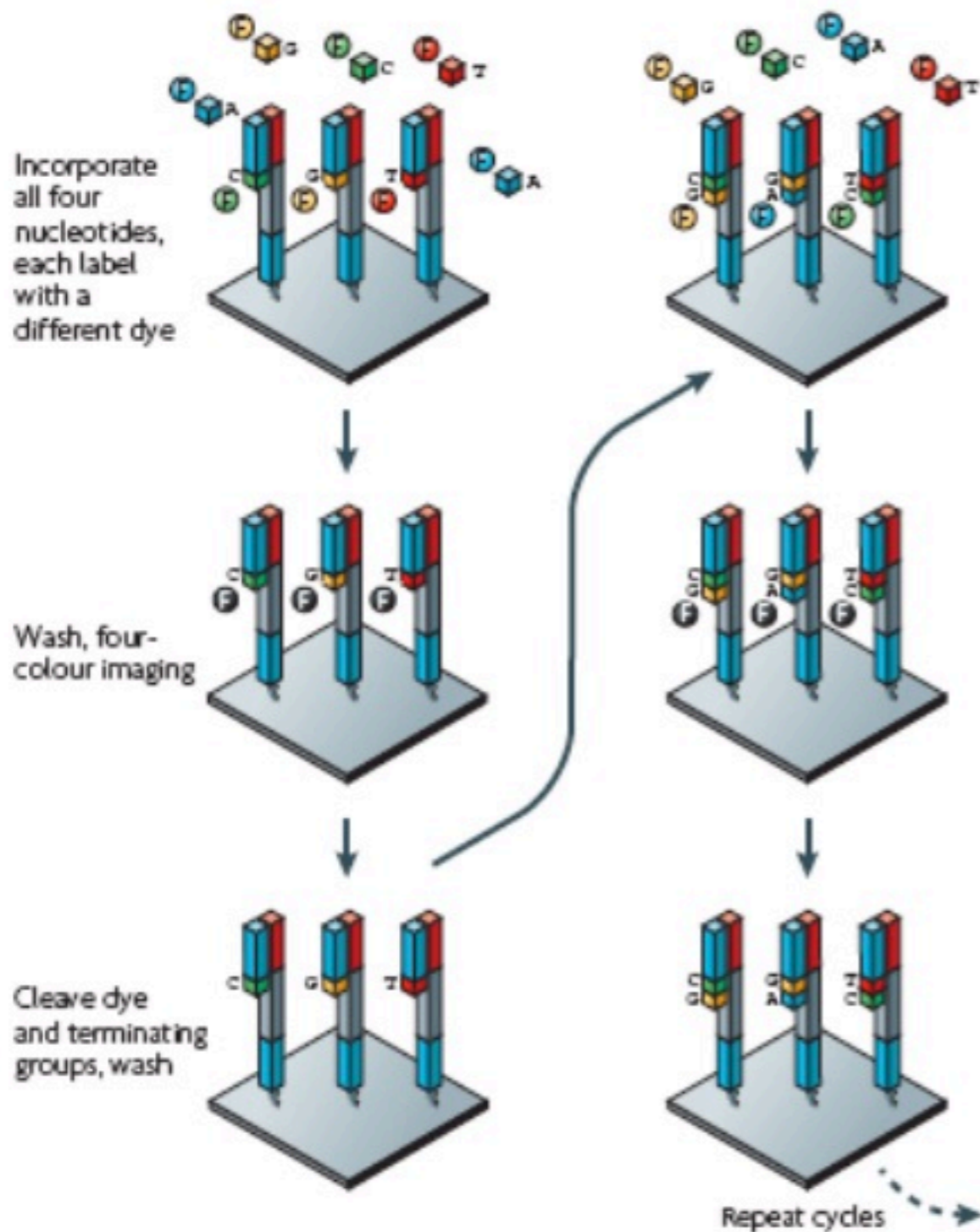
Distribution of % heads from 1000 experiments with 50 tosses each



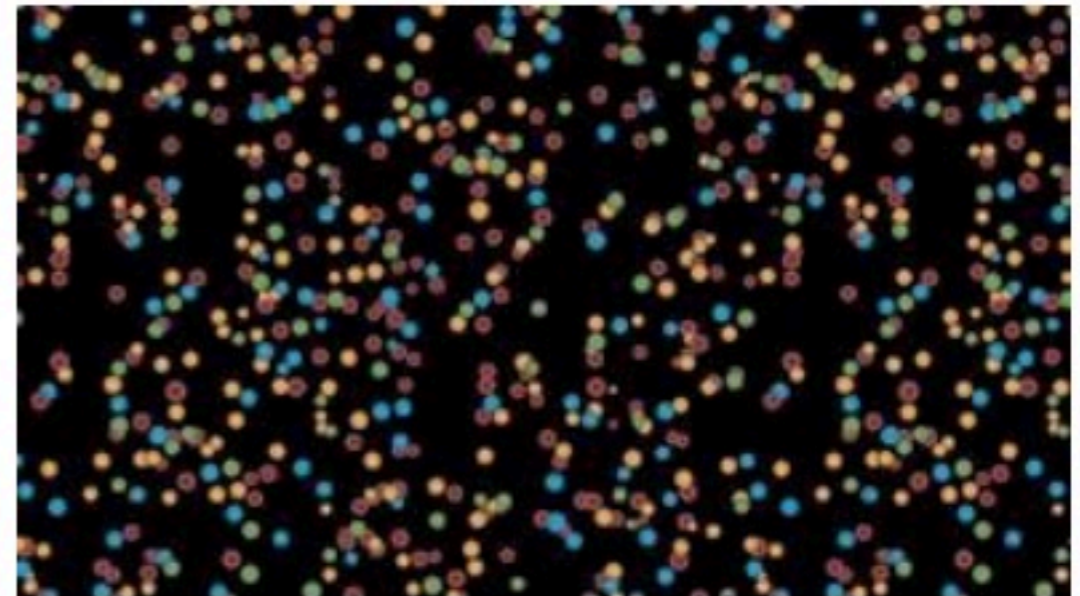
Distribution of % heads from 1000 experiments with 200 tosses each



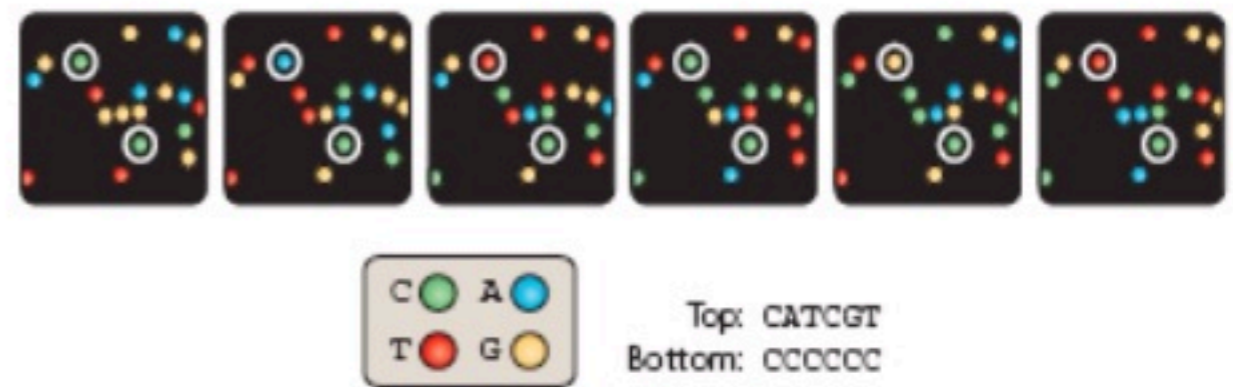
Not surprising: Solexa sequencing is ~stargazing.
Think about this. It is amazing. Error rate is ~0.1%!!!



4 images merged



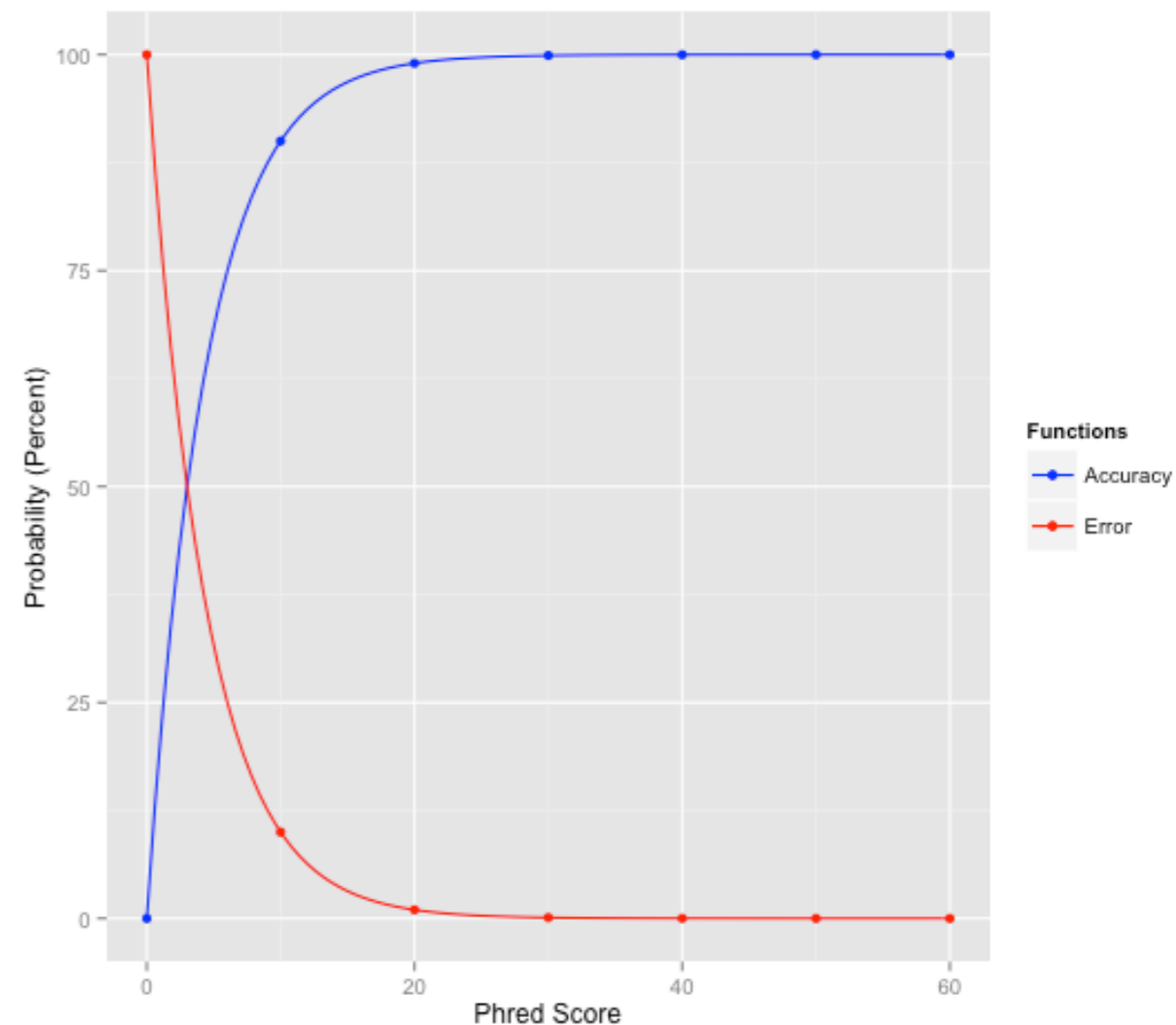
6 cycles w/ base-calling



“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

1. The algorithm

```

TCTGACCAATCTA A A A A T A C C T G T G A T T A A
TCTGACCAATCTA A C A A T A C C T G T G A T T A A
TCTGACCAATCTA A C A A T A C C T G T G A T T A A
-----
TCTGACCAATCTA A A A A T A C C T G T G A T T A A
tctgaccaatctaa c a a t a c c t g t g a t t a a
    
```

probability of polymorphism

base call, base quality

a priori polymorphism rate

$$P(SNP) = \sum_{\text{all variable } S} \frac{P(S_1 | R_1) \frac{P(S_N | R_N)}{P_{prior}(S_N)} \cdot P_{prior}(S_1, \dots, S_N)}{\sum_{S_{i_1} \in \{A,C,G,T\}} \dots \sum_{S_{i_N} \in \{A,C,G,T\}} \frac{P(S_{i_1} | R_1)}{P_{prior}(S_{i_1})} \dots \frac{P(S_{i_N} | R_N)}{P_{prior}(S_{i_N})} \cdot P_{prior}(S_{i_1}, \dots, S_{i_N})}$$

base composition

depth of coverage

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, POLY-BAYES, 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 g g AATT
-----
TGAAA t GAATT
    
```

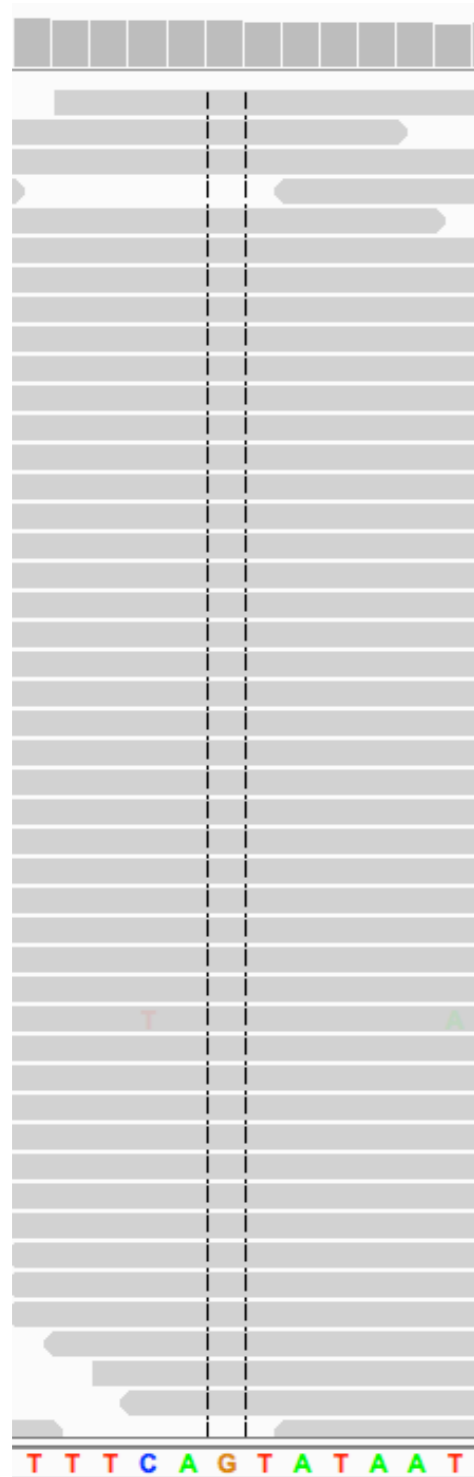
sequencing error

```

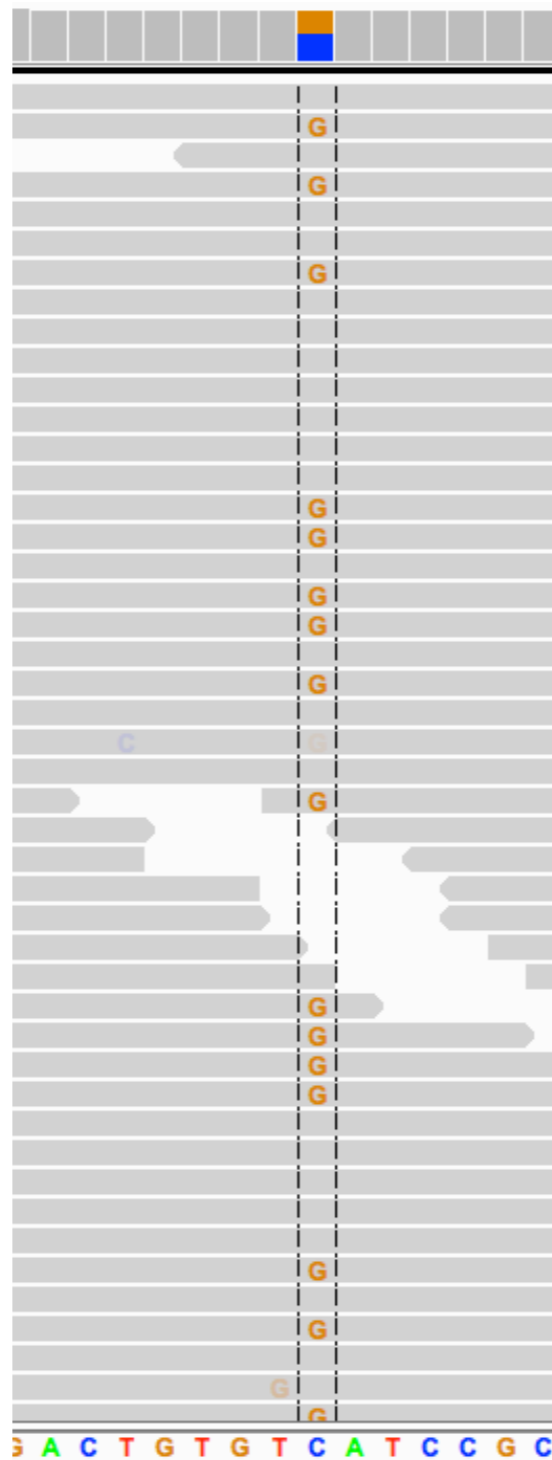
TTGAT C CCTGT
-----
TTGAT T CCTGT
    
```

true polymorphism

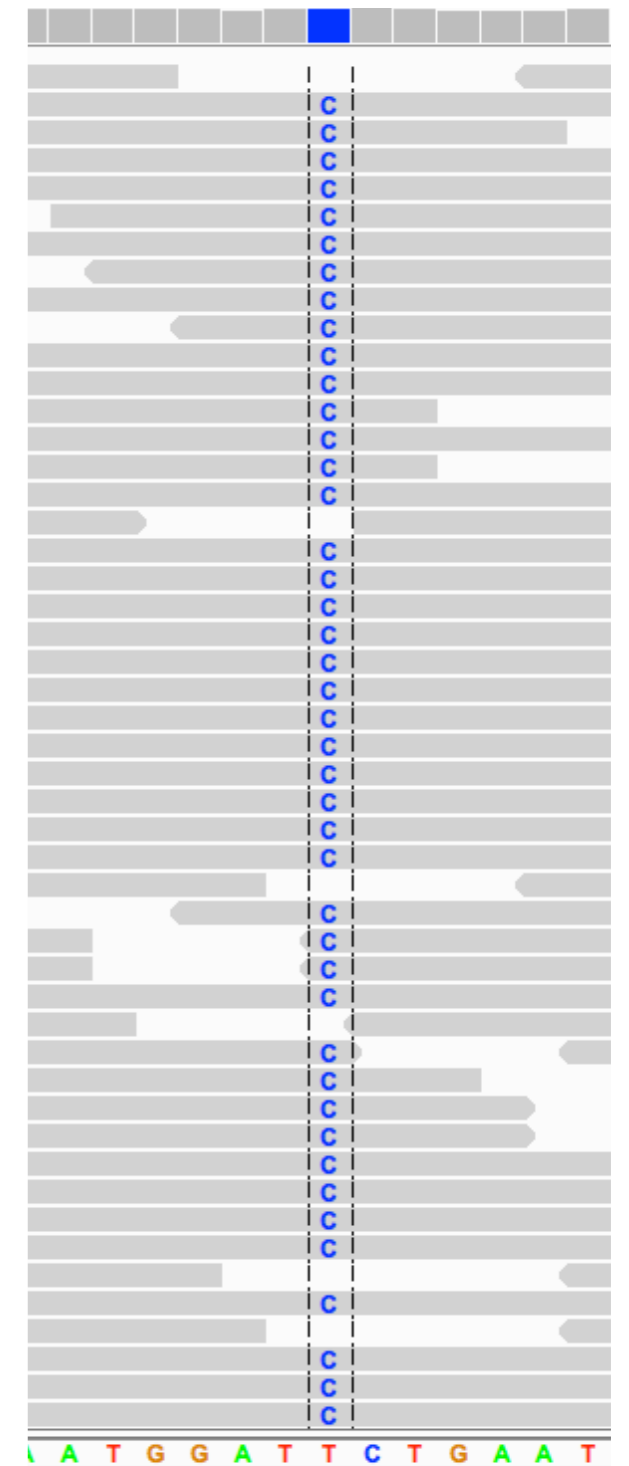
Different (diploid) SNP genotypes



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

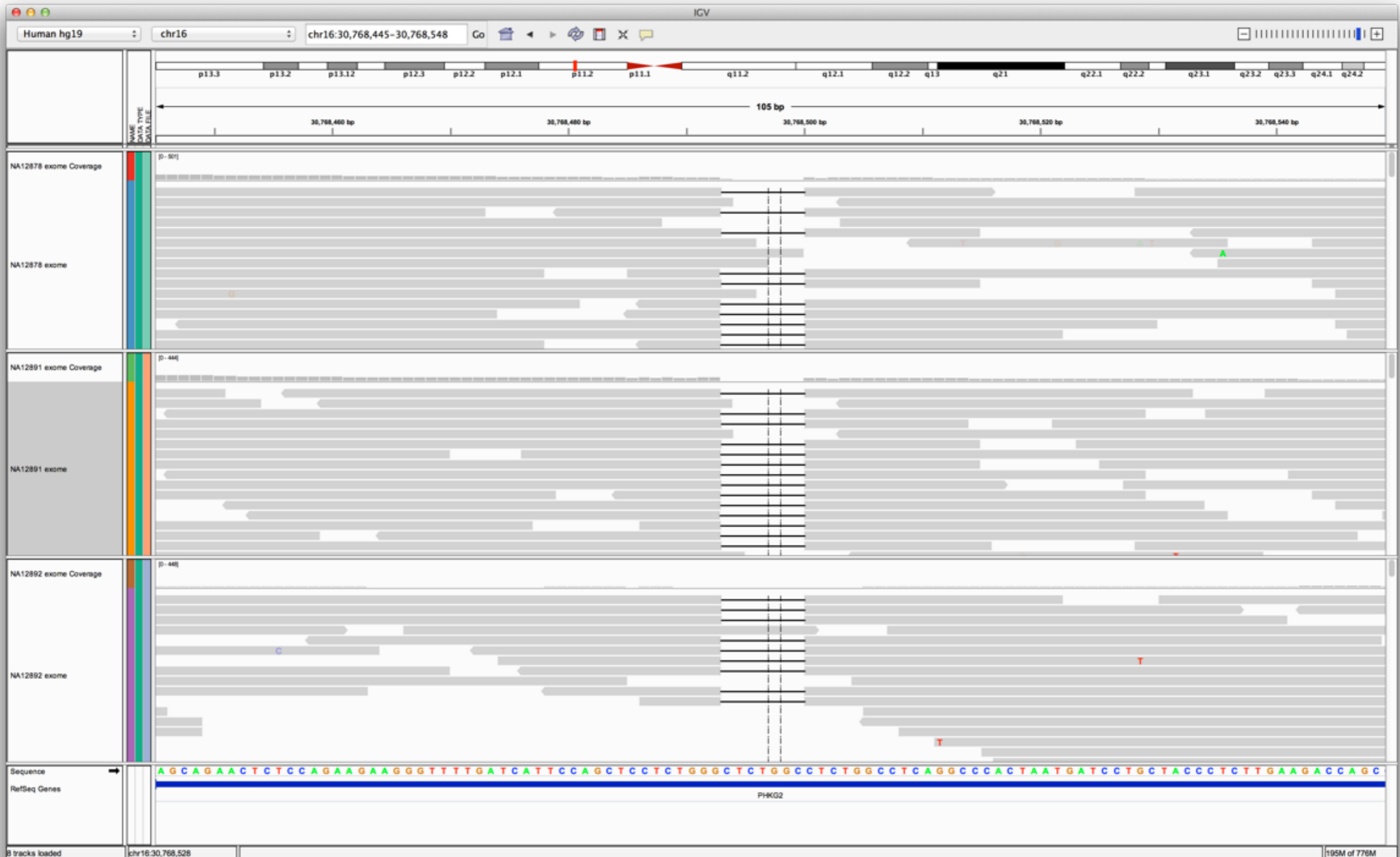


Heterozygous
(i.e., 1 chrom same as ref, 1 diff.)

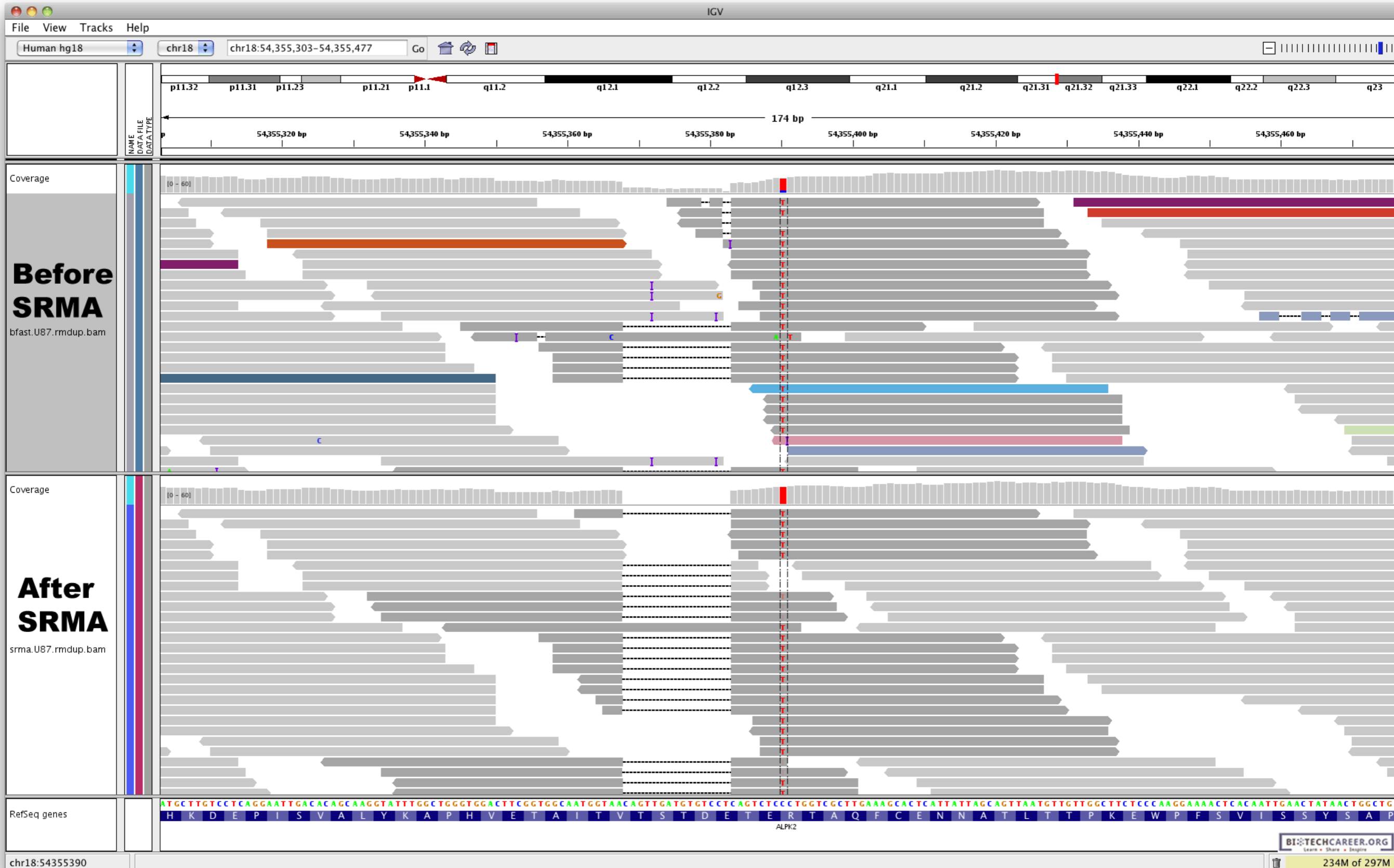


Homozygous for reference
(i.e., both chroms diff than ref)

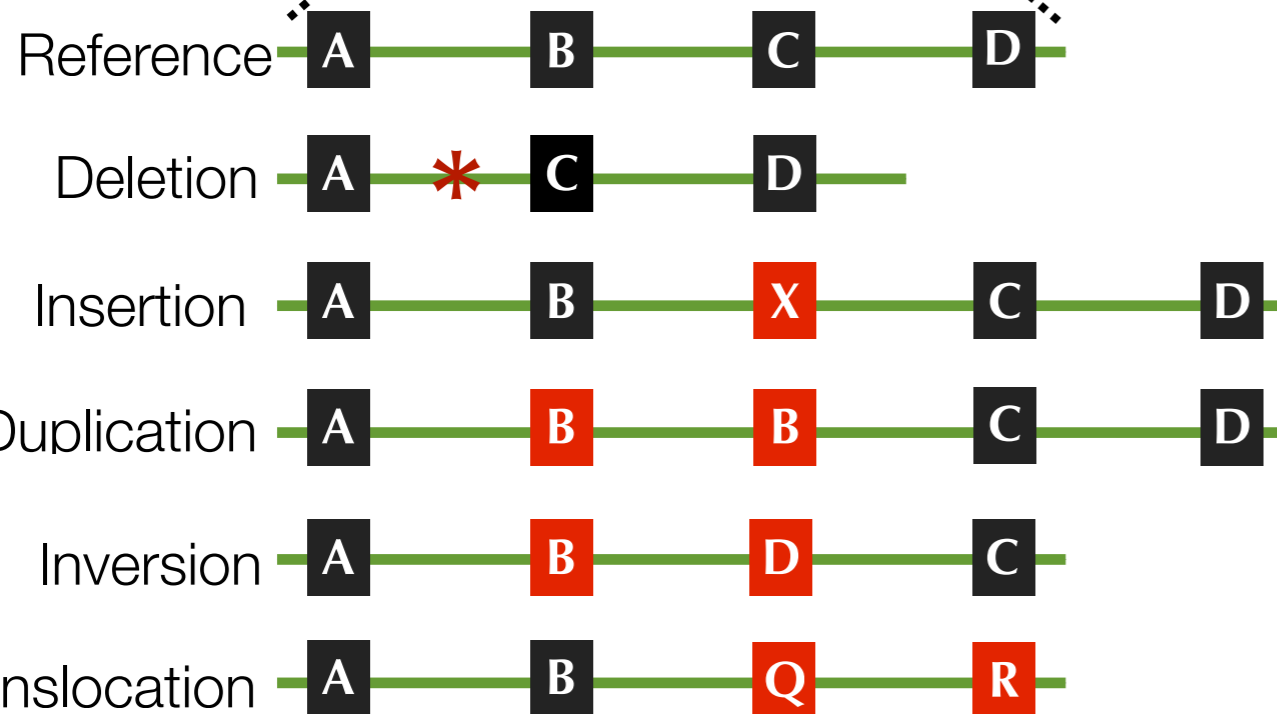
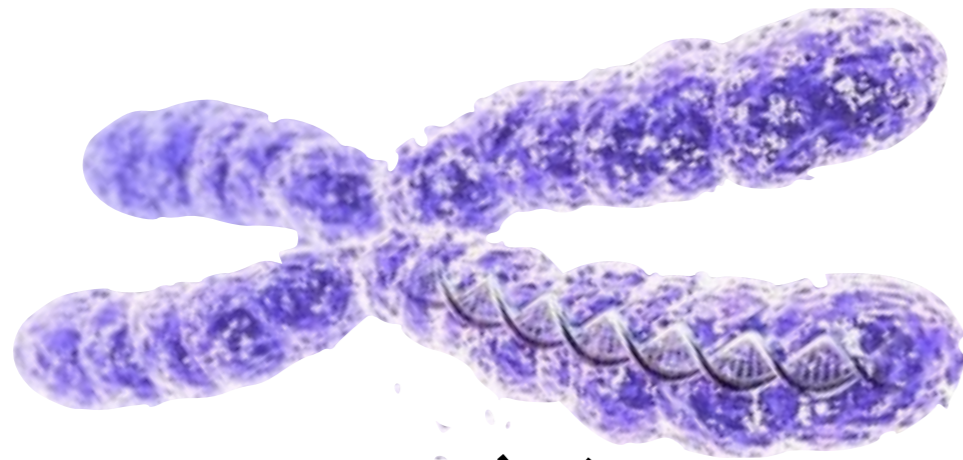
Insertion-deletion polymorphisms (INDELs)



Not always so simple...



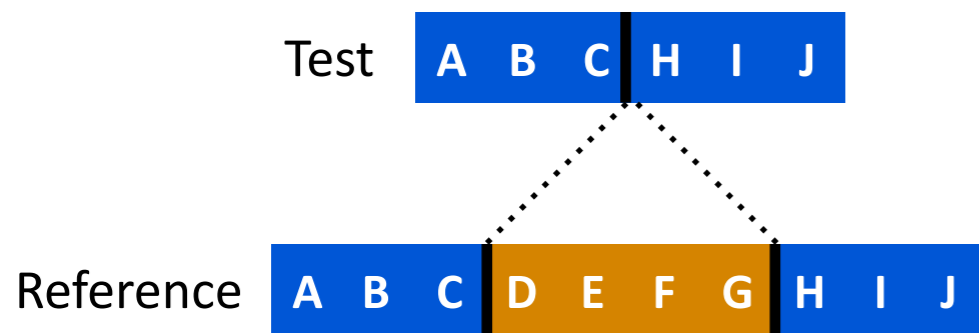
Structural Variation



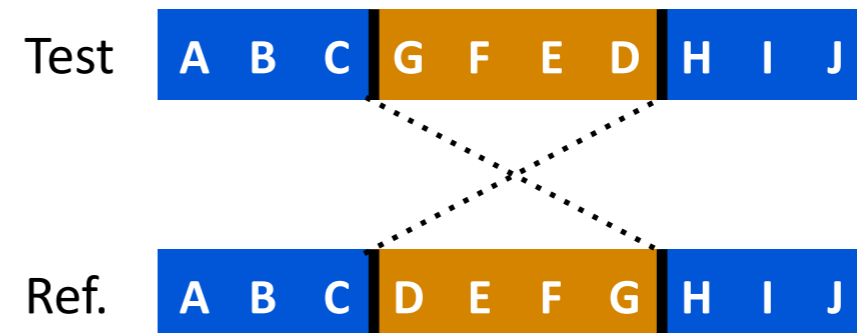
- large (>1kb) differences that affect the copy number, orientation, or location of genomic segments
- Common in mammalian genomes (~3-5 thousand between two people)
- A hallmark of cancer
- A major cause of spontaneous disease
- more are functional than SNPs
- very challenging to identify

SV Breakpoints

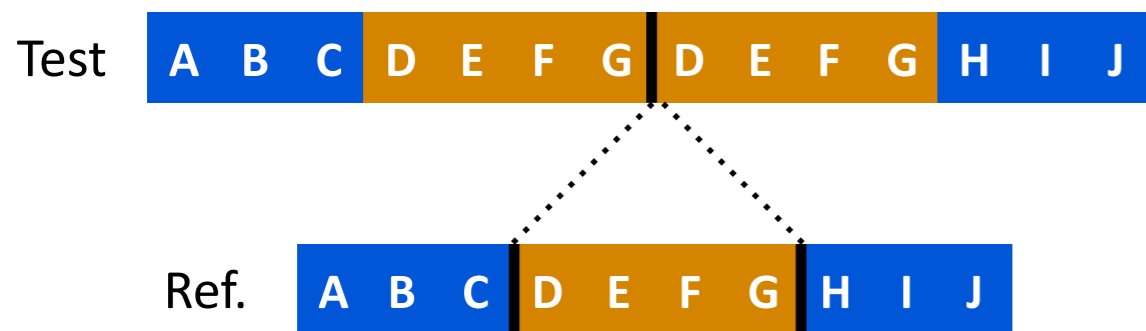
Deletion



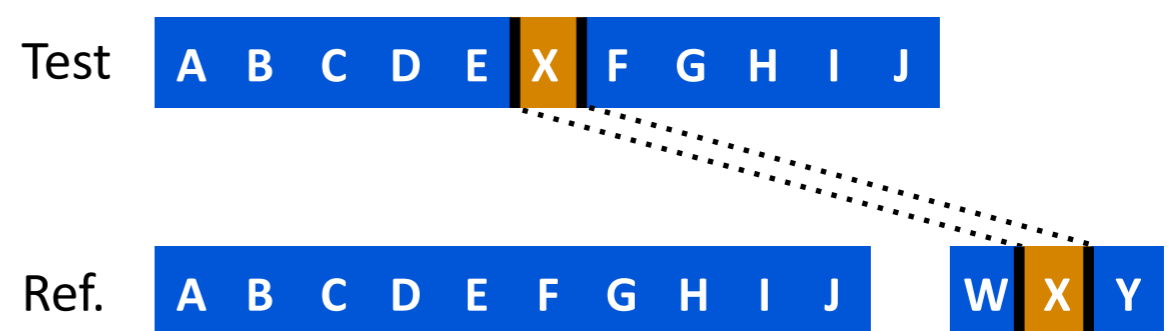
Inversion



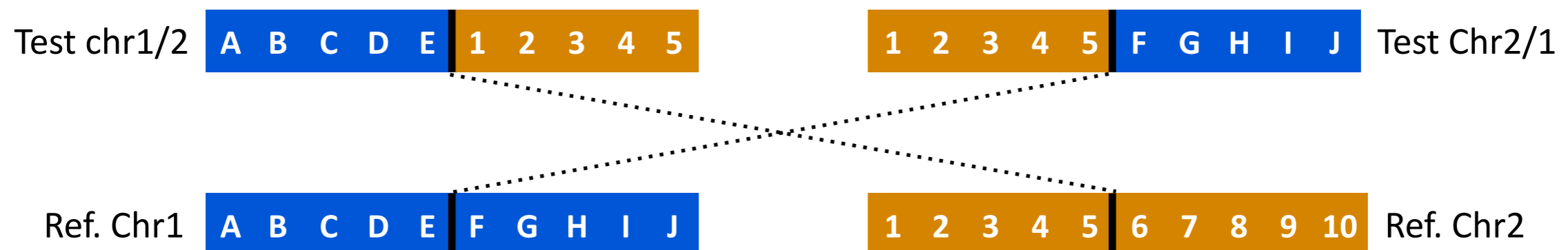
Tandem Duplication



Distant Insertion

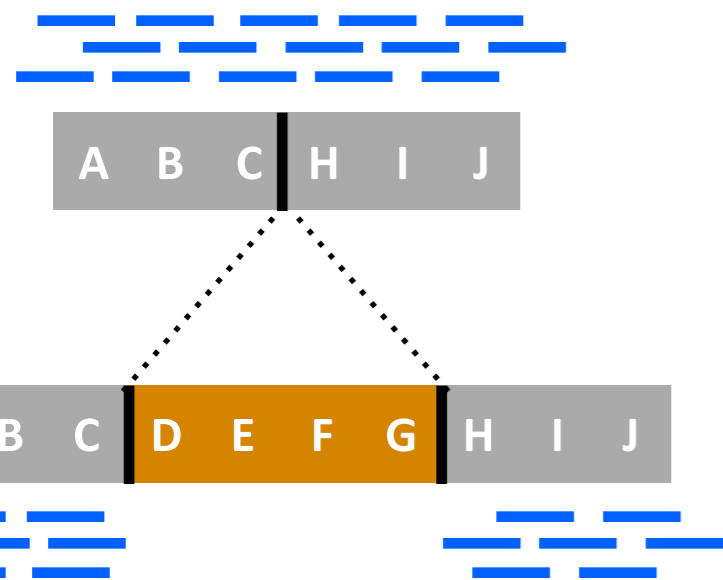


Reciprocal translocation



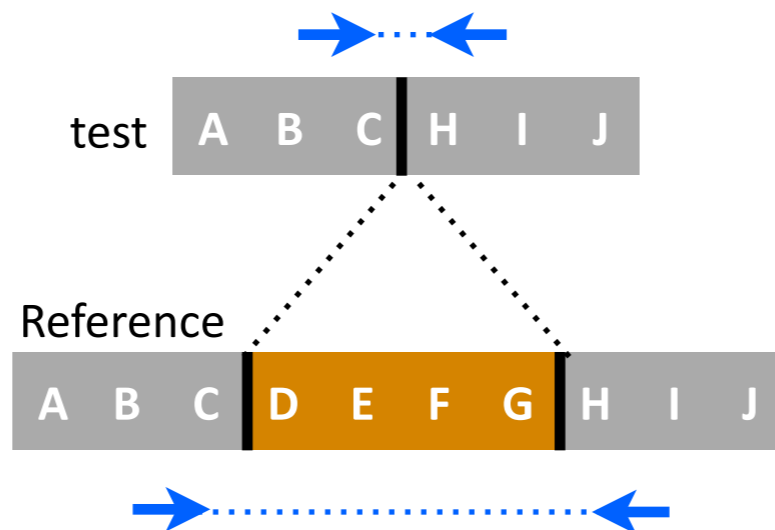
Detecting SVs from alignments

1) Read depth



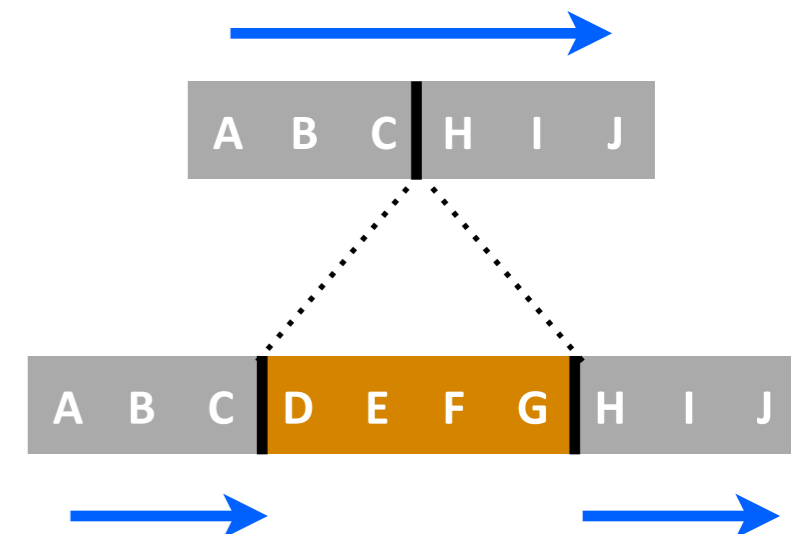
2) Paired-end mapping

paired-reads (or strobe)

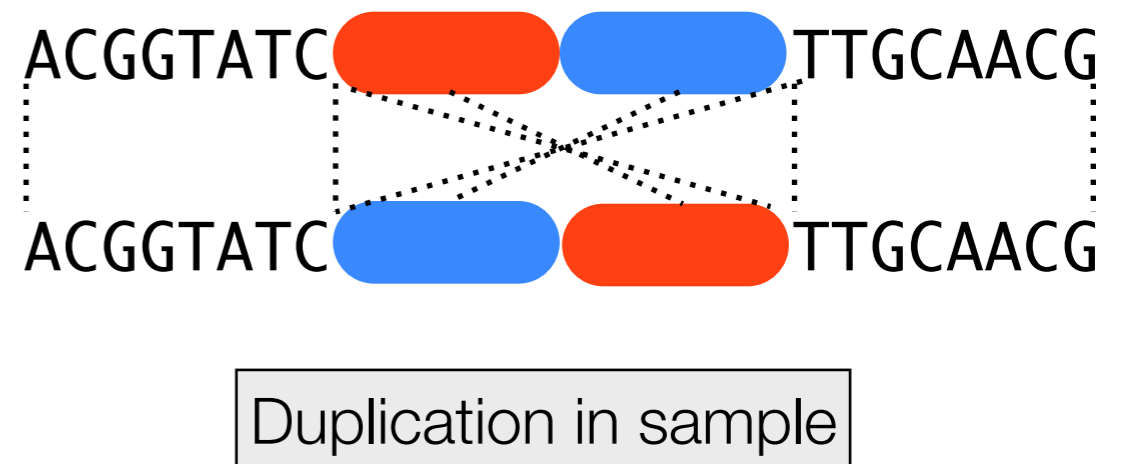
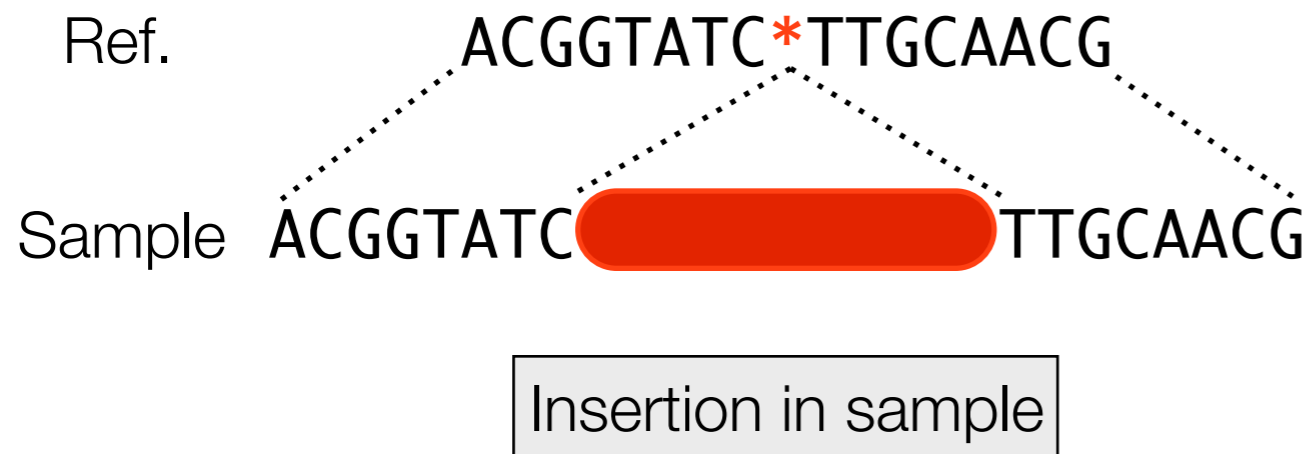
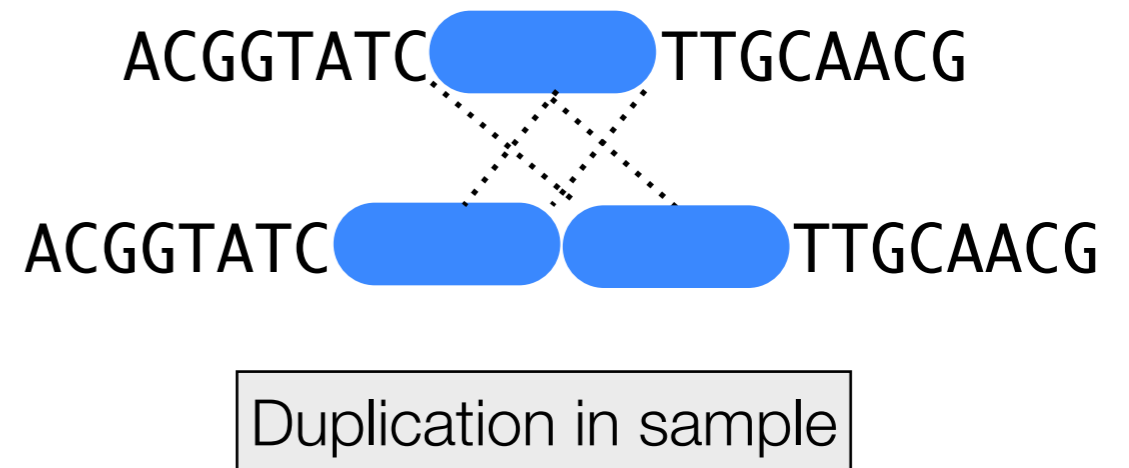
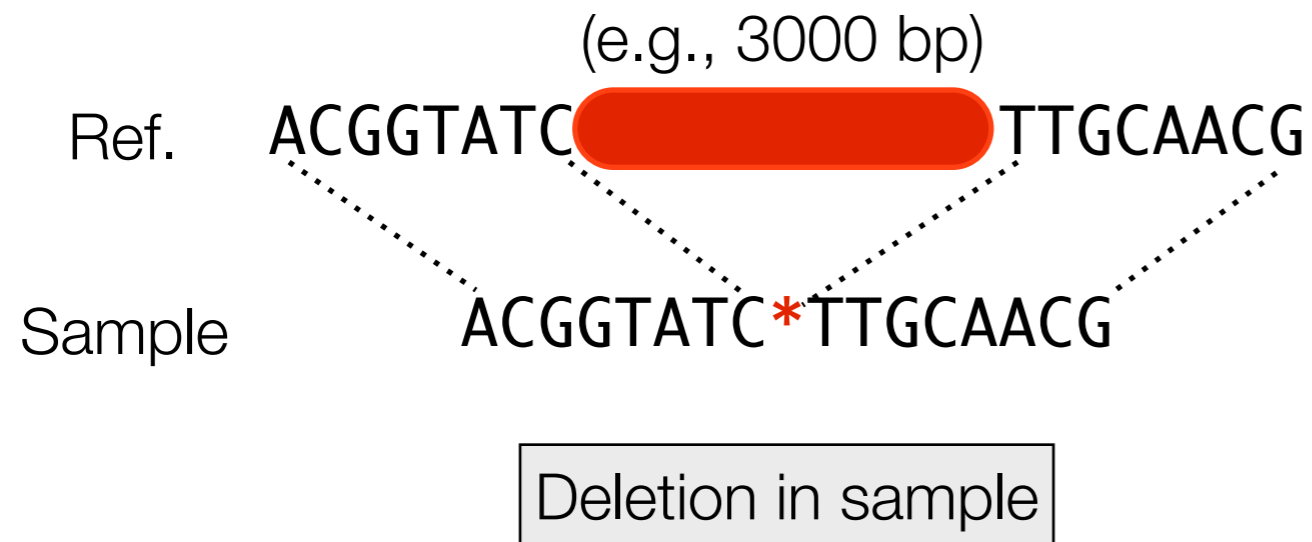


3) Split mapping

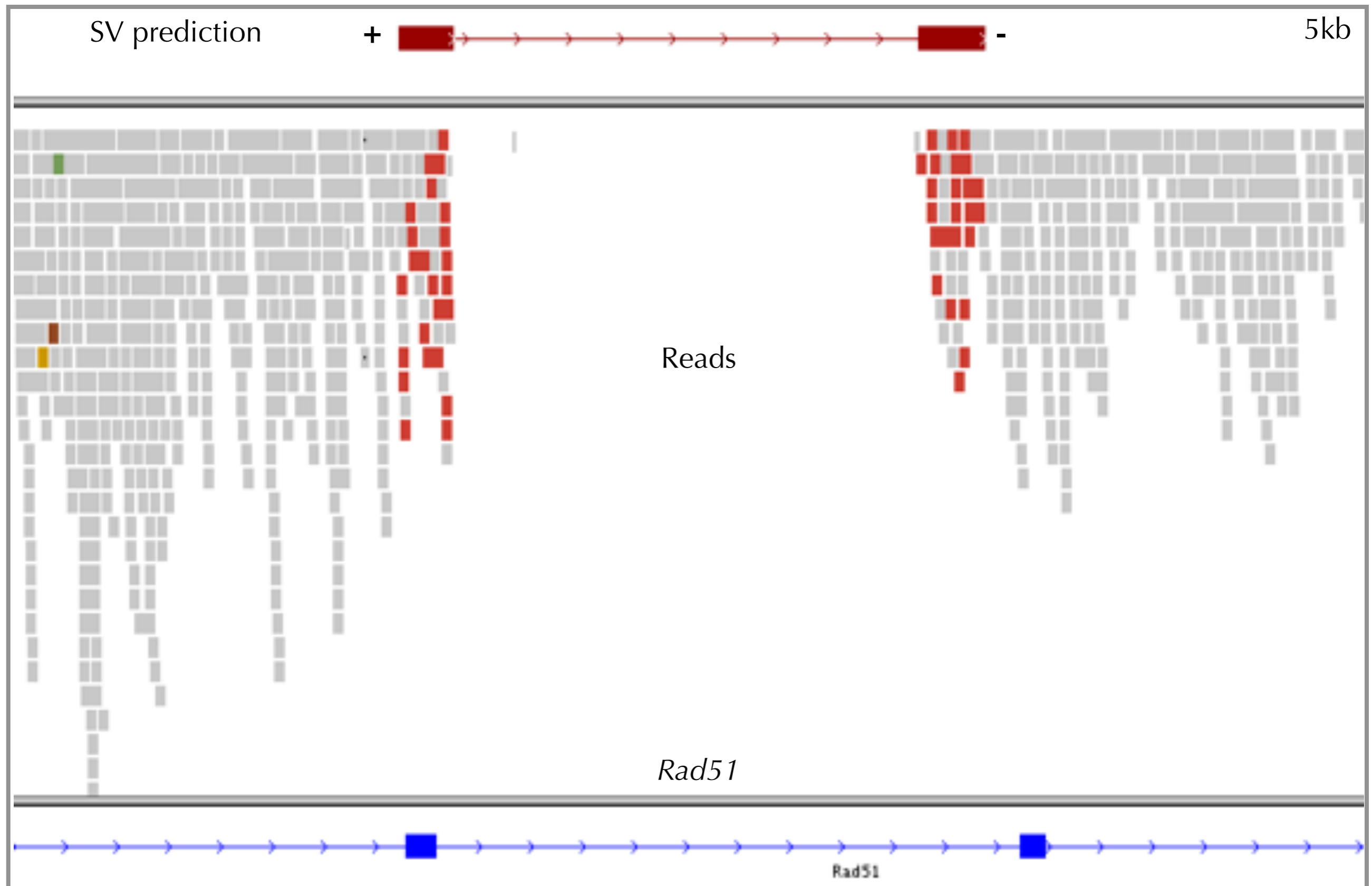
long reads or contigs



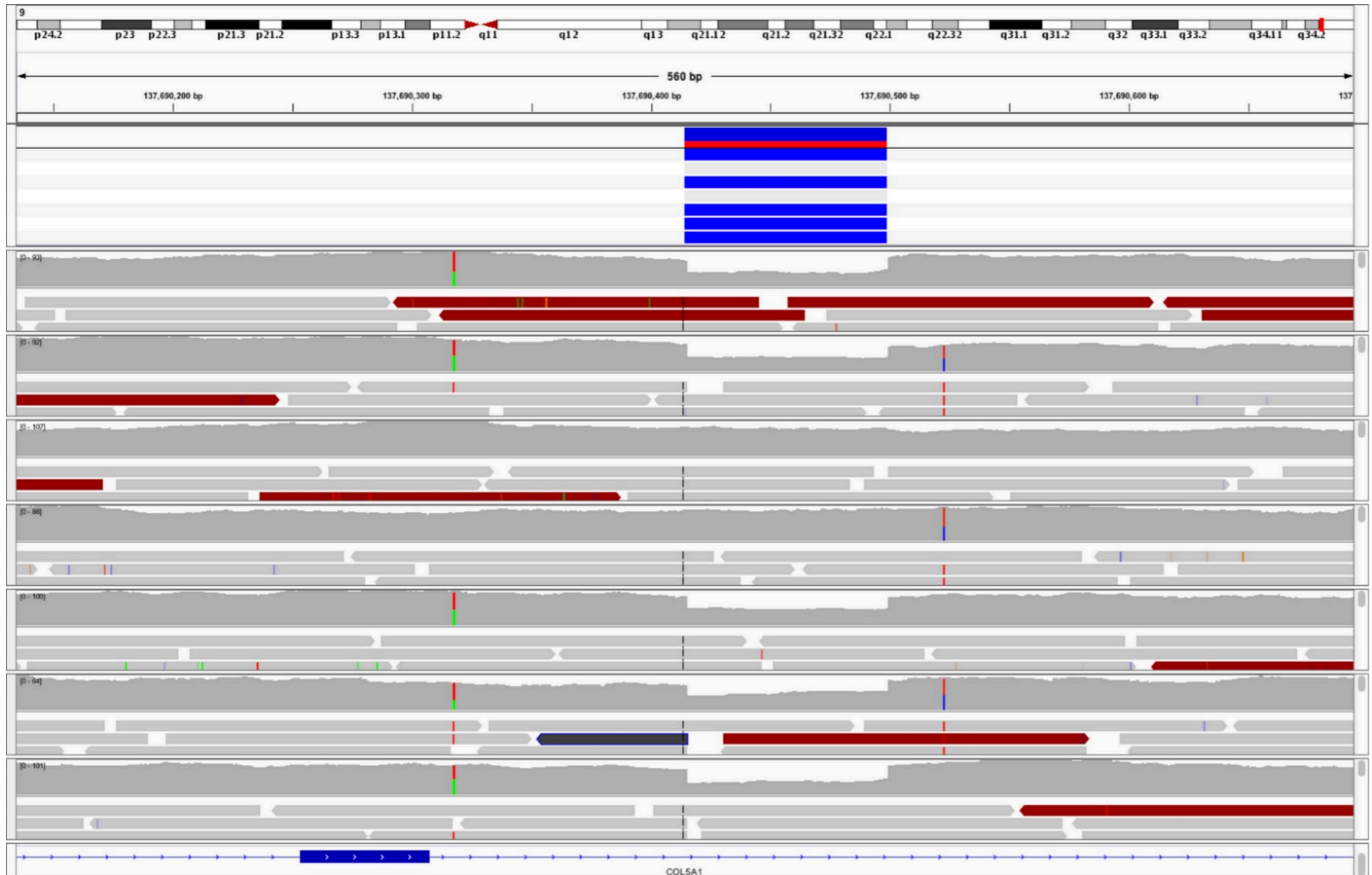
SV Alignment patterns



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.