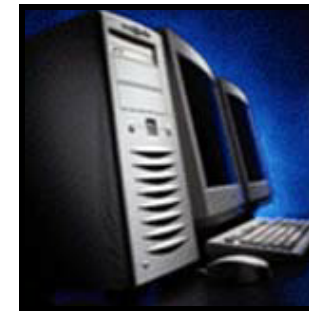
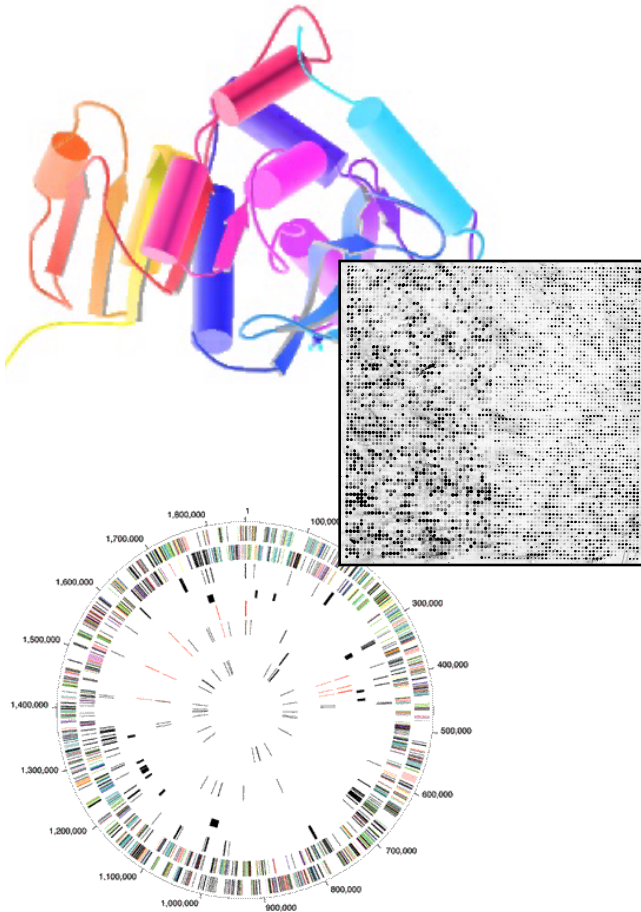


# Variant Identification, Focusing on SVs



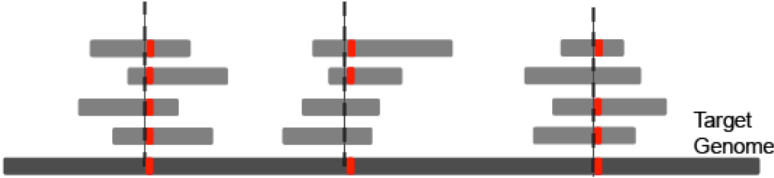
Mark Gerstein, Yale University  
[gersteinlab.org/courses/452](http://gersteinlab.org/courses/452)  
(last edit in spring '20, pack #6)

### Step 0: Generate Reads



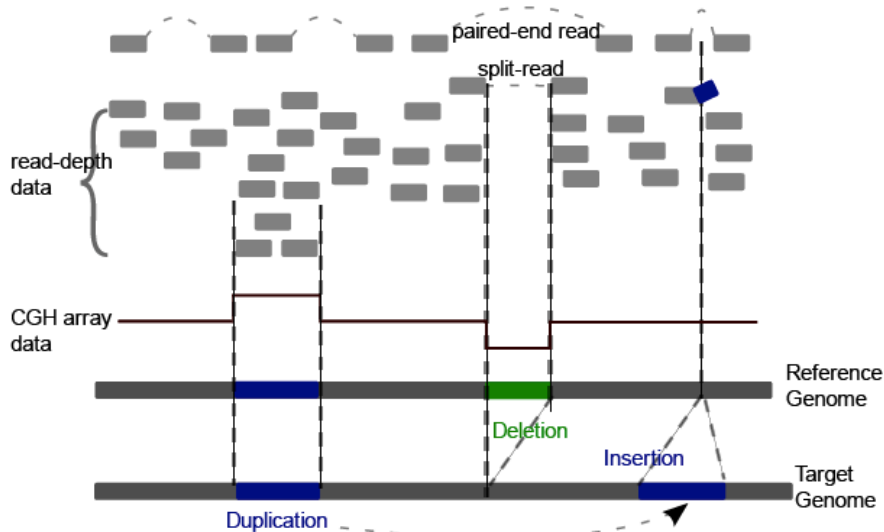
### Step 1: Call SNPs

using uniquely and correctly mapped reads



### Step 2: Find SVs

with aberrant paired-end reads, split-reads, read-depth analysis and CGH array data

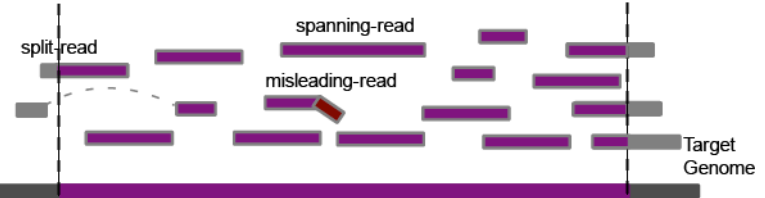


# Main Steps in Genome Resequencing

[Snyder et al. Genes & Dev. ('10)]

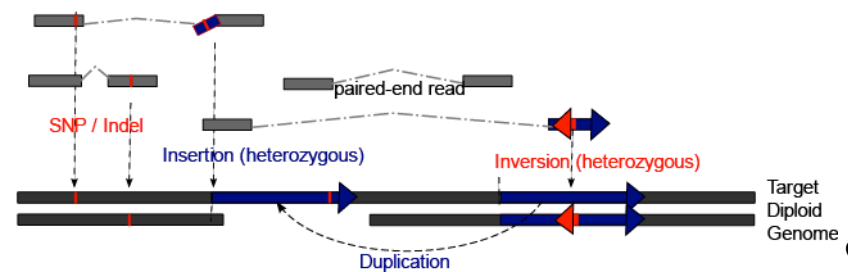
### Step 3: Assemble New Sequences

with split-, spanning- and misleading-reads



### Step 4: Phasing

mostly with paired-end reads



# Main Steps in Genome Resequencing

[Snyder et al. Genes & Dev. ('10)]

Step 0: Generate Reads



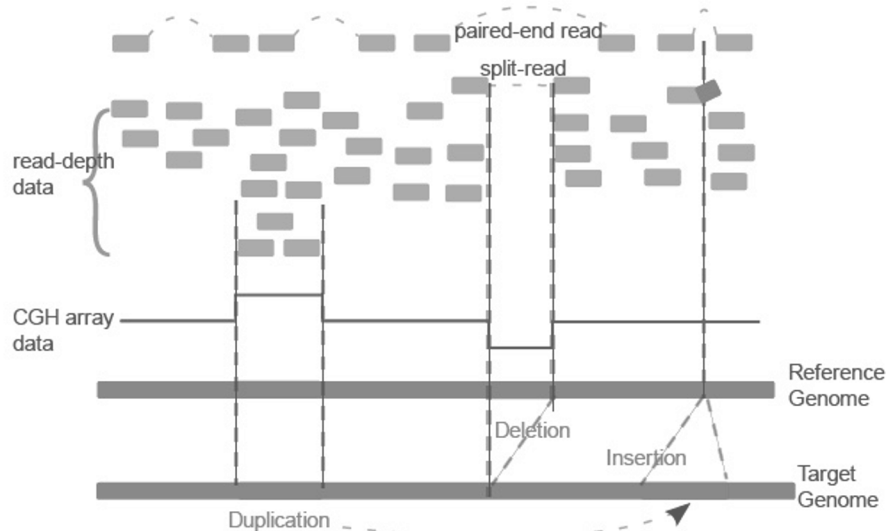
Step 1: Call SNPs

using uniquely and correctly mapped reads



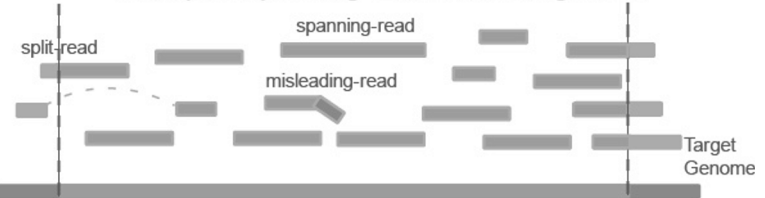
Step 2: Find SVs

with aberrant paired-end reads, split-reads, read-depth analysis and CGH array data



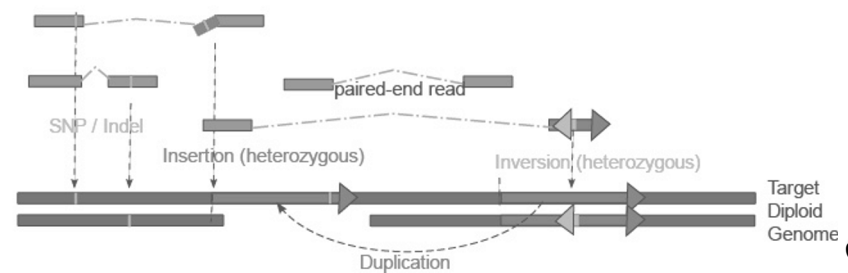
Step 3: Assemble New Sequences

with split-, spanning- and misleading-reads

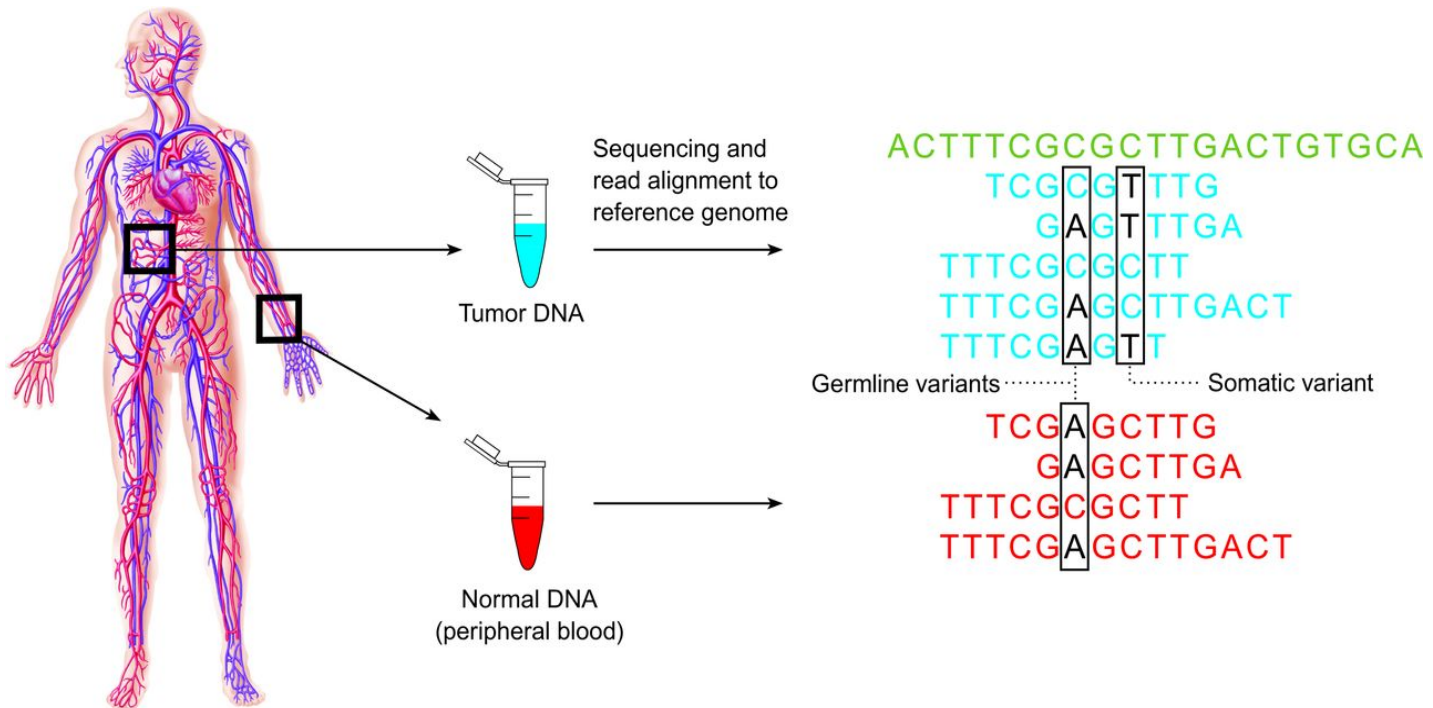


Step 4: Phasing

mostly with paired-end reads



# Characterization of genomic variations: somatic vs germline



**Sequencing tumor and normal samples from cancer patients provide insight into somatic and germline variation profile.**

# Bayes' Theorem to detect genomic variant

A	AGCTTGAC	TCCA	TGATGATT
B	AGCTTGAC	GCCA	TGATGATT
C	AGCTTGAC	TCCC	TGATGATT
D	AGCTTGAC	GCCC	TGATGATT
E	AGCTTGAC	TCCA	TGATGATT
F	AGCTTGAC	GCCA	TGATGATT
G	AGCTTGAC	TCCC	TGATGATT
H	AGCTTGAC	GCCC	TGATGATT

$$P(G|D) = \frac{P(D|G)P(G)}{P(D)}$$
$$= \frac{P(D|G) P(G)}{\sum_{i=1}^n P(D|G_i) P(G_i)}$$

In the above equation:

- $D$  refers to the observed data
- $G$  is the genotype whose probability is being calculated
- $G_i$  refers to the  $i$ th possible genotype, out of  $n$  possibilities

Calculating the conditional distribution  $P(D|G)$ :

Assuming an error free model, for each heterozygous SNP site of the diploid genome, covered by  $K$  reads, the number of reads  $i$  representing one of the two alleles follows binomial distribution.

$$P_{err\_free}(D|G) = f(i|k, 0.5) = \binom{k}{i} 0.5^k$$

With errors, the calculation is more complicated.

In general:

$$P(D|G) = P_{err\_free}(D|G) + P_{err}(D|G)$$

# Main Steps in Genome Resequencing

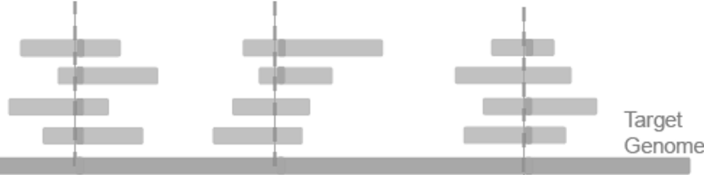
[Snyder et al. Genes & Dev. ('10)]

Step 0: Generate Reads



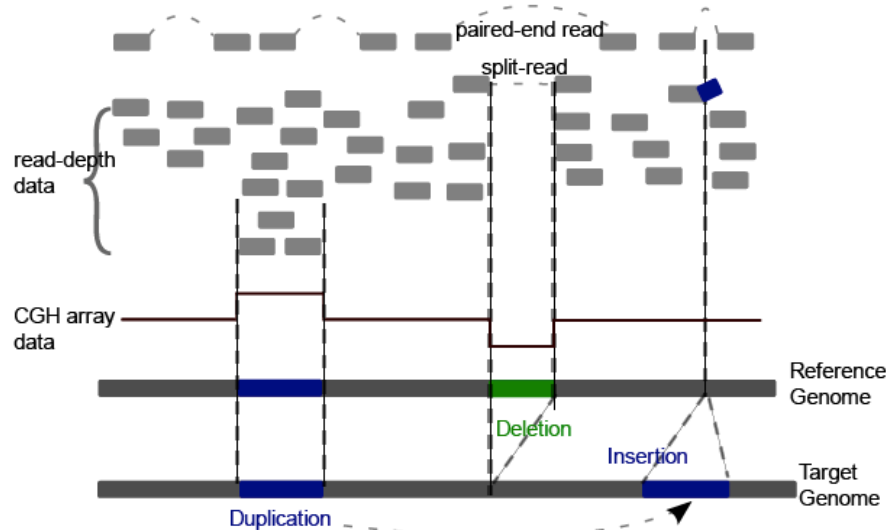
Step 1: Call SNPs

using uniquely and correctly mapped reads



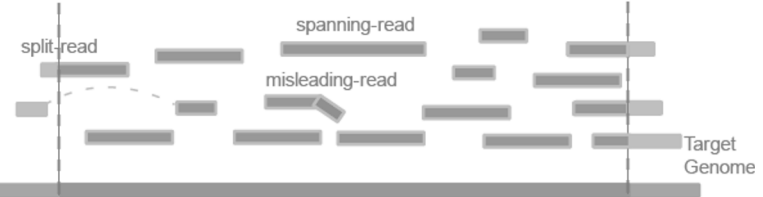
Step 2: Find SVs

with aberrant paired-end reads, split-reads, read-depth analysis and CGH array data



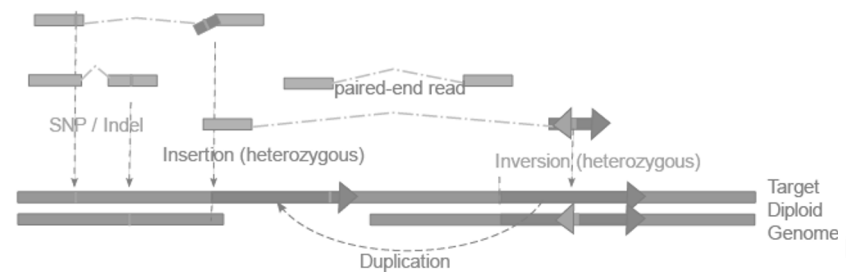
Step 3: Assemble New Sequences

with split-, spanning- and misleading-reads

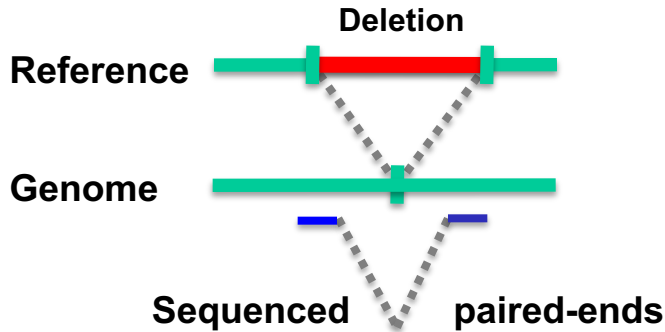


Step 4: Phasing

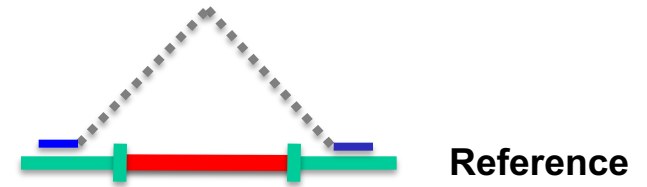
mostly with paired-end reads



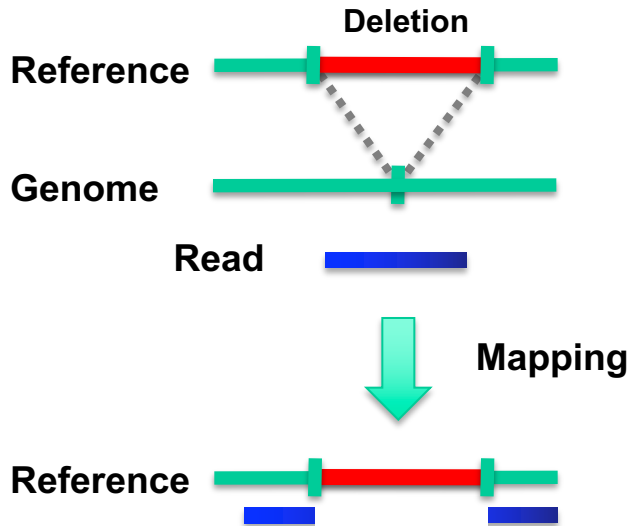
# 1. Paired ends



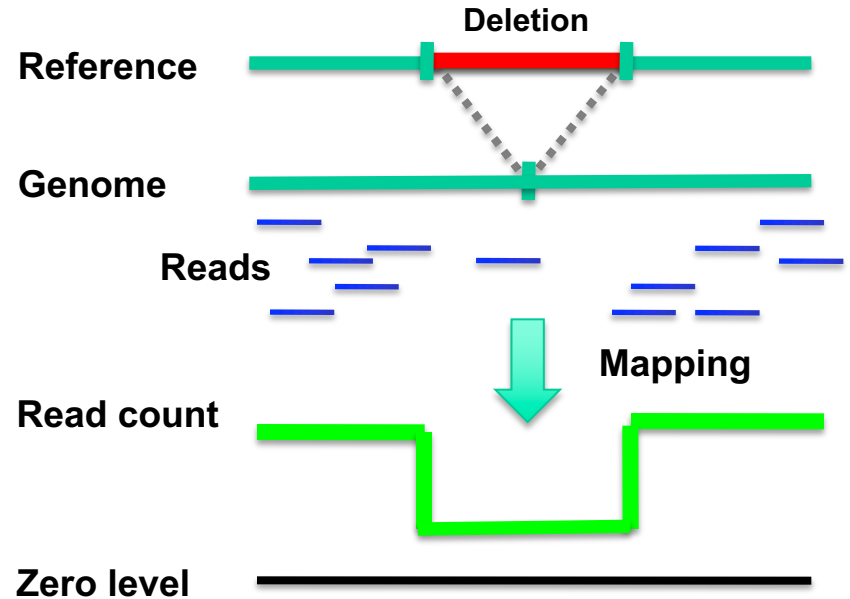
Mapping  
→



# 2. Split read



# 3. Read depth (or aCGH)



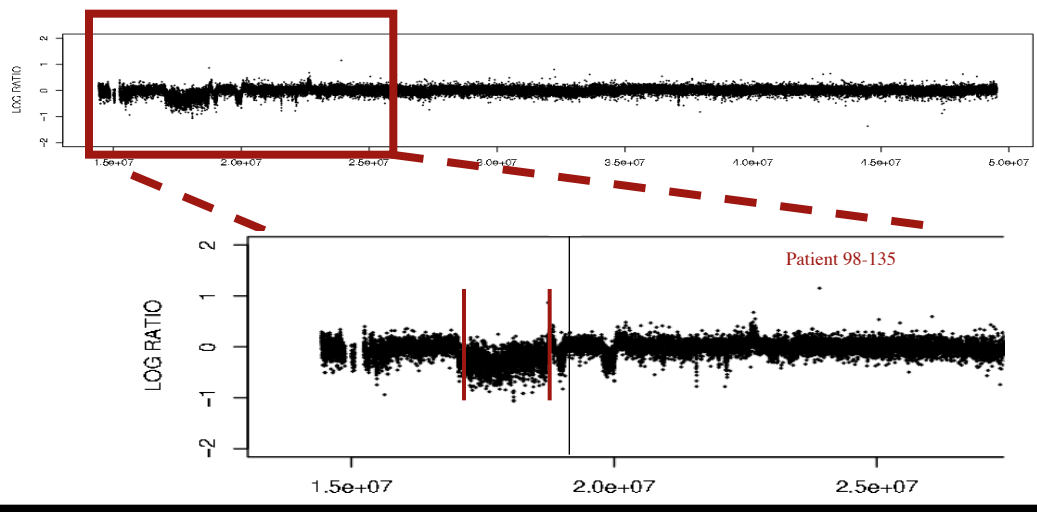
# 4. Local Reassembly

[Snyder et al. Genes & Dev. ('10)]



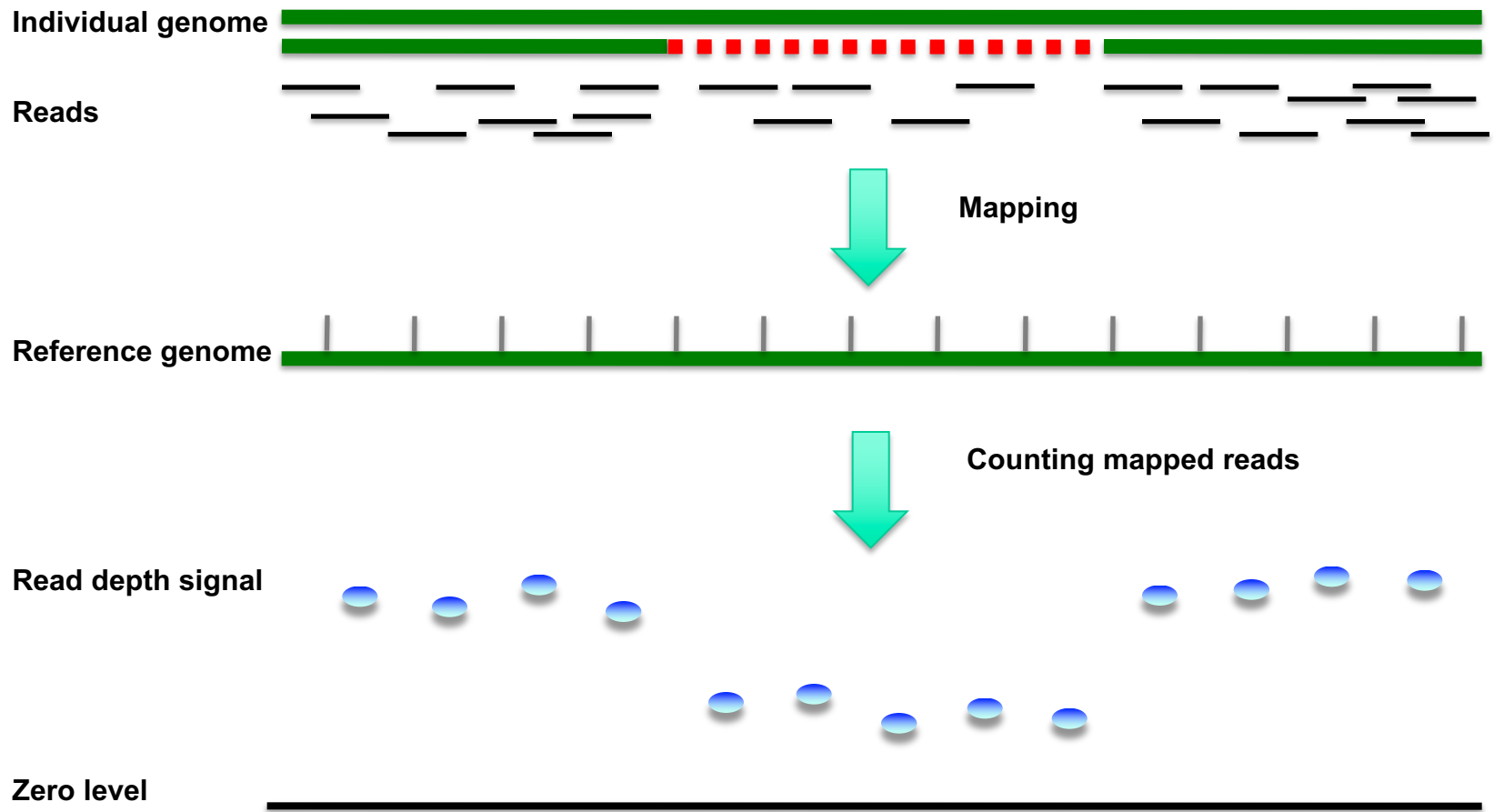
# Read Depth

[Urban et al. ('06) PNAS; Wang et al. Gen. Res. ('09);  
Abyzov et al. Gen. Res. ('11)]

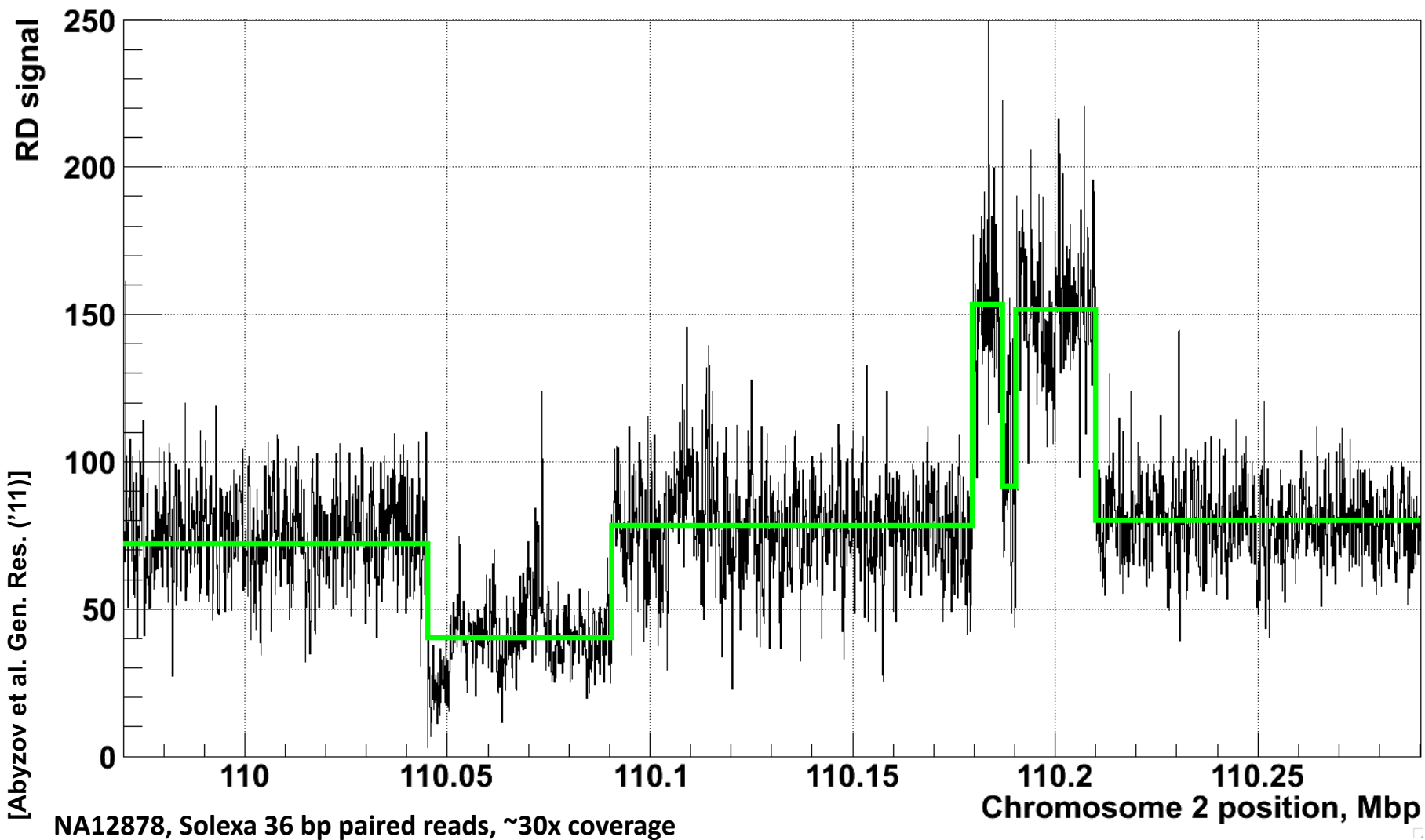


# Array Signal

# Read depth

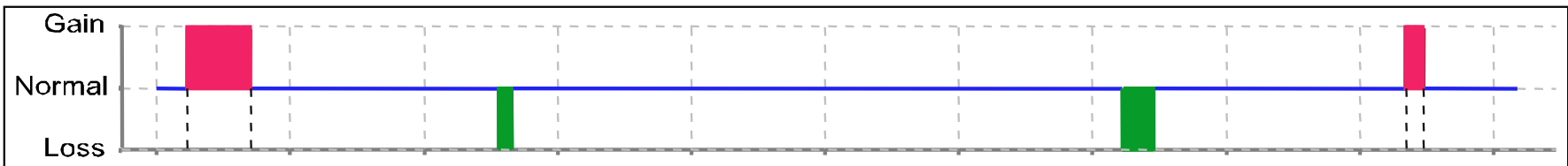
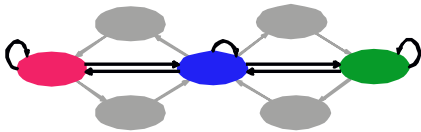
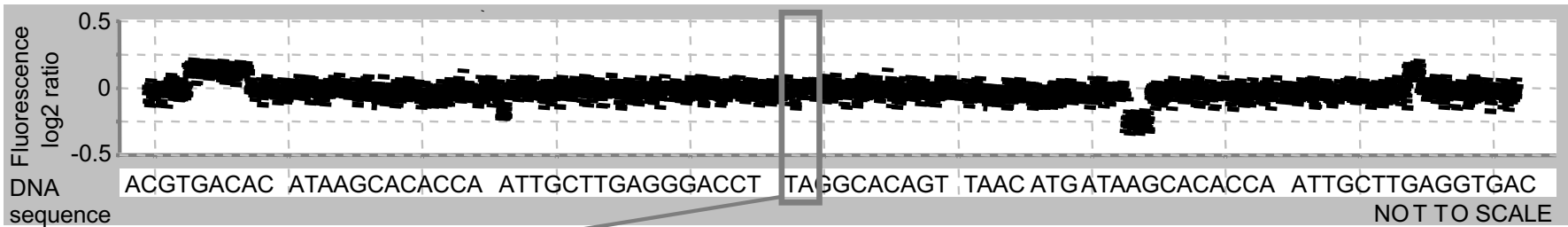


# Example of Application to RD data

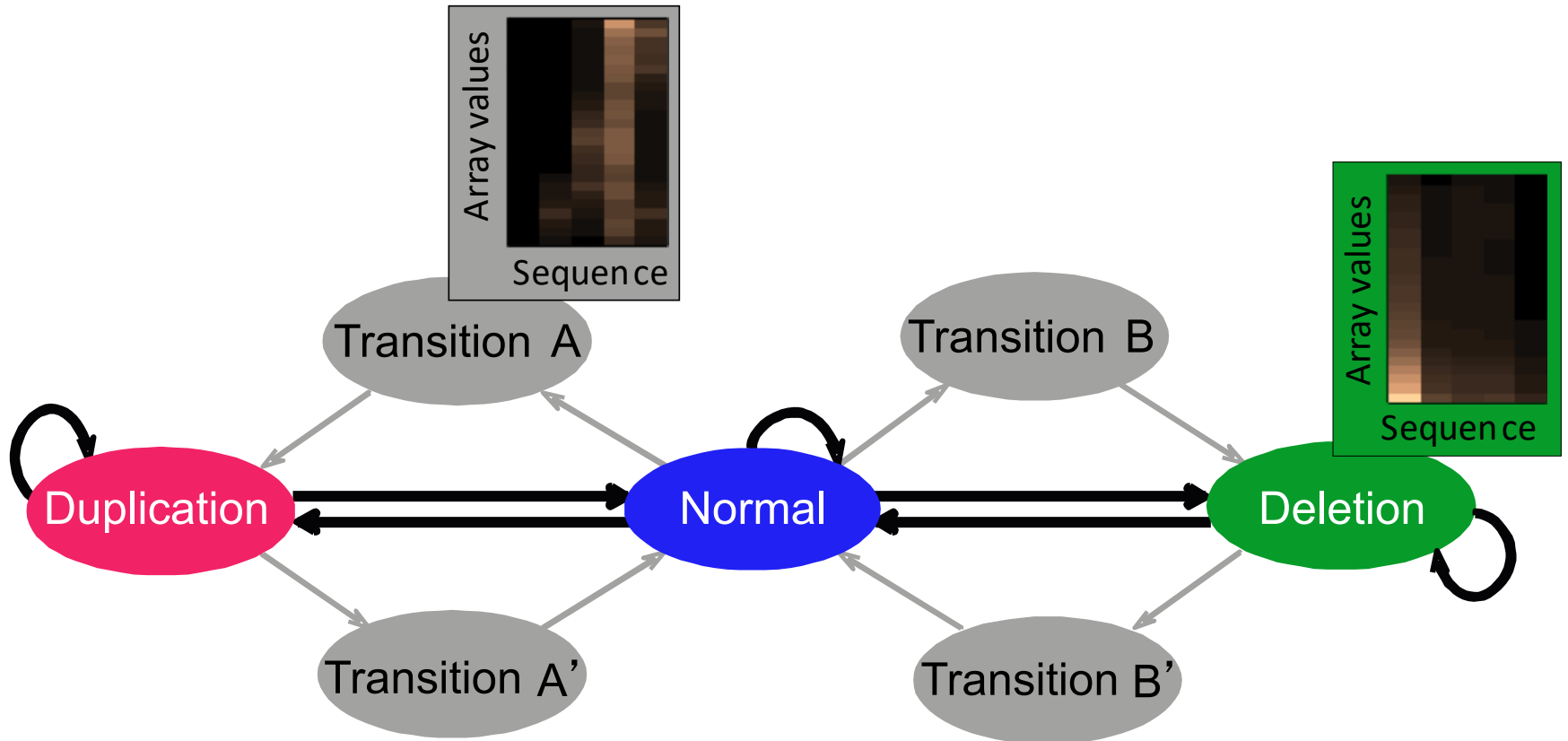


# HMM

- To get highest resolution on breakpoints need to smooth & segment the signal
- BreakPtr: prediction of breakpoints, dosage and cross-hybridization using a system based on Hidden Markov Models

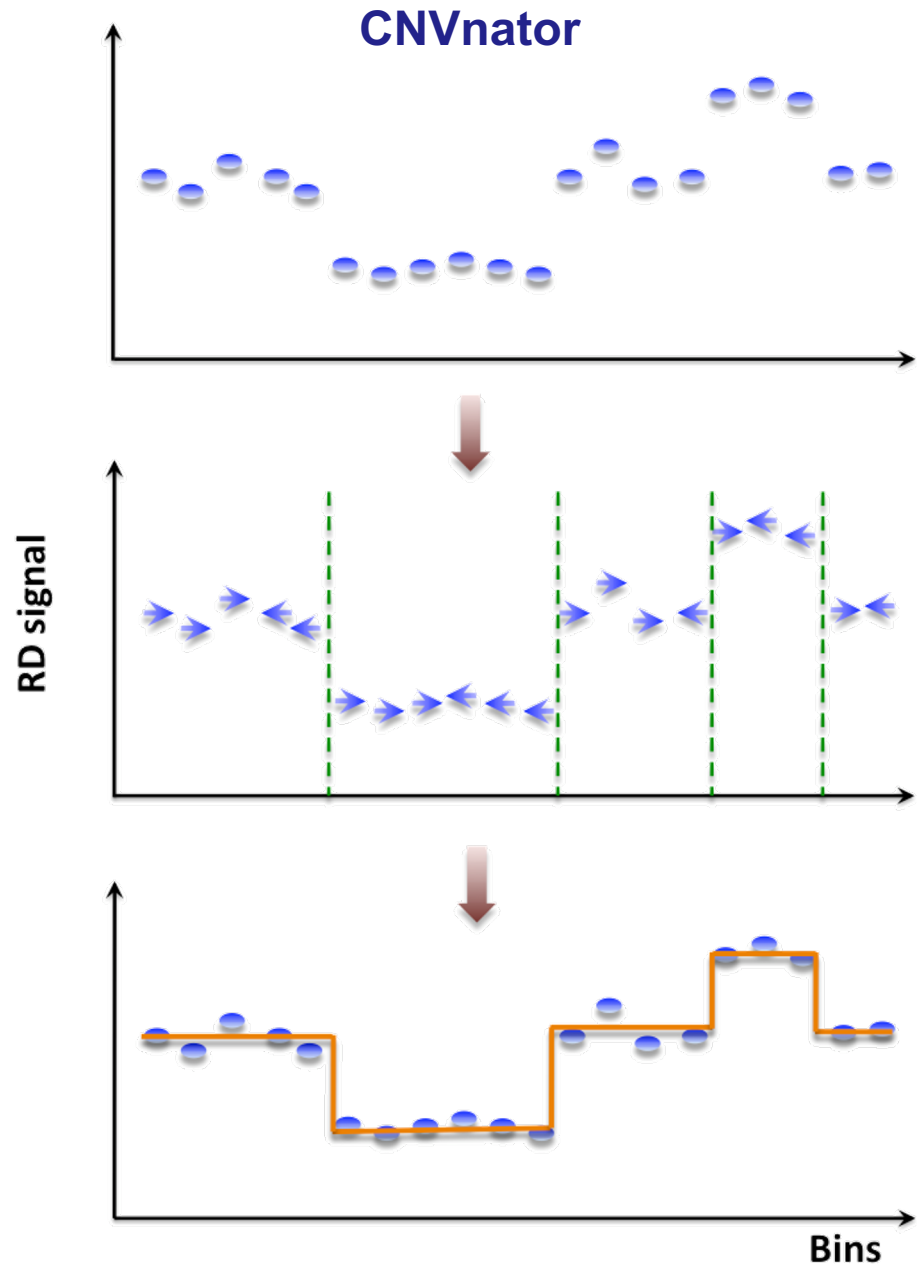


Statistically integrates array signal and DNA sequence signatures  
(using a discrete-valued bivariate HMM)



# Mean-shift-based (MSB) segmentation: no explicit model

- For each bin attraction (mean-shift) vector points in the direction of bins with most similar RD signal
- No prior assumptions about number, sizes, haplotype, frequency and density of CNV regions
- Not Model-based (e.g. like HMM) with global optimization, distr. assumption & parms. (e.g. num. of segments).
- Achieves discontinuity-preserving smoothing
- Derived from image-processing applications



[Abyzov et al. Gen. Res. ('11)]

# Intuitive Description of MSB

● Observed depth of coverage counts as samples from PDF

➔ Kernel-based approach to estimate local gradient of PDF

⊕ Iteratively follow grad to determine local modes

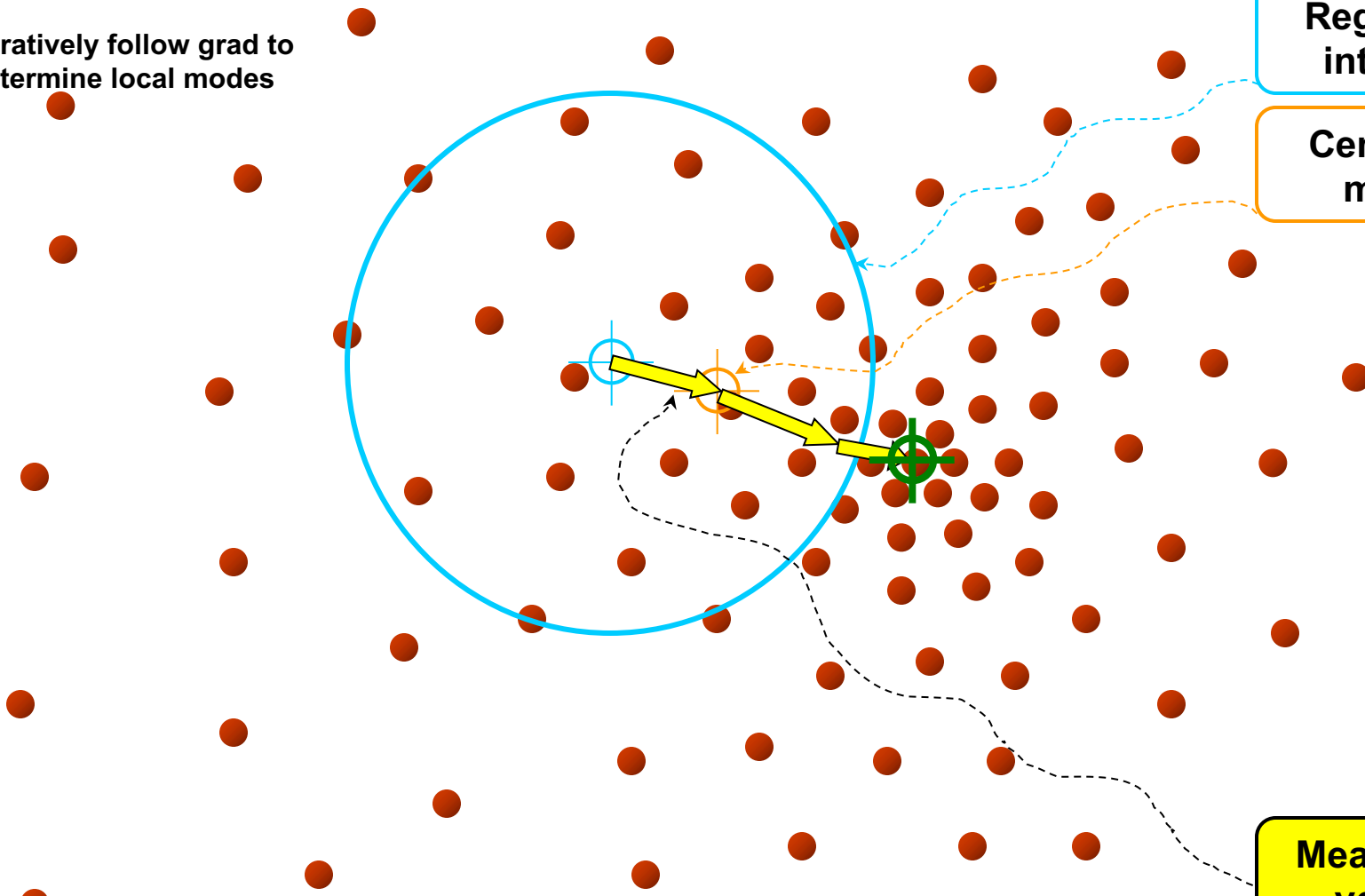
Region of interest

Center of mass

Mean Shift vector

**Objective : Find the densest region**  
**Distribution of identical billiard balls**

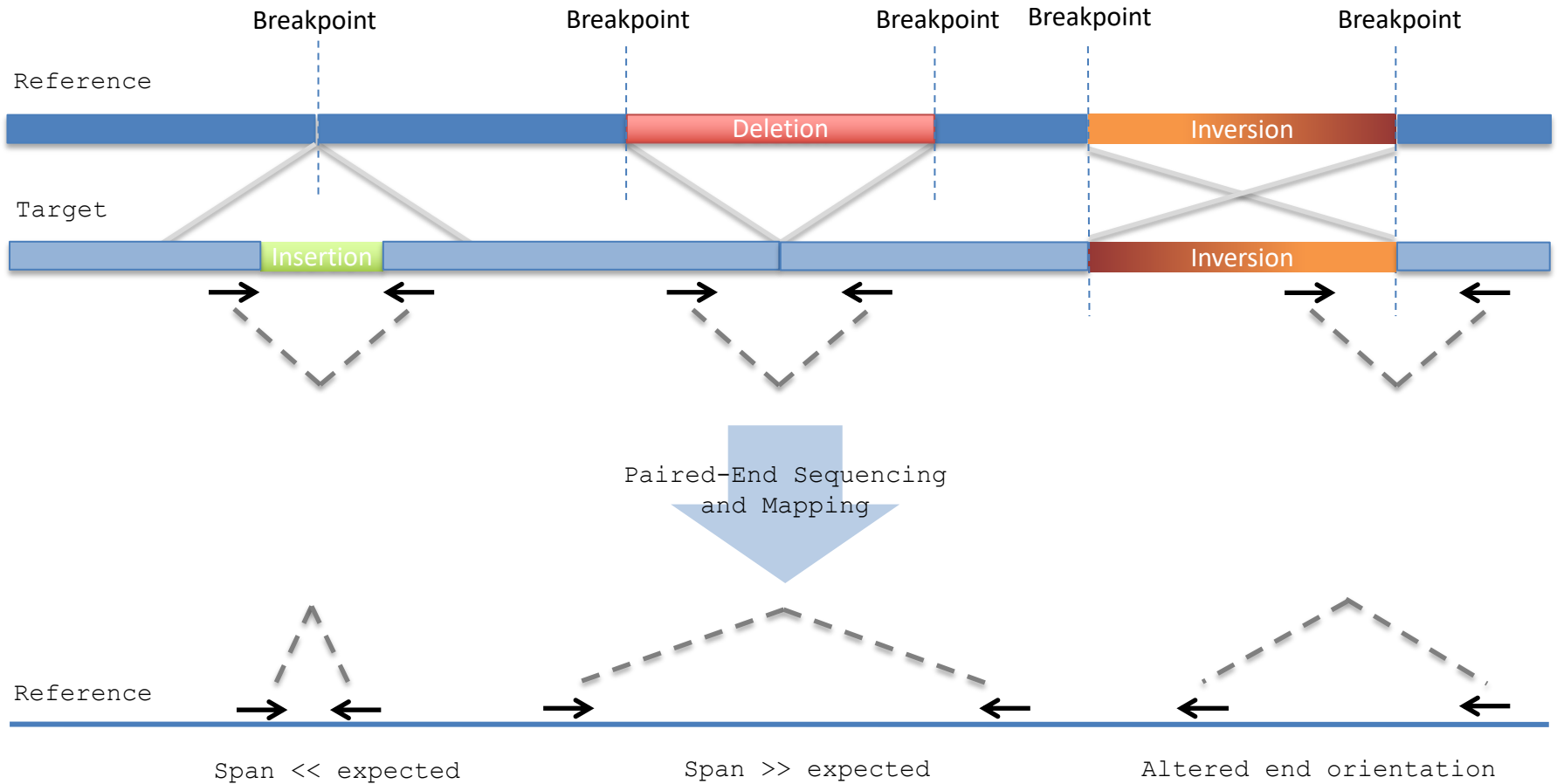
[ Adapted from S Ullman et al. "Advanced Topics in Computer Vision,"  
www.wisdom.weizmann.ac.il/~vision/courses/2004\_2 ]



# Paired-End



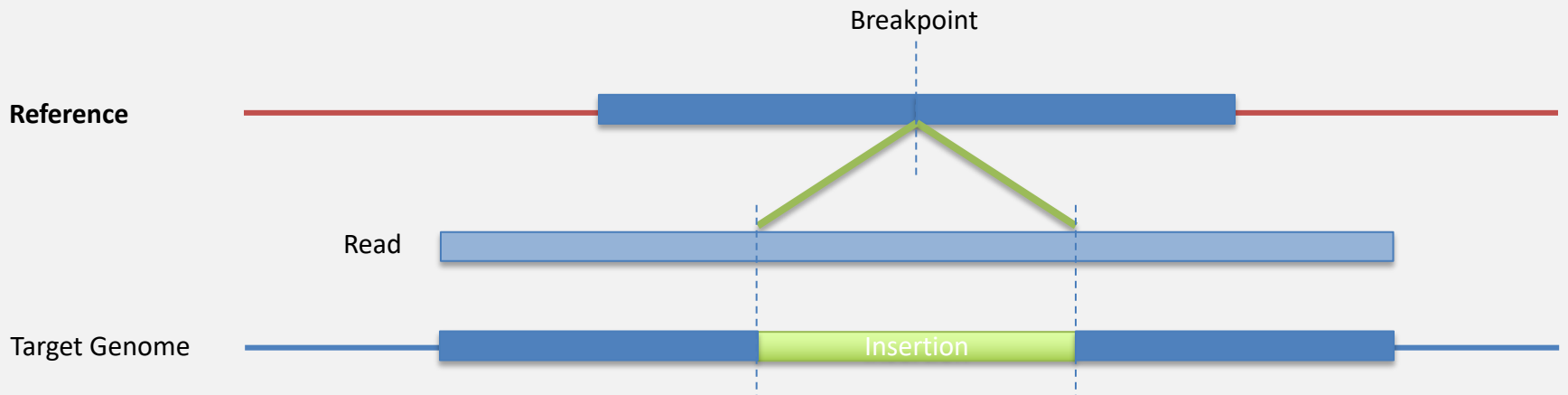
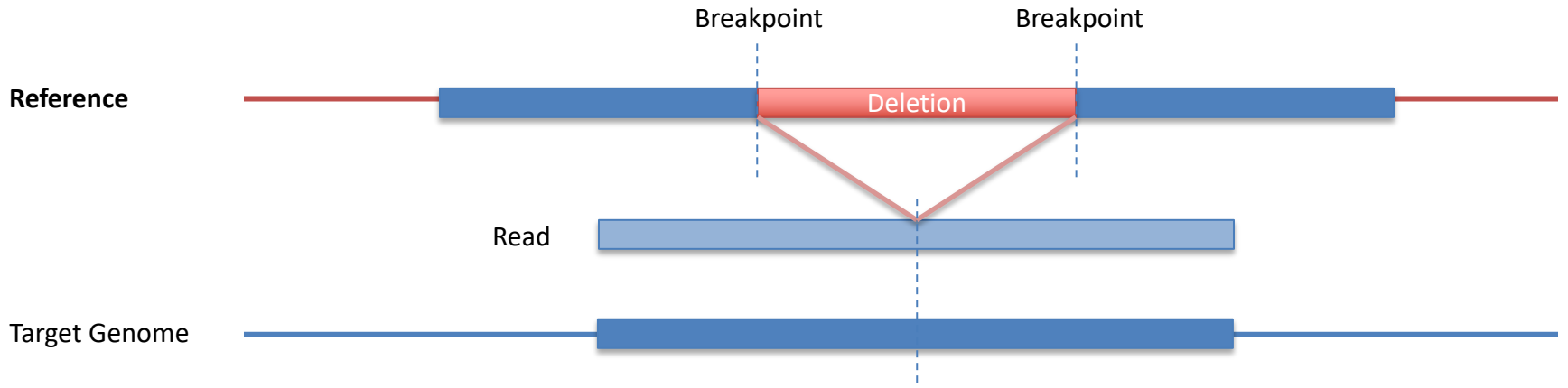
# Paired-End Mapping



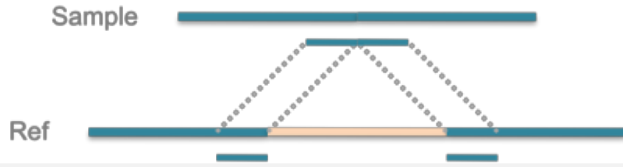
- Both paired-ends map within repeats.
- Limited the distance between pairs; therefore, neither large nor very small rearrangements can be detected

# Split Read

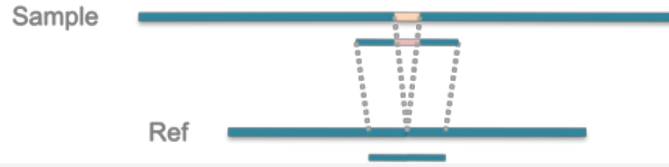
# Split-read Analysis



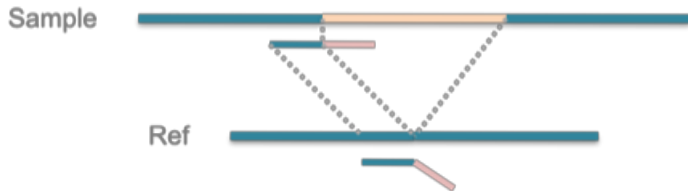
*Deletion*



*Insertion, small*

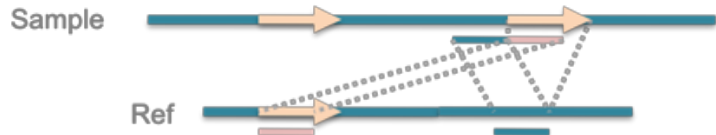


*Insertion, large*

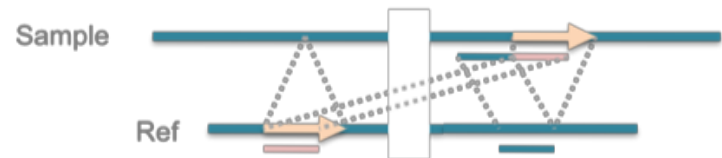
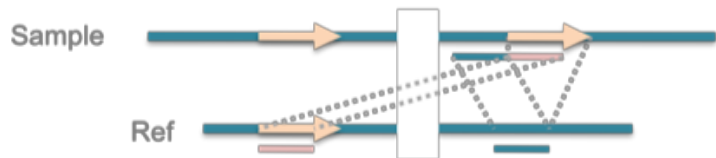
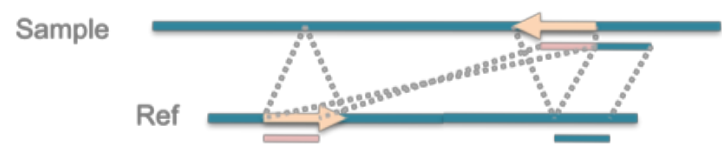
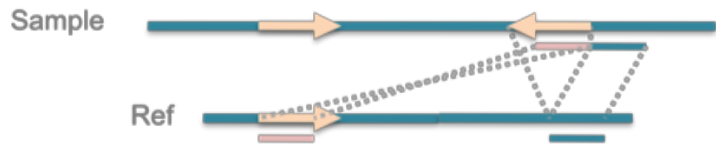
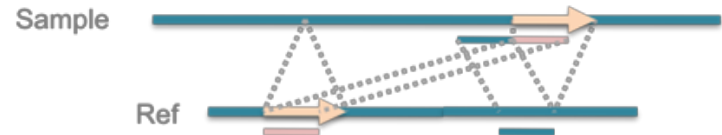


Deletions are the Easiest to Identify

*Duplication*

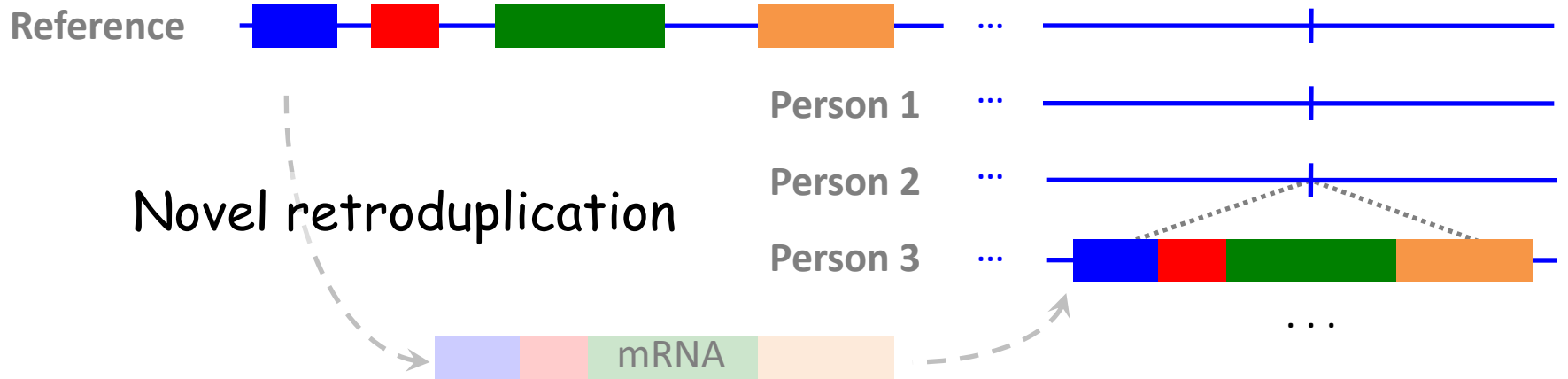
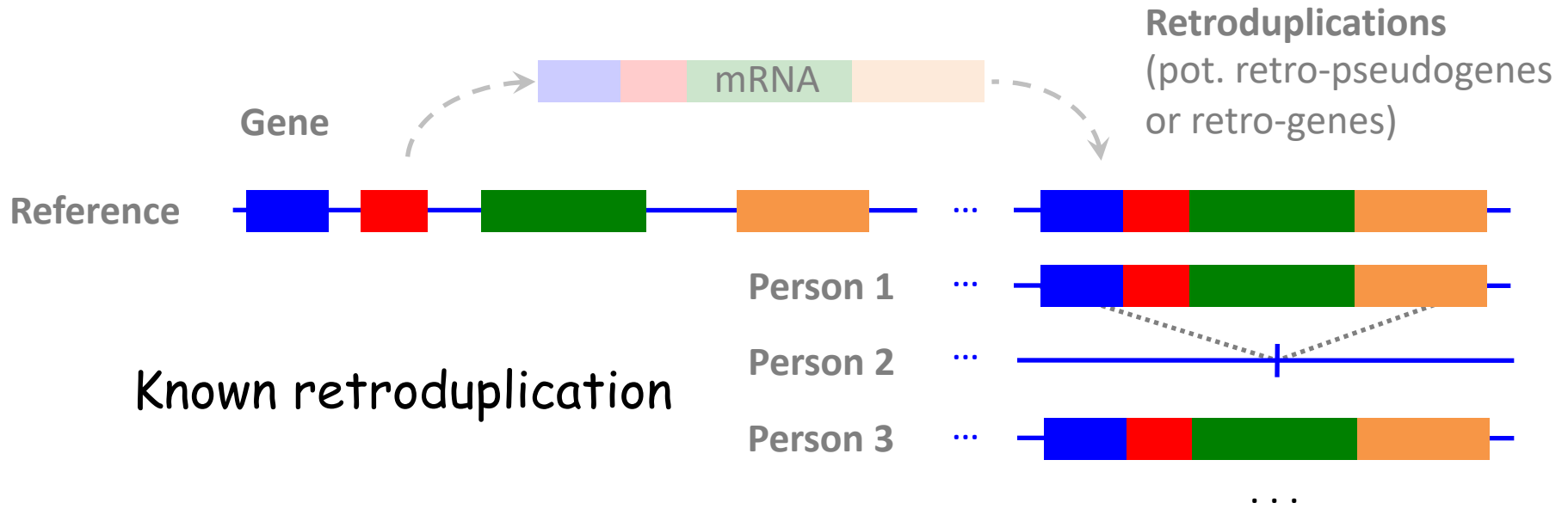


*Translocation*

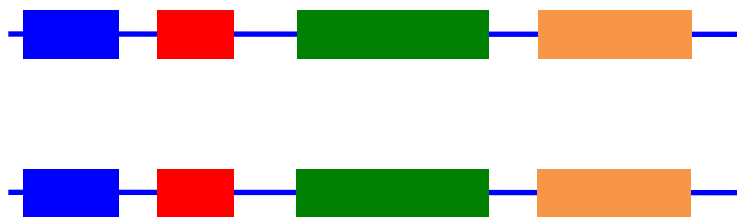


# RDV & Mobile Elements

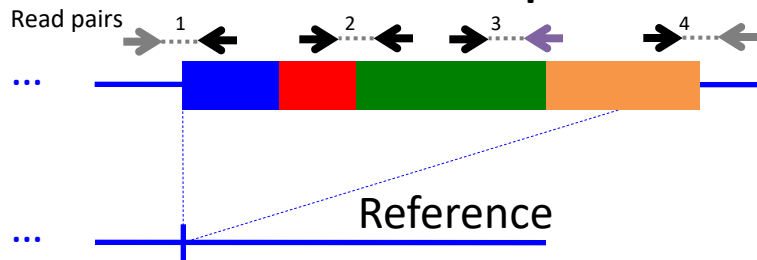
# Retroduplication variation (RDV)



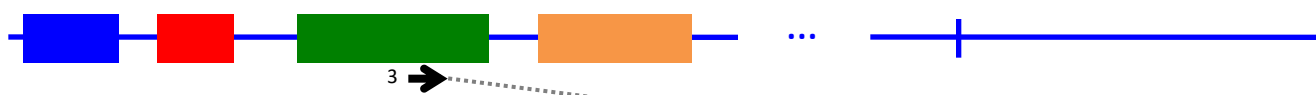
# Gene



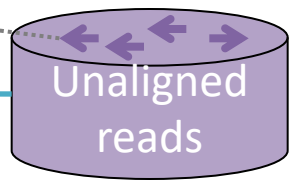
# Novel retroduplication



Alignment to the reference

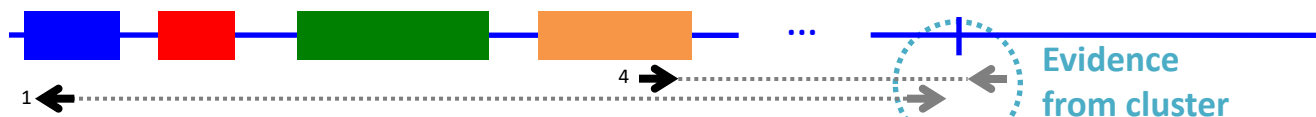


Evidence from alignment



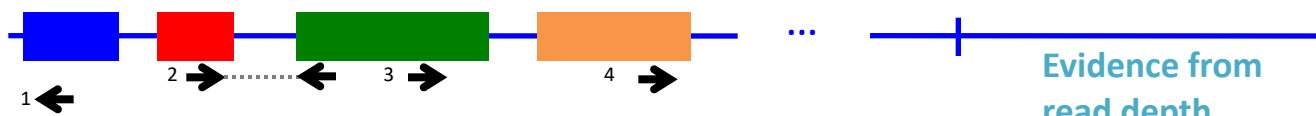
1

Aligned reads



Evidence from cluster

2



Evidence from read depth

3



Zero level

Pipeline to identify novel retrodups. from 3 evidence sources