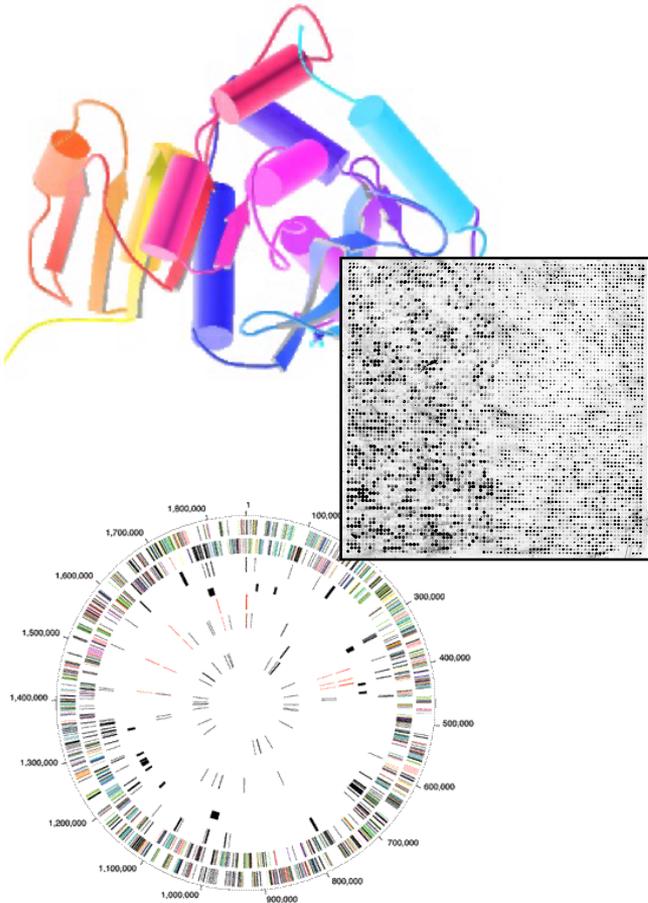


Variant Identification, Focusing on SVs



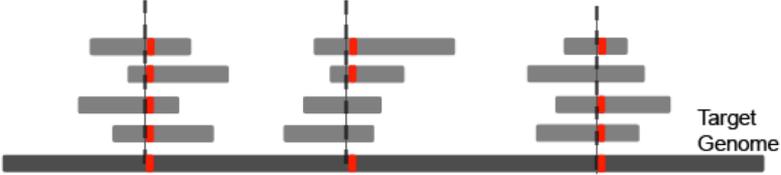
Mark Gerstein, Yale University
gersteinlab.org/courses/452
(last edit in spring '19, pack #6)

Step 0: Generate Reads



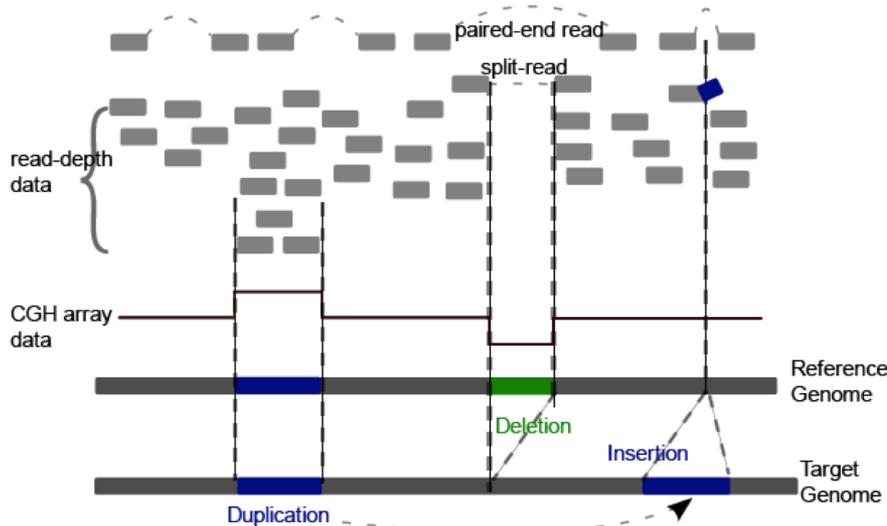
Step 1: Call SNPs

using uniquely and correctly mapped reads



Step 2: Find SVs

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

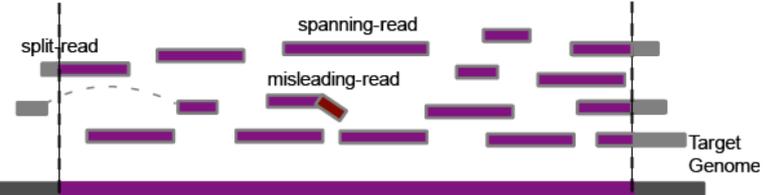


Main Steps in Genome Resequencing

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

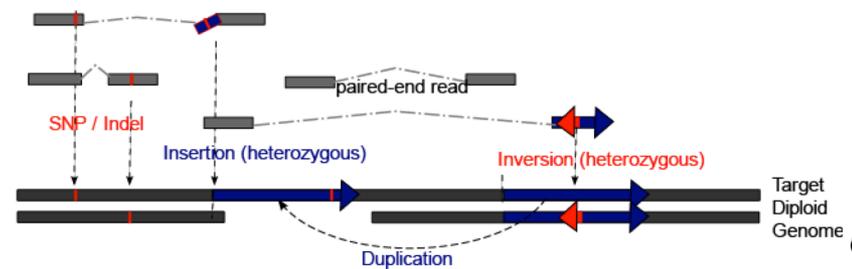
Step 3: Assemble New Sequences

with split-, spanning- and misleading-reads



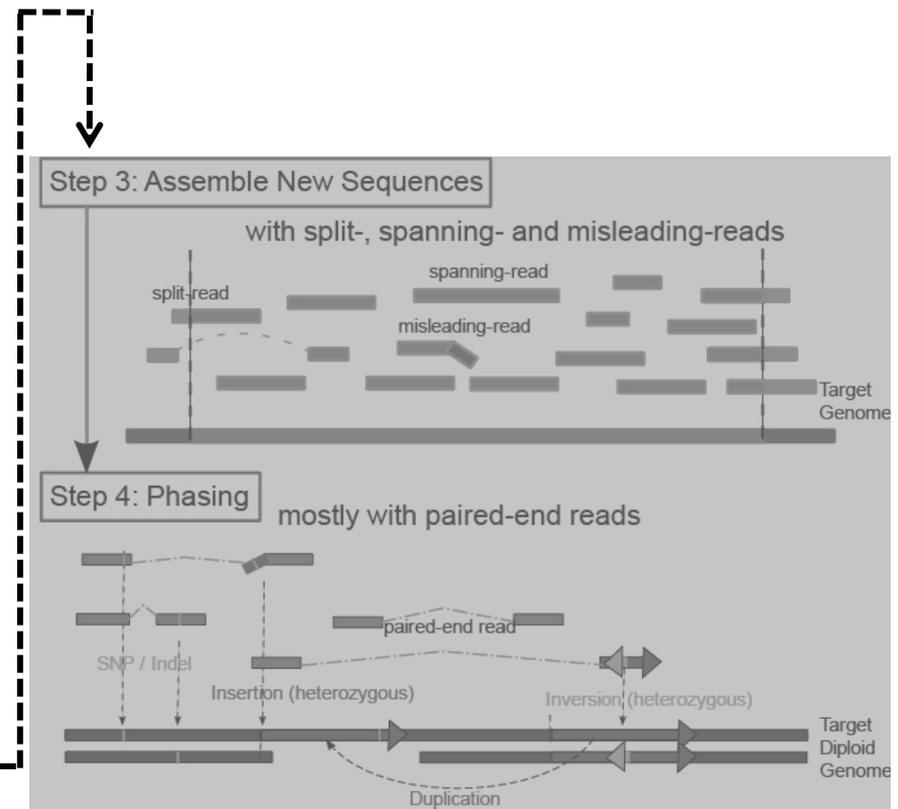
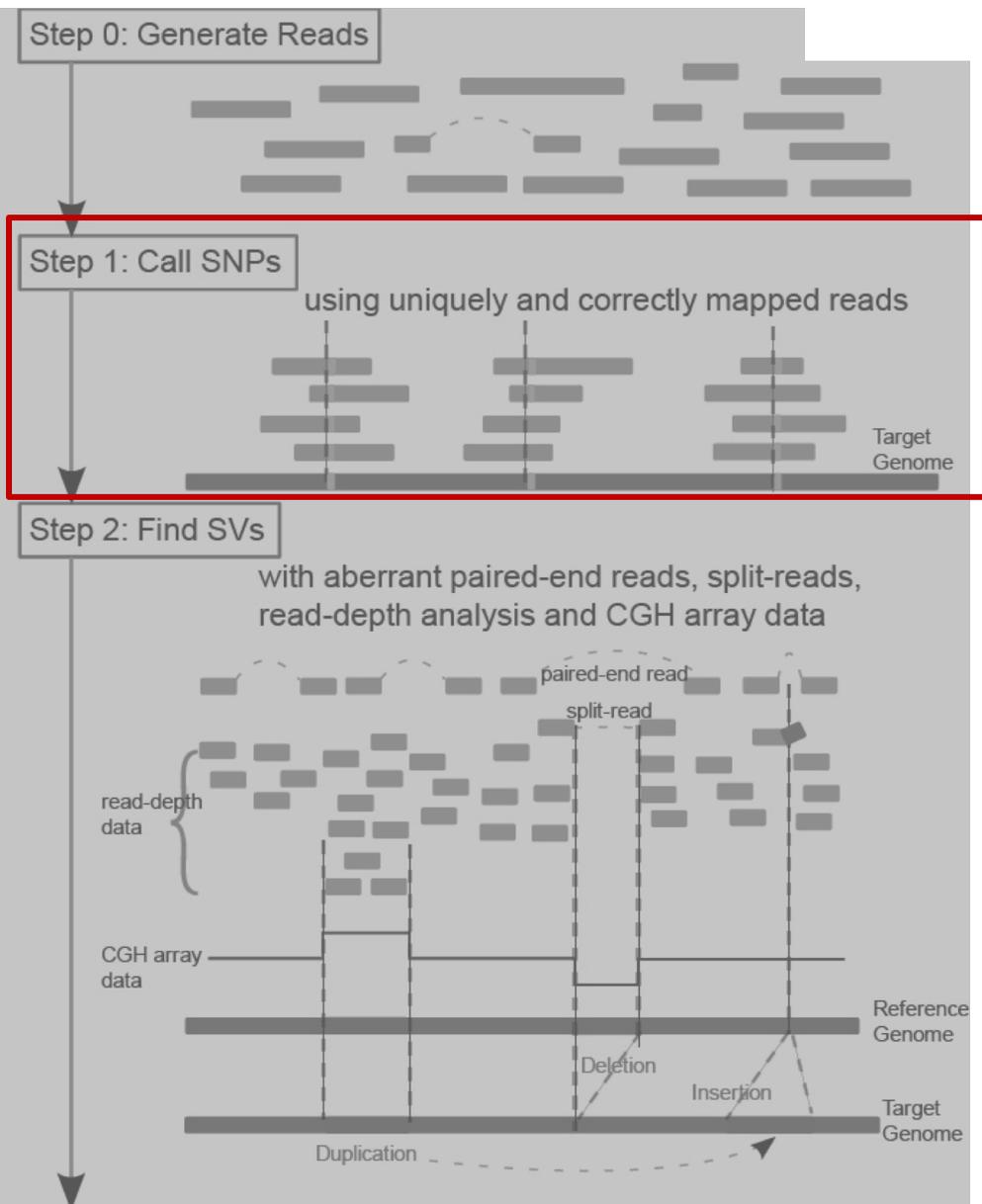
Step 4: Phasing

mostly with paired-end reads



Main Steps in Genome Resequencing

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



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

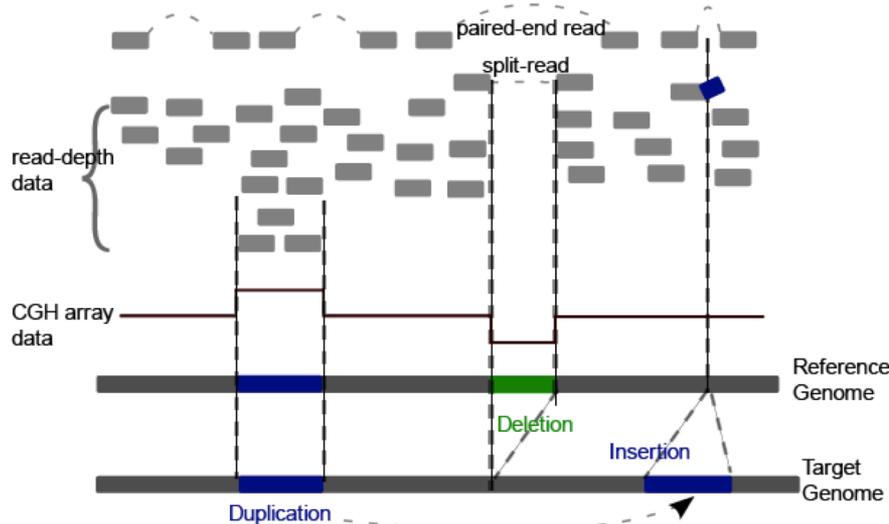
Step 1: Call SNPs

using uniquely and correctly mapped reads

Target Genome

Step 2: Find SVs

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



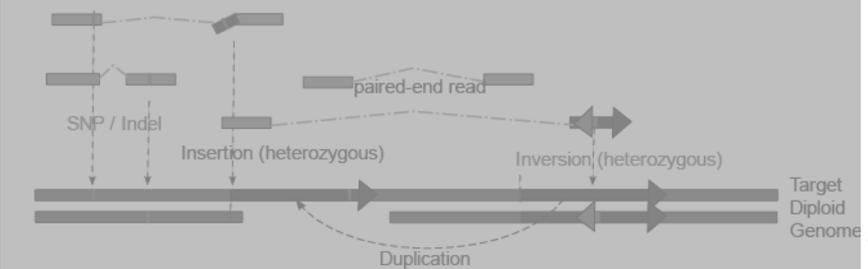
Step 3: Assemble New Sequences

with split-, spanning- and misleading-reads

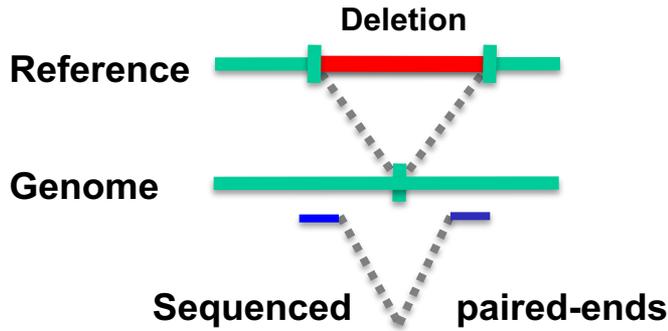
Target Genome

Step 4: Phasing

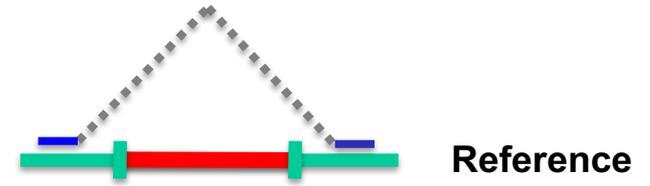
mostly with paired-end reads



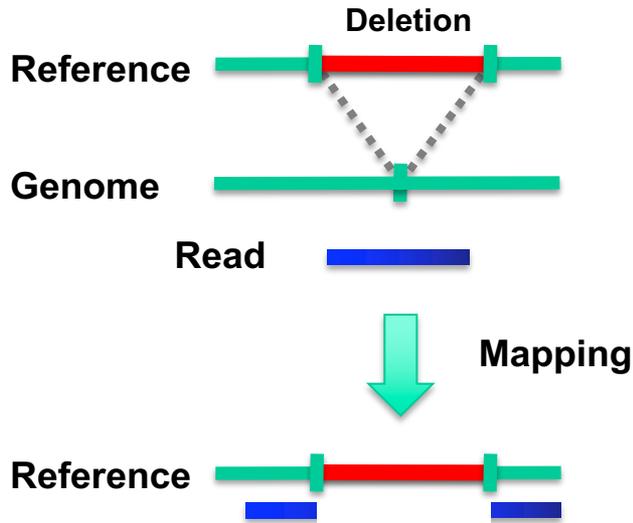
1. Paired ends



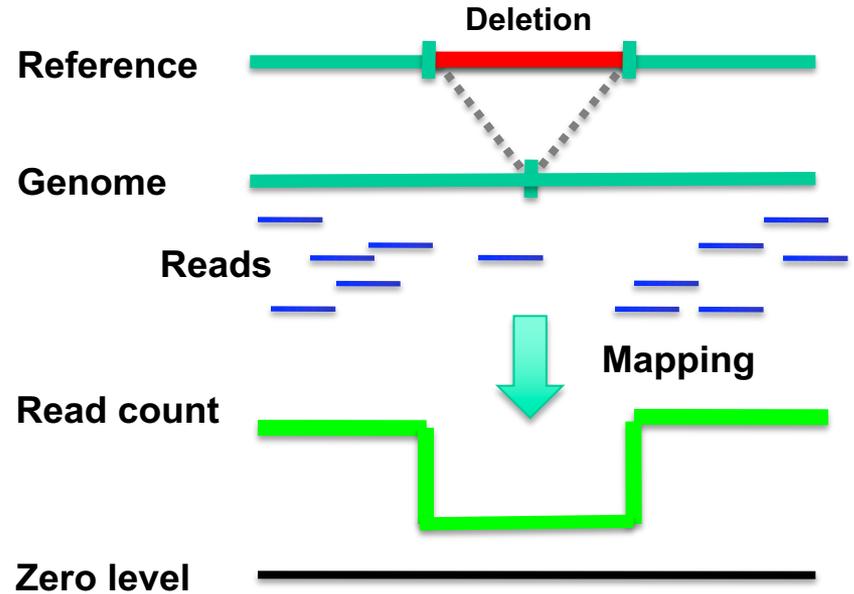
Mapping
→



2. Split read



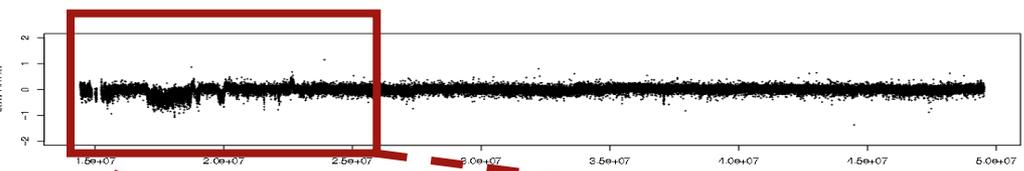
3. Read depth (or aCGH)



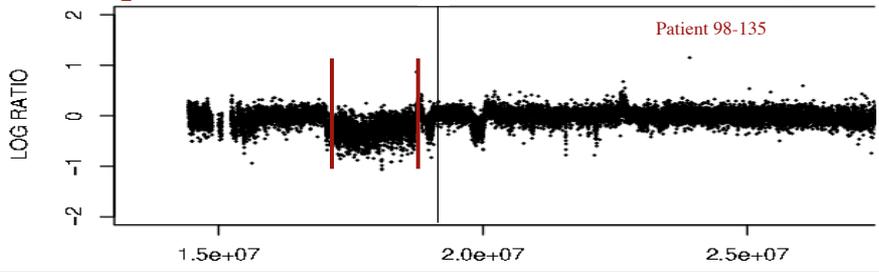
4. Local Reassembly

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

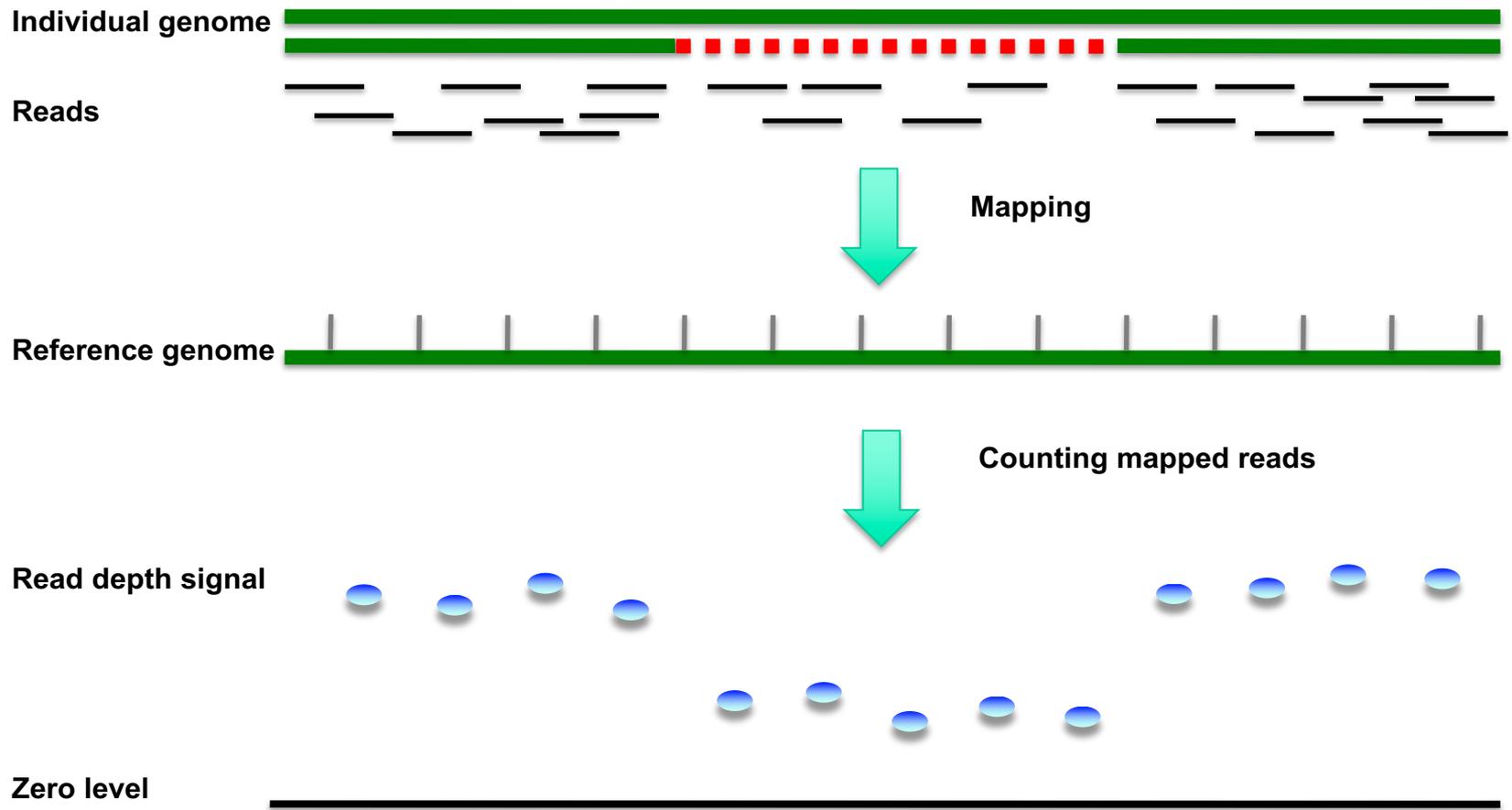
Read Depth



Array Signal



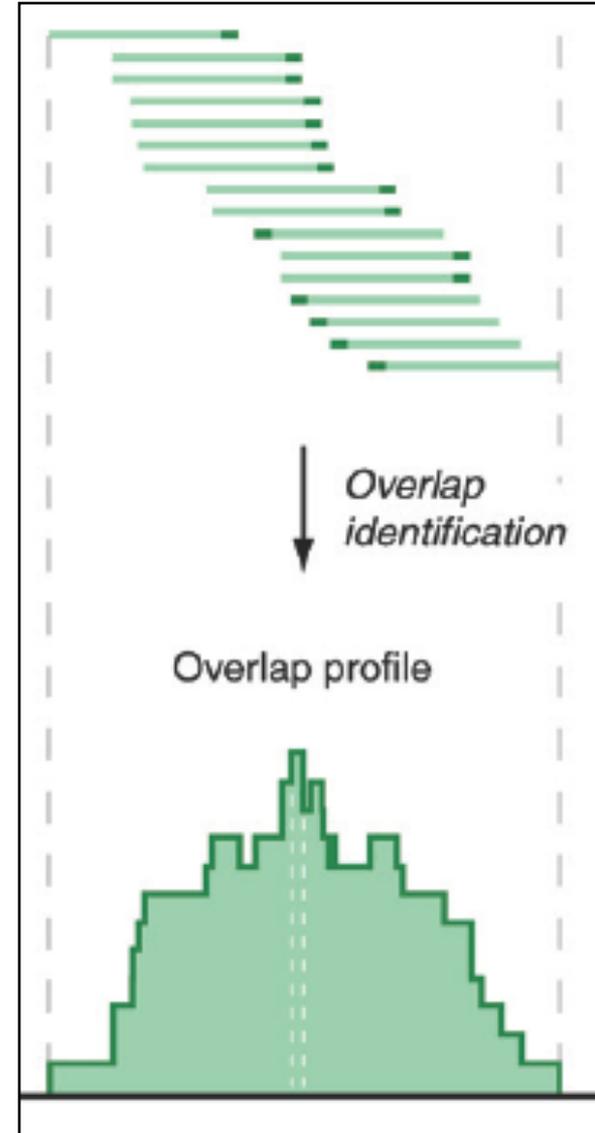
Read depth



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

Reads to Signal Track

```
@ILMN-GA001 3 208HWAAXX 1 1 110 812
ATACAAGCAAGTATAAGTTCGTATGCCGTCTT
+ILMN-GA001 3 208HWAAXX 1 1 110 812
hhhYhh]NYhhhhhhYIhhaZT[hYHNSPKXR
@ILMN-GA001 3 208HWAAXX 1 1 111 879
GGAGGCTGGAGTTGGGGACGTATGCGGCATAG
+ILMN-GA001 3 208HWAAXX 1 1 111 879
hSWhRNJ\hFhLdhVOhAIB@NFKD@PAB?N?
```



Reads (fasta)

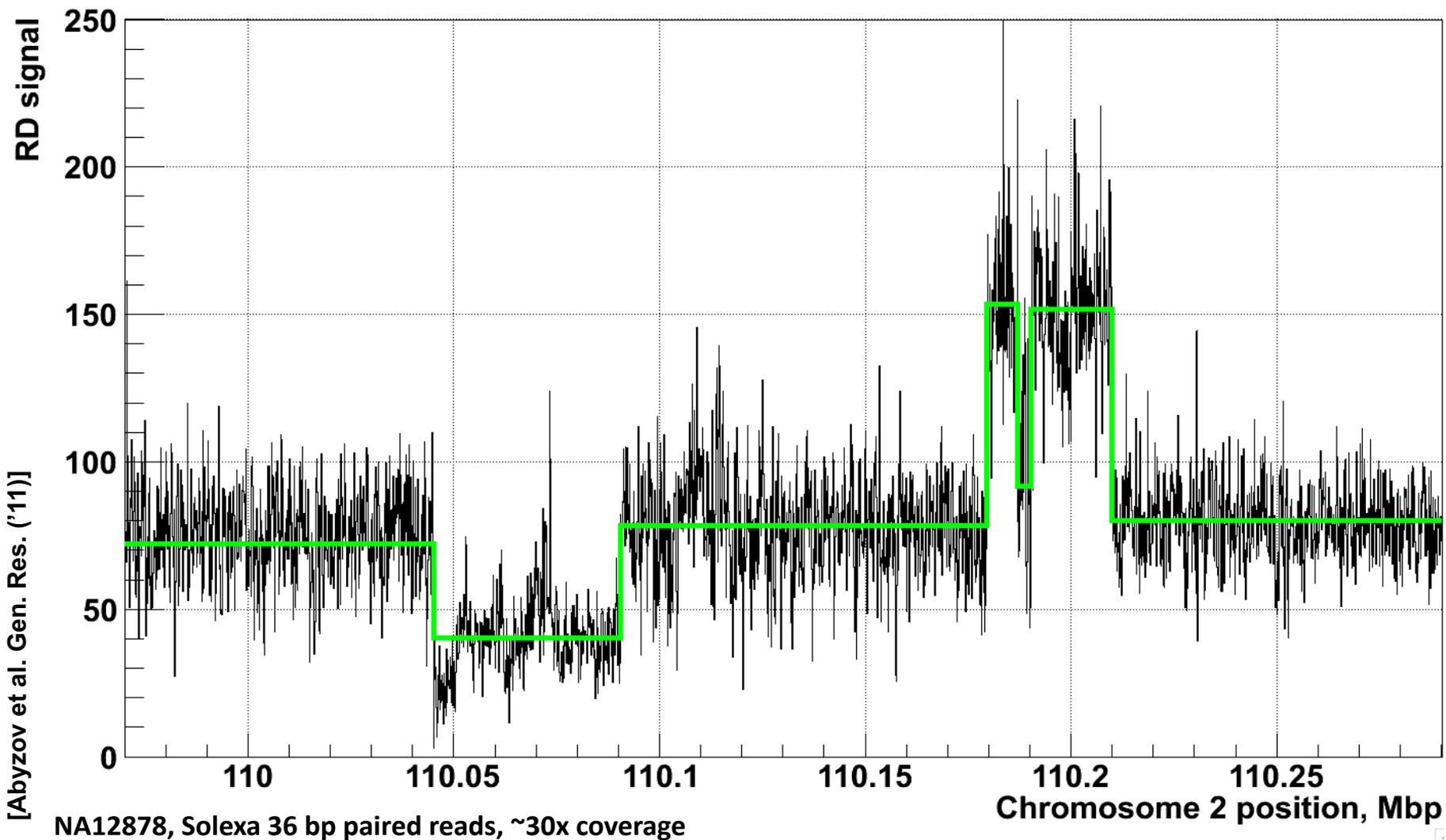
+ quality scores (fastq)

+ mapping (BAM)

Reads => Signal (Intermediate file)

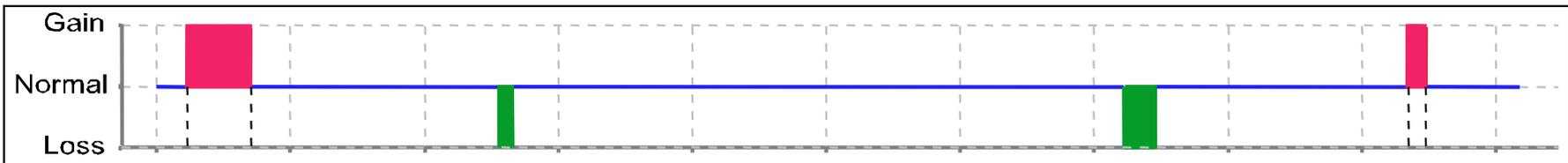
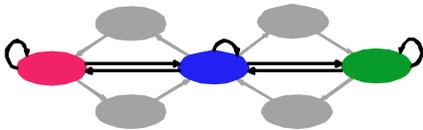
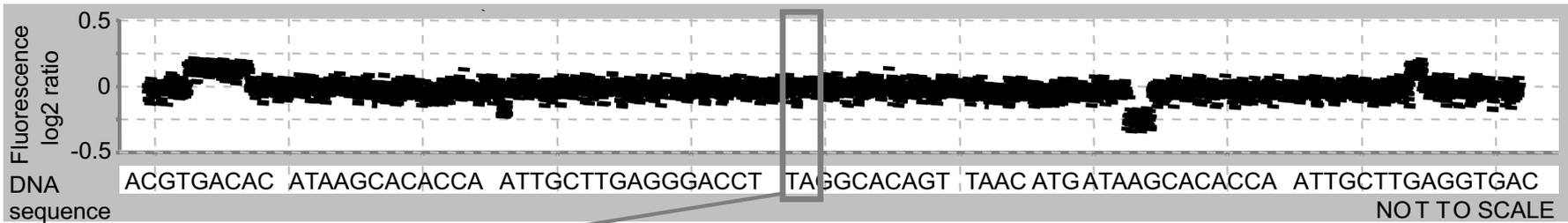
Accumulating @ >1 Pbp/yr (currently),
~20% of tot. HiSeq output

Example of Application to RD data

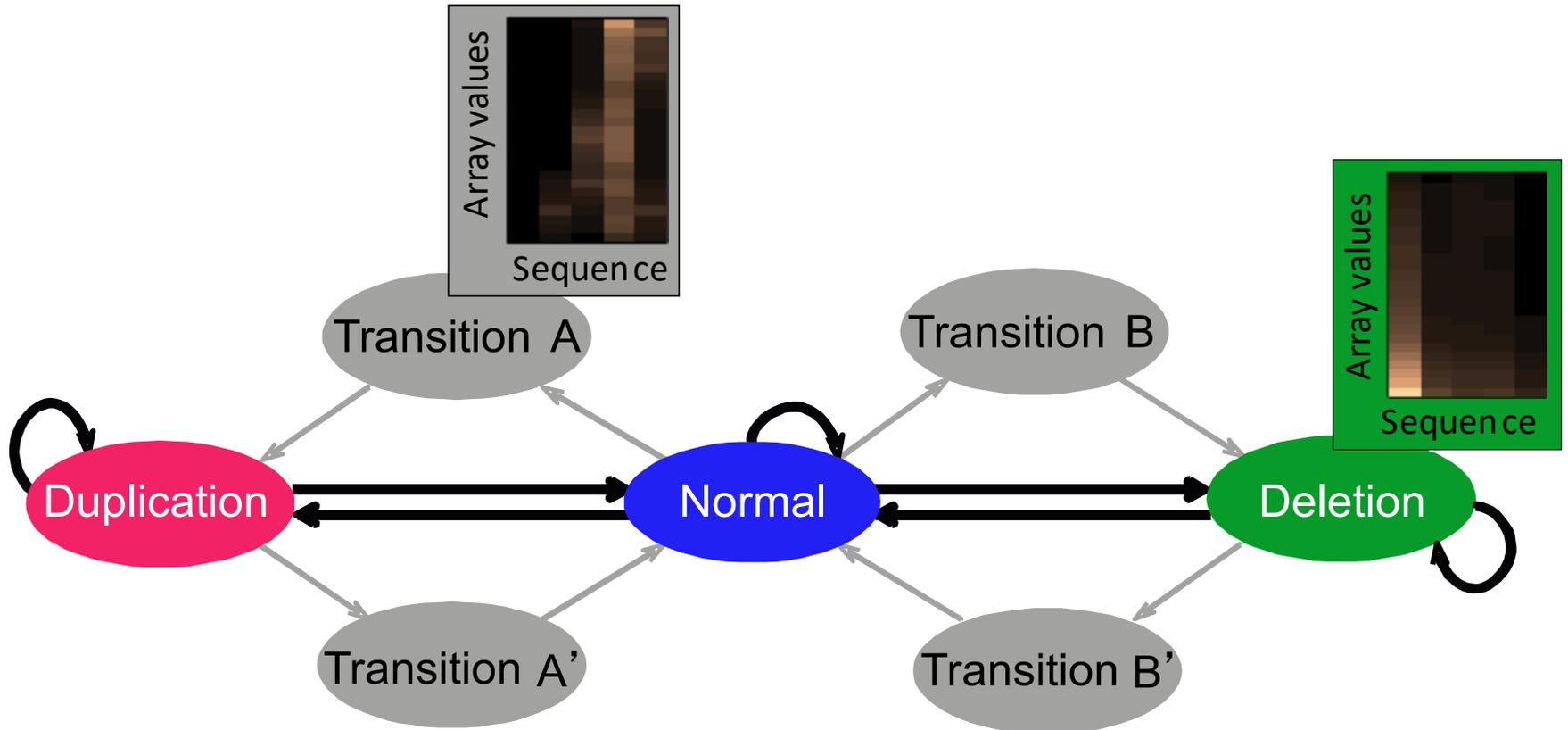


HMM

- 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

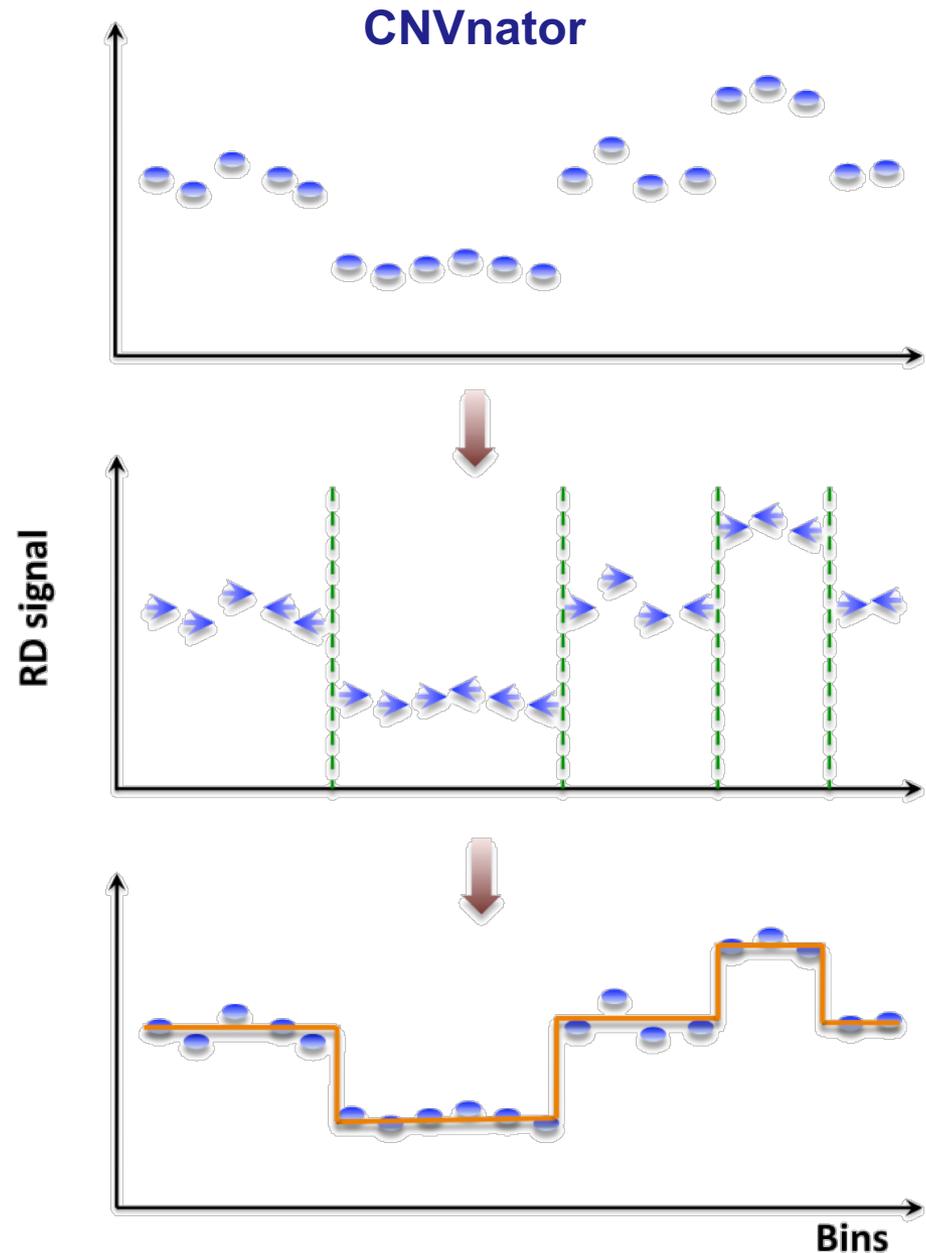


Statistically integrates array signal and DNA sequence signatures
(using a discrete-valued bivariate HMM)



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

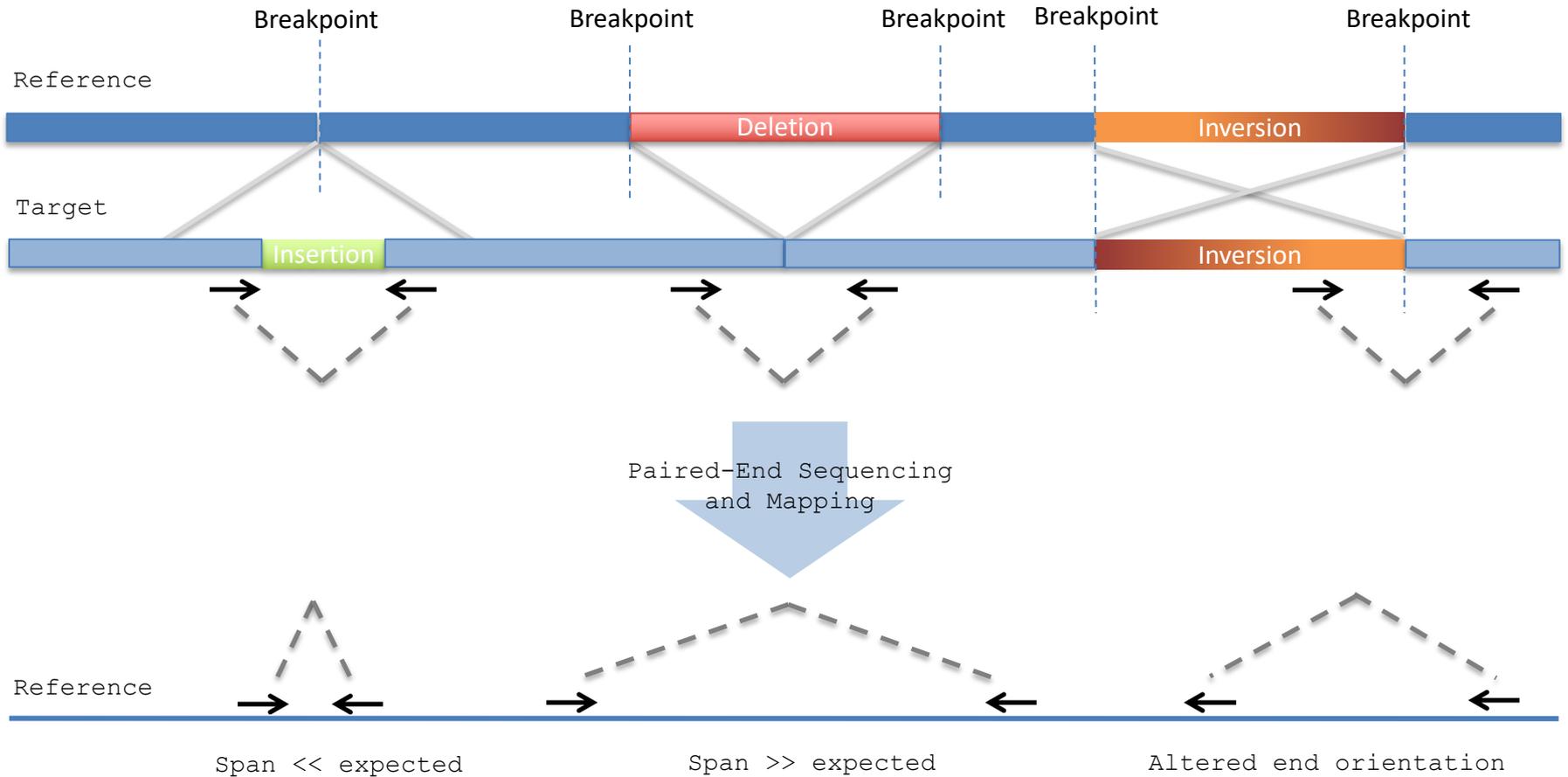
Mean Shift vector

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]

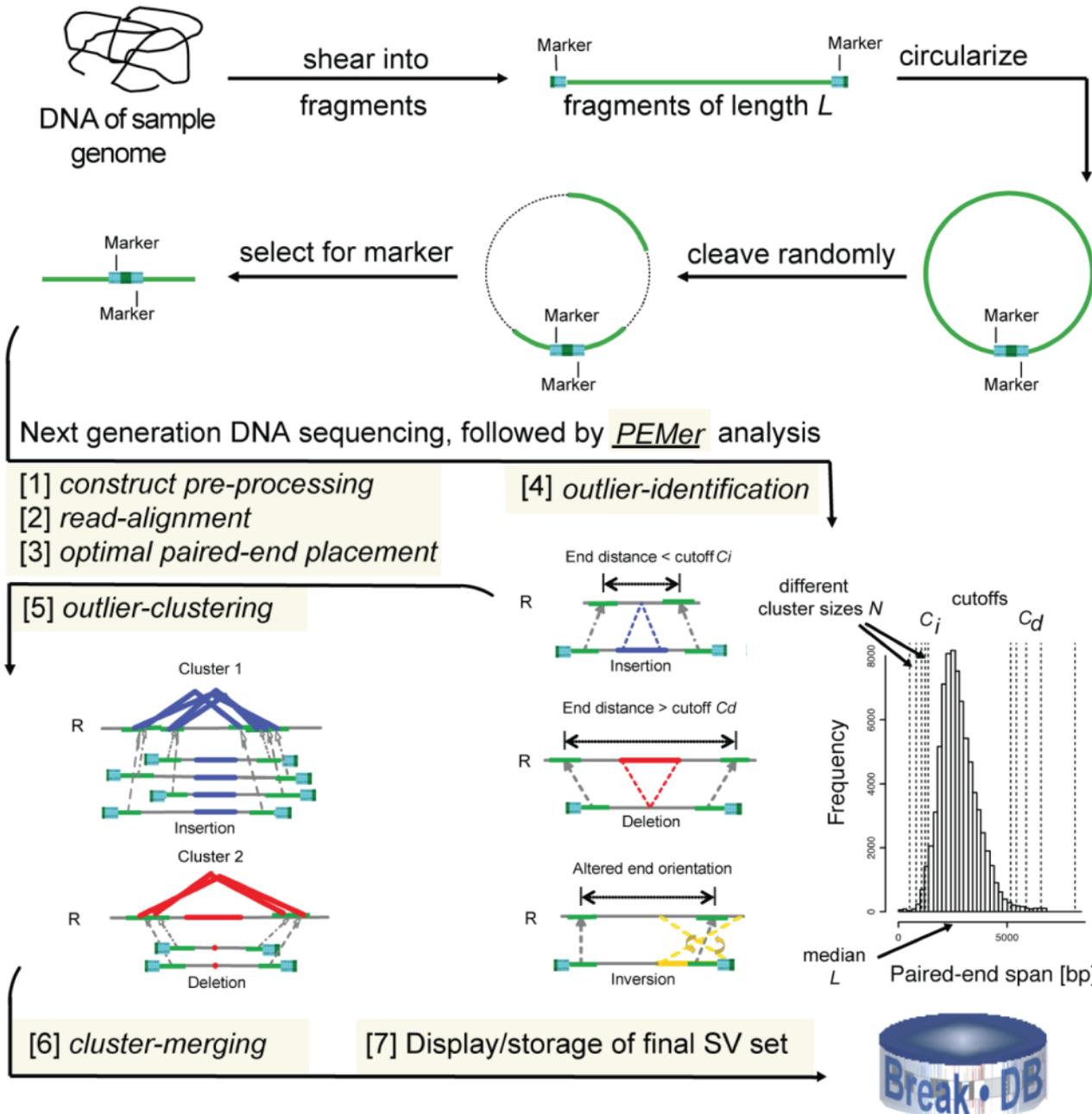
Paired-End

Paired-End Mapping



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

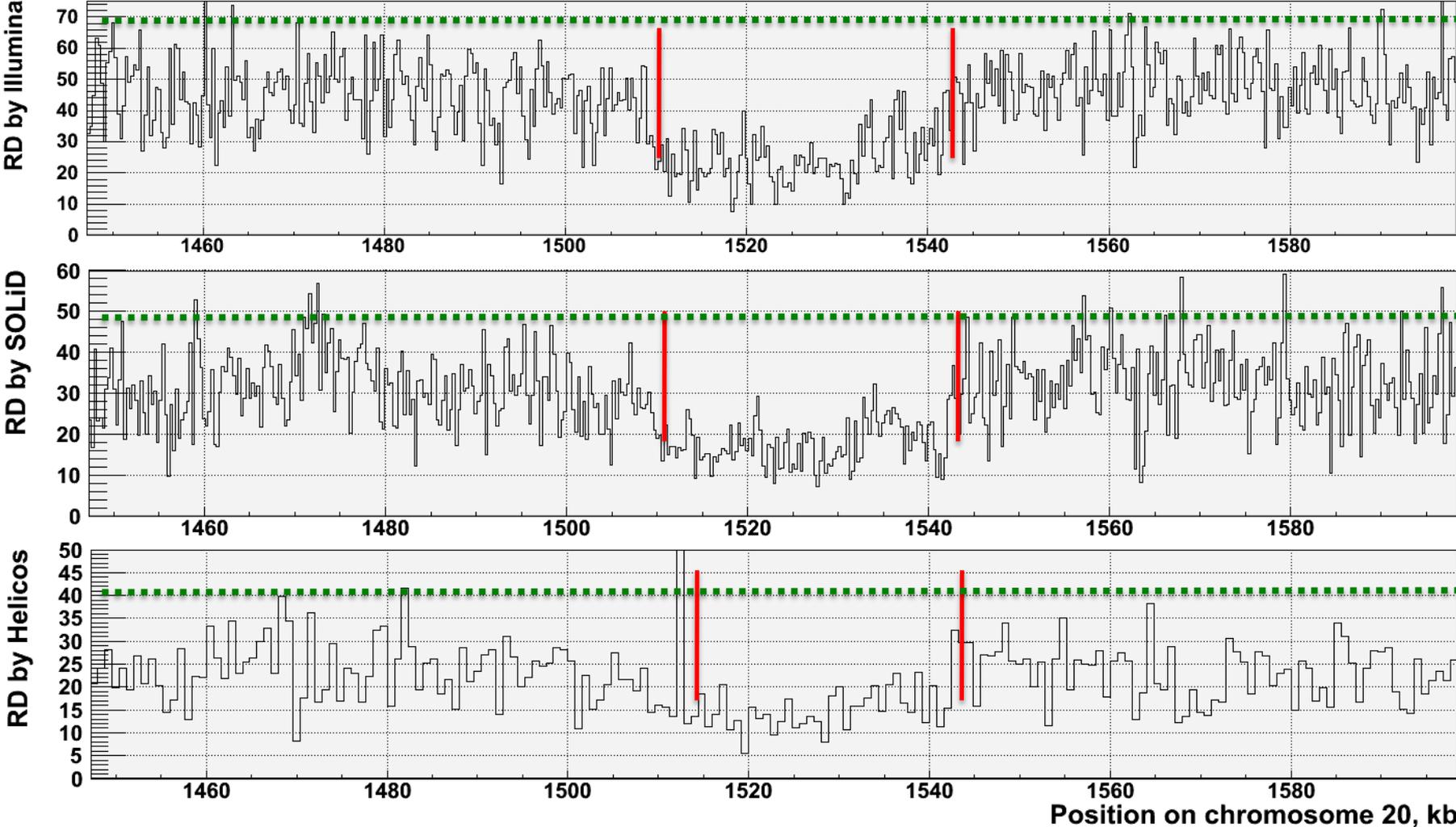
Overall Strategy for Analysis of NextGen Seq. Data to Detect Structural Variants



[Korbel et al.,
 Science ('07);
 Korbel et al.,
 GenomeBiol. ('09)]

Split Read

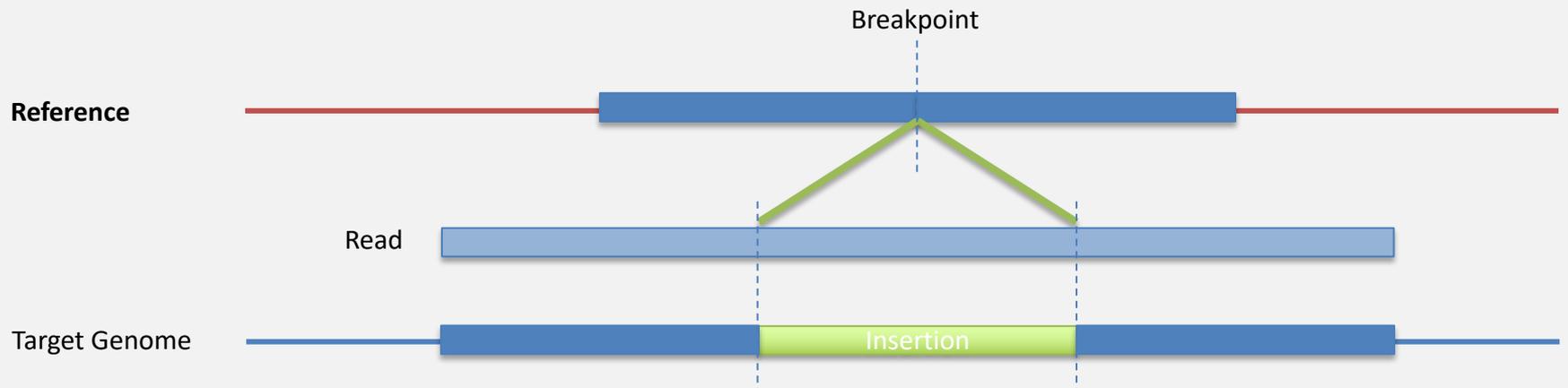
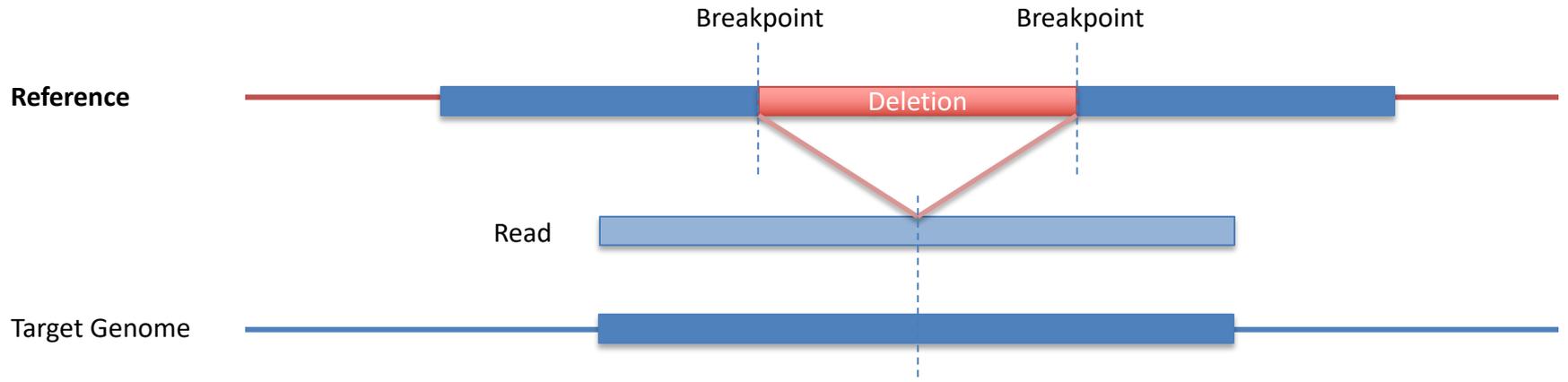
Read-depth works well on a variety of sequencing platforms but provides imprecise breakpoints



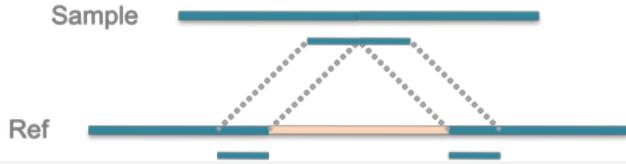
[Abyzov et al. Gen. Res. ('11)]

[NA18505]

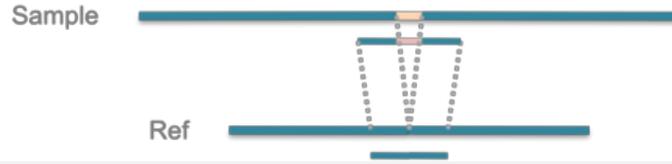
Split-read Analysis



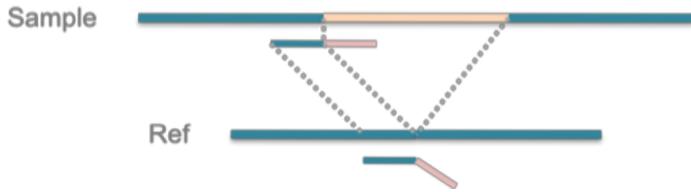
Deletion



Insertion, small

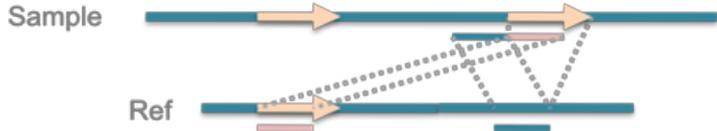


Insertion, large

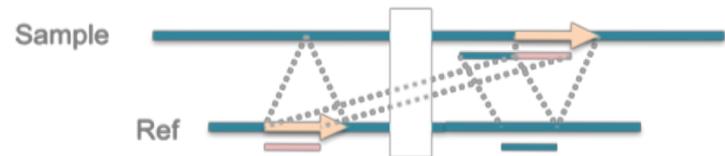
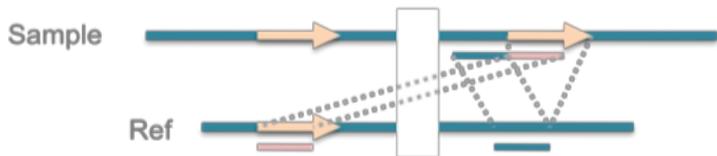
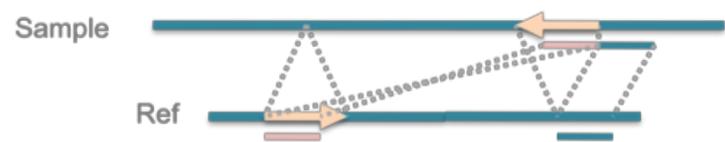
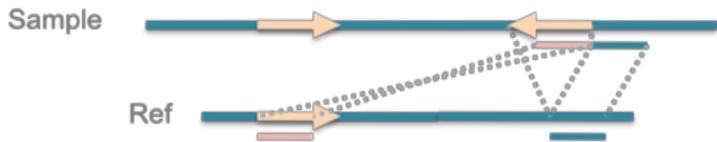
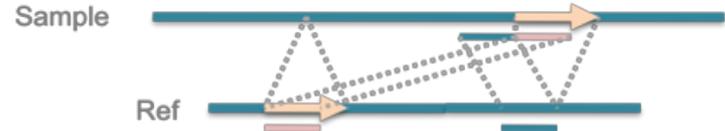


Deletions are the Easiest to Identify

Duplication

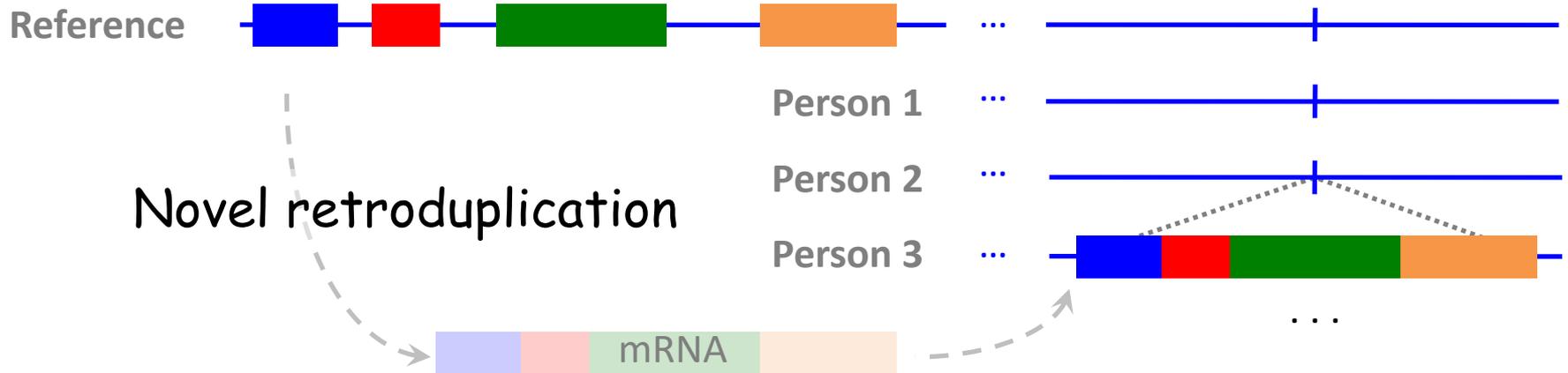
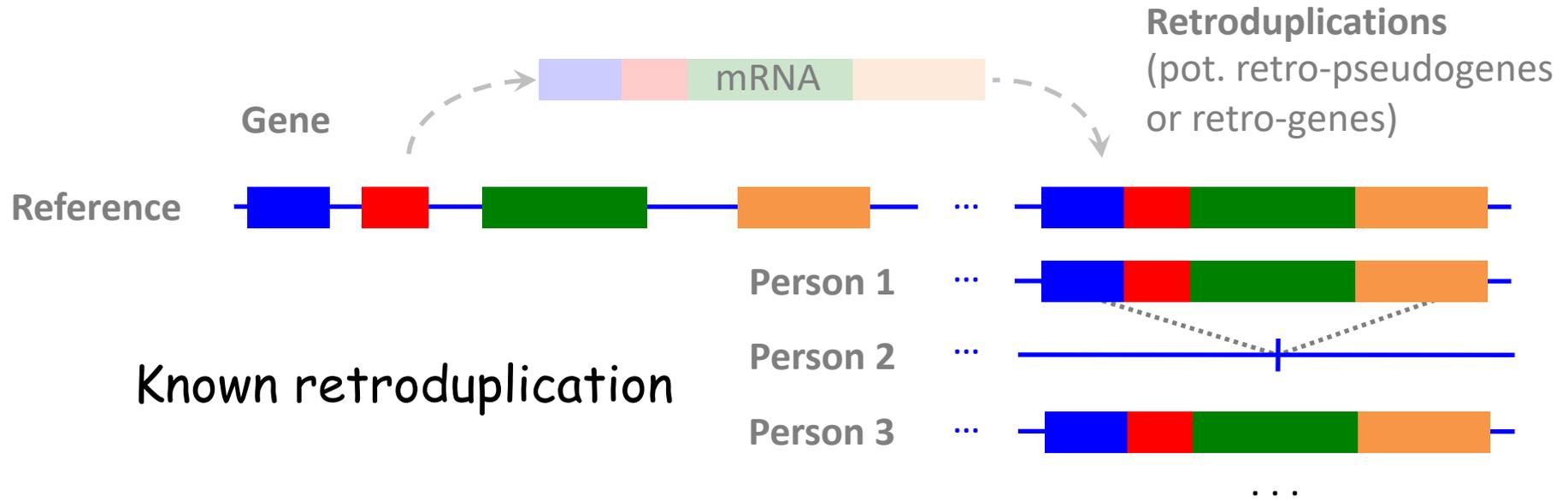


Translocation

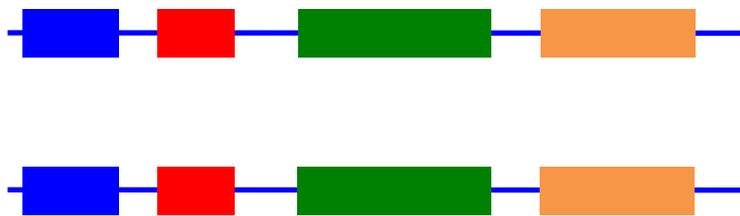


RDV & Mobile Elements

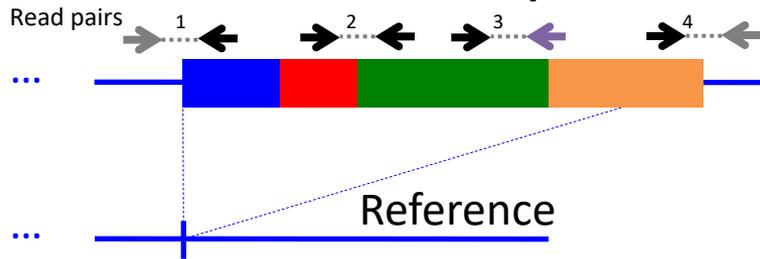
Retroduplication variation (RDV)



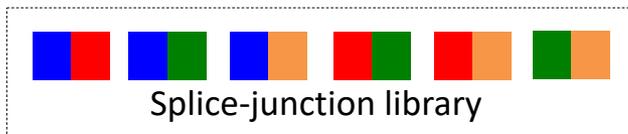
Gene



Novel retroduplication



Alignment to the reference

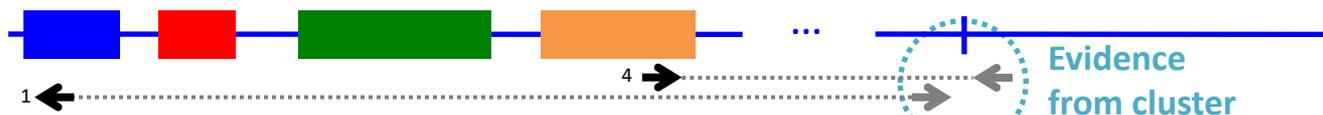


Evidence from alignment



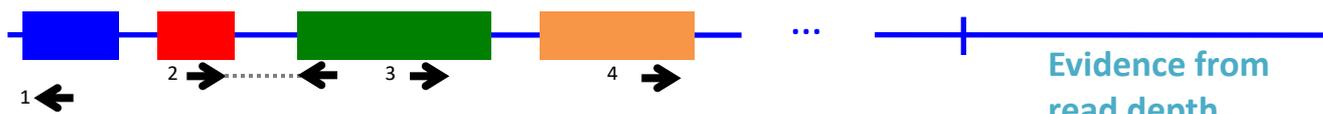
1

Aligned reads



Evidence from cluster

2



Evidence from read depth

3

Pipeline to identify novel retrodups. from 3 evidence sources

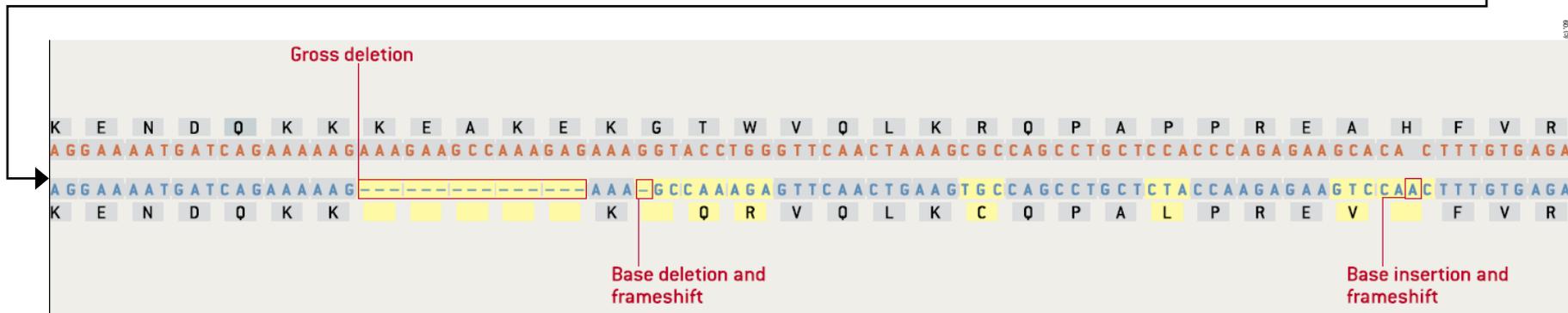
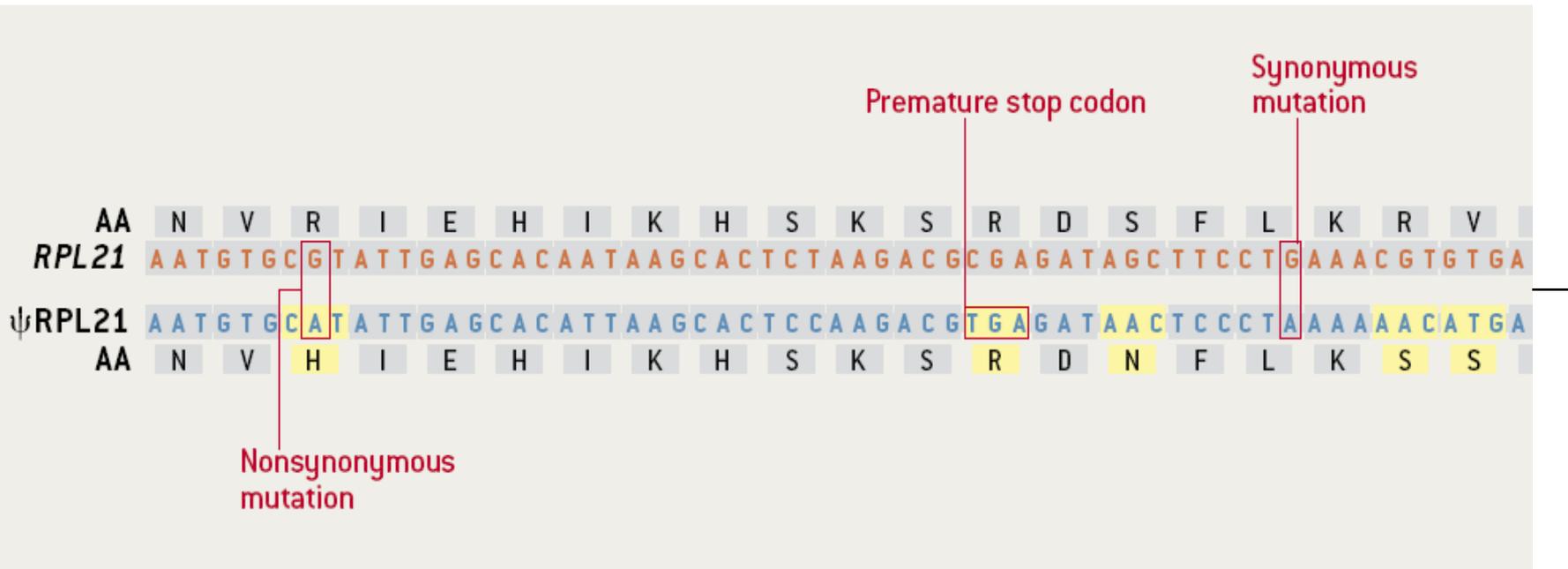
Zero level

Pseudogenes & Genomic Duplications

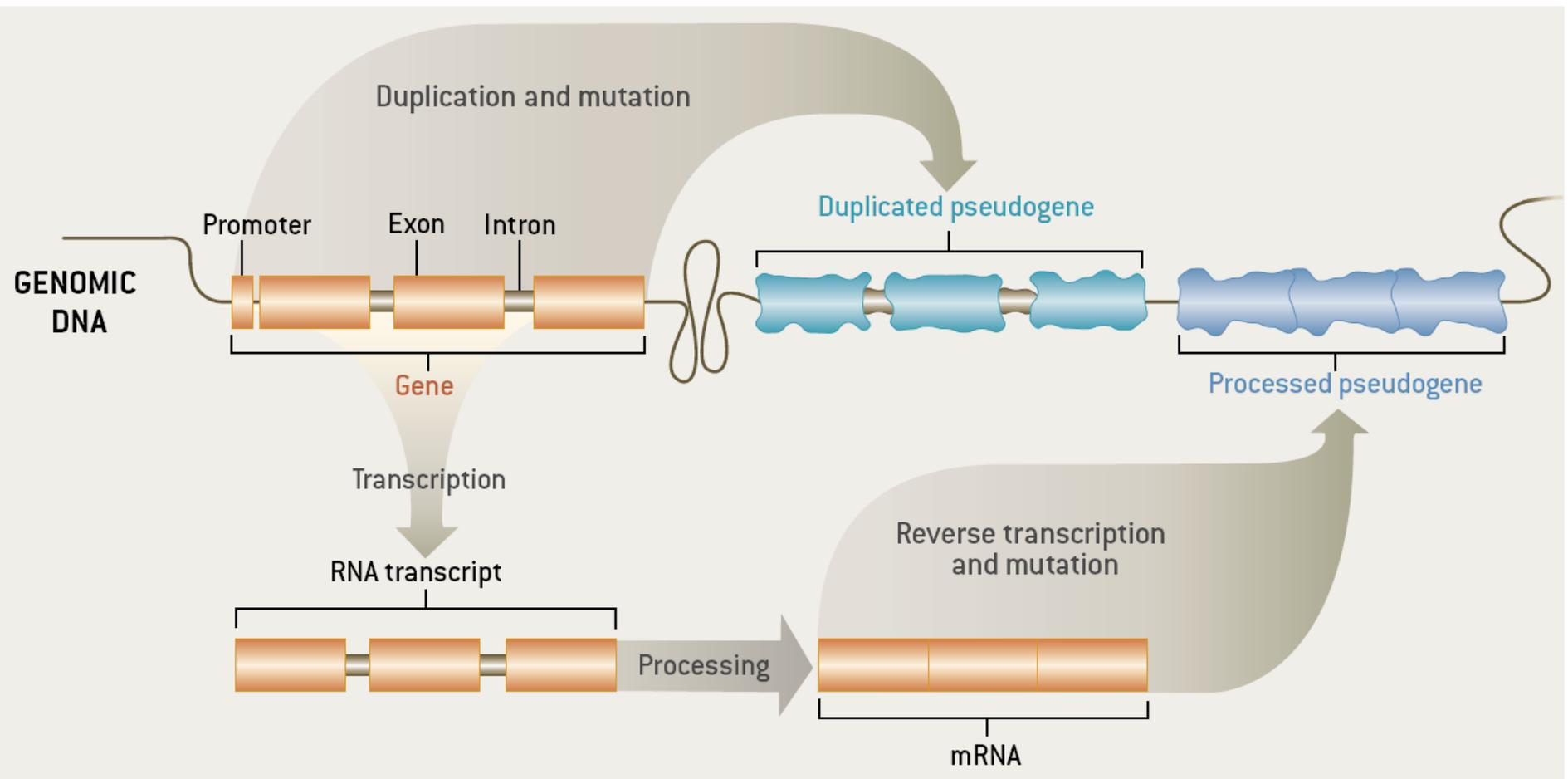
Pseudogenes are among the most interesting intergenic elements

- Formal Properties of Pseudogenes (Ψ G)
 - Inheritable
 - Homologous to a functioning element – ergo a repeat!
 - Non-functional
 - No selection pressure so free to accumulate mutations
 - Frameshifts & stops
 - Small Indels
 - Inserted repeats (LINE/Alu)
 - **What does this mean?** no transcription, no translation?...

Identifiable Features of a Pseudogene (ψ RPL21)



Two Major Genomic Remodeling Processes Give Rise to Distinct Types of Pseudogenes



Impact of Genetic Variability: Loss-of-function

Gene

Polymorphic

Pseudogene

- - Truncating nonsense SNPs
- - Splice-disrupting SNPs
- - Frameshift-causing indels
- - Disrupting structural variants

- Previous LoFs are considered as having high probability of being deleterious
- Surprisingly, ~ 100 LoF variants per genome, 20 genes are completely inactivated
- Among ~100 LoFs, we estimate 2 recessive, close to 0 dominant disease nonsense variants per healthy genome.

Genomic Variation



Al
u

Gene

Ancestral State

Gene

Al
u

Gene

The Genome Remodeling Process

THE GENOME REMODELING PROCESS

Genomic Variation



Non-allelic homologous recombination (NAHR)

Ancestral State



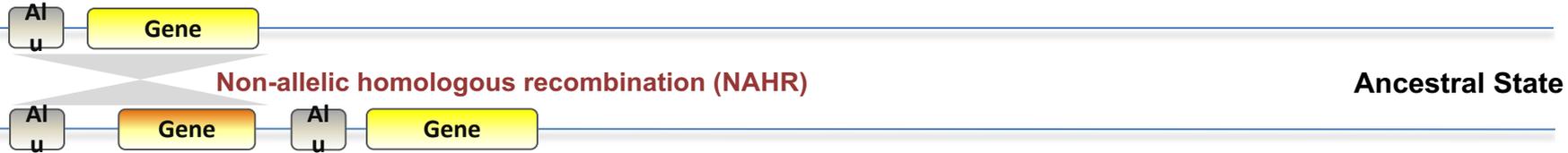
The Genome Remodeling Process

THE GENOME REMODELING PROCESS

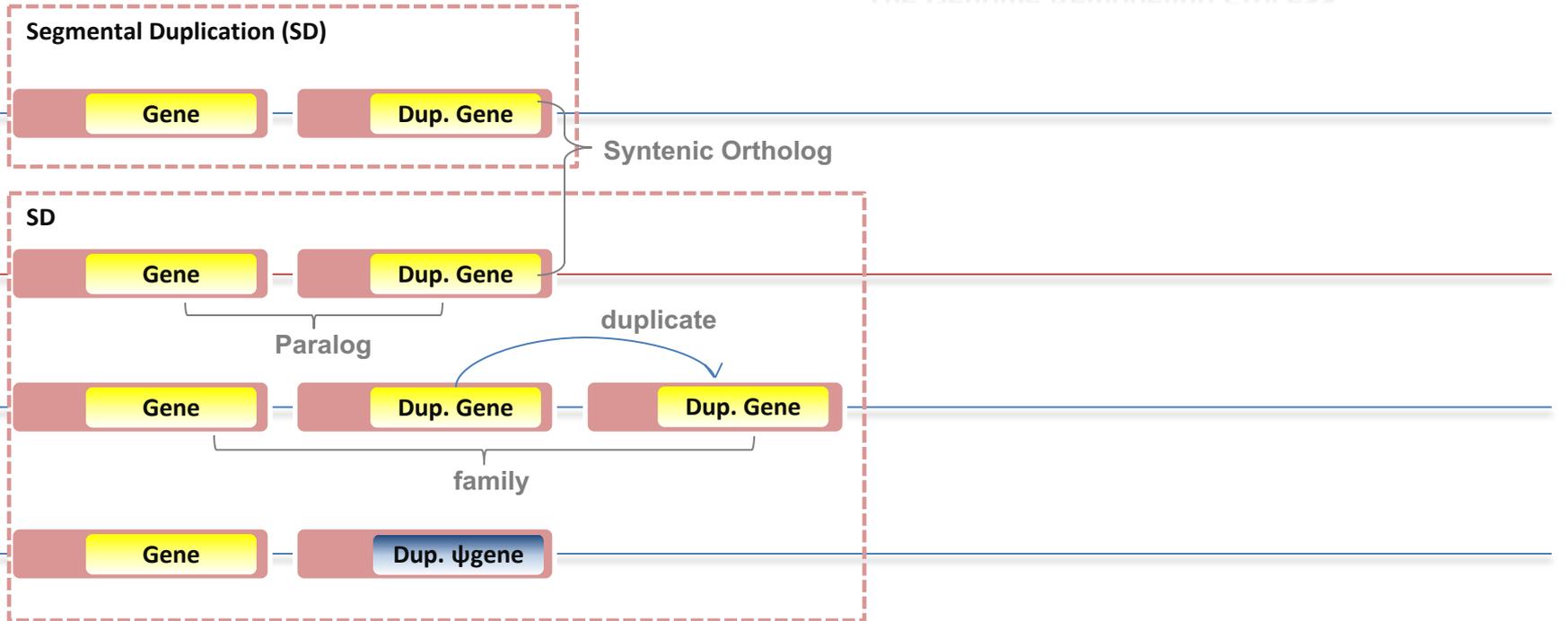
Segmental Duplication (SD)



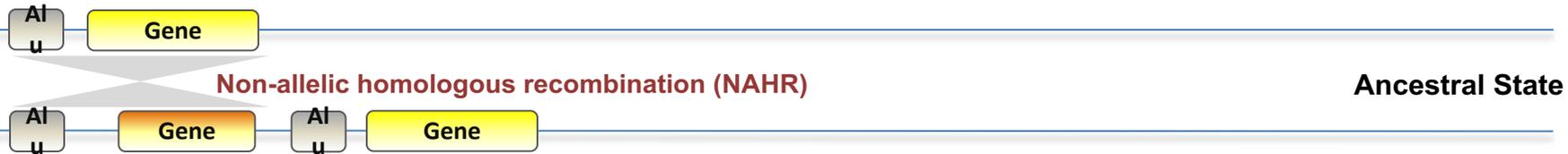
Genomic Variation



The Genome Remodeling Process

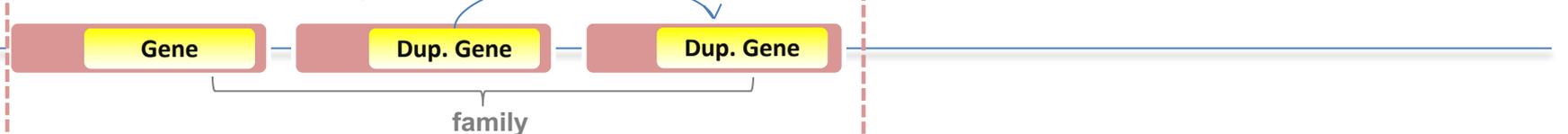
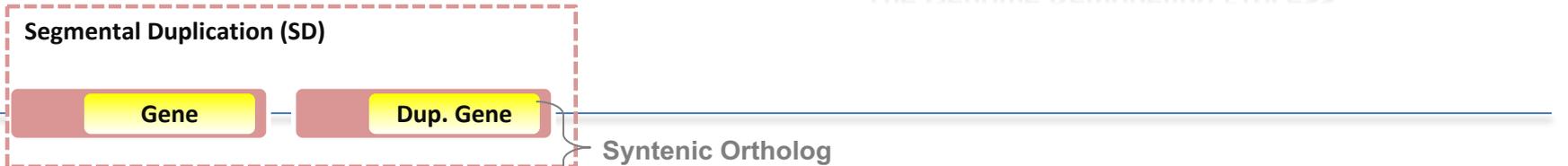


Genomic Variation



Ancestral State

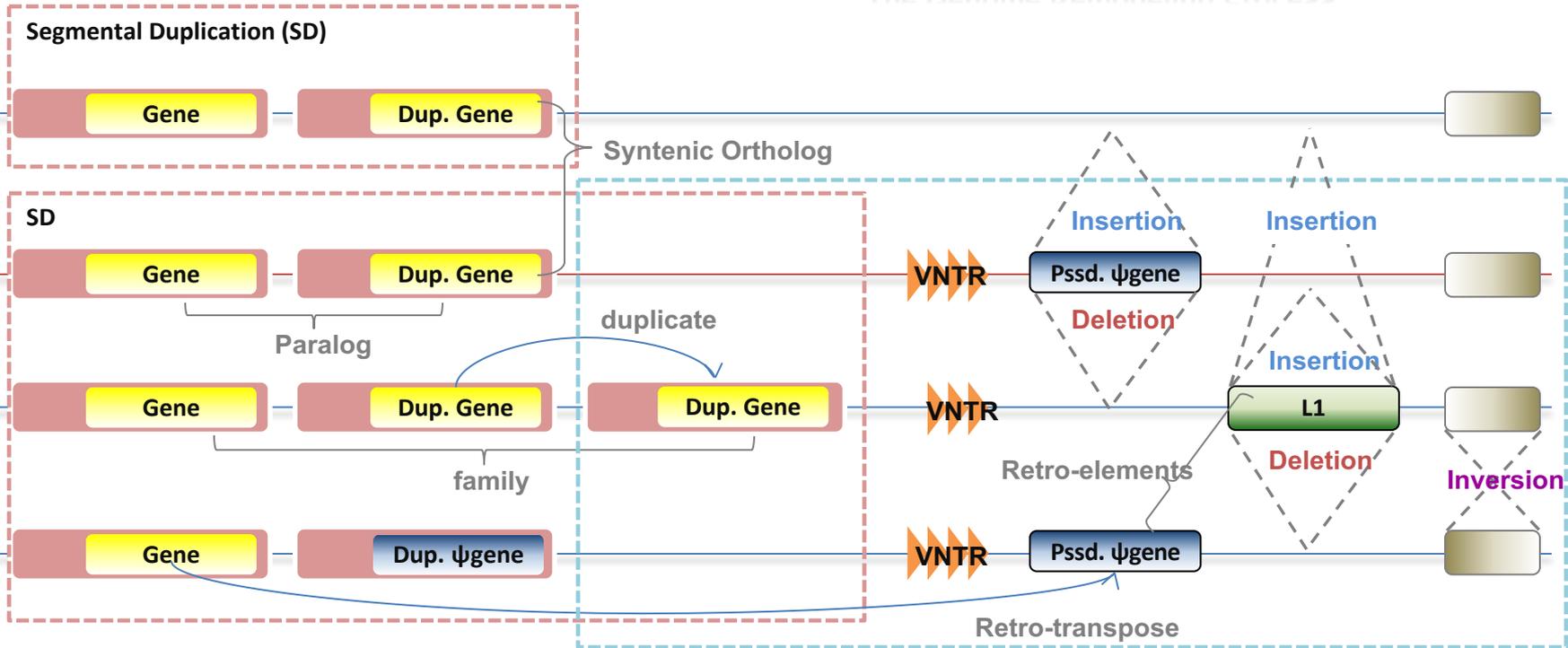
The Genome Remodeling Process



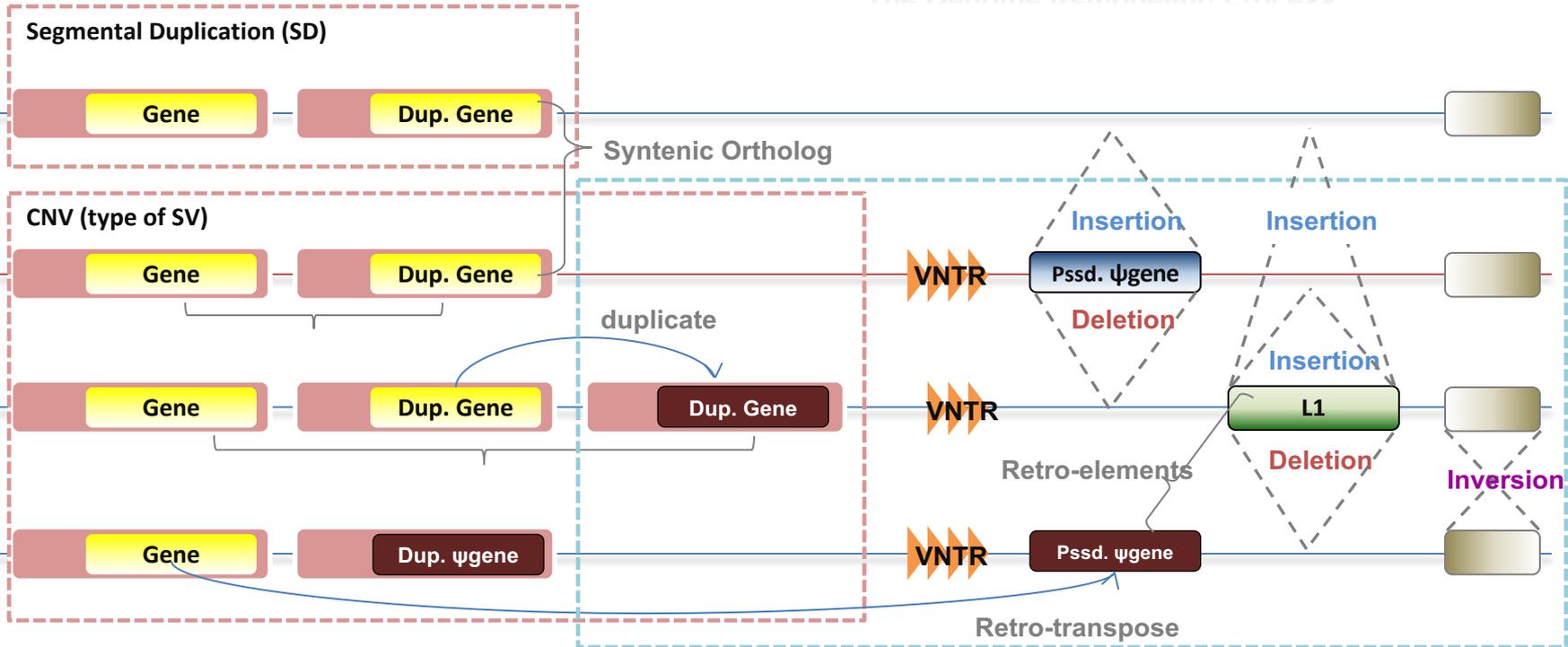
Genomic Variation



The Genome Remodeling Process



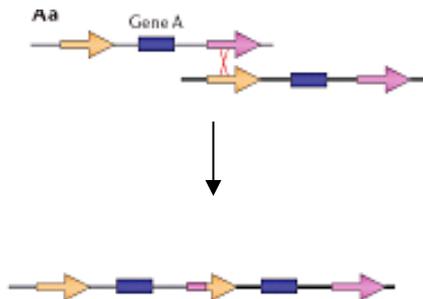
Genomic Variation



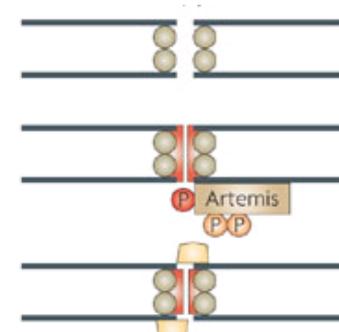
"Polymorphic" Genes & Pseudogenes

Exact Breakpoints & Mechanism Classification

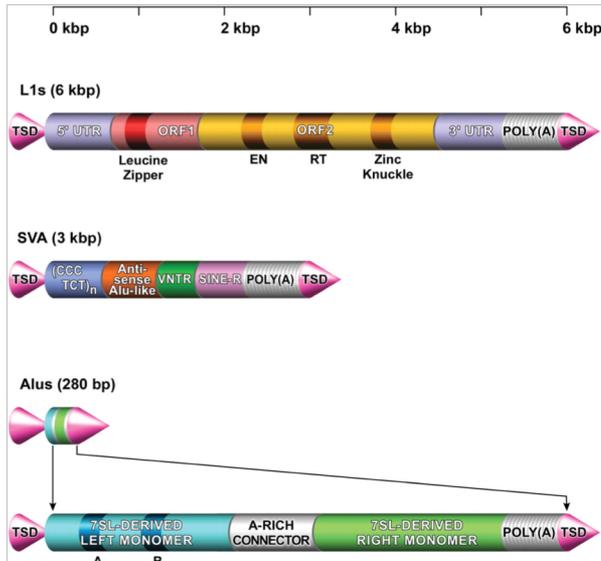
4 mechanisms for SV formation



NAHR
 (Non-allelic homologous recombination)
 Flanking repeat
 (e.g. Alu, LINE...)

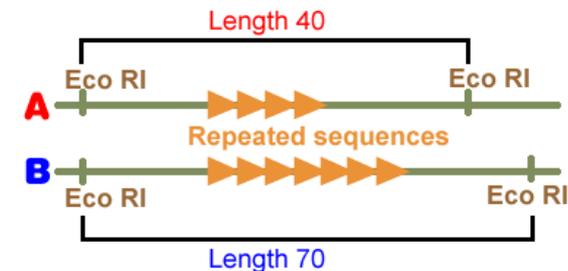


NHEJ (NHR)
 (Non-homologous-end-joining)
 No (flanking) repeats.
 In some cases <4bp microhomologies



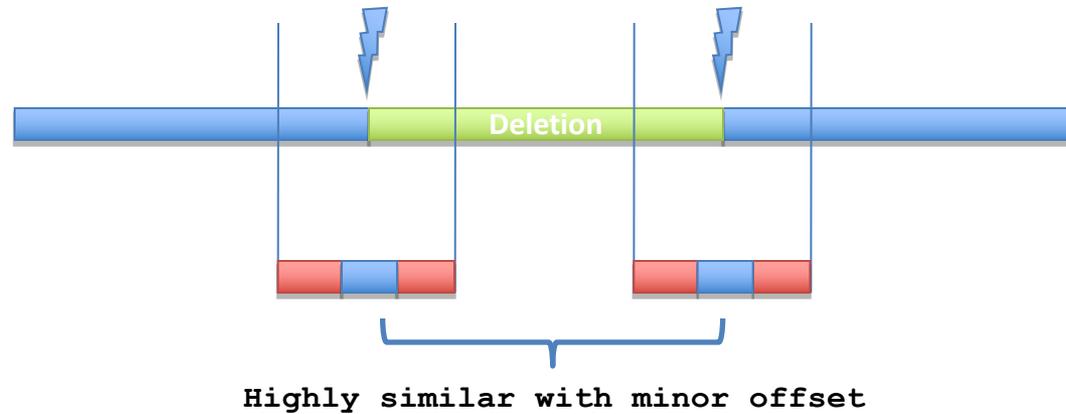
TEI
 (Transposable element insertion)
 L1, SVA, Alus

VNTR
 (Variable Number Tandem Repeats)
 Number of repeats varies between different people



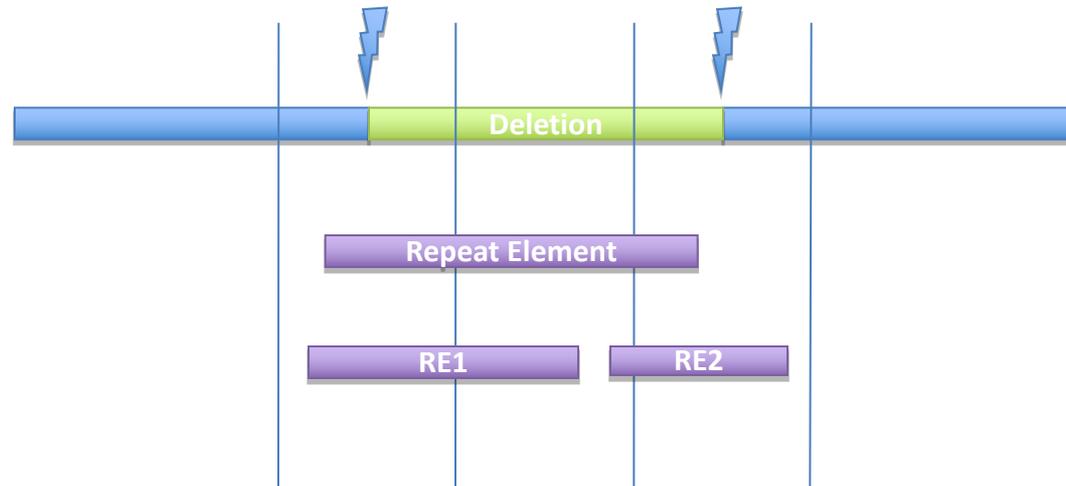
SV Mechanism Classification

NAHR



Single RETRO

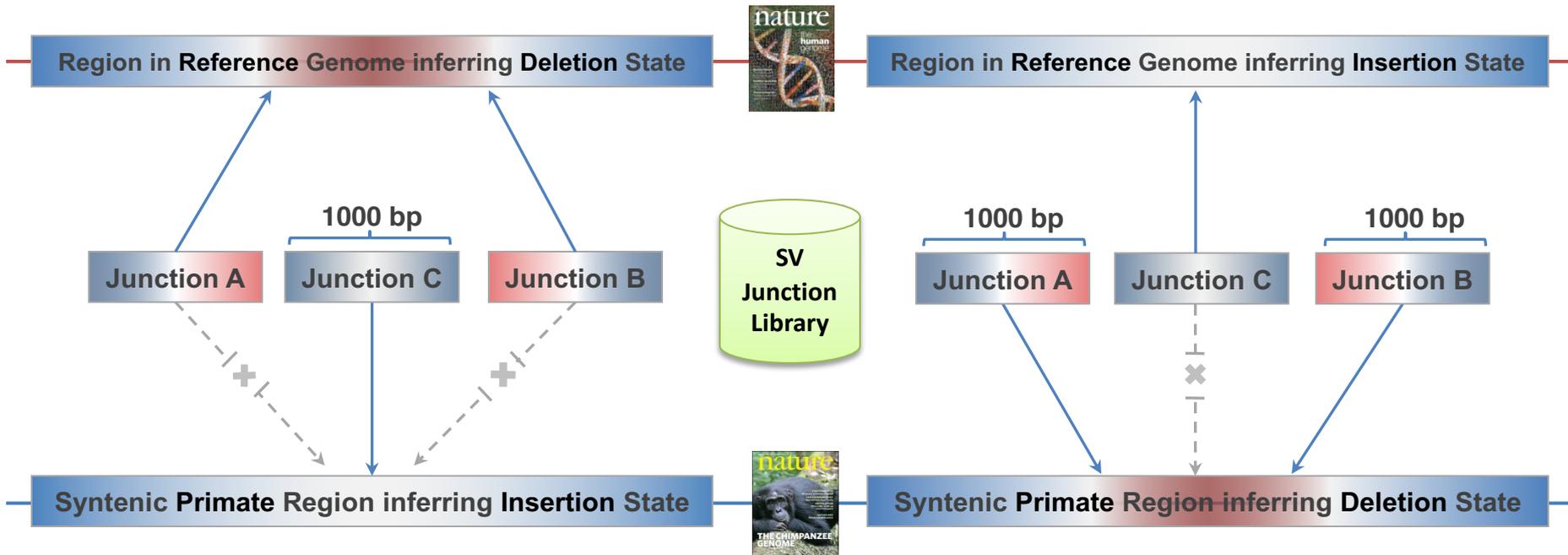
Multiple RETRO



SV Ancestral State Analysis

Inferring **Insertion** according to **Ancestral State**

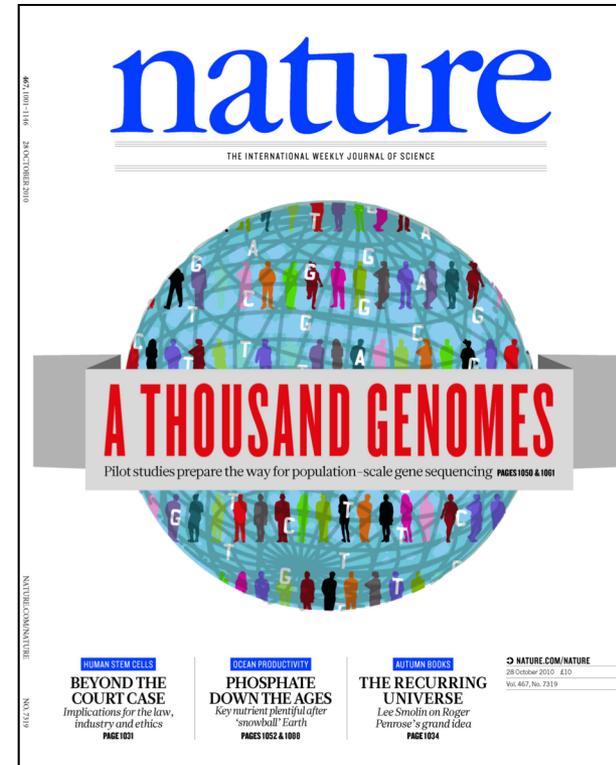
Inferring **Deletion** according to **Ancestral State**



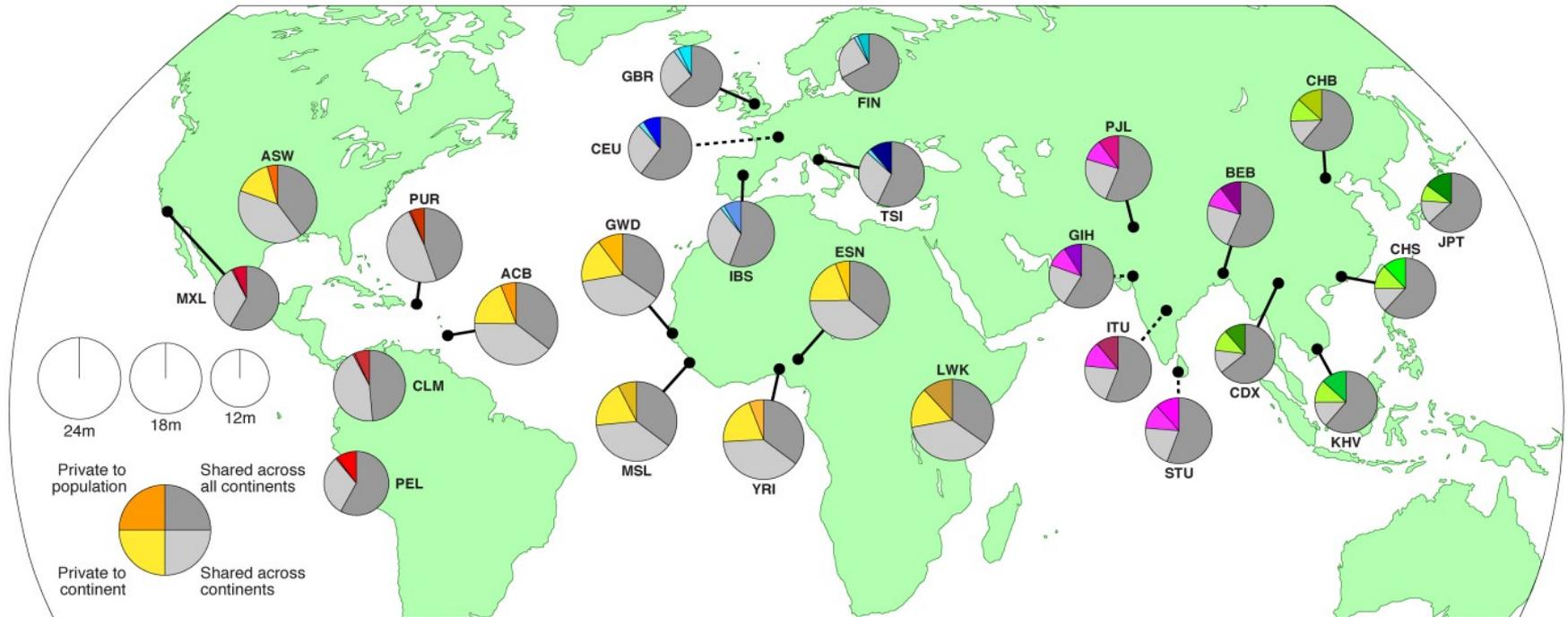
1000G summary

1000G SV (Pilot, Phase I & III)

- **Many different callers compared & used**
 - including SRiC & CNVnator but also VariationHunter, Cortex, NovelSeq, PEMer, BreakDancer, Mosaik, Pindel, GenomeSTRiP, mrFast....
- Merging
- Genotyping (GenomeSTRiP)
- Breakpoint assembly (AGE & Tigras_V)
- Mechanism Classification



Summary Stats of 1000GP SV Phase3



- 68,818 SVs
- 2,504 unrelated individuals
- 26 populaSons
- 37,250 SVs with resolved breakpoints

[2] 1000GP Phase3 SV paper. Submided to Nature, 2015.

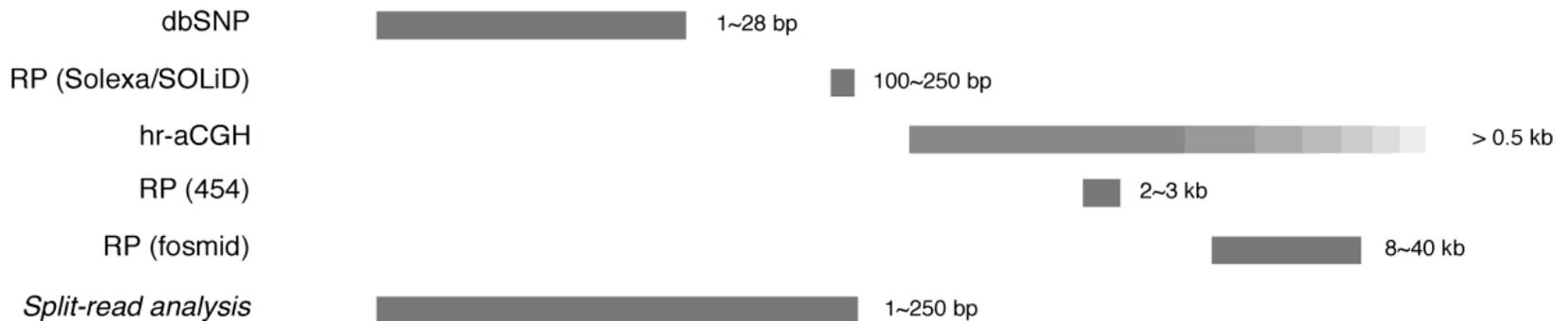
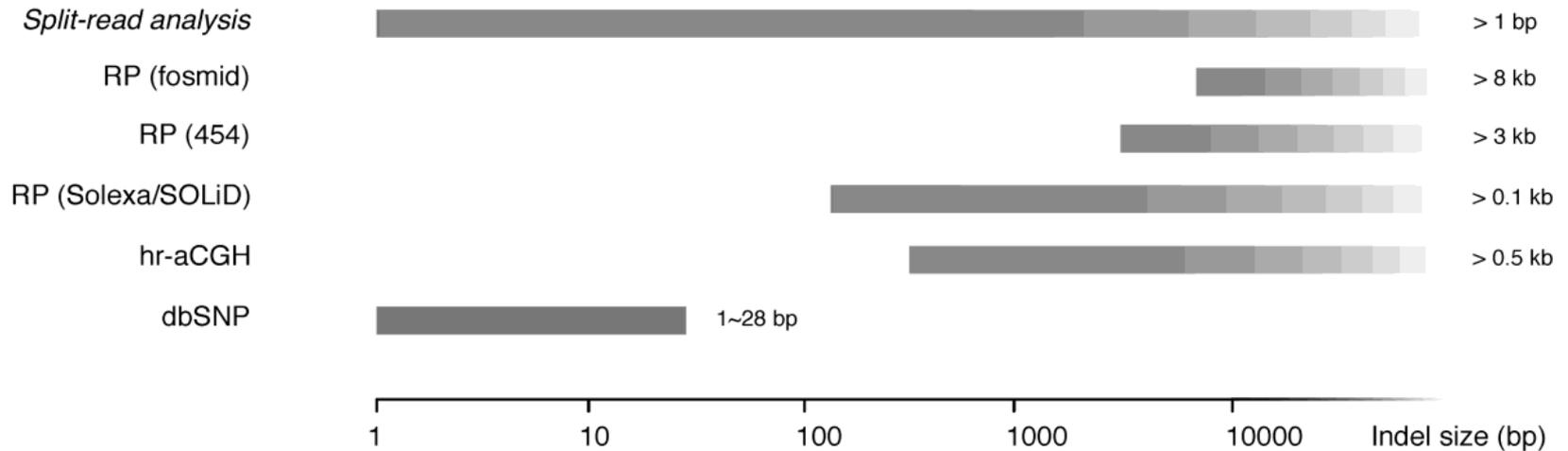
[3] 1000GP ConsorSum. Submided to Nature, 2015.

Phase 3: Median Autosomal Variant Sites Per Genome

	AFR		AMR		EAS		EUR		SAS	
Samples	661		347		504		503		489	
Mean Coverage	8.2		7.6		7.7		7.4		8.0	
	Var. Sites	Singletons								
SNPs	4.31M	14.5k	3.64M	12.0k	3.55M	14.8k	3.53M	11.4k	3.60M	14.4k
Indels	625k	-	557k	-	546k	-	546k	-	556k	-
Large Deletions	1.1k	5	949	5	940	7	939	5	947	5
CNVs	170	1	153	1	158	1	157	1	165	1
MEI (Alu)	1.03k	0	845	0	899	1	919	0	889	0
MEI (LINE1)	138	0	118	0	130	0	123	0	123	0
MEI (SVA)	52	0	44	0	56	0	53	0	44	0
MEI (MT)	5	0	5	0	4	0	4	0	4	0
Inversions	12	0	9	0	10	0	9	0	11	0
NonSynon	12.2k	139	10.4k	121	10.2k	144	10.2k	116	10.3k	144
Synon	13.8k	78	11.4k	67	11.2k	79	11.2k	59	11.4k	78
Intron	2.06M	7.33k	1.72M	6.12k	1.68M	7.39k	1.68M	5.68k	1.72M	7.20k
UTR	37.2k	168	30.8k	136	30.0k	169	30.0k	129	30.7k	168
Promoter	102k	430	84.3k	332	81.6k	425	82.2k	336	84.0k	430
Insulator	70.9k	248	59.0k	199	57.7k	252	57.7k	189	59.1k	243
Enhancer	354k	1.32k	295k	1.05k	289k	1.34k	288k	1.02k	295k	1.31k
TFBS	927	4	759	3	748	4	749	3	765	3
Filtered LoF	182	4	152	3	153	4	149	3	151	3
HGMD-DM	20	0	18	0	16	1	18	2	16	0
GWAS	2.00k	0	2.07k	0	1.99k	0	2.08k	0	2.06k	0
ClinVar	28	0	30	1	24	0	29	1	27	1

Different Approaches Work Differently on Different Events

Deletions



Insertions