

A circular genome plot (Circos plot) showing genomic data. The plot consists of concentric tracks. The outermost track is a dense network of colored arcs (blue, green, orange, red) representing genomic interactions or structural variants. Inner tracks show numerical values and colored bars, likely representing copy number variations or other genomic metrics. The plot is divided into segments, possibly representing chromosomes or genomic regions.

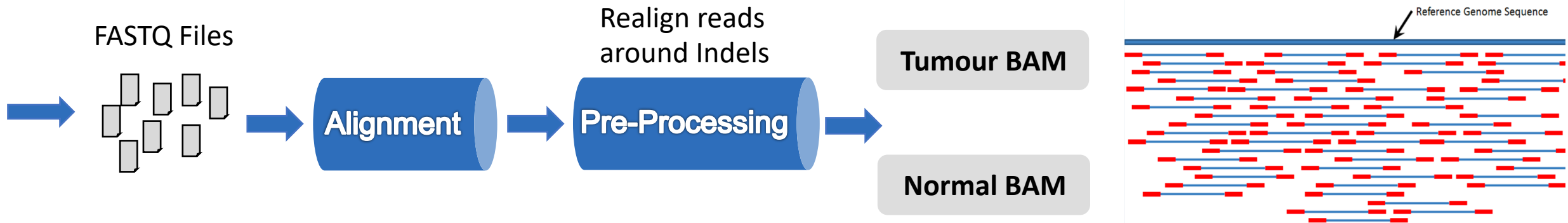
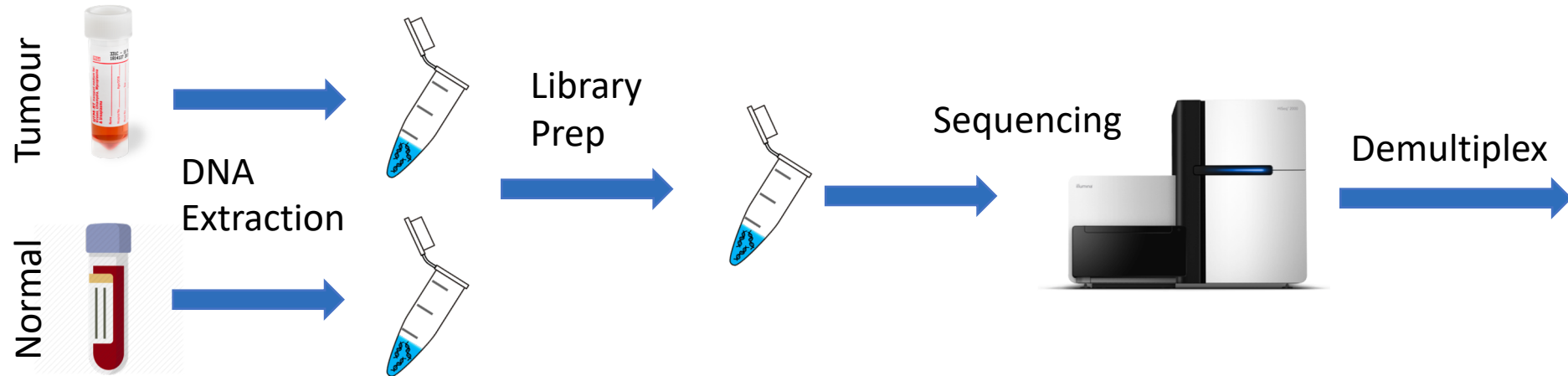
Detecting variants in DNA sequencing data

MBP Bioinformatics Tech Talk

Jeff Bruce, PhD

jbruce@uhnresearch.ca

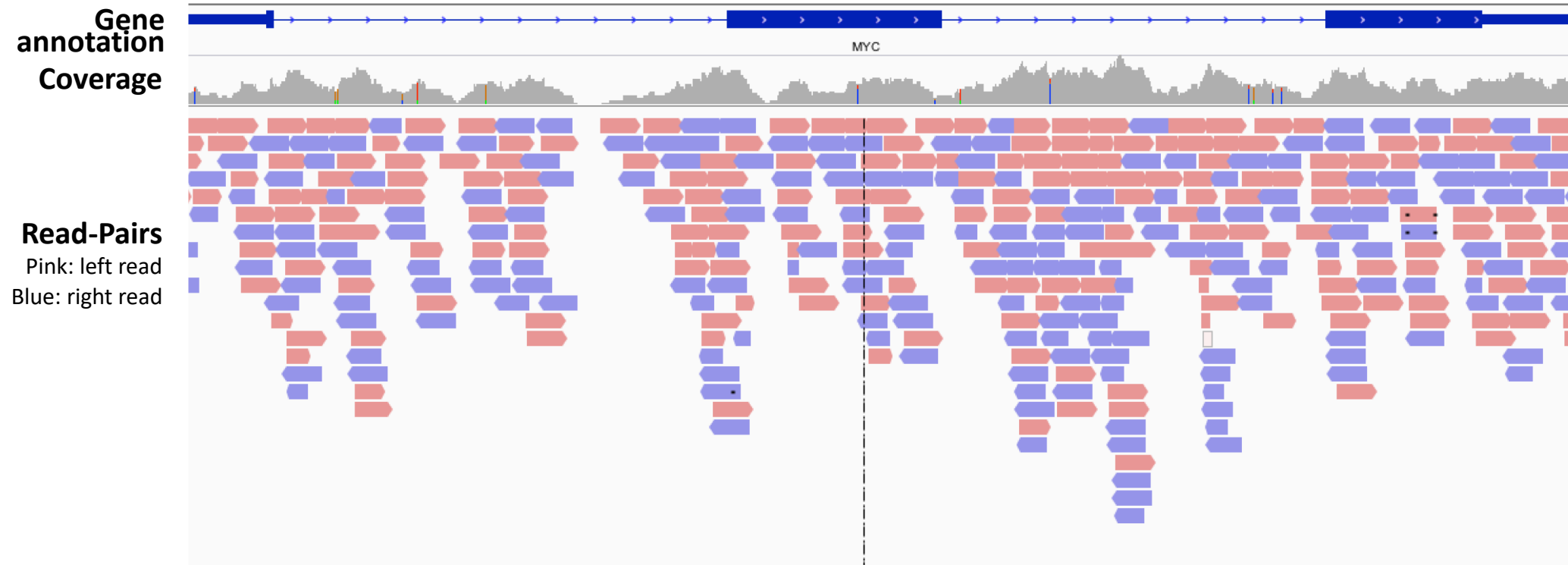
Brief Overview: Sequencing and Alignment



Brief Overview: Sequencing and Alignment

Sequence Alignment/Map (SAM)

Binary representation of SAM = BAM

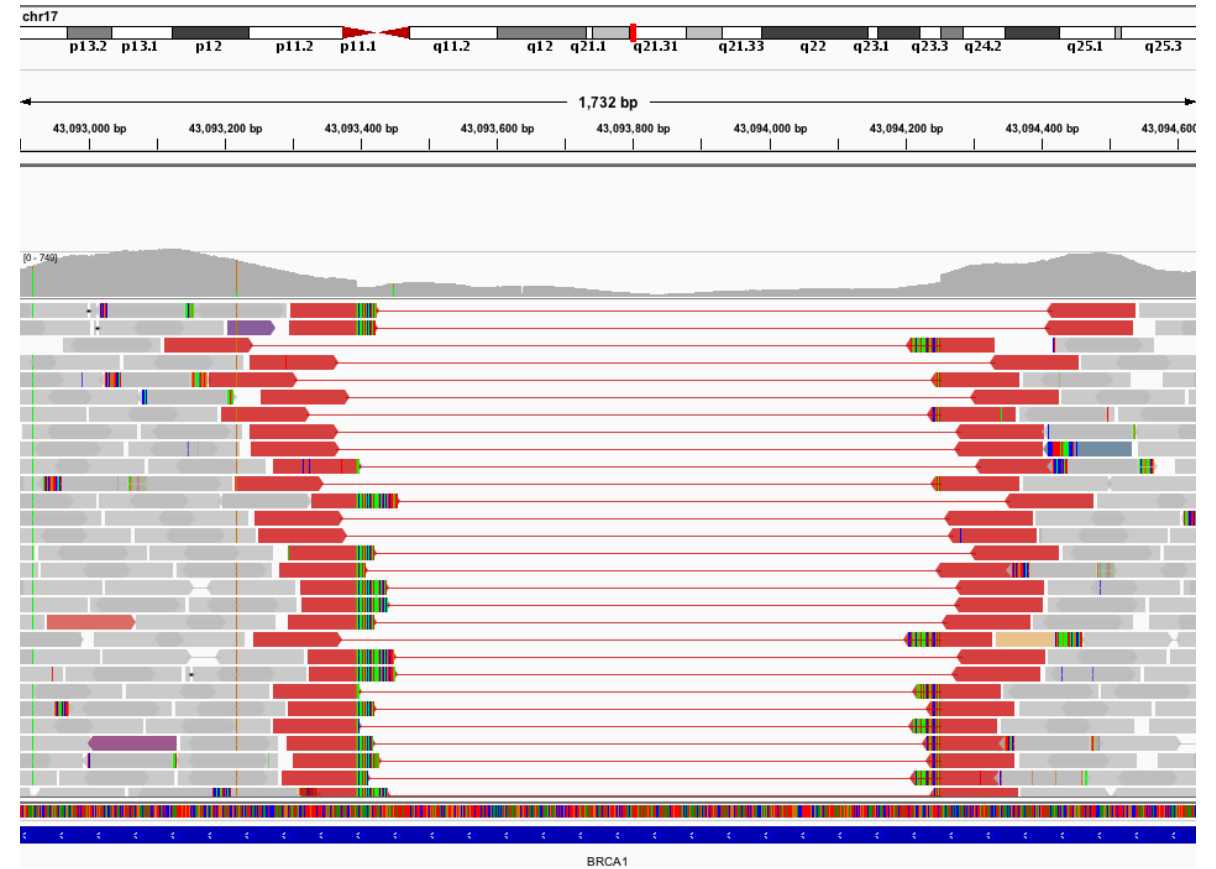
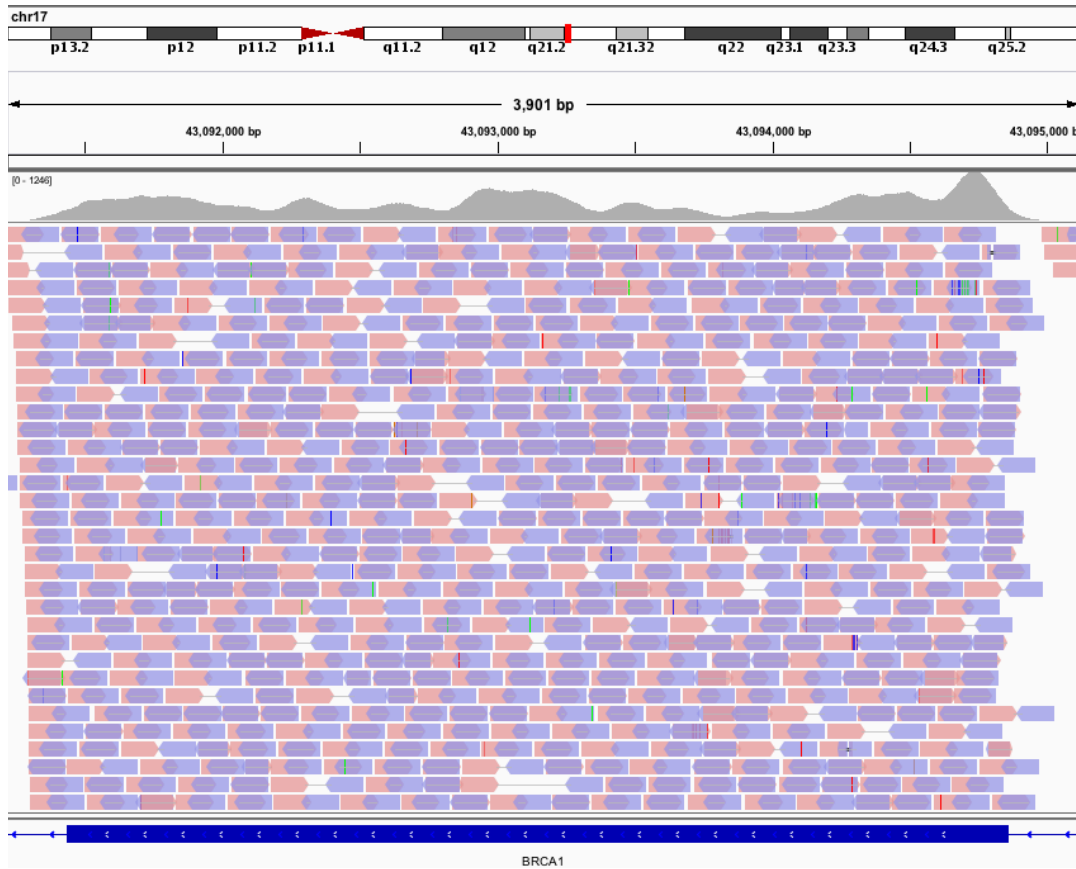


Brief Overview: Integrative Genomics Viewer



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

Brief Overview: Integrative Genomics Viewer (IGV)



Brief Overview: HPC/Linux/command line

- **Some HPC options in Toronto/Ontario**
 - [Mordor](#), [HPC4Health](#), [SciNet](#), [SickKids HPF](#), [SharkNet](#)
- **Cloud HPC options**
 - [AWS \(amazon\)](#)/[Google cloud](#)/[Microsoft Azure](#)
 - [Cancer Genomics Cloud](#)
 - [FireCloud](#) (broad institute)
- **Why use command-line tools for computational biology?**

The field is always changing so we need tools that are:

- *Relatively* simple/quick/inexpensive to create
- Open source/easily modifiable
- Distributable across HPC cluster/cloud nodes
- Lightweight and portable

Brief Overview: HPC/Linux/command line

Tools to access remote HPC clusters

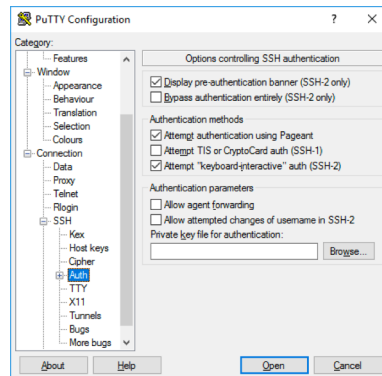
- MacOS and Linux:

[Terminal](#)

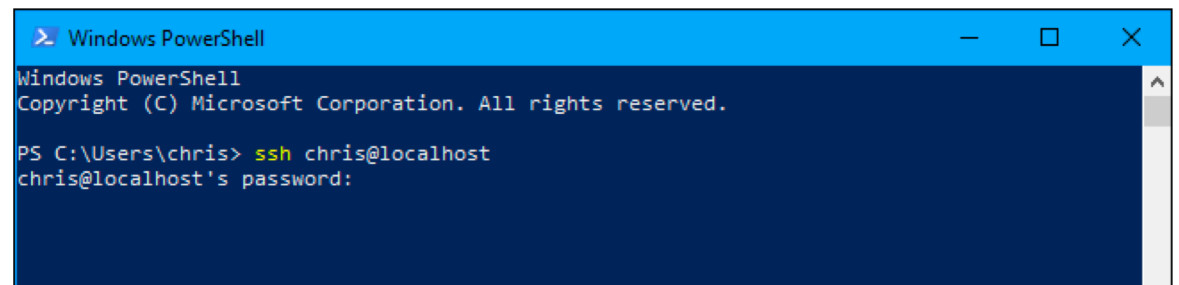


```
jbruce — -bash — 67x7
Last login: Thu Oct 31 19:30:22 on ttys001
locals-MacBook-Pro:~ jbruce$
```

- Windows: [PuTTY](#)



[Windows PowerShell](#)



Terminology:

- **Variant:** Any genomic sequence that differs from a given reference
- **Germline Variant:** Inherited; present in all* of the cells of the individual
- **Somatic Variant:** Variants that are not inherited or passed on to offspring through the germline. In cancer, these are tumour specific
- **Mutation:**
 - Germline - Based on population frequency (<1% of a given population)
 - **The physical event resulting in a change to the genome**
 - Sometimes used Interchangeably with “variant”
- **Polymorphism:**
 - Germline variants present in >1% of the population

* Or a large proportion in the case of mosaicism

Types of Variants

Single Nucleotide Variant



Deletion



Insertion



Tandem Duplication



Interspersed Duplication



Inversion



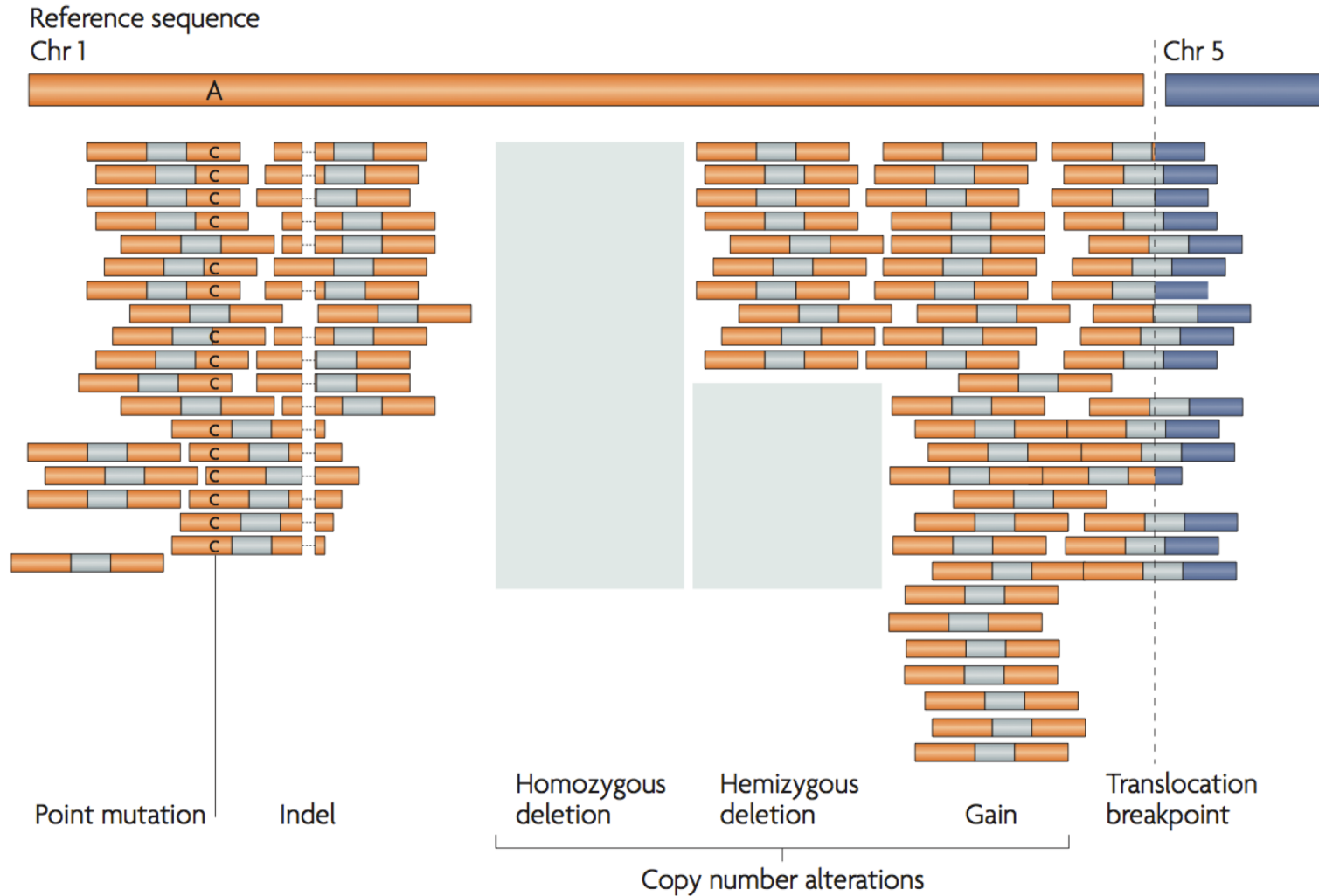
Translocation



Copy Number Variant




Types of Variants



Meyerson *et al.* *Nat Rev Genet.* 2010 Oct;11(10):685-96.

Sequencing Techniques

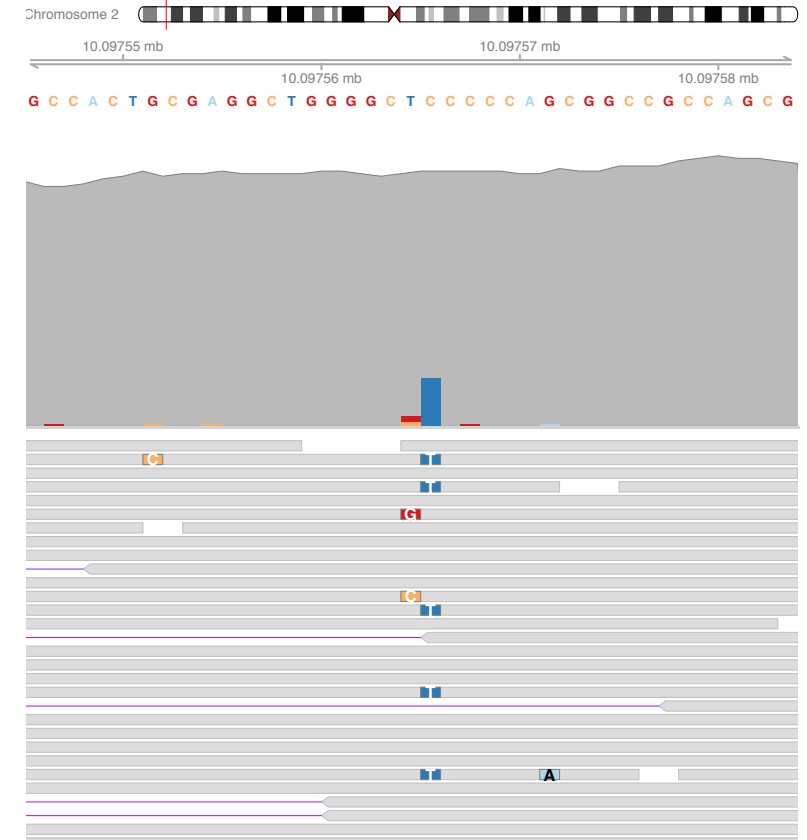


Sequencing Methodology	Detectable Variant Types				
	Coding SNVs/Indels	Non-Coding SNVs/Indels	Exon/Gene CNVs	Mid-Large/Complex SVs	Large CNVs
Sanger					
Targeted Panel		*	*	*	*
Whole Exome		*			
Whole Genome					

Yes
 * With Caveats

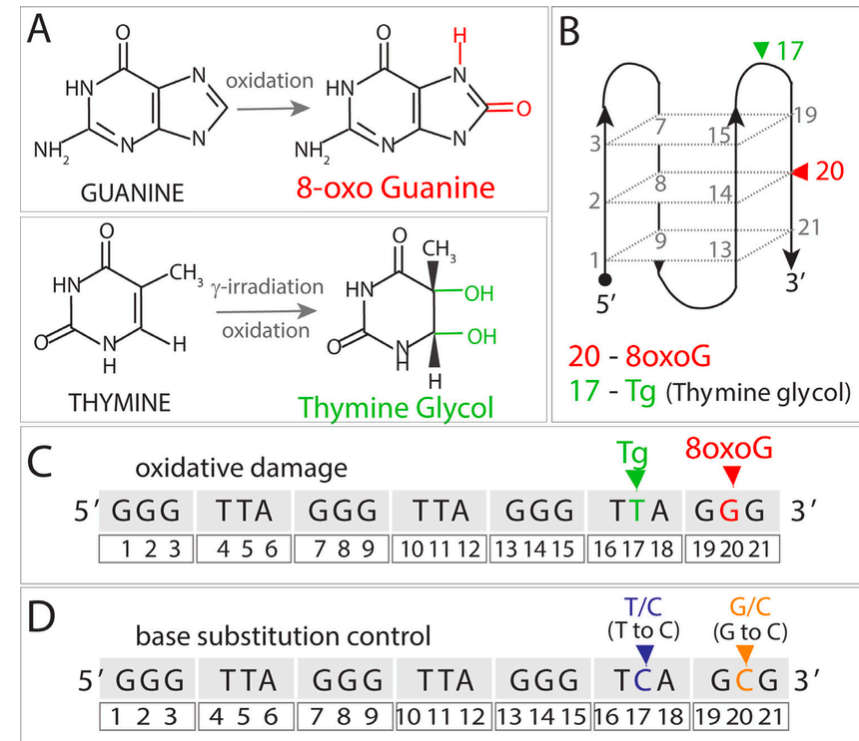
Challenges for variant detection:

- **Low variant read proportion:**
 - Depth of sequencing
 - Normal contamination
 - Subclonality
 - Overlapping copy-number variants
 - Difficult read mapping
- **Sources of false positives:**
 - DNA damage due to processing
 - PCR errors
 - Sequencing errors
 - Mapping artifacts
- **Sources of false negatives:**
 - Poor mapping regions
 - Depth of sequencing
 - Local error rate
 - Variant complexity and size



Challenges for variant detection:

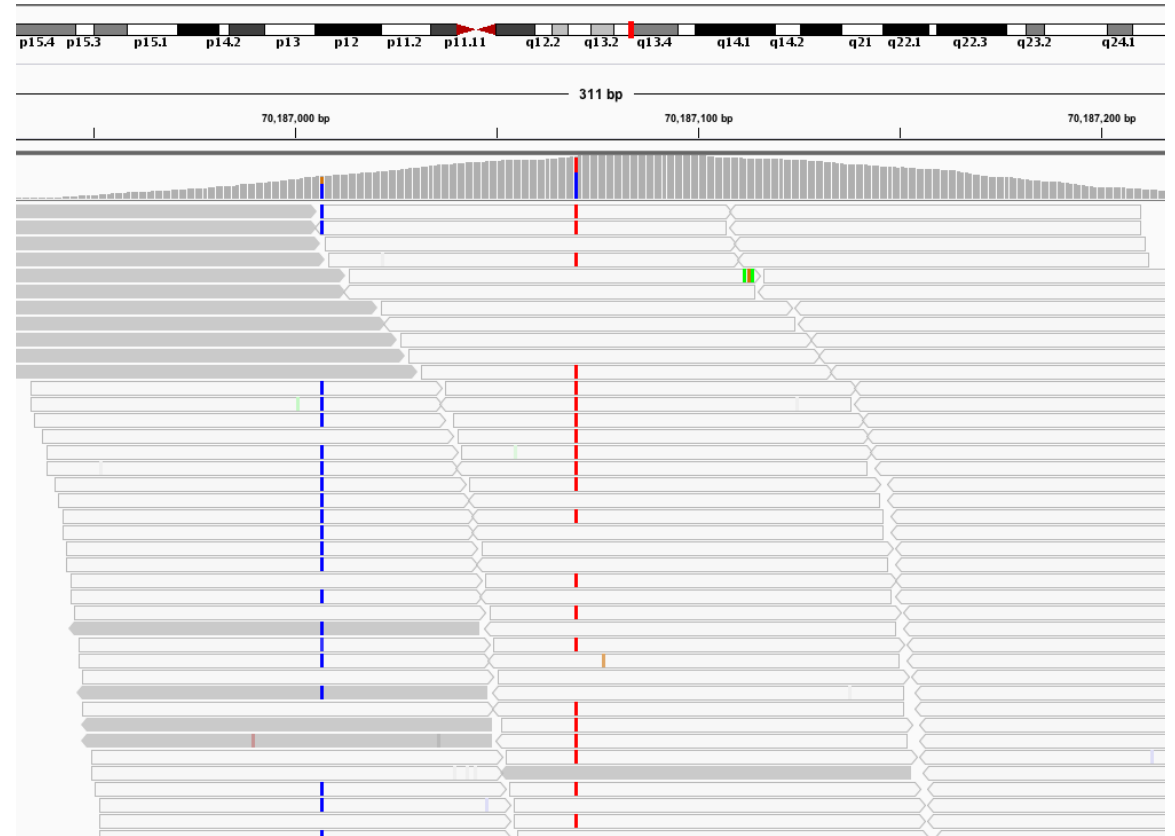
- **Low variant read proportion:**
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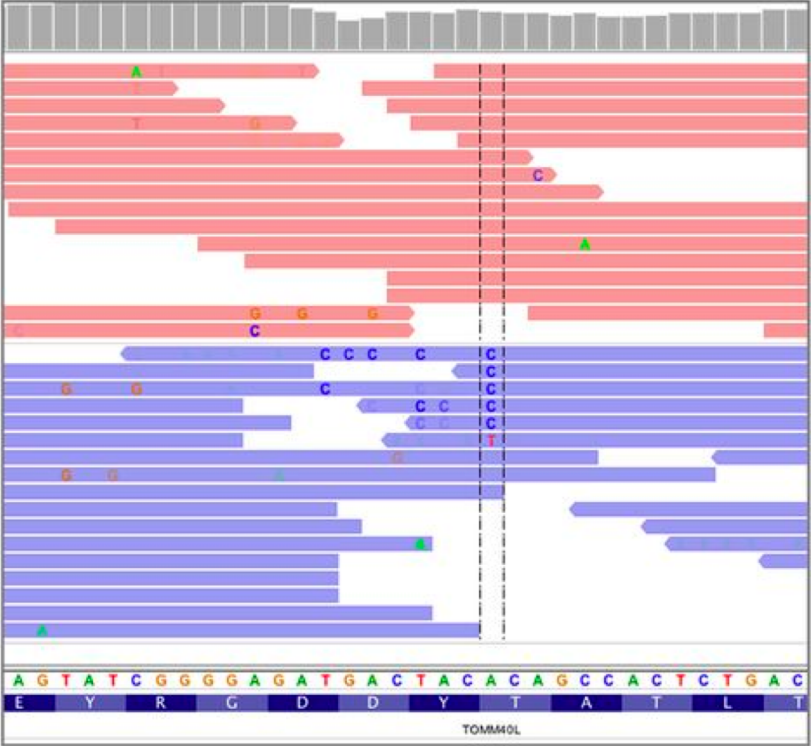
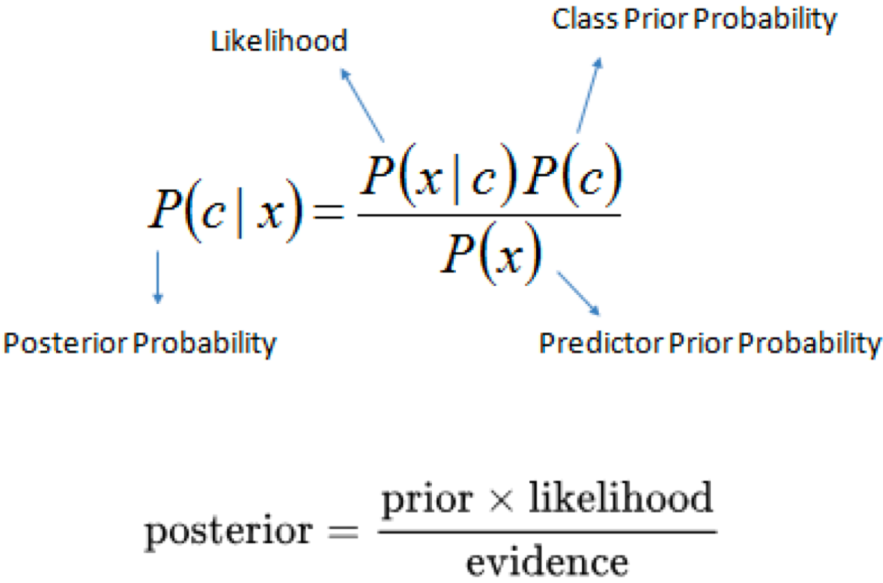
Nucleic Acids Research 45(20) · September 2017

Challenges for variant detection:

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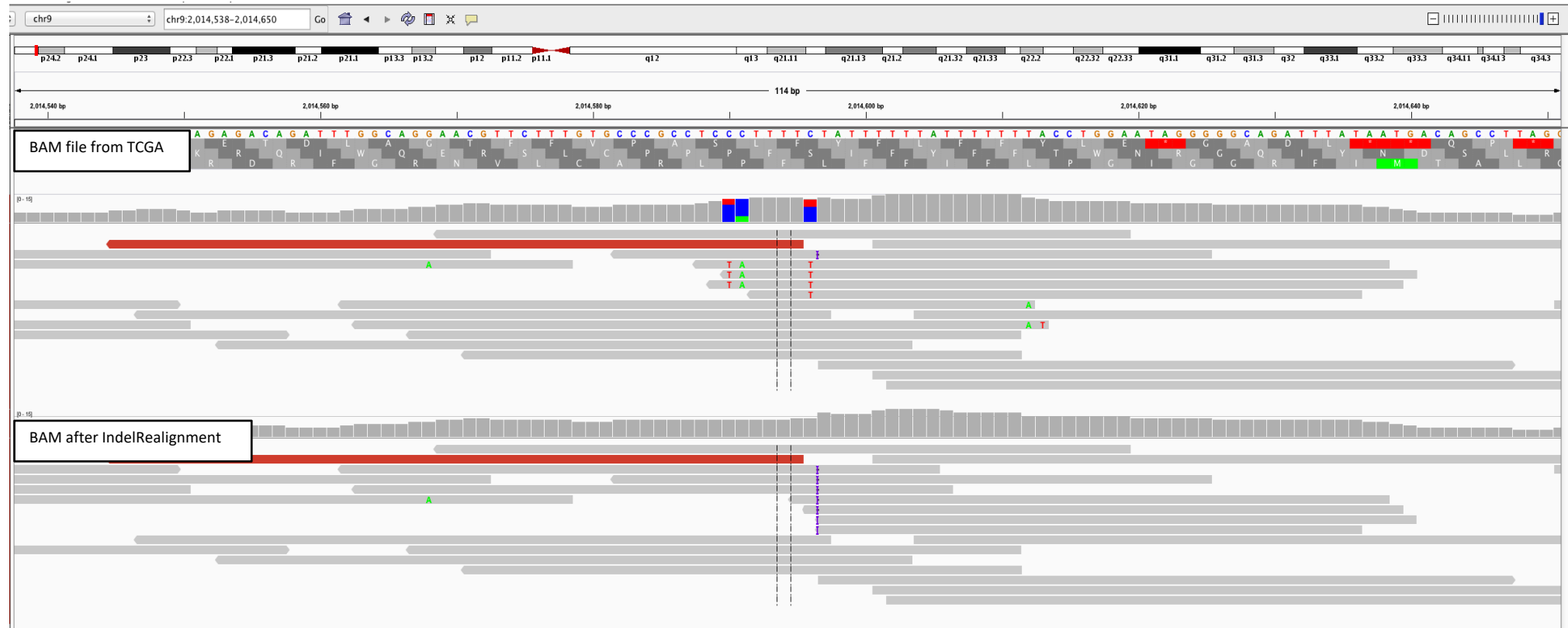
These challenges, or “priors” lend themselves to the application of Bayesian statistics



Robinson, *et al.* Cancer Research 2017

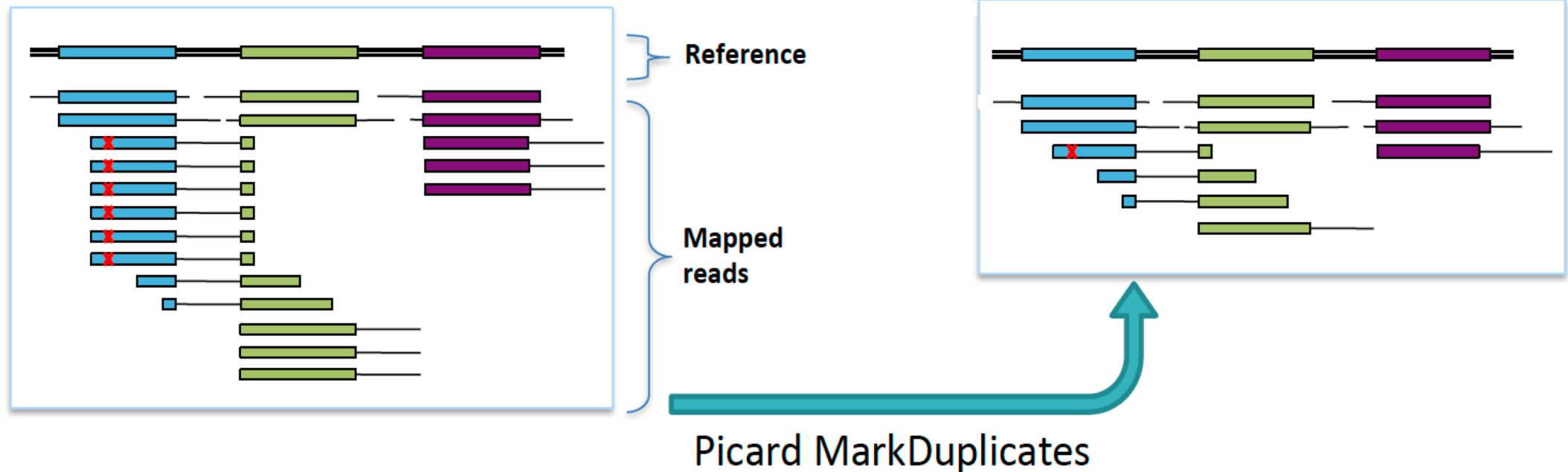
Some strategies to remove/limit errors

- Indel realignment (Local/genome-wide)



Some strategies to remove/limit errors

- Indel realignment (Local/genome-wide)
- Duplicate-read marking/filtering

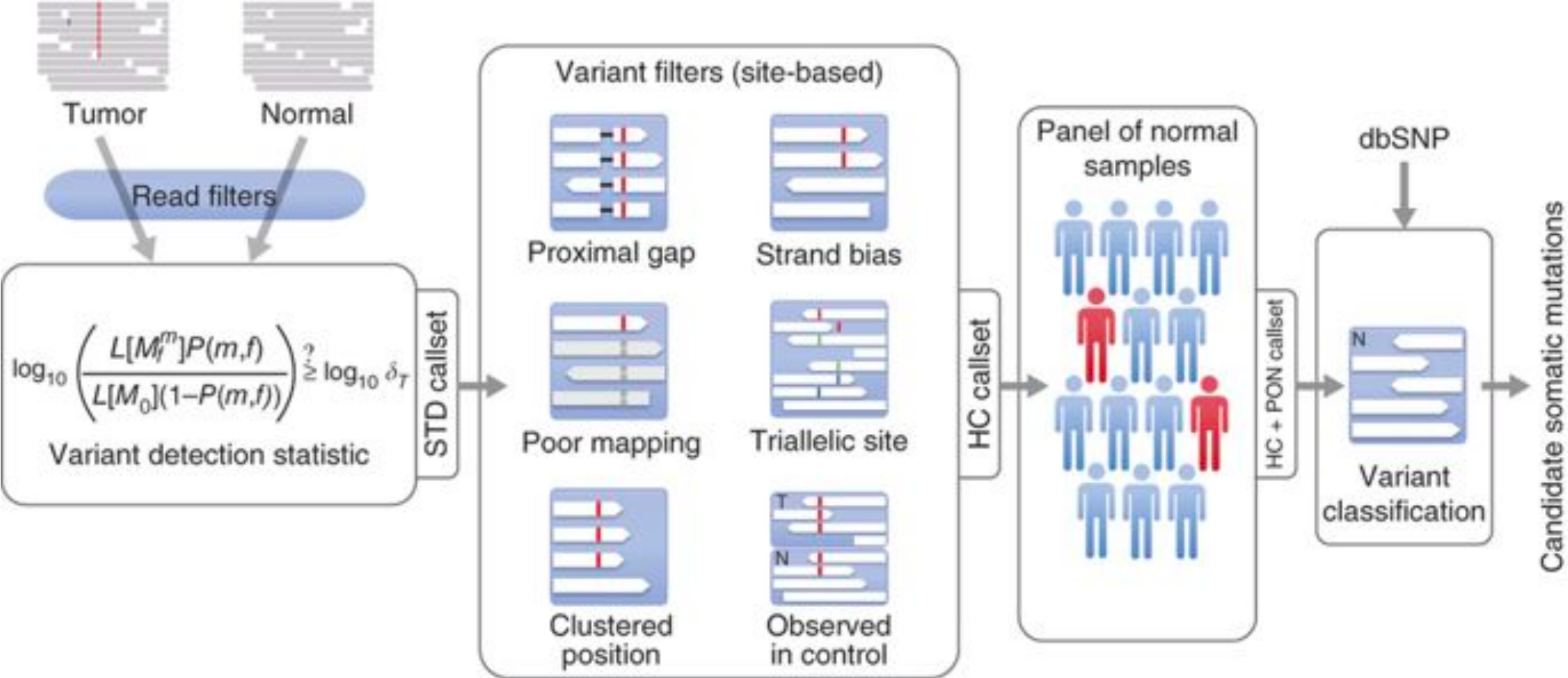


✘ = sequencing error propagated in duplicates

Some strategies to remove/limit errors

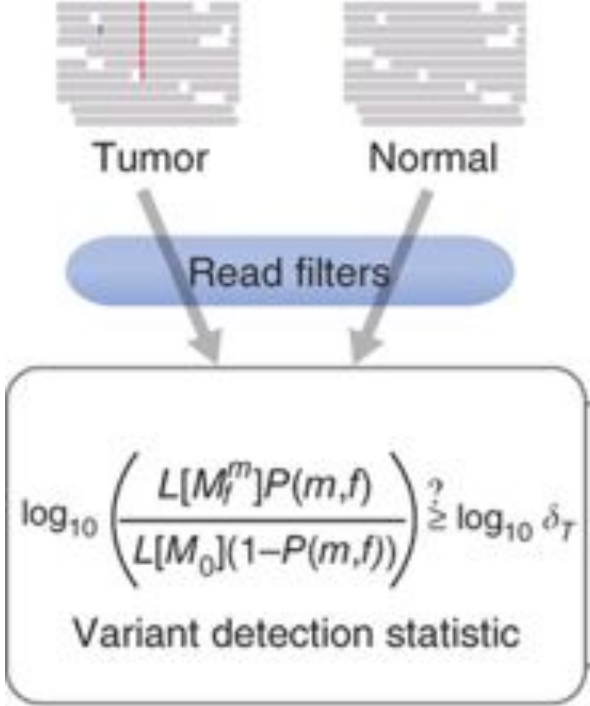
- Indel realignment (Local/genome-wide)
- Duplicate-read marking/filtering
- Base quality recalibration
- Joint calling (germline)
- Panel of Normals (somatic)

SNV detection algorithm example: MuTect



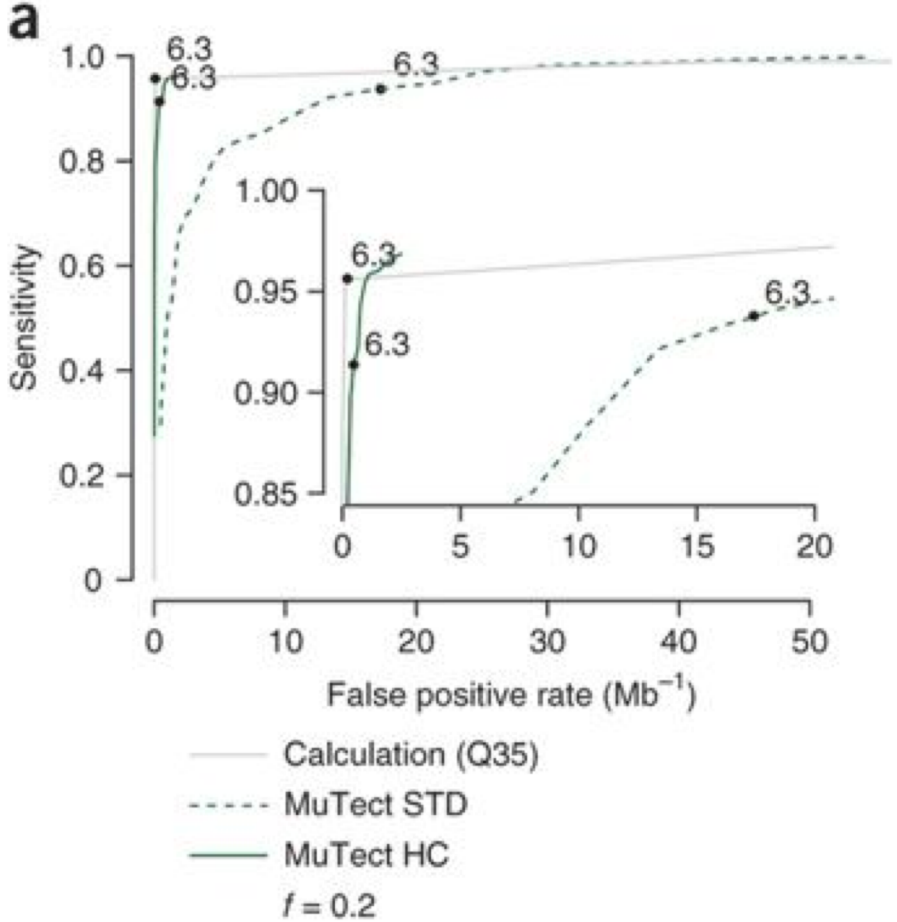
Cibulskis *et al.* Nature Biotechnology 2013

SNV detection algorithm example: MuTect



M_0 : reference model, alt allele is due to error

M_1 : variant model, alt allele is a true variant



Cibulskis *et al.* Nature Biotechnology 2013

Practical application example: MuTect v.1.1.4

Available Parameters:

```
usage: java -jar muTect-1.1.4.jar -T <analysis_type> [-args <arg_file>] [-I <input_file>] [-rbs <read_buffer_size>] [-et
<phone_home>] [-K <gatk_key>] [-tag <tag>] [-rf <read_filter>] [-L <intervals>] [-XL <excludeIntervals>] [-isr
<interval_set_rules>] [-im <interval_merging>] [-ip <interval_padding>] [-R <reference_sequence>] [-ndrs]
[--disableRandomization] [-maxRuntime <maxRuntime>] [-maxRuntimeUnits <maxRuntimeUnits>] [-dt <downsampling_type>]
[-dfrac <downsample_to_fraction>] [-dcov <downsample_to_coverage>] [-baq <baq>] [-baqGOP <baqGapOpenPenalty>] [-PF
<performanceLog>] [-OQ] [-BQSR <BQSR>] [-DIQ] [-EQQ] [-preserveQ <preserve_qscores_less_than>] [-DBQ
<defaultBaseQualities>] [-S <validation_strictness>] [-rpr] [-kpr] [-U <unsafe>] [-nt <num_threads>] [-nct
<num_cpu_threads_per_data_thread>] [-mte] [-bfh <num_bam_file_handles>] [-rgbl <read_group_black_list>] [-ped
<pedigree>] [-pedString <pedigreeString>] [-pedValidationType <pedigreeValidationType>] [-l <logging_level>] [-log
<log_to_file>] [-h] sample_list);do
```

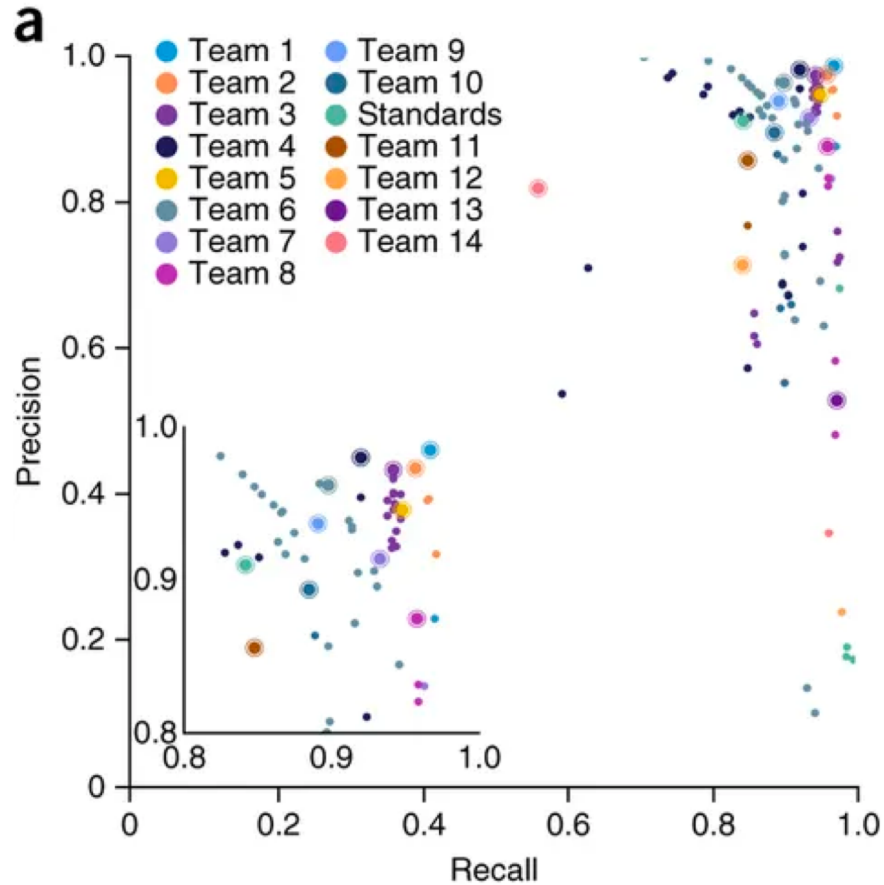
Example Command:

```
module load mutect/1.1.4
module load igenome-human/hg19

java -Djava.io.tmpdir=./tmp/ -Xmx8g -jar $mutect_dir/muTect-1.1.4.jar --analysis_type MuTect \
--enable_extended_output --fraction_contamination 0.02 -dt NONE -L Interval.bed --reference_sequence $REF \
--input_file:normal /path/to/normal/file.bam --input_file:tumor /path/to/tumour/file.bam \
--out /path/to/output.call_stats --vcf /path/to/output.vcf --coverage_file /path/to/output_coverage.wig.txt
```

Combining tumor genome simulation with crowdsourcing to benchmark somatic single-nucleotide-variant detection

Ewing ... Boutros *et al.*, ICGC-TCGA Network Nature Methods volume 12, pages 623–630 (2015)



Winner: MuTect – Broad Institute



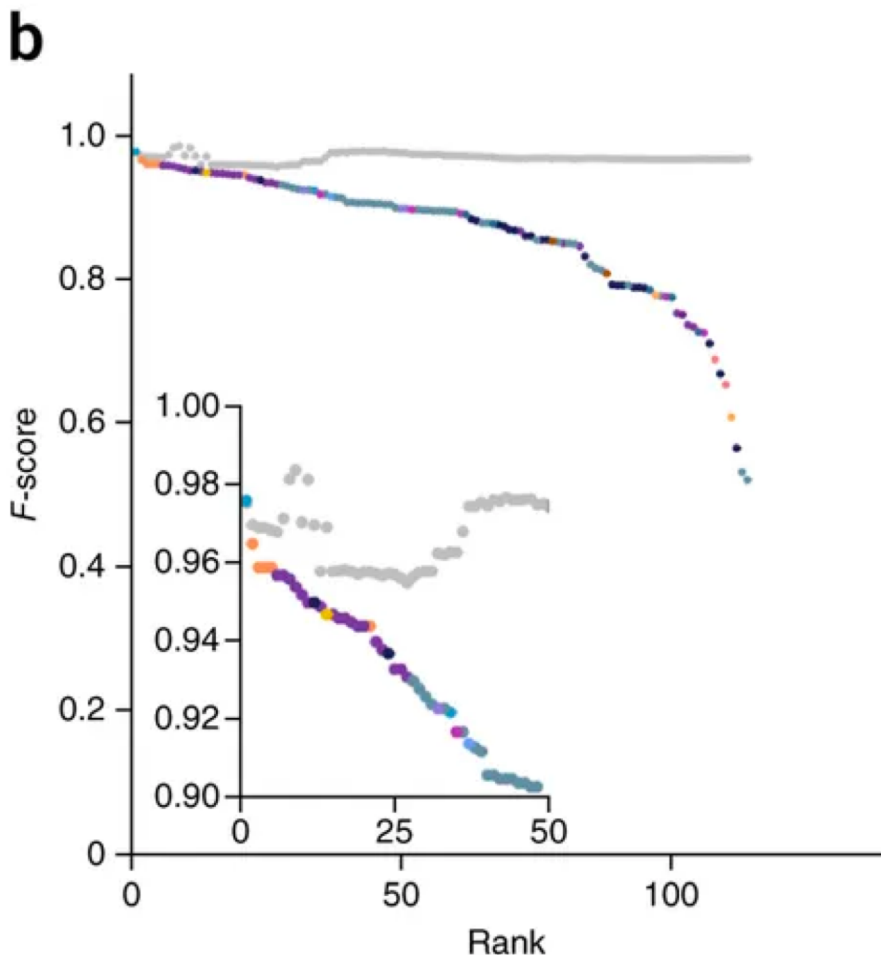
Combining tumor genome simulation with crowdsourcing to benchmark somatic singlenucleotide-variant detection

Ewing ... Boutros *et al.*, ICGC-TCGA Network Nature Methods volume 12, pages 623–630 (2015)

Name	Entity ID	Team	# True Positives	# False Positives	Recall	Precision	F-score
MuTect - L10	syn2343084	Broad SMC	3421	57	0.967204	0.983611	0.975339
MuTect - Stock	syn2343082	Broad SMC	3431	547	0.970031	0.862494	0.913107

Combining tumor genome simulation with crowdsourcing to benchmark somatic singlenucleotide-variant detection

Ewing ... Boutros *et al.*, ICGC-TCGA Network Nature Methods volume 12, pages 623–630 (2015)



“...and in subclonality, an ensemble of pipelines outperforms the best individual pipeline in all cases”

Tumor	Cell line	Number of somatic SNVs	Cellularity (%)	Subclone VAFs
<i>In silico</i> 1	HCC1143 BL	3,537	100	N/A
<i>In silico</i> 2	HCC1954 BL	4,332	80	N/A
<i>In silico</i> 3	HCC1143 BL	7,903	100	50%, 33%, 20%

ICGC-TCGA DREAM Mutation Calling challenge

1. Registration

DREAM8 Challenges Official Rules

2. News

3. Data Description

3.1 Synthetic Tumours

4. Data Access

4.1 Gene Torrent download for local compute

4.2 Using Google Cloud

4.3 GeneTorrent Client Installation in Amazon EC2

4.4 S3 based download from Bionimbus

5. Comparing Algorithms Performance

6. Important Dates

7. Submitting Results

8. Leaderboards

Participation Overview

Overall Participation

Participants	Count
Total Challenge Participants	539
ICGC DACO Approved Participants	110

Synthetic Tumours

Tumour	SNV Submissions (Challenge Eligible)	SNV Teams (Challenge Eligible)	SNV Submissions (Post Challenge)	SNV Teams (Post Challenge)	SV Submissions (Challenge Eligible)	SV Teams (Challenge Eligible)	SV Submissions (Post Challenge)	SV Teams (Post Challenge)	INDEL Submissions (Challenge Eligible)	INDEL Teams (Challenge Eligible)

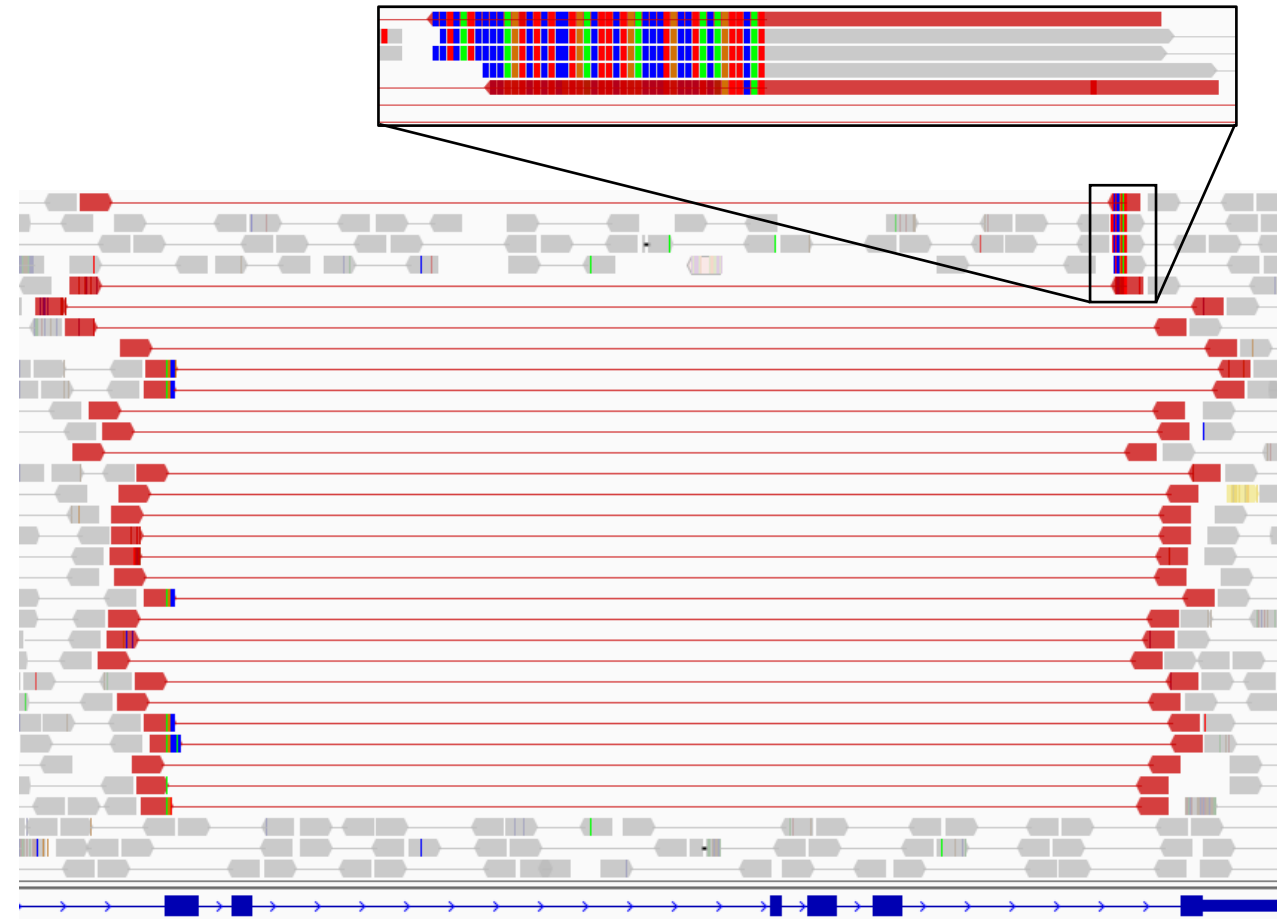
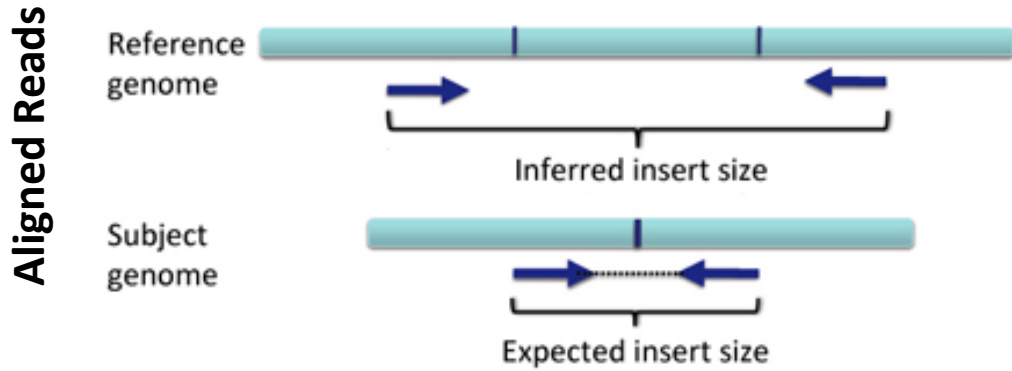
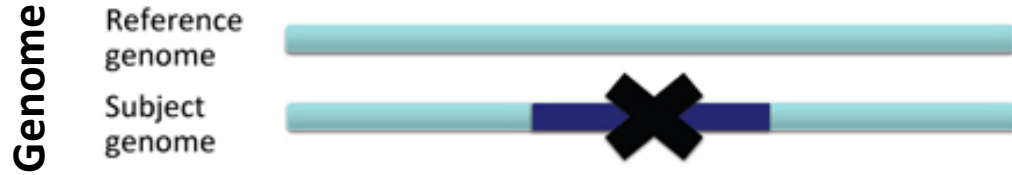
“We will keep leaderboards open indefinitely to allow rapid comparison of methods”

<https://www.synapse.org/#!/Synapse:syn312572/wiki/61509>

Small (SNV/Indel) Variant Annotation

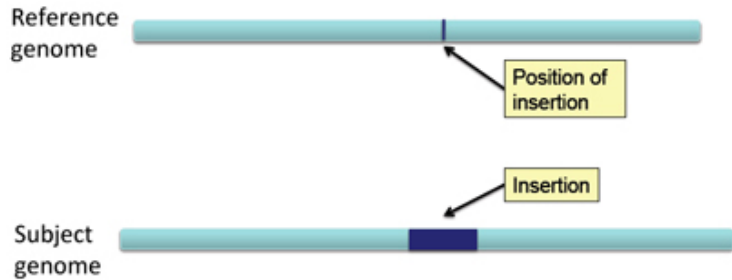
- **Purpose**
 - To aid in the interpretation of variants
- **Annotations**
 - Classification (Missense, frameshift etc.)
 - Predicted amino acid change
 - Predicted impact (ex. SIFT, Polyphen)
 - Occurrence in public databases (dbSNP, COSMIC, ExAC, Gnomad)
 - +++
- **Some available tools:**
 - Ensembl Variant Effect Predictor ([VEP](#))
 - [Annovar](#)
 - [Oncotator](#)

Structural variants affecting insert size: Deletions

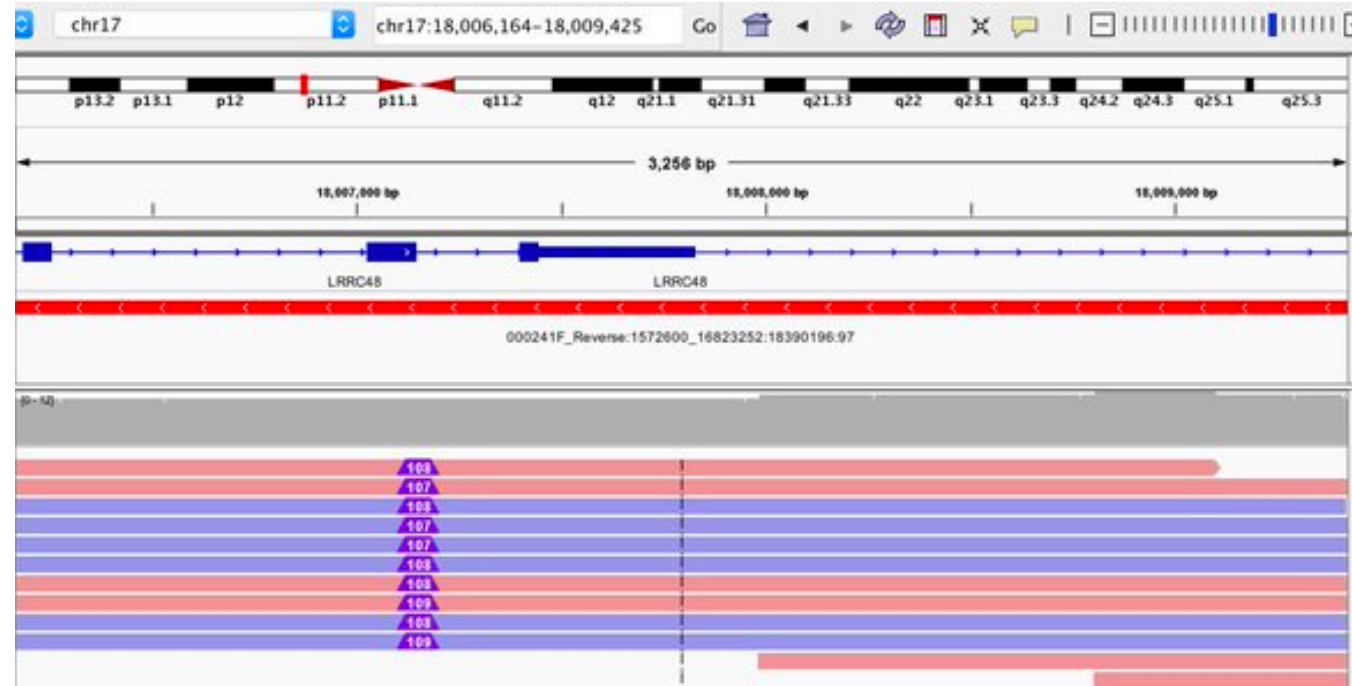
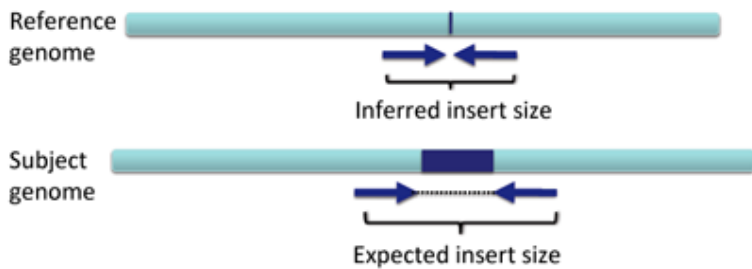


Structural variants affecting insert size: Insertions

Genome



Aligned Reads



Pacbio long-reads <https://twitter.com/infoecho/>

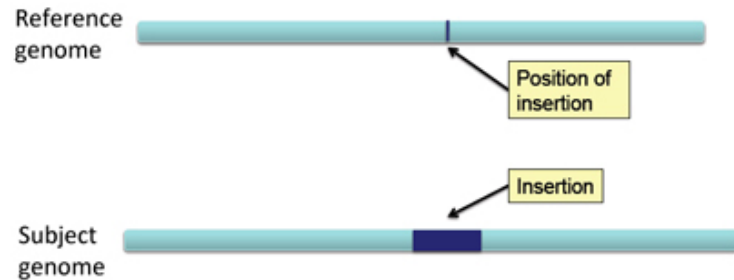
Note: The maximum size of an insertion detectable by variant bases is limited by read length

The maximum detectable size is approximately equal to:

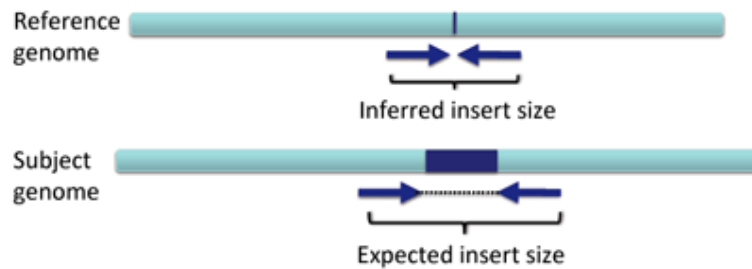
$\text{read length}/2$ <- pushing it

Structural Variants: Insertions

Genome



Aligned Reads



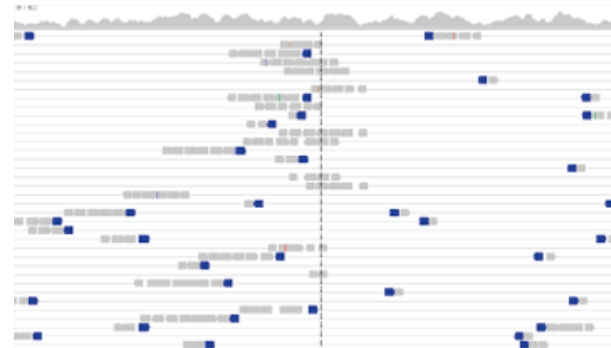
Note: The maximum size of an insertion detectable by insert size anomaly is limited by the size of the fragments.

They must be long enough to span the insertion and include sequences on both ends that are mapped to the reference.

The maximum detectable size is approximately equal to:

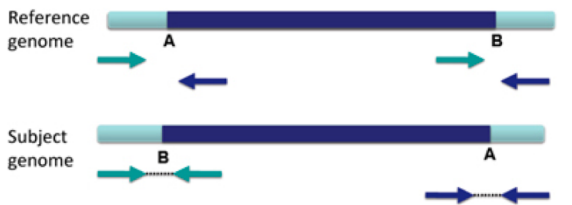
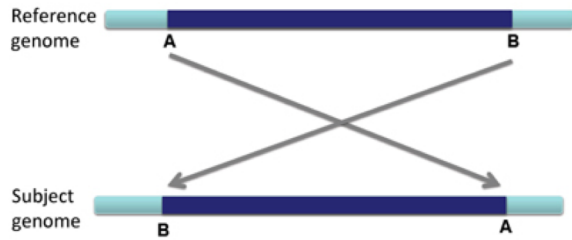
$$\text{Fragment length} - (2 \times \text{read length})$$

Detection of this event is therefore more likely with larger fragment libraries, such as Illumina mate-pair (not paired-end) and SOLID.

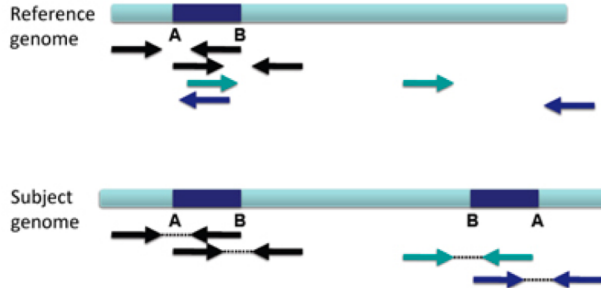
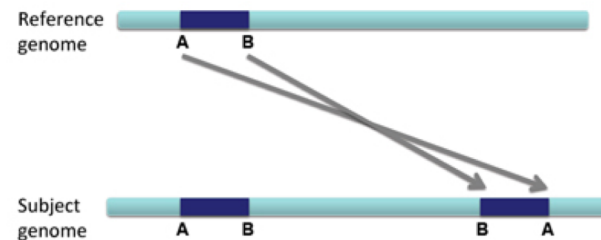


Structural Variants affecting mapped-read direction:

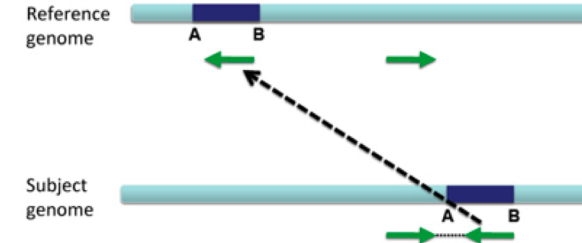
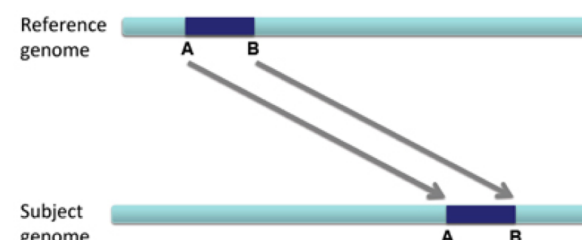
Inversion:



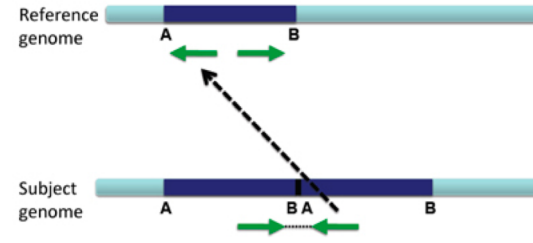
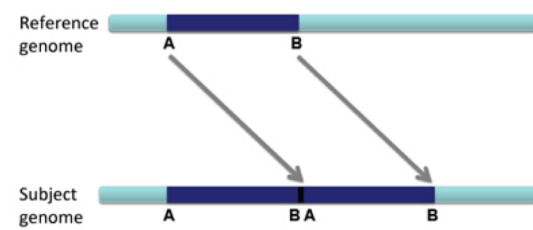
Inverted Duplication:



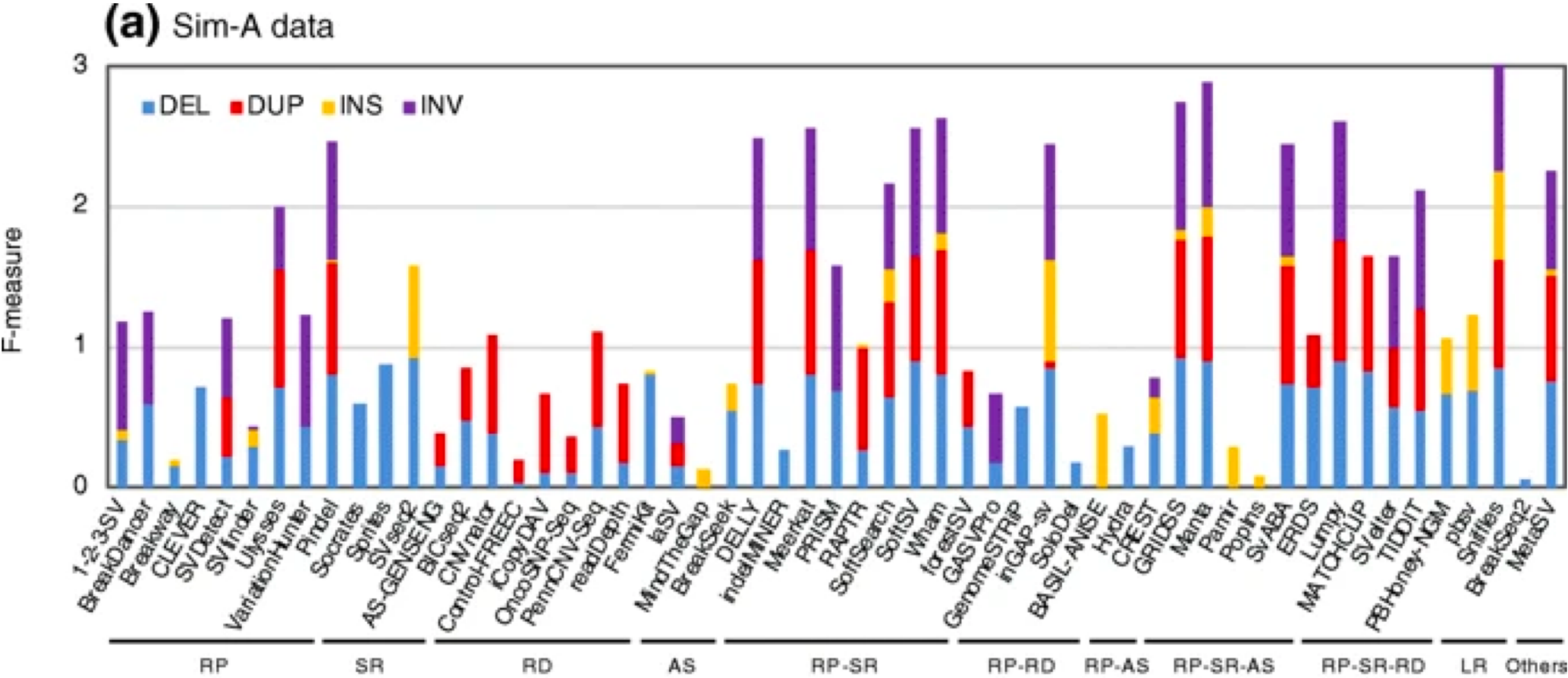
Translocation (intra chr):



Tandem Duplication:



Structural variant caller Performance varies depending on SV type



Kosugi *et al.* Genome Biology 2019

Structural Variant Annotation

- **Purpose**
 - To aid in the interpretation of variants
- **Annotations**
 - Classification (Deletion, translocation, inverted translocation etc.)
 - Genes/regions affected
 - Predicted amino acid change
 - +++
- **Some available tools:**
 - [MAVIS](#)
 - [SVAnnotator](#)

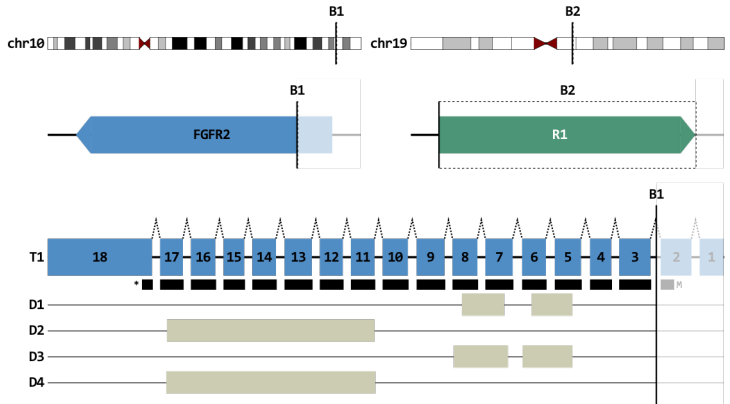
Structural Variant Annotator Example: MAVIS

Merging,
Annotation,
Validation, and
Illustration of
Structural Variants

> Clusters breakpoints and SVs the same or multiple tools

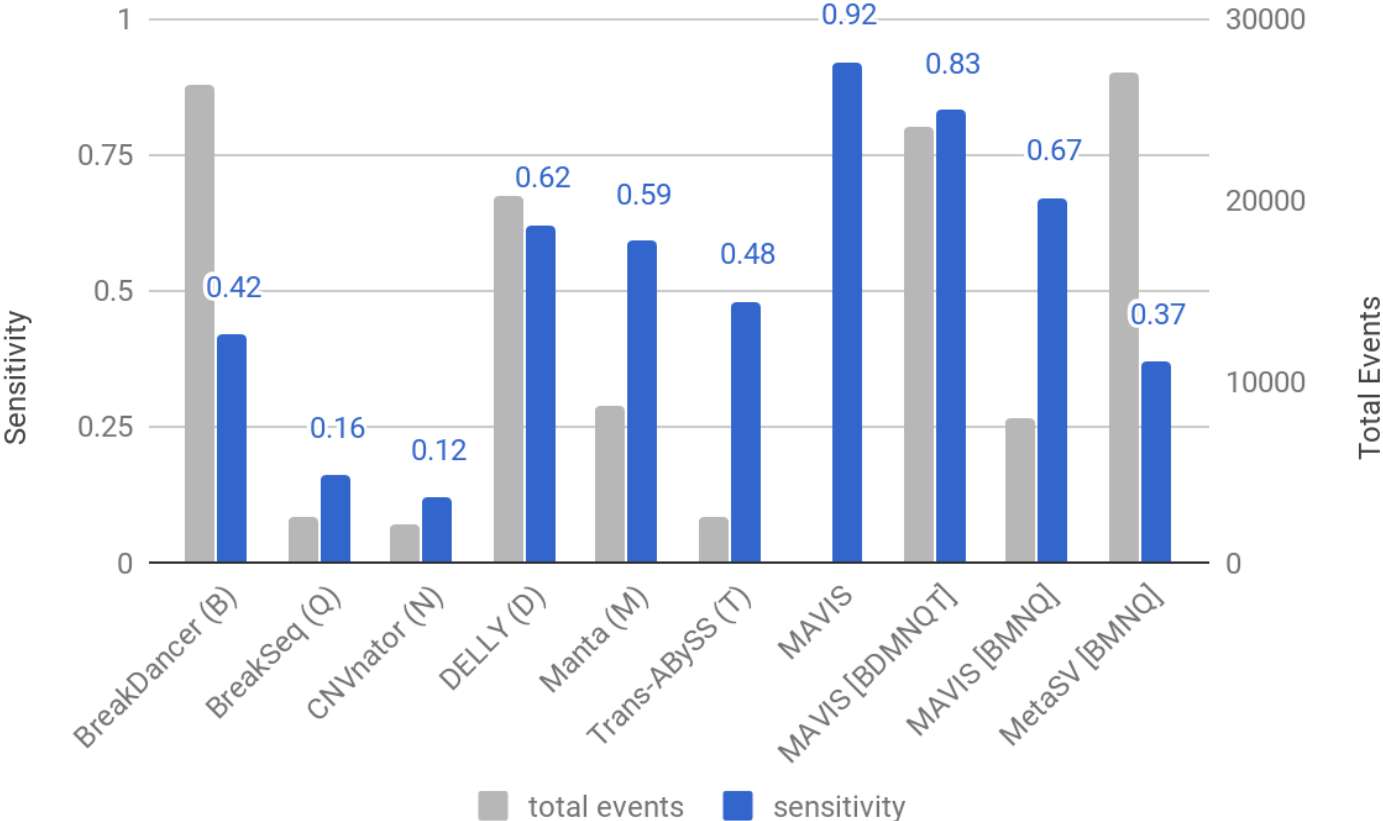
> Annotates SV with genes, somatic/germline status, AA effect etc.

> Performs in-bam validation to further polish/filter results



Structural Variant Annotator Example: MAVIS

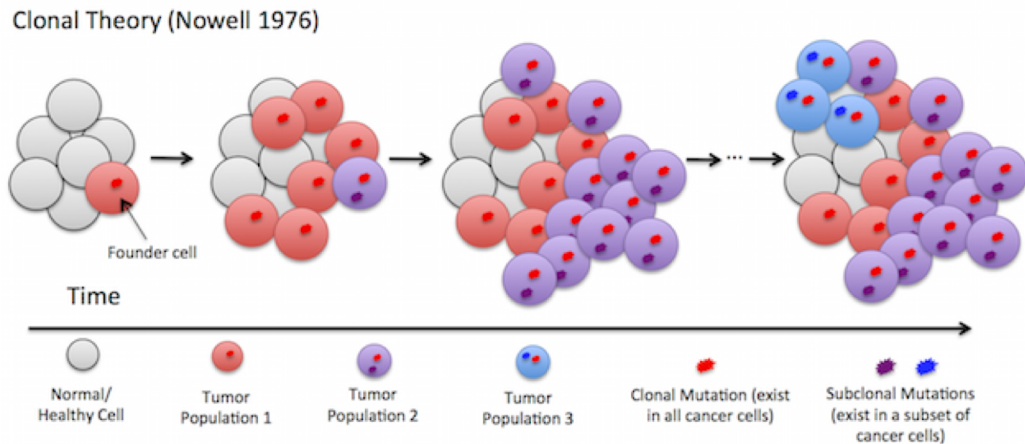
*Adding additional callers appears to improve structural variant detection as well



Reisle *et al.* Bioinformatics 2019

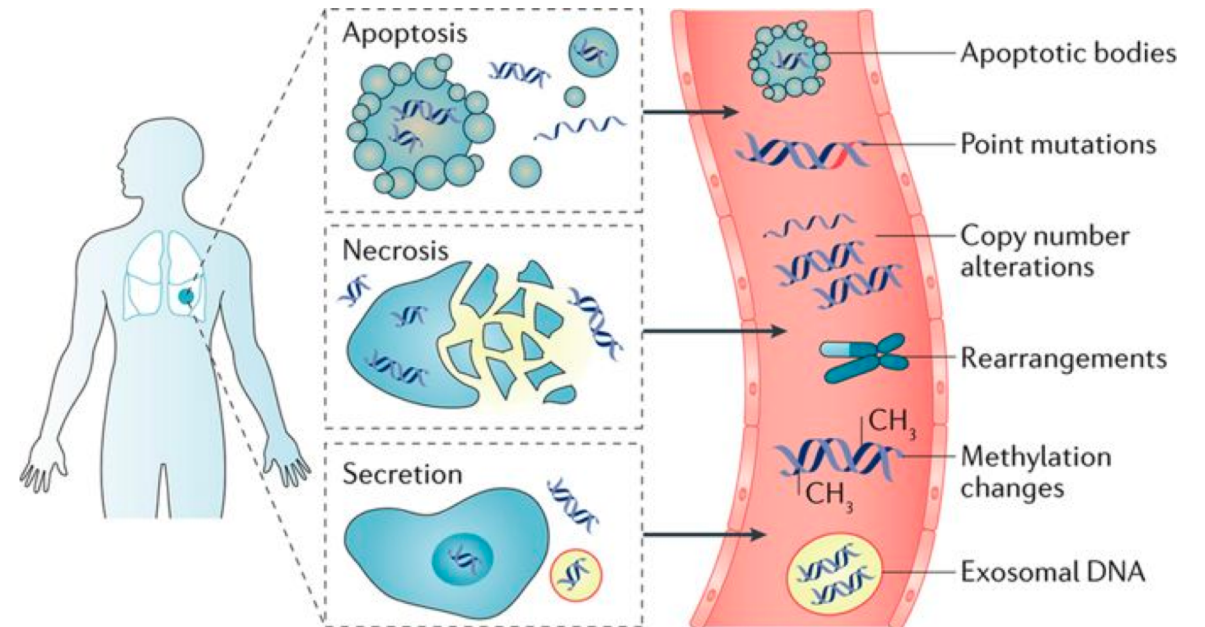
Challenges facing modern day small variant calling algorithms: Low variant allele frequencies (VAF)

Subclonal Mutations



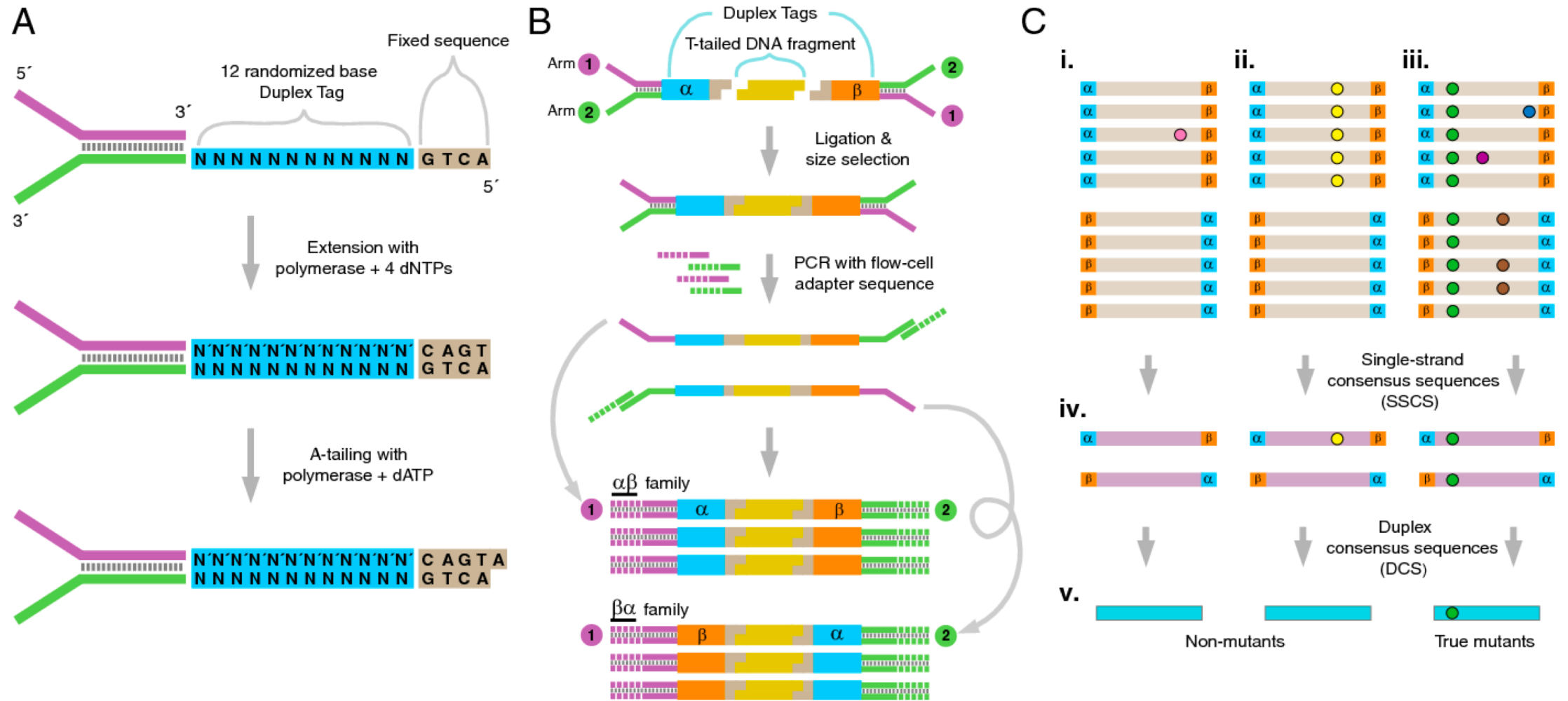
<http://www.cs.carleton.edu/faculty/loesper/research.html>

Cell-Free DNA



Nature Reviews Cancer volume17, pages223–238 (2017)

Detecting rare variants using unique molecular indices and duplex sequences



Challenges facing modern day small variant calling algorithms: Difficult to align regions

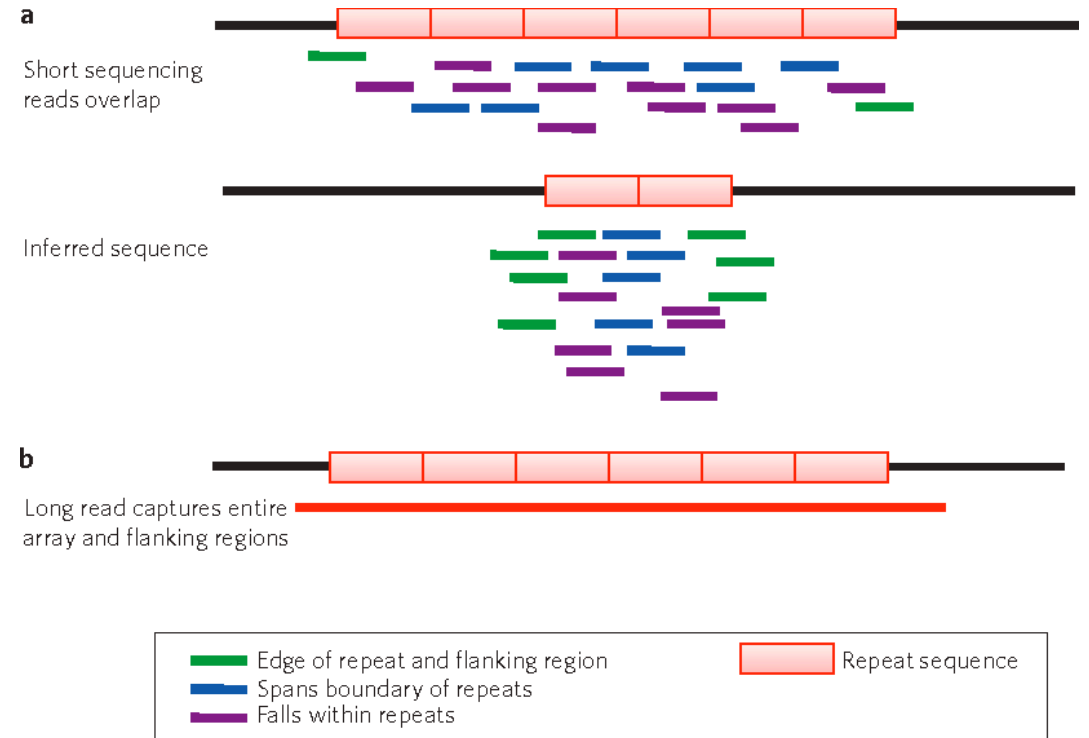
Linked-Reads from 10X genomics

Barcodes recruit short-reads into paralogous gene loci



<https://www.10xgenomics.com/solutions/genome/>

Long Reads from Pacbio



Nature plants 2015 DOI:10.1038/nplants.2015.169

Questions?

Break

Hands on Exercise

Practical Exercise: Calling Somatic SNVs and Indels

- **Getting started:**

- I have e-mailed the group a link with the necessary files: also Here
- Those **with** access to Mordor: scp (copy) these data to a directory of your choosing
- Those **without** access to Mordor: pair up with a) a user with access or b) another user without access. I will provide a laptop and log you in

- **Your task:**

- 1) Alter the two bash scripts in the `~/scripts/` directory to correspond with the location of the files you are processing
- 2) Submit these jobs to the cluster (ex. `qsub runMutect2_mordor.sh`)
- 3) Compare the resulting final output from each tool
 - hint:** final Mutect2 vcf has 'filtered' in the file name
 - final Strelka vcfs have 'passed.somatic' in the file name
- 4) Find the two variants that were found by one tool and not by the other
- 5) Load the two provided BAM files into IGV and take a snapshot of these two loci

Bonus: Why do you think one tool did not report each of these?

Practical Exercise: Calling Somatic SNVs and Indels

- **Useful Linux Commands:**

- Change directory: **cd** Ex: **cd ./a_subdirectory**
- Go back one directory level: **cd ../**
- List files in current directory: **ls** or **ll**
- View the contents of a file: **cat**
- Edit a file on the cluster: nano or vim Ex: **nano runMutect2_mordor.sh**

Thanks!