

Calling DNA variants – SNVs, CNVs, and SVs

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Variant Effect Predictor

Training Course

Prague, 6th November 2017

cnag

centre nacional d'anàlisi genòmica
centro nacional de análisis genómico



RD  **Connect**





Calling DNA variants – SNVs, CNVs, SVs

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1. What is a variant?
2. Paired End read mapping
3. Calling Single Nucleotide Variants (SNVs) and InDels
4. Calling Copy Number Variants (CNVs)
 - From Whole Genome Sequencing data
 - From Whole Exome Sequencing data
5. Calling Structural Variants (SVs)

- Internationally recognised state-of-the-art sequencing centre situated in the Parc Científic de Barcelona. **Publically funded, not-for-profit.**
- 60 staff, **over 50% informatics/computer engineers**

Mission

Carry out projects in genome analysis that will lead to **significant improvements in people's health and quality of life**, in collaboration with the Spanish, European and International Research Community.

Research interests

- Disease Gene Identification and **Personalised Medicine**
- Cancer Genomics
- Single Cell RNAseq
- Agrogenomics and Model Organisms (e.g. genome assembly and gene prediction of various primate *spp.*, Iberian Lynx, Olive ...)



CNAG Genomehenge (version 2017)



Sequencing capacity

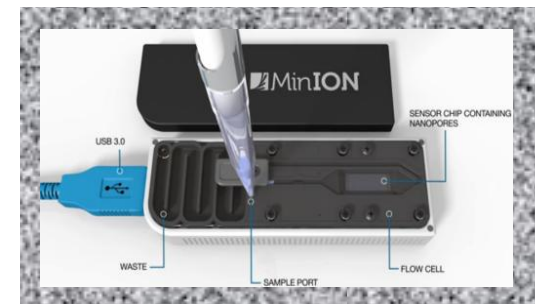
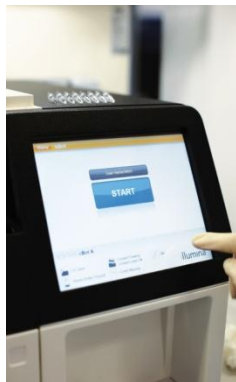
- >1000 Gbases/day = 10 human genomes per day at 30x coverage

Sequencing

- 3 Illumina HiSeq2000
- 3 Illumina HiSeq2500
- 1 Illumina HiSeq4000
- 1 Illumina MiSeq
- 4 Illumina cBots
- 3 Oxford Nanopore Minlons

Computing

- 3552 cores
- 3.7 PB disk + 3 PB tape archive
- 35.5 TB RAM
- Barcelona SuperComputing Center - 10 x 10 Gb/s



CNAG QA Certification



✓ December 2013

“Illumina **CPro** recognizes that CNAG provides customers with industry-leading data quality and service in genetic analysis.”



✓ May 2014

“CNAG has successfully completed **Agilent Certified Services Training** for Target Enrichment System for NGS.”



✓ December 2014

“**ISO 9001 certified** for management and performance of high throughput sequencing and genomic analysis projects and services.”

✓ April 2016

“**ISO 17025 accreditation** for DNA & RNA Analysis using high throughput sequencing (NGS)”



✓ May 2017

Roche- Nimblegen SeqCap EZ Certified Service Providers
CNAG is the first and only Nimblegen certified provider in Europe



Active participant in many international biomedical initiatives

Member of the **Global Alliance for Genomics and Health (GA4GH)**



Participation, through the National Bioinformatics Institute (INB), in **ELIXIR**, the European bioinformatics infrastructure.



Participation in the **International Human Epigenome Consortium (IHEC)**



Participation in the **International Cancer Genome Consortium (ICGC)**



Participation in the **International Rare Diseases Research Consortium (IRDIRC)**





Calling DNA variants – SNVs, CNVs, SVs

7

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What is a **variant**?

8

A **variant** is any position/region in our sample **which differs from the haploid reference genome** to which we are comparing.

There are 4 basic classes:

- **Single Nucleotide Variants (SNVs)**
 - e.g. A → G – note **diploid** individual may be “AA”, “AG”, or “GG”
- **Short (<50nt) insertions and deletions (InDels)**
 - e.g. TA → TATA (insertion of “TA”)
 - e.g. CT → C (deletion of the “T” at the second position)
- **Copy Number Variants (CNVs)** – generally tandem duplications of typically longer regions (~1-100kb) that are often polymorphic within the population e.g. AMY1
- **Structural Variants (SVs)** – often larger still, and often complex in nature



What is a **variant**?

9

A **variant** is any position/region in our sample **which differs from the haploid reference genome** to which we are comparing.

There are 4 basic classes:

- **Single Nucleotide Variants (SNVs)**
~ 3,750,000-4,500,000 (Yuen *et al*, Nat. Neuro. 2017)
- **Short (<50nt) insertions and deletions (InDels)**
~ 700,000-1,000,000 (Yuen *et al*, Nat. Neuro. 2017)
- **Copy Number Variants (CNVs)** – generally tandem duplications
~ 11.3Mbp per individual (1kGP);
5-9% of genome 50bp-3Mbp (Zarrei *et al*, NRG, 2015)
- **Structural Variants (SVs)** – often larger still, and often complex in nature
~ 10Mbp per individual (1kGP) – 59Mbp (English *et al*, 2015)



Summary of Variant Types

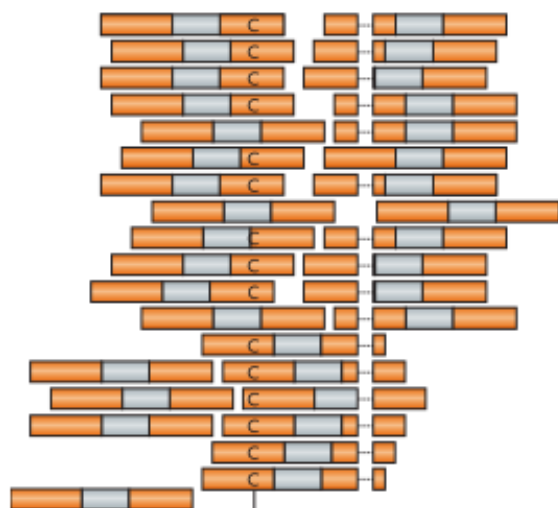
10

Reference sequence
Chr 1



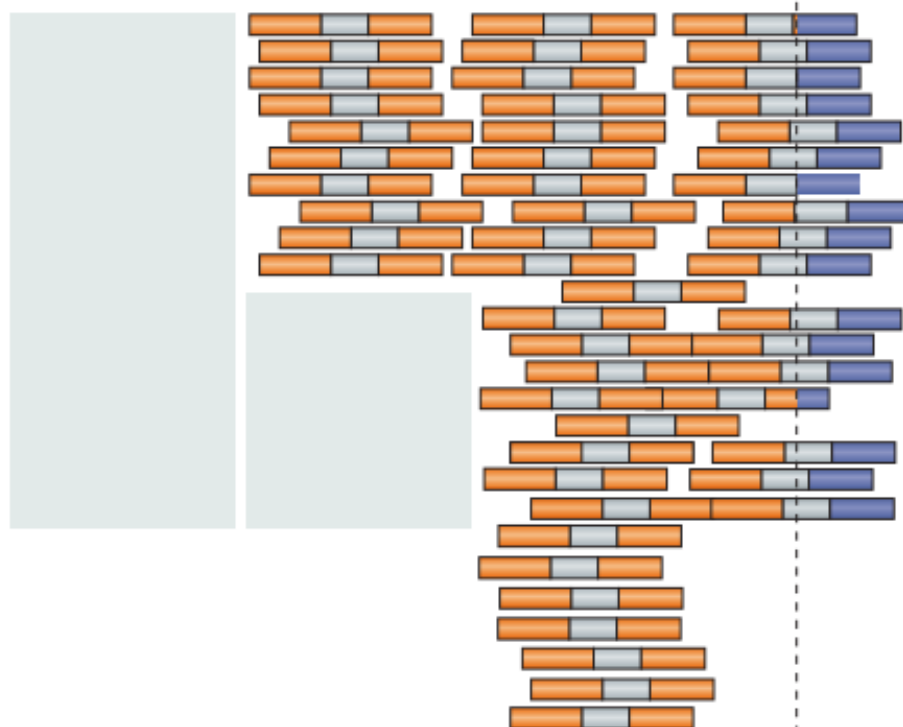
Chr 5

Non-human
sequence



Point mutation

Indel



Homozygous
deletion

Hemizygous
deletion

Gain

Translocation
breakpoint

Pathogen

Copy number alterations

Meyerson et al. . 2010. Advances in understanding cancer genomes through second-generation sequencing. Nature Reviews Genetics 11, no. 10 (October): 685-696



Calling DNA variants – SNVs, CNVs, SVs

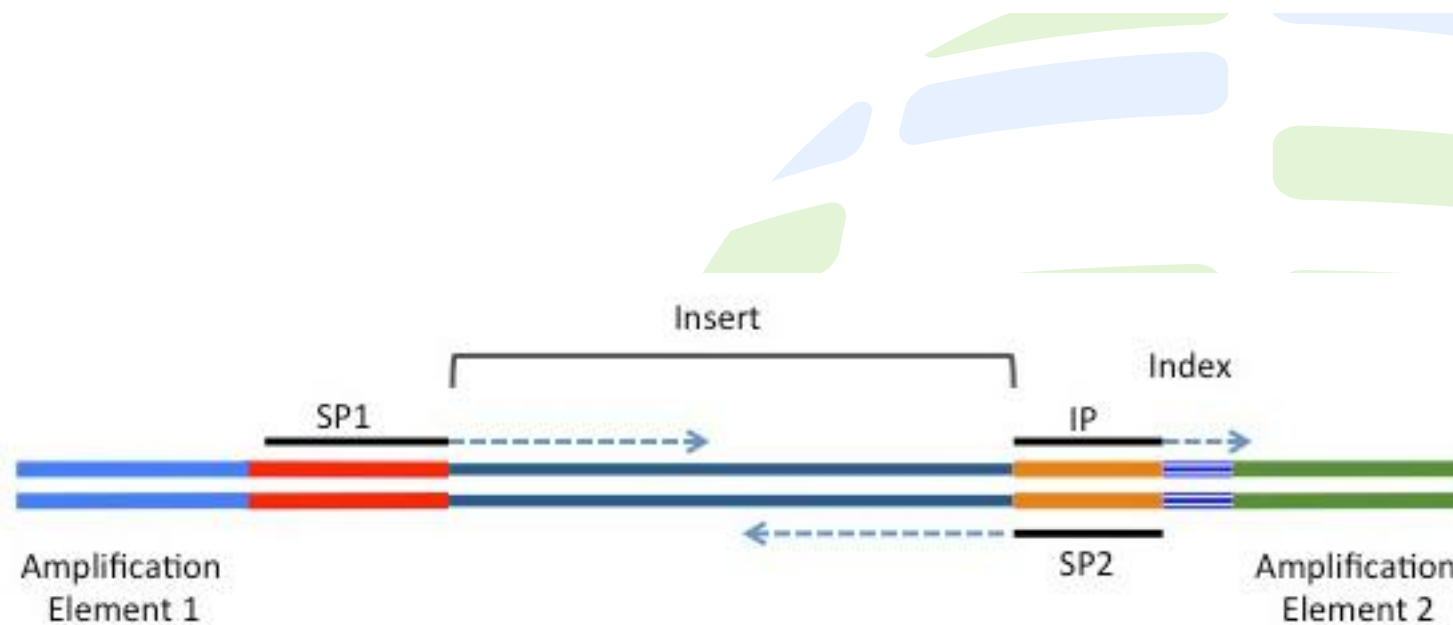
11

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Paired-end Read Mapping

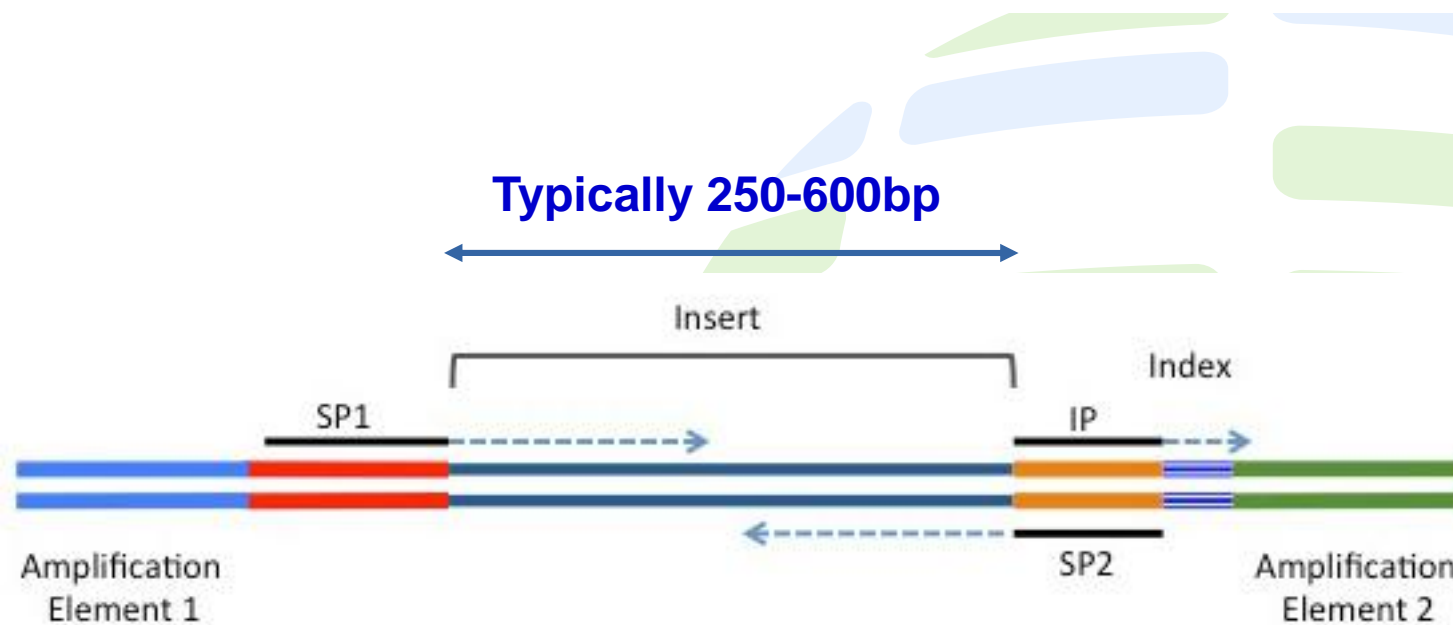
12





Paired-end Read Mapping

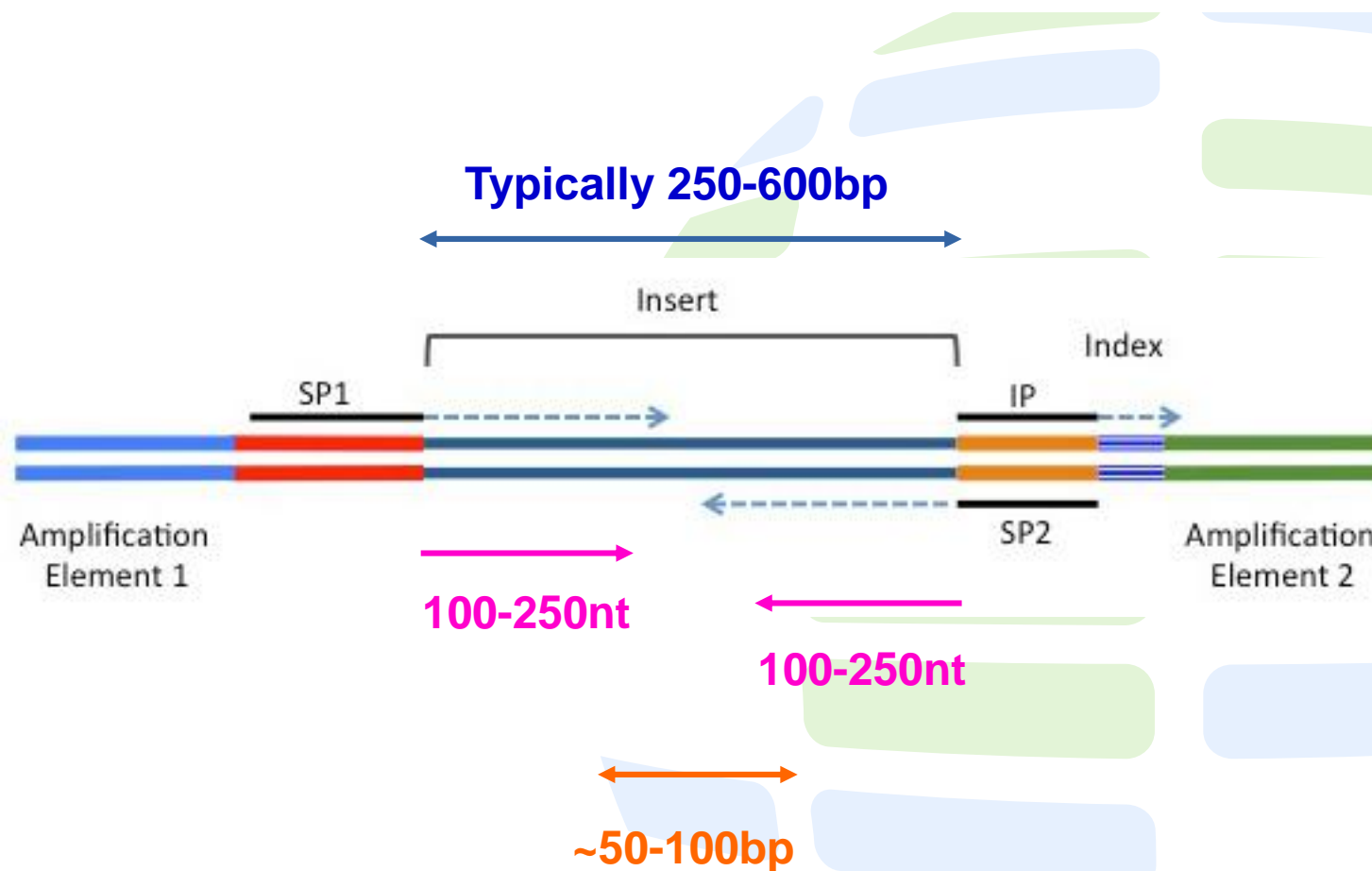
13





Paired-end Read Mapping

14





Paired-end Read Mapping

15

Reference Genome Sequence

100nt read

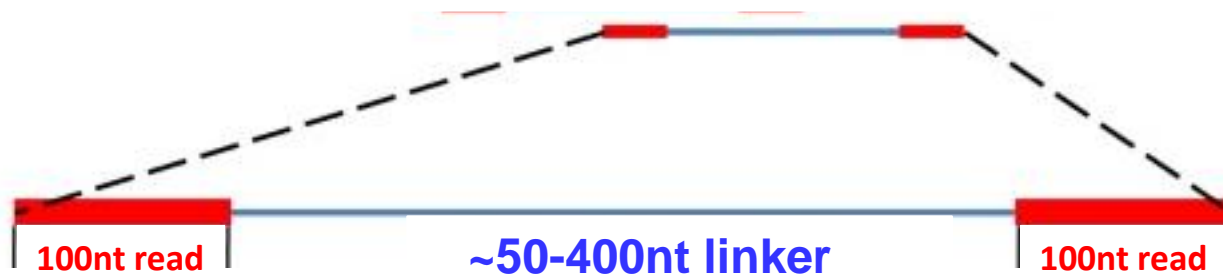
~50-400nt linker

100nt read



Paired-end Read Mapping

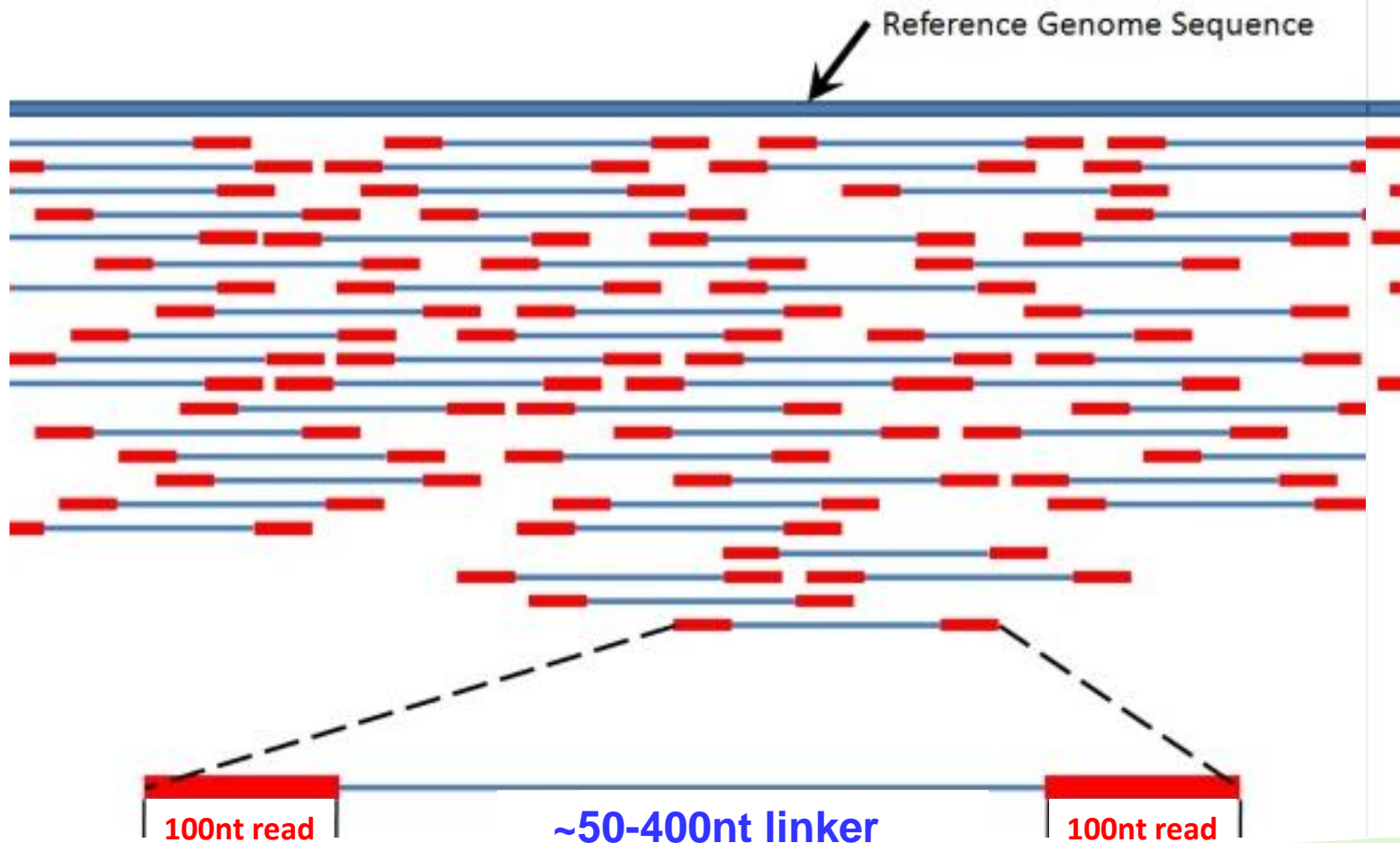
16





Paired-end Read Mapping

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Mapped reads viewed in IGV

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Coverage

Reads

Exons





What is a **variant**?

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A **variant** is any position/region in our sample **which differs from the haploid reference genome** to which we are comparing.

There are 4 basic classes:

- **Single Nucleotide Variants (SNVs)**
~ 4,000,000
- **Short (<50nt) insertions and deletions (InDels)**
~ 400,000
- **Copy Number Variants (CNVs)** – generally tandem duplications
~ 5-10% of genome
- **Structural Variants (SVs)** – often larger still, and often complex in nature
~ 13% of genome



Calling DNA variants – SNVs, CNVs, SVs

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Calling SNVs and InDels

21



Calling SNVs and InDels

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Reference

TGGACCATCTGGTTGAGCATGTGGGGGTCAACTCCCACATTCCCAGGGAGCCCCCGG



Calling SNVs and InDels – dream future?

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Reference

TGGACCATCTGGTTGAGCATGTGGGGGTCAACTCCCACATTCCCAGGGAGCCCCCGG

TGG **A**CCATCTGGTTGAGCAT **T**GTGGGGGTCAACT **T**CCACATTCCCAGGGAG **C**CCCCCGG
TGG **A**CCATCTGGTTGAGCA **C**GTGGGGGTCAACT **T**CCACATTCCCAGGGAG **G**CCCCCGG

ref/ref
0/0

homozygous reference

ref/alt
0/1

heterozygous

alt/alt
1/1

homozygous alternative

ref/alt
0/1

heterozygous

Sequence each
chromosome from
start to end without
errors



Calling SNVs and InDels – **back to reality**

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Reference

TGGACCATCTGGTTGAGCATGTGGGGGTCAACTCCCACATTCCCAGGGAGCCCCCGG
TGG**A**CCATCTGGTTGAGCA**T**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**C**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**G**CCCCCGG

ref/ref
0/0
homozygous reference

ref/alt
0/1
heterozygous

alt/alt
1/1
homozygous alternative

ref/alt
0/1
heterozygous

Aligned reads
derived from
the sample

TGG**A**CCATCTGGTTGAGCA**T**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**C**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**C**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**T**GTGGGGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**T**GTGG**G**GGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**C**GTGG**G**GGGTCAACT**T**CCACATTCCCAGGGAG**G**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**T**GTGG**G**GGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
TGG**A**CCATCTGGTTGAGCA**C**GTGG**G**GGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG
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TGG**A**CCATCTGGTTGAGCA**C**GTGG**G**GGGTCAACT**T**CCACATTCCCAGGGAG**C**CCCCCGG

ref/ref
0/0
homozygous reference
0% alternative allele

ref/alt
heterozygous
50% alternative allele

alt/alt
1/1
homozygous alternative
100% alternative allele

?
?
?
20% alternative allele



Calling SNVs & InDels

25





Tools for Calling SNVs & InDels

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BIOINFORMATICS ORIGINAL PAPER

Vol. 27 no. 21 2011, pages 2987–2993
doi:10.1093/bioinformatics/btr509

Sequence analysis

Advance Access publication September 8, 2011

A statistical framework for SNP calling, mutation discovery, association mapping and population genetical parameter estimation from sequencing data

Heng Li

Medical Population Genetics Program, Broad Institute, 7 Cambridge Center, Cambridge, MA 02142, USA

Associate Editor: Jeffrey Barrett

SAMtools, 2011



Tools for Calling SNVs & InDels

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Heng Li

Medical
Association

A framework for variation discovery and genotyping using next-generation DNA sequencing data

SAMtools, 2011

GATK, 2011

Mark A DePristo¹, Eric Banks¹, Ryan Poplin¹, Kiran V Garimella¹, Jared R Maguire¹, Christopher Hartl¹, Anthony A Philippakis^{1–3}, Guillermo del Angel¹, Manuel A Rivas^{1,4}, Matt Hanna¹, Aaron McKenna¹, Tim J Fennell¹, Andrew M Kernytsky¹, Andrey Y Sivachenko¹, Kristian Cibulskis¹, Stacey B Gabriel¹, David Altshuler^{1,3,4} & Mark J Daly^{1,3,4}



Tools for Calling SNVs & InDels

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Tim J Fennell¹, Anne
David Altshuler^{1,3,4}

Haplotype-based variant detection from short-read sequencing

Erik Garrison and Gabor Marth

July 24, 2012

FreeBayes, 2012



Variant Calling Tools (SNVs & InDels)

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Whole-genome haplotype-based variant detection from short-read sequencing
Integrating mapping-, assembly- and
haplotype-based approaches for calling variants in
clinical sequencing applications

Andy Rimmer, Hang Phan, Iain Mathieson, Zamin Iqbal, Stephen R F Twigg, WGS500
Consortium, Andrew O M Wilkie, Gil McVean & Gerton Lunter

Affiliations | Contributions | Corresponding author

Nature Genetics 46, 912–918 (2014) | doi:10.1038/ng.3036

Received 22 November 2013 | Accepted 23 June 2014 | Published online 13 July 2014

SAMtools, 2011

GATK, 2011

FreeBayes, 2012

Platypus, 2014



Calling SNVs & InDels

30

Variant calling tools will start by calling **every** potential variant they observe

- This will include true variants, and false-positives due to:
 - Sample quality/Library preparation issues
 - PCR artefacts
 - Sequencing errors
 - Mapping issues
 - Variant Calling algorithm issues
- Subsequently they apply a number of mechanisms to attempt to help identify the false-positives.



Calling SNVs & InDels

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Variant calling tools will start by calling **every** potential variant they observe

- This will include true variants, and false-positives due to:
 - Sample quality/Library preparation issues
 - PCR artefacts
 - Sequencing errors
 - Mapping issues
 - Variant Calling algorithm issues
- Subsequently they apply a number of mechanisms to attempt to help identify the false-positives
- **Currently, you will always encounter some false positives, and some false negatives**



Calling SNVs & InDels

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- There are **3 key metrics** that can give us a good idea as to whether to trust a variant call
 - **Read Depth (DP)**
 - A general rule is the deeper, the better – ideally >20 supporting reads
 - **Genotype Quality (GQ)**
 - A value produced by variant calling algorithms indicating the probability that the call is wrong. Scaled from 1-99 (30 means 1/1000)
 - **Allele Balance (aka. Alternative/Beta Allele frequency)**
 - For heterozygote positions this should be close to 0.5
 - 0.25-0.75 is generally reliable
 - <0.15 or >0.85 is highly suspicious
 - For homozygote positions this should be very close to 0 or 1



InDel identification

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1 2 3

4



Raw BWA mapped reads

DePristo, M. *et al.* (2011)



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Strand Bias

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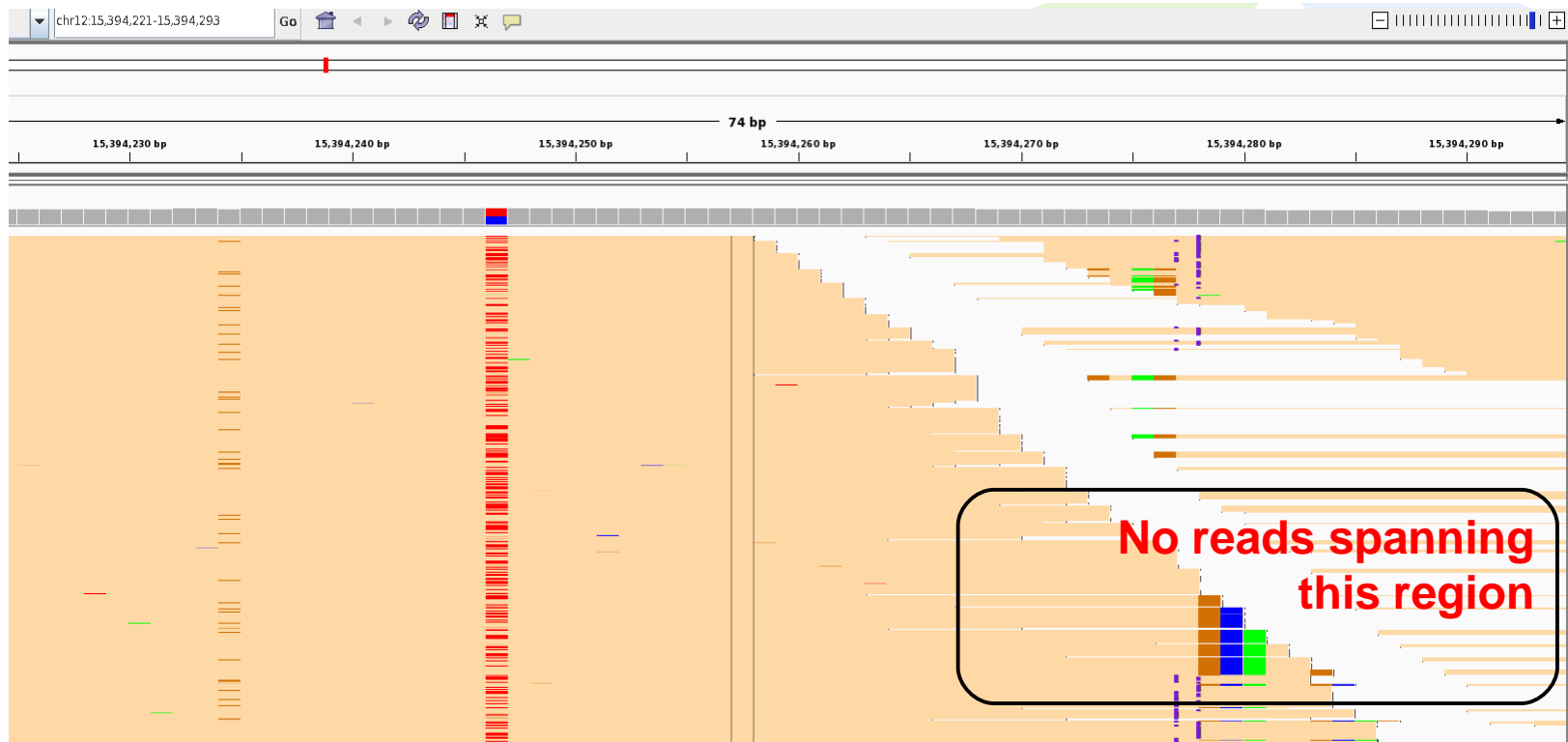
GATK: FS field (Phred-scaled p-value)

SAMtools: PV4 field (p-value)



Tail Distance/Variant Position Bias

36



ReadPosRankSum = 1.635

ReadPosRankSum = - 0.434

ReadPosRankSum = - 9.805

SAMtools equivalent: PV4 field (p-value)



NIST/GIAB – Analyses of NA12878

37

- The NIST is attempting to produce “Gold Standard” call sets for all variants in NA12878, and other samples, through integration of results from a variety of pipelines

Integrating human sequence data sets provides a resource of benchmark SNP and indel genotype calls

Justin M Zook, Brad Chapman, Jason Wang, David Mittelman, Oliver Hofmann, Winston Hide & Marc Salit

[Affiliations](#) | [Contributions](#) | [Corresponding author](#)

Nature Biotechnology **32**, 246–251 (2014) | doi:10.1038/nbt.2835

Received 14 December 2013 | Accepted 27 January 2014 | Published online 16 February 2014



Benchmarking of VC Pipelines

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NA12878 50xWGS FastQs (Illumina Platinum), analysed with several pipelines. Concordance with **Gold Standard VC set** from GIAB/NIST (Zook *et al.*, 2014) **for the reliably-callable region of the genome (70%)**

Dataset	Total calls	Specificity	Sensitivity
Whole genome SNVs			
NIST v2.18 Gold Standard	2,740,732		
BWA-MEM-MEM + FreeBayes	2,744,545	0.99769	0.99908
BWA-MEM + HaplotypeCaller	2,748,582	0.99631	0.99916
BWA-MEM + SAMtools fast	2,748,866	0.99622	0.99918
BWA-MEM + SAMtools normal	2,736,410	0.99871	0.99714
GEM3 + FreeBayes	2,742,937	0.99732	0.99812
GEM3 + HaplotypeCaller	2,745,423	0.99745	0.99915
GEM3 + SAMtools fast	2,749,554	0.99533	0.99854
GEM3 + SAMtools normal	2,736,871	0.99833	0.99693

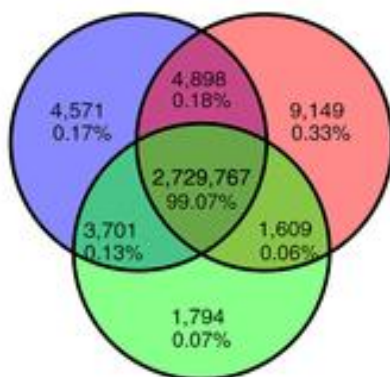


Benchmarking of VC Pipelines

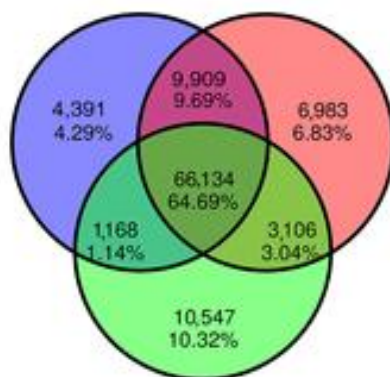
39

Reliably
Callable

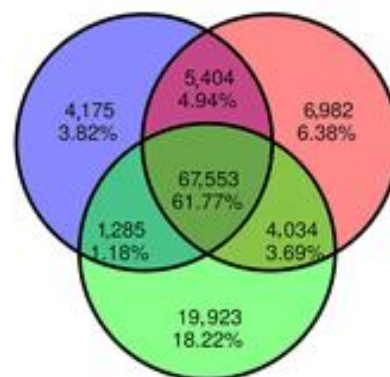
SNVs: NIST reliable



Dels: NIST reliable



Ins: NIST reliable



● FreeBayes ● HaplotypeCaller ● SAMtools

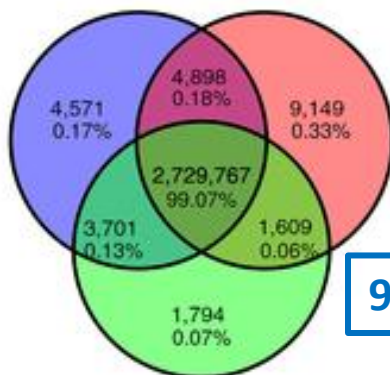
Laurie *et al.* Human Mutation, 2016



Benchmarking of VC Pipelines

40

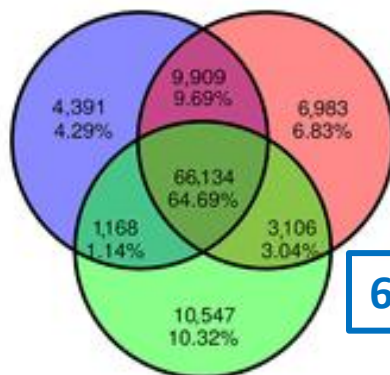
SNVs: NIST reliable



Reliably
Callable

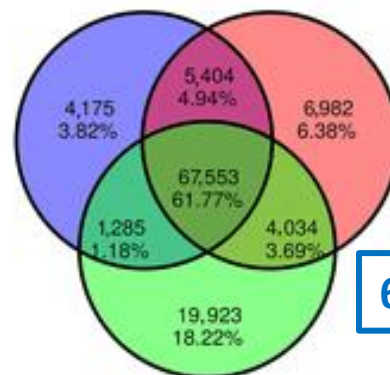
99%

Dels: NIST reliable



65%

Ins: NIST reliable



62%

● FreeBayes ● HaplotypeCaller ● SAMtools

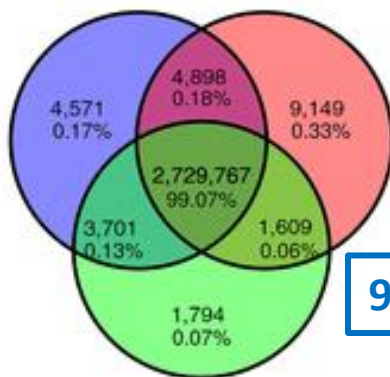
Laurie *et al.* Human Mutation, 2016



Benchmarking of VC Pipelines

41

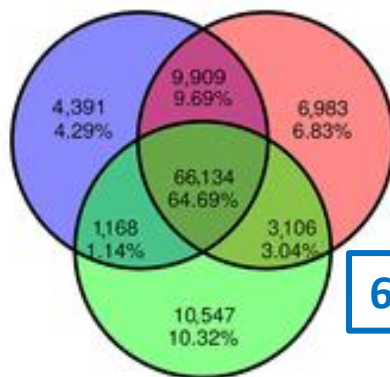
SNVs: NIST reliable



Reliably
Callable

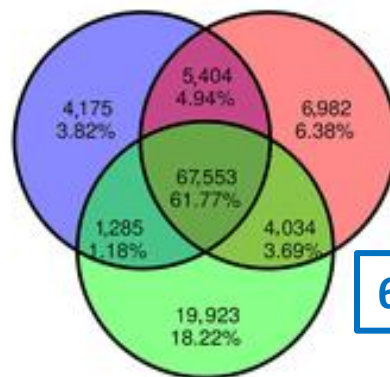
99%

Dels: NIST reliable



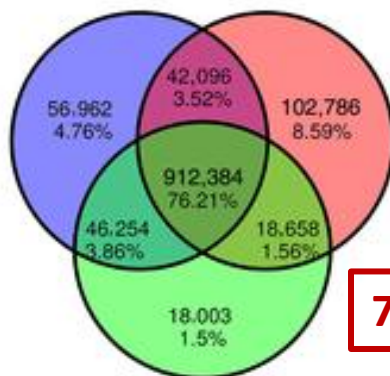
65%

Ins: NIST reliable



62%

SNVs: NIST non-reliable



Not
Reliably
Callable

76%

Dels: NIST non-reliable



31%

Ins: NIST non-reliable



31%

● FreeBayes ● HaplotypeCaller ● SAMtools

Laurie *et al.* Human Mutation, 2016



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Calling CNVs and SVs

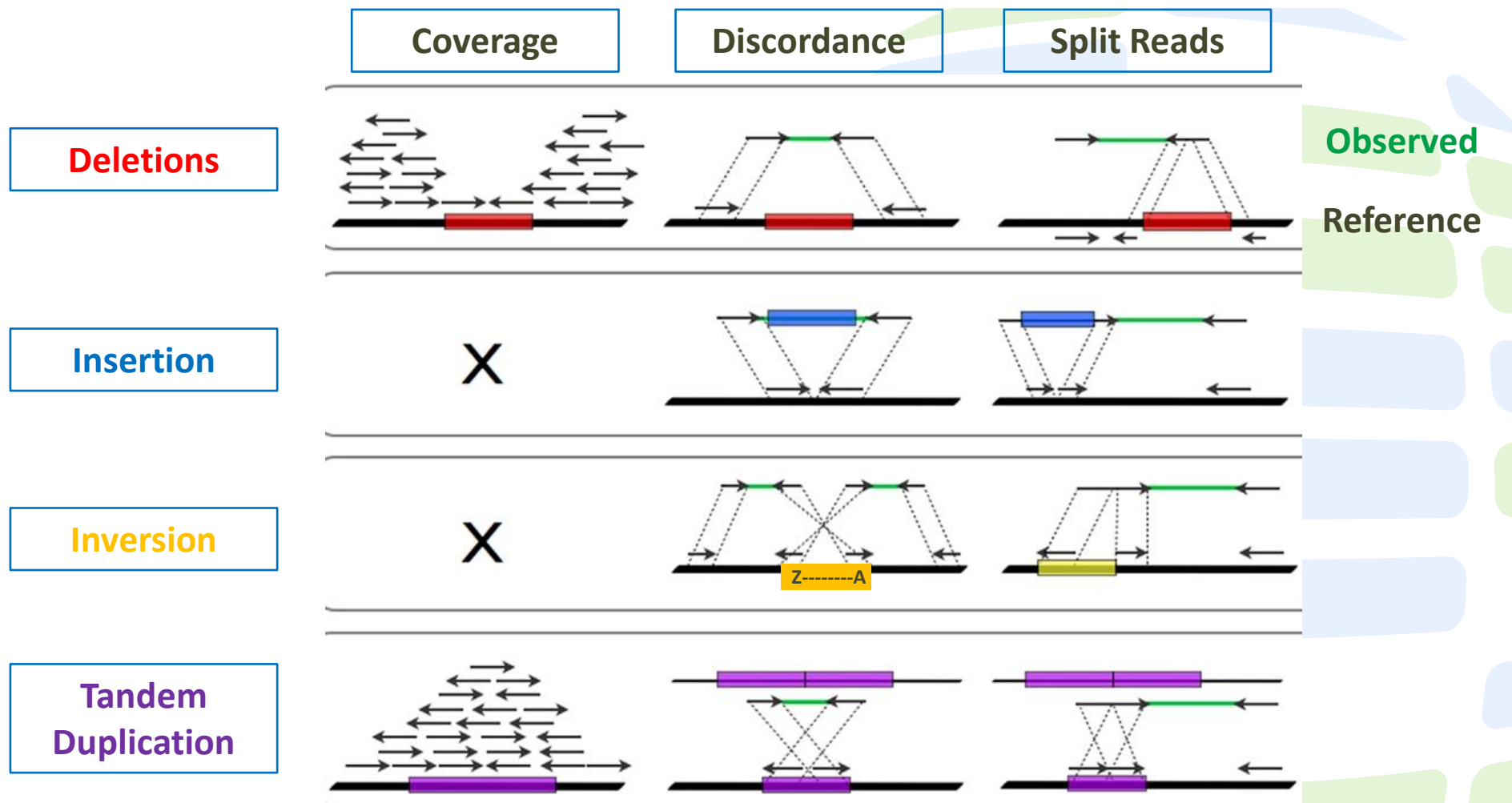
43

- There are **3 main classes of signal** that tools use when attempting to identify the presence of a Copy Number or Structural Variant
 - **Discordant Read Pair Mapping**
 - The gap between the two reads is significantly longer/shorter than expected → Insertion or deletion respectively
 - The orientation of the reads is different from that expected → inversion
 - **Split Read Mapping**
 - The ends of an **individual read** map to different locations
 - **Depth of Coverage (Read Count) Metrics**
 - The depth of coverage in a particular region is significantly more than, or less than expected → copy number gain or copy number loss respectively



Calling CNVs and SVs – The Signals

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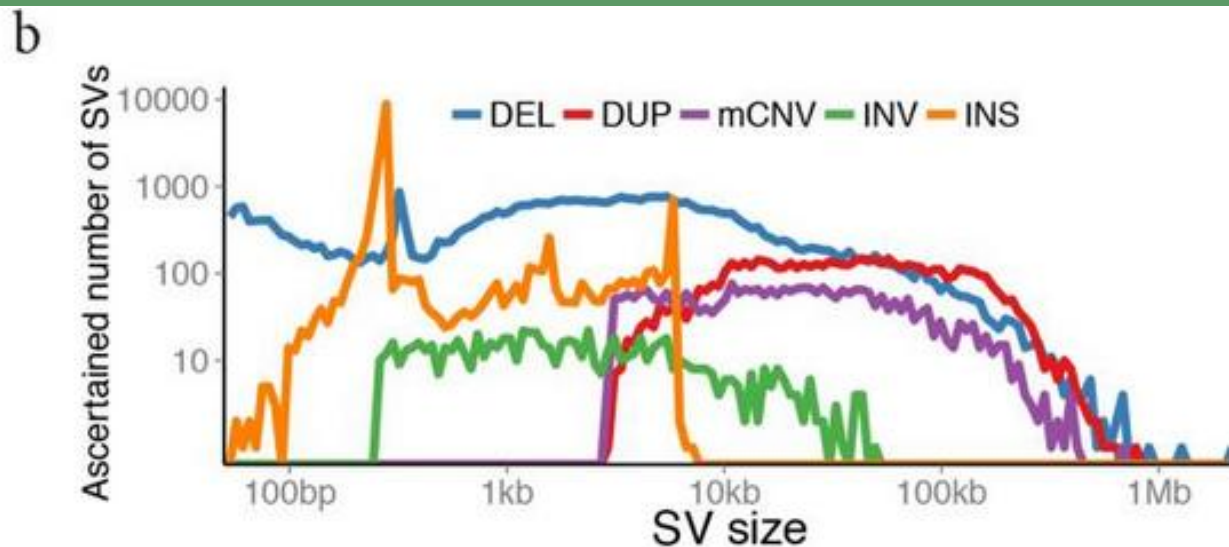
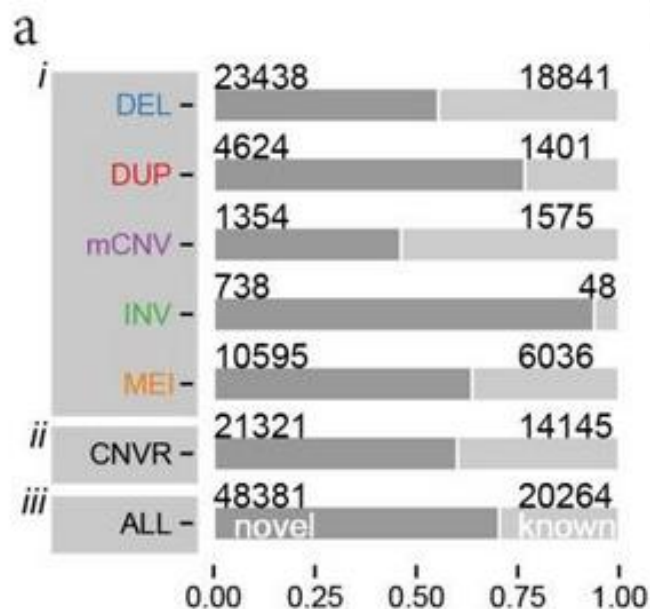


Adapted from Tattini *et al*, 2015



Calling CNVs and SVs - overview

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SV Class	Median Size	Median alleles	Median Kbp
DEL	2455	2788	5615
DUP	35890	17	518
mCNV	19466	340	11346
Inversion	1697	37	78
MEI	297	1218	691

Sudmant, P. *et al.* (2015)



Calling CNVs from WGS data

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- Popular tools include
 - cm.mops
 - CNVnator
 - Control-FreeC
 - Delly
 - ERDS
 - GenomeSTRiP
 - Lumpy





Calling CNVs from WGS data

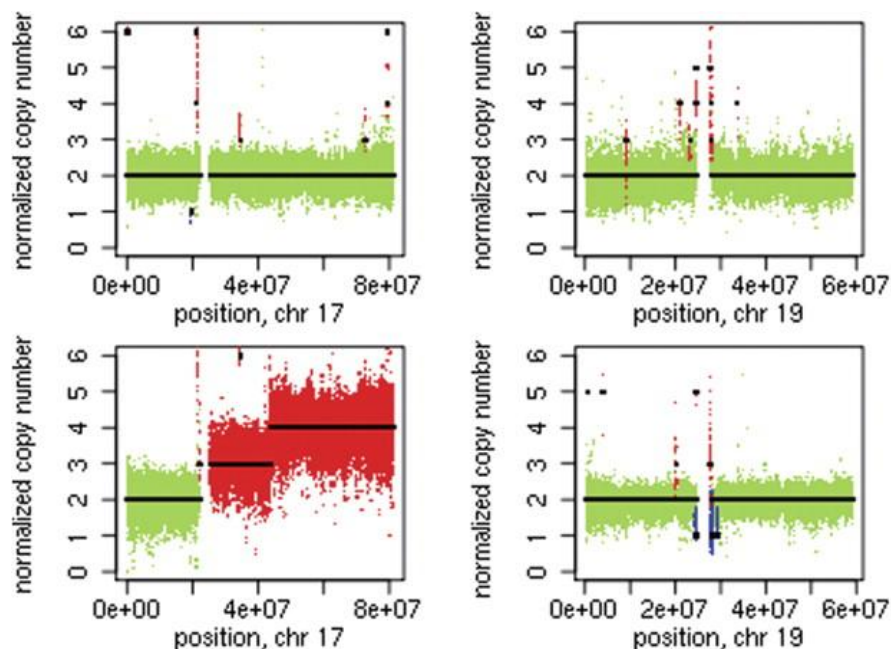
47

- In general easier than for WES data
- Can typically be used on a single sample
- Account for sources of bias such as GC content, and low complexity regions
- Sensitive to stochastic coverage effects

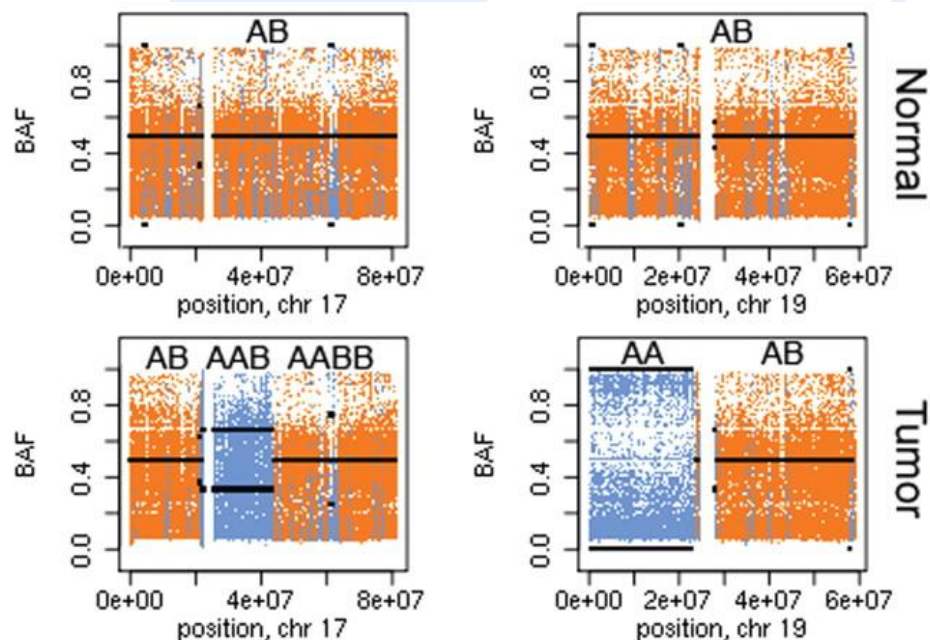


Calling CNVs from WGS data

48



Normalised
Copy Number



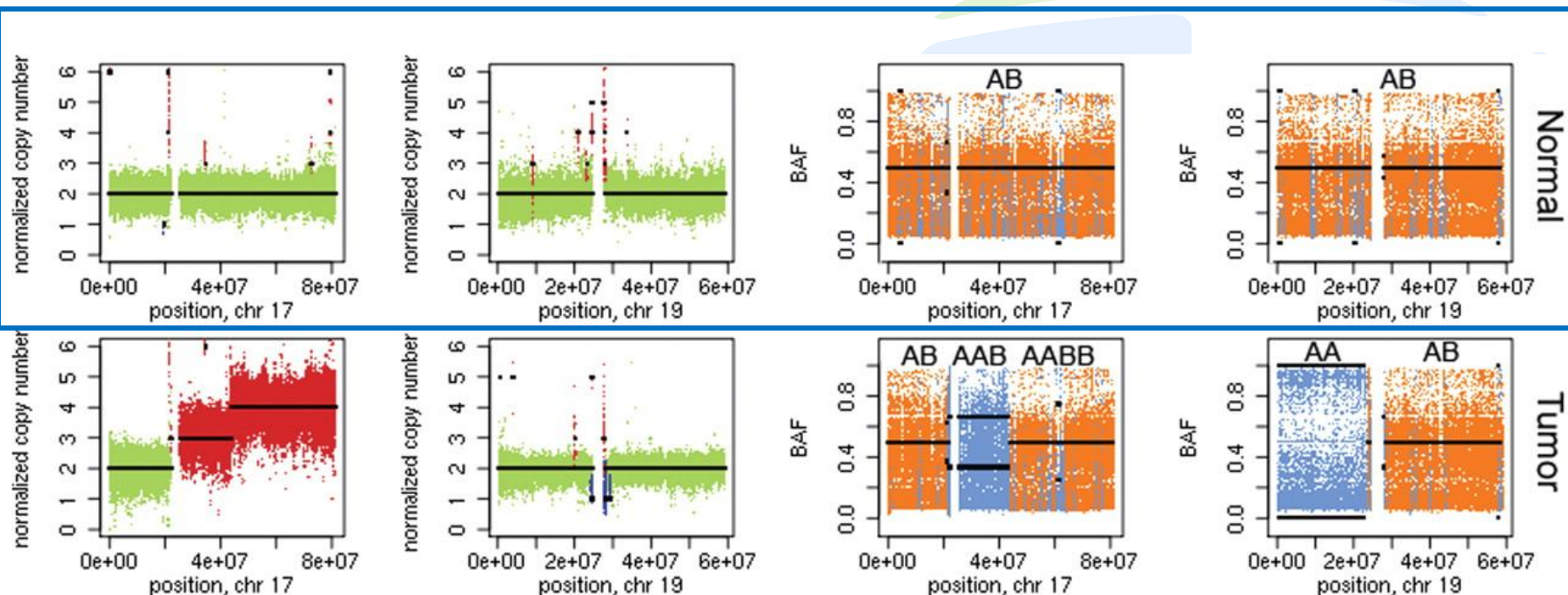
Beta Allele
Frequency

Boeva, V. et al. (2012)



Calling CNVs from WGS data

49



Normalised
Copy Number

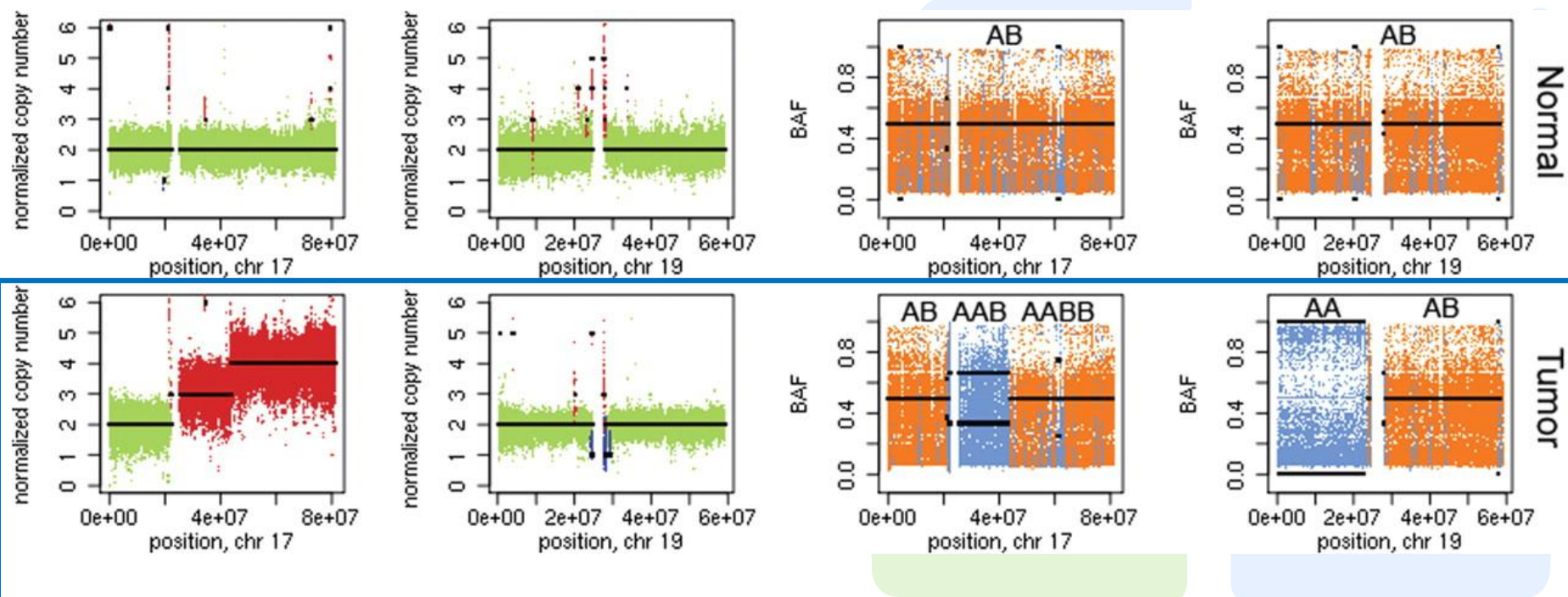
Beta Allele
Frequency

Boeva, V. et al. (2012)



Calling CNVs from WGS data

50



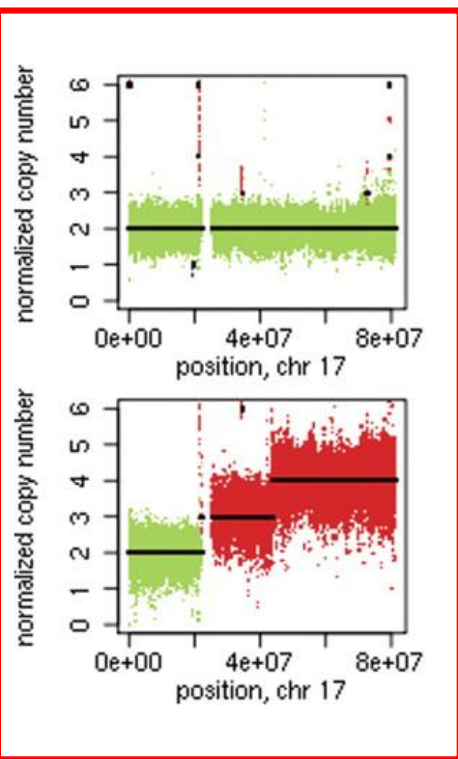
Normalised
Copy Number

Beta Allele
Frequency

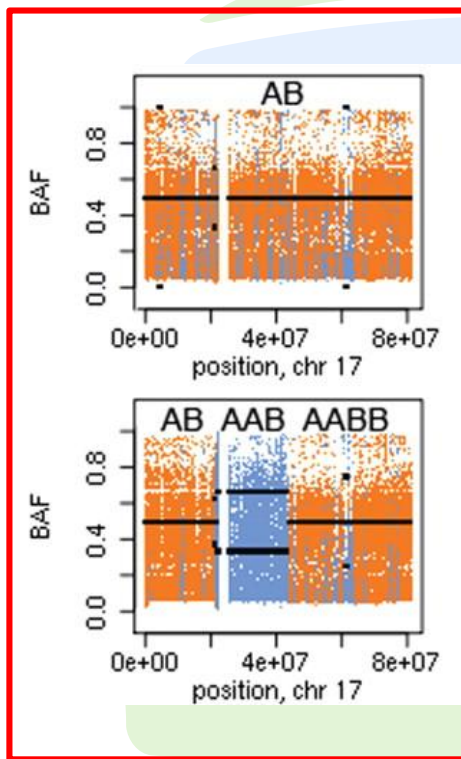
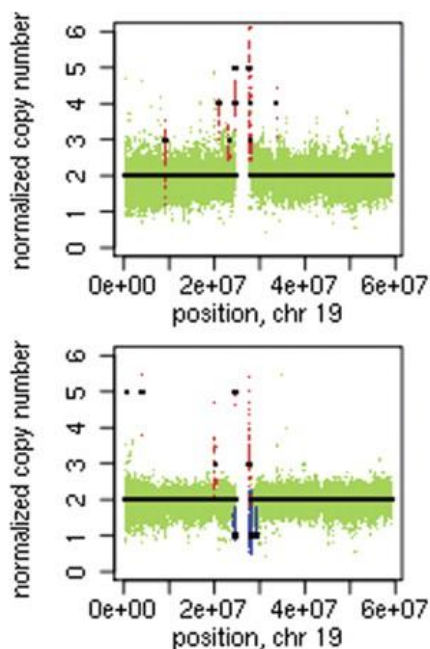


Calling CNVs from WGS data

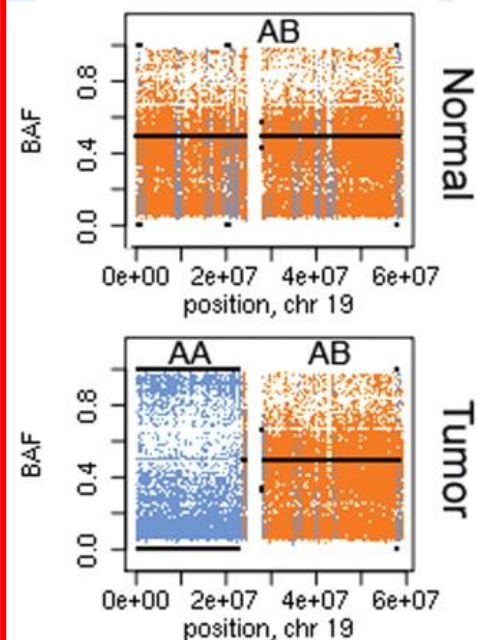
51



Normalised
Copy Number



Beta Allele
Frequency



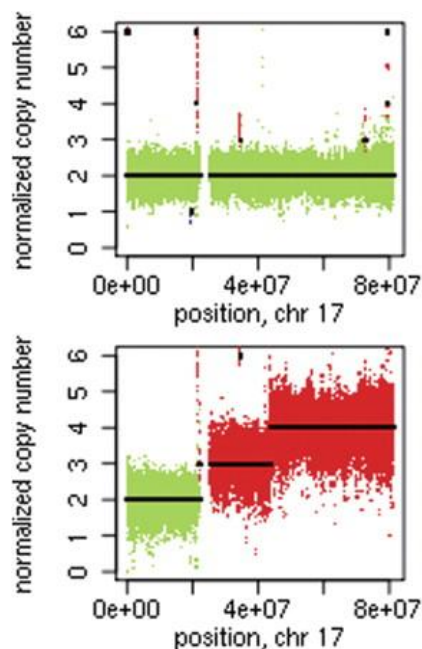
Normal

Tumor

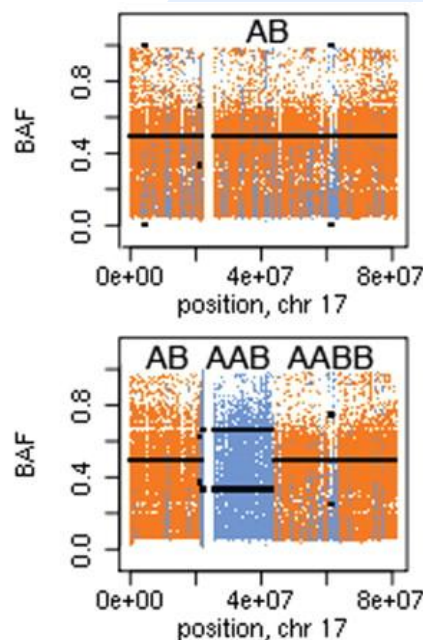
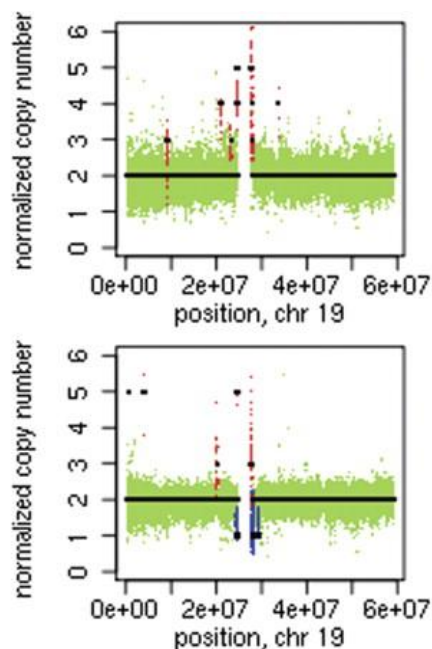


Calling CNVs from WGS data

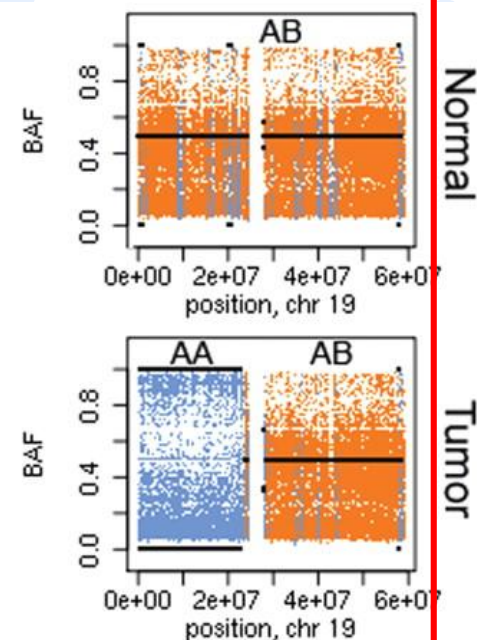
52



Normalised
Copy Number



Beta Allele
Frequency

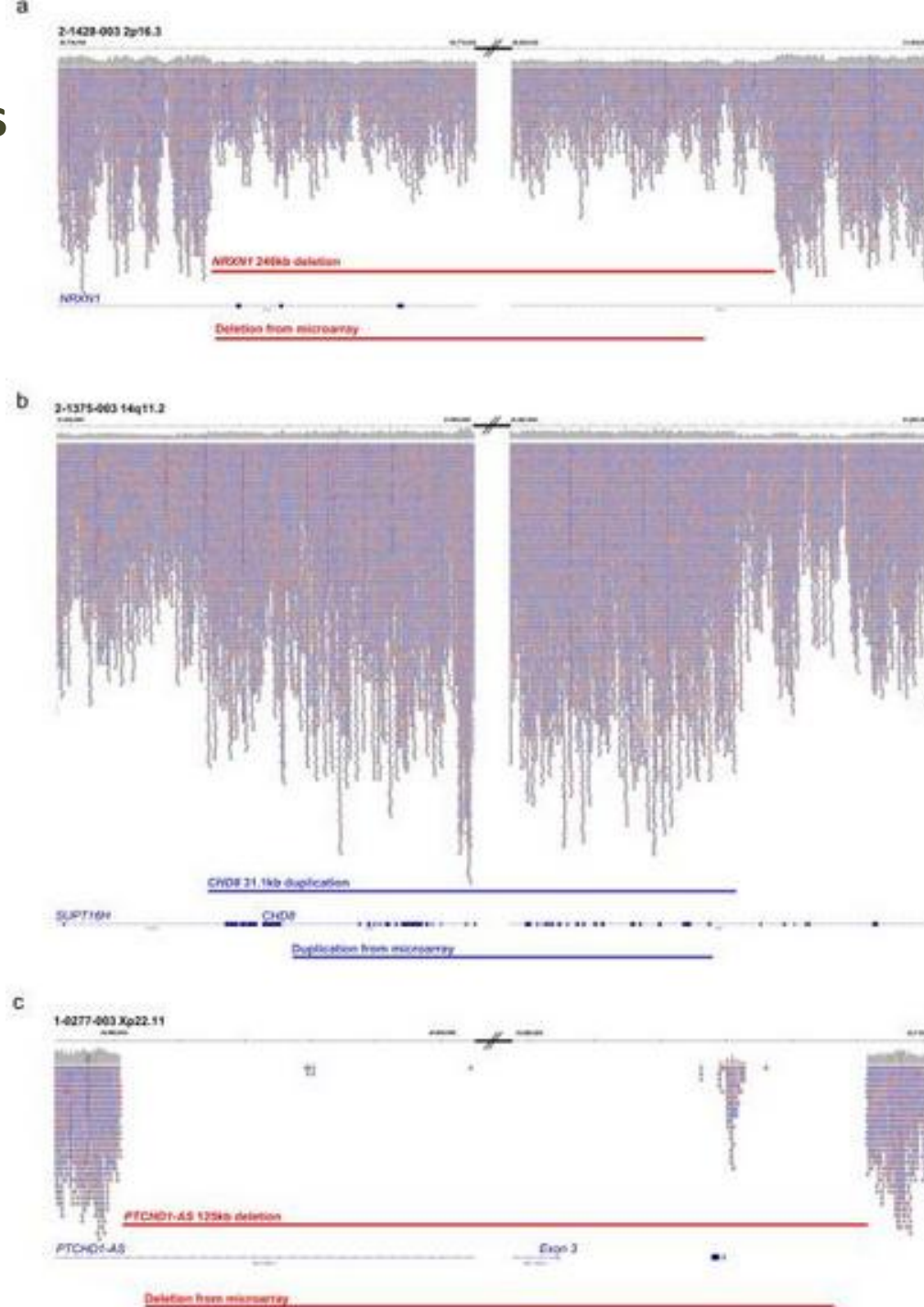


Normal

Tumor

Boeva, V. et al. (2012)

Calling CNVs from WGS data

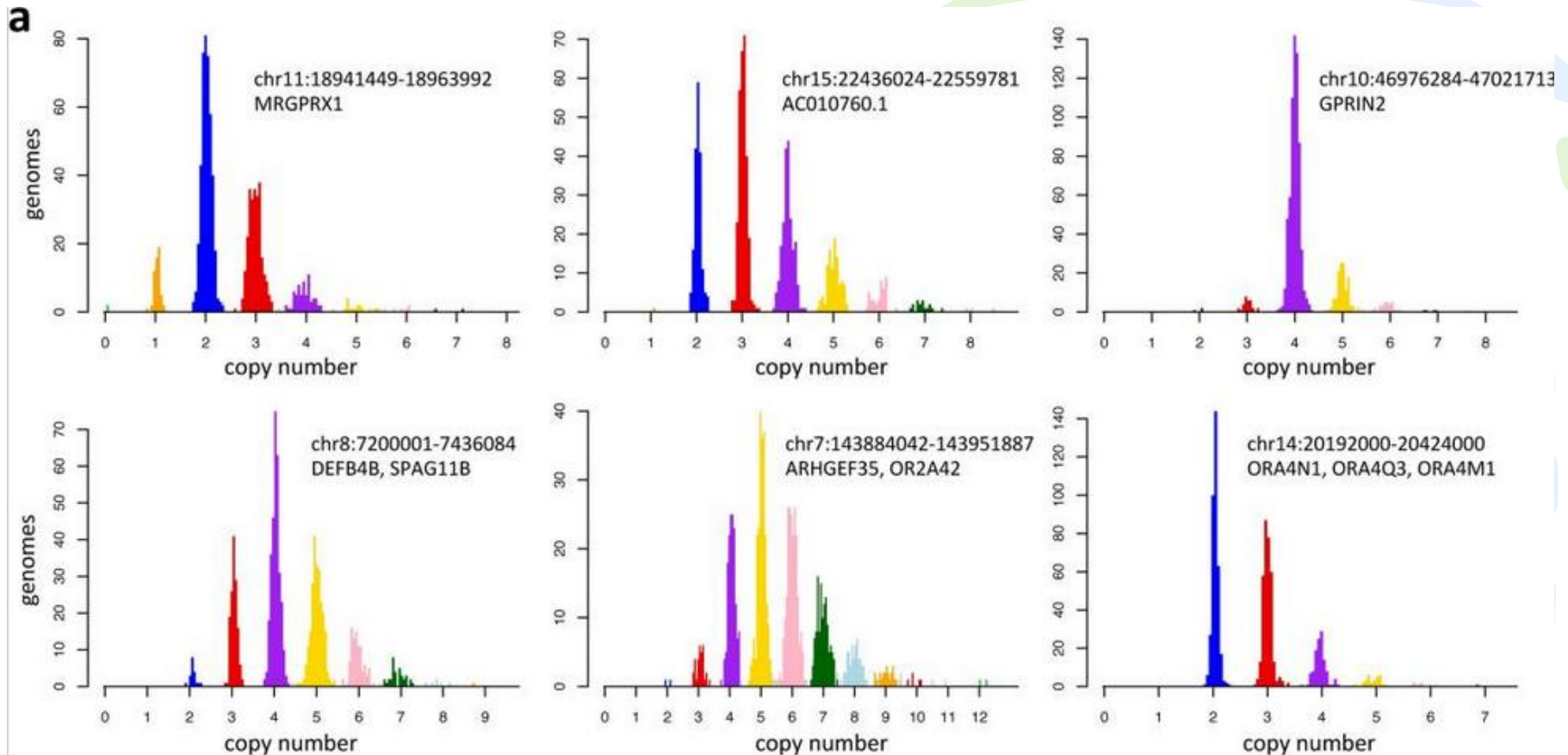


Yuen, RK *et al.*
(2017)



mCNVs are segregating in the wild

54



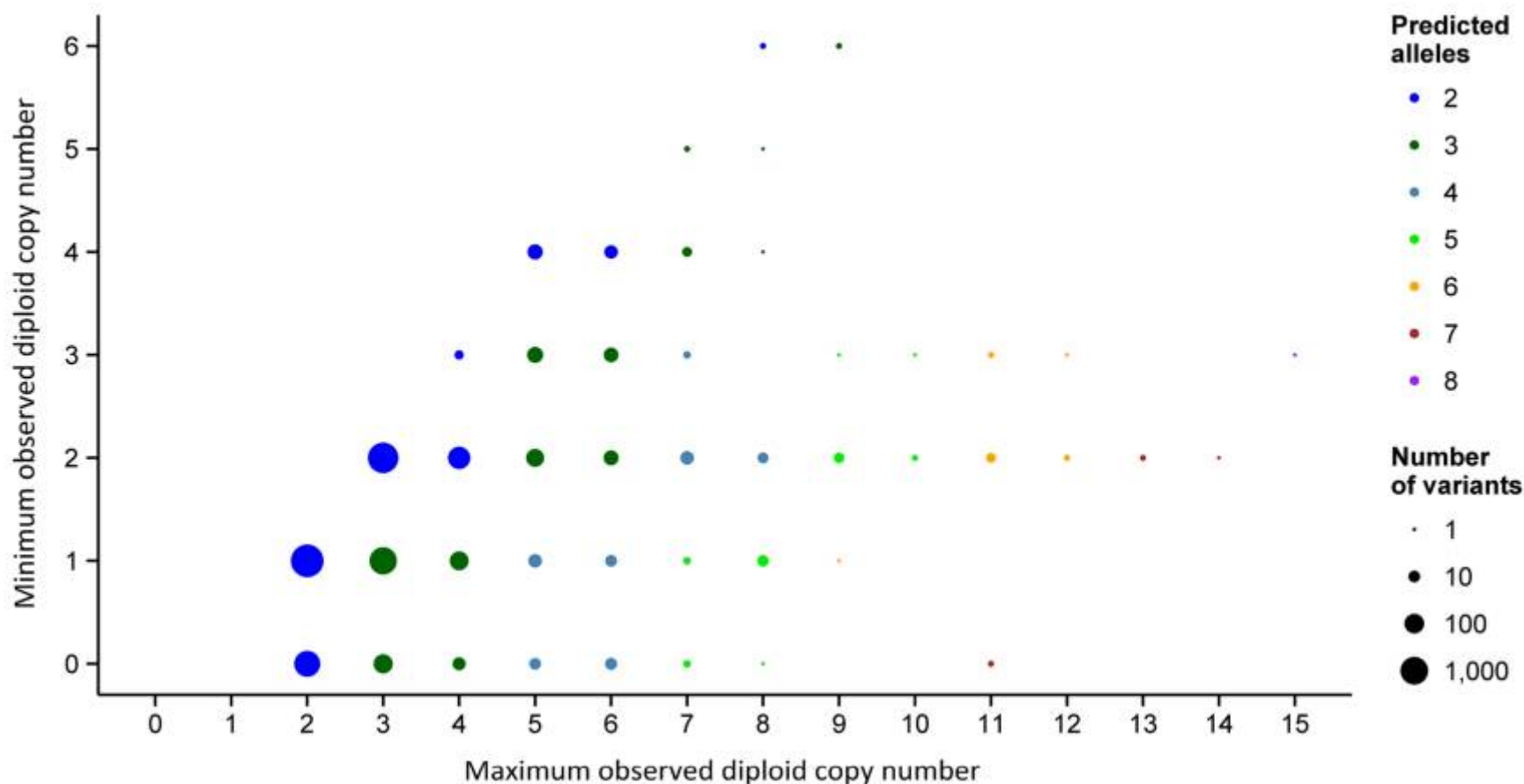
Handsaker, RA et al. (2015)



mCNVs are segregating in the wild

55

b



Handsaker, RA *et al.* (2015)



Calling CNVs from WES data

56

- Unlike WGS data, WES data is discontinuous meaning it is virtually impossible to detect large SVs, other than large deletions
- The discontinuity also makes it difficult, but not impossible to detect Copy Number Variants. Most tools require a minimum of 3 exons to be affected to make a reliable call
- Detection is further complicated by the fact that coverage is not uniformly distributed across the capture regions, with peaks in the middle, dropping off to the sides



Coverage

Exons





Calling CNVs from WES data

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- Tools have to normalise, both horizontally, and vertically
 - Comparison to a reference set
 - Account for factors such as GC content, low complexity regions
 - Account for batch-type effects, by removing sources of extreme variance using PCA/SVD
- Make calls, typically using a Hidden Markov Model (HMM)
- Identify regions that appear significantly different in a specific sample when compared to the reference set



Calling CNVs from WES data

59

- Tools have to normalise, both horizontally, and vertically
 - Comparison to a reference set
 - Account for factors such as GC content, low complexity regions
 - Account for batch-type effects, by removing sources of extreme variance using PCA/SVD
- Make calls, typically using a Hidden Markov Model (HMM)
- Identify regions that appear significantly different in a specific sample when compared to the reference set
- **Even when detected, we don't know where they are**



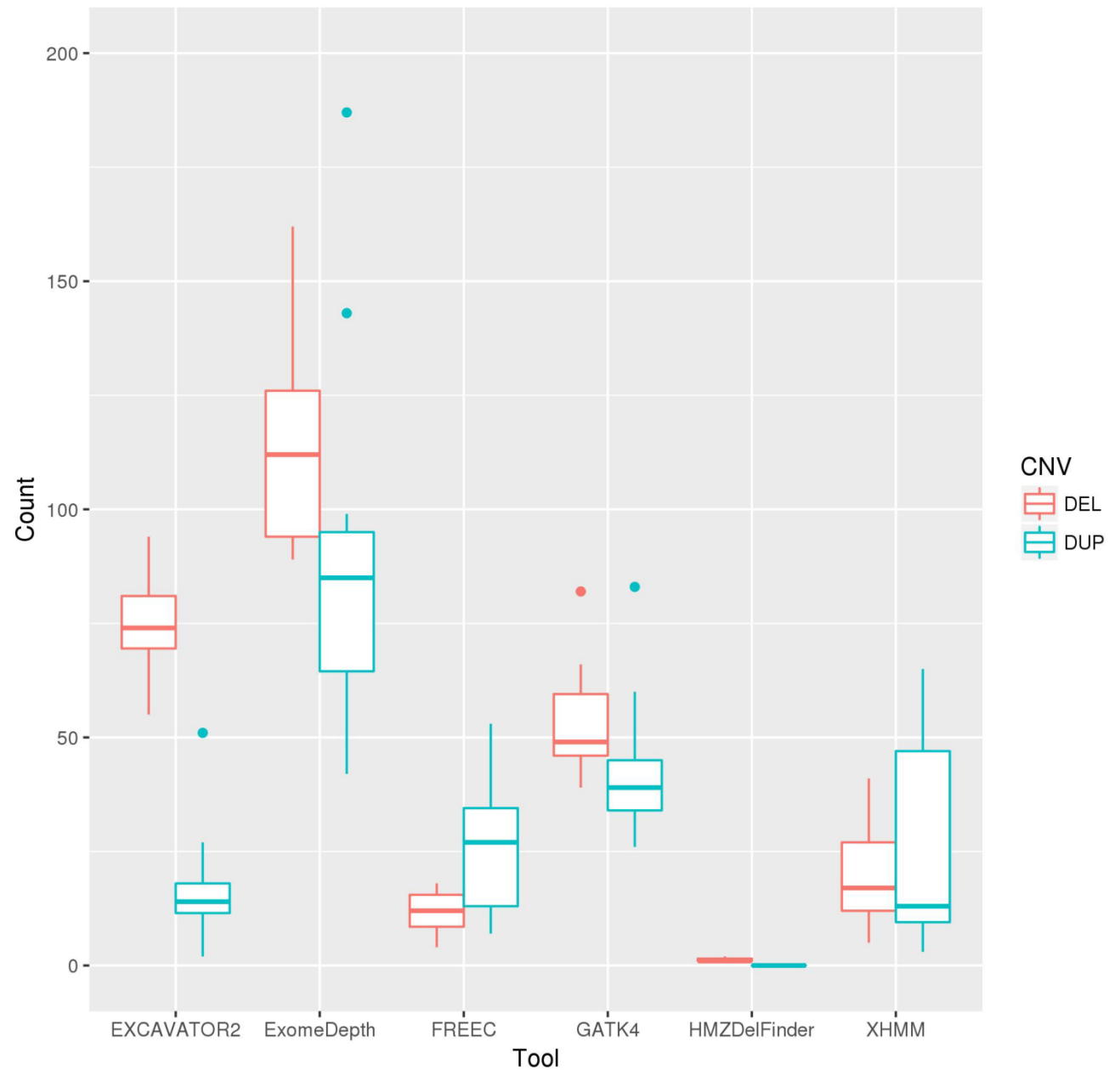
Calling CNVs from WES data

60

- Popular tools include
 - ExomeDepth (1 versus 10)
 - Conifer (All v All 8+)
 - XHMM (All v All – rare)
- Other notable options
 - Control-FreeC (ongoing development)
 - GATK-4 (Coming soon ...)
 - For all tools, **the more standardised your data, the better they will perform** i.e. Capture kit, sequencing depth, sequencing lab etc.

Calling CNVs from WES data

Number of events per tool at coverage 90



Sandra Rédo



Large Structural Variants – WGS

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- Popular tools include
 - BreakDancer
 - cm.mops
 - CNVnator
 - Control-FreeC
 - Delly
 - ERDS
 - GenomeSTRiP
 - Lumpy
 - Pindel





Large Structural Variant Classes

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Translocation



Inversion



Large Insertions
and Deletions



Reference
Chromosome



In principle should be easy – **lots of signal** 😊

cxSV Subclass	Abbreviation	SVs	Validation Rate	Total Observations	Subjects with ≥ 1 SV	Median Size (kb)	DEL INS DUP INV	Rearrangement Schematic	Simulated Copy Number Profile
Inversion with 5' Flanking Deletion	delINV	38	100% (19/19)	2,301	99.7%	12.2		5' ref. DEL A INV B ref. 3'	
Inversion with 3' Flanking Deletion	INVdel	40	100% (19/19)	2,242	100.0%	9.3		5' ref. INV A DEL B ref. 3'	
Paired-Deletion Inversion	delINVdel	58	96% (25/26)	2,288	98.0%	15.2		5' ref. DEL A INV B DEL C ref. 3'	
Inversion with 5' Flanking Duplication	dupINV	7	75% (3/4)	62	8.7%	54.6		5' ref. DUP A INV B ref. 3'	
Inversion with 3' Flanking Duplication	INVdup	3	100% (1/1)	6	0.9%	82.9		5' ref. INV A DUP B ref. 3'	
Paired-Duplication Inversion	dupINVdup	45	96% (27/28)	151	19.0%	112.5		5' ref. DUP A INV B DUP C ref. 3'	
Inversion with 5' Flanking Duplication and 3' Flanking Deletion	dupINVdel	6	100% (2/2)	9	1.3%	27.3		5' ref. DUP A INV B DEL C ref. 3'	
Inversion with 5' Flanking Deletion and 3' Flanking Duplication	delINVdup	10	100% (5/5)	90	12.2%	67.5		5' ref. DEL A INV B DUP C ref. 3'	
Inverted Duplication with Flanking Triplication	dupTRIPdup-INV	5	100% (5/5)	5	0.7%	113.9		5' ref. DUP A B C INV A B C ref. 3'	
Inverted Repeat / Inverted Tandem Duplication	IR	11	88% (7/8)	36	5.1%	73.8		5' ref. DUP A INV A ref. 3'	
Compound CNV	cpdCNV	22	100% (17/17)	2,085	99.4%	29.1		Various	
Dispersed Duplication	dDUP	10	100% (2/2)	42	6.1%	17.5		5' ref. DUP A ref. INS A' ref. 3'	
Dispersed Duplication with Deletion	dDUPdel	9	100% (4/4)	60	8.5%	32.1		5' ref. DUP A ref. DEL B INS A' ref. 3'	
Insertion with Deletion	INSdel	4	100% (1/1)	12	1.7%	5.9		5' ref. A ref. DEL B INS A ref. 3'	
Compound Insertion	cpdINS	5	100% (2/2)	251	36.0%	3.1		Various	
Compound Insertion with Deletion	cpdINSdel	1	NA (0/0)	5	0.7%	9.6		Various	
Compound/Complex Rearrangement (or Other)	CCR	15	100% (11/11)	21	3.1%	239.8	Various		
All cxSV	-	289	97% (150/154)	9,666 (14/subject)	100.0%	27.3	DEL: 61.8% DUP: 47.8% INV: 84.8% INS: 11.8%		

Collins, RL et al. (2017)



Large Structural Variants – WGS

65

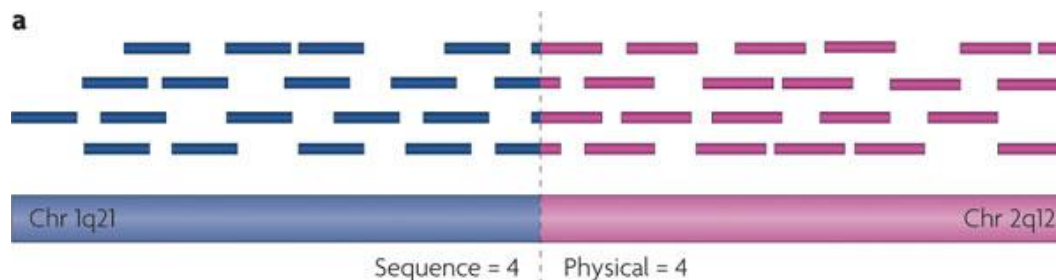
- Most tools have been tailored to best identify specific classes of SV
 - Therefore may want to use more than one tool
- More recently developed tools tend to look at more than one type of evidence, and thus can call different classes
- To optimise discovery of SVs, ideally want to use **a mix of library strategies** and/or technologies i.e. short-read and long-read simultaneously



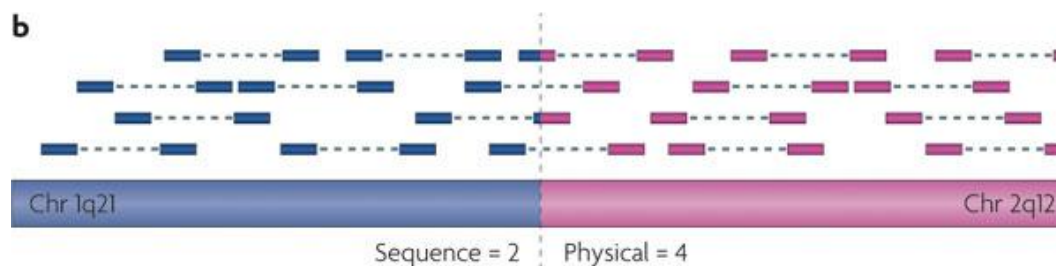
Large Structural Variants – WGS

66

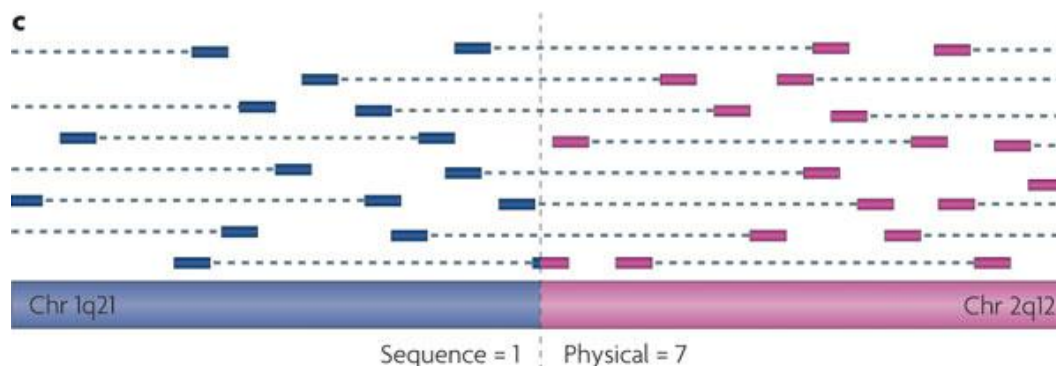
Single End



Paired End



“Mate Pairs”





Other interesting topics

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- Assembly approaches to structural variant detection
- Long read technologies e.g. PacBio and Oxford Nanopore
- Somatic variant calling
- Balance cytogenic abnormalities



Acknowledgements

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cnag

centre nacional d'anàlisi genòmica
centro nacional de análisis genómico

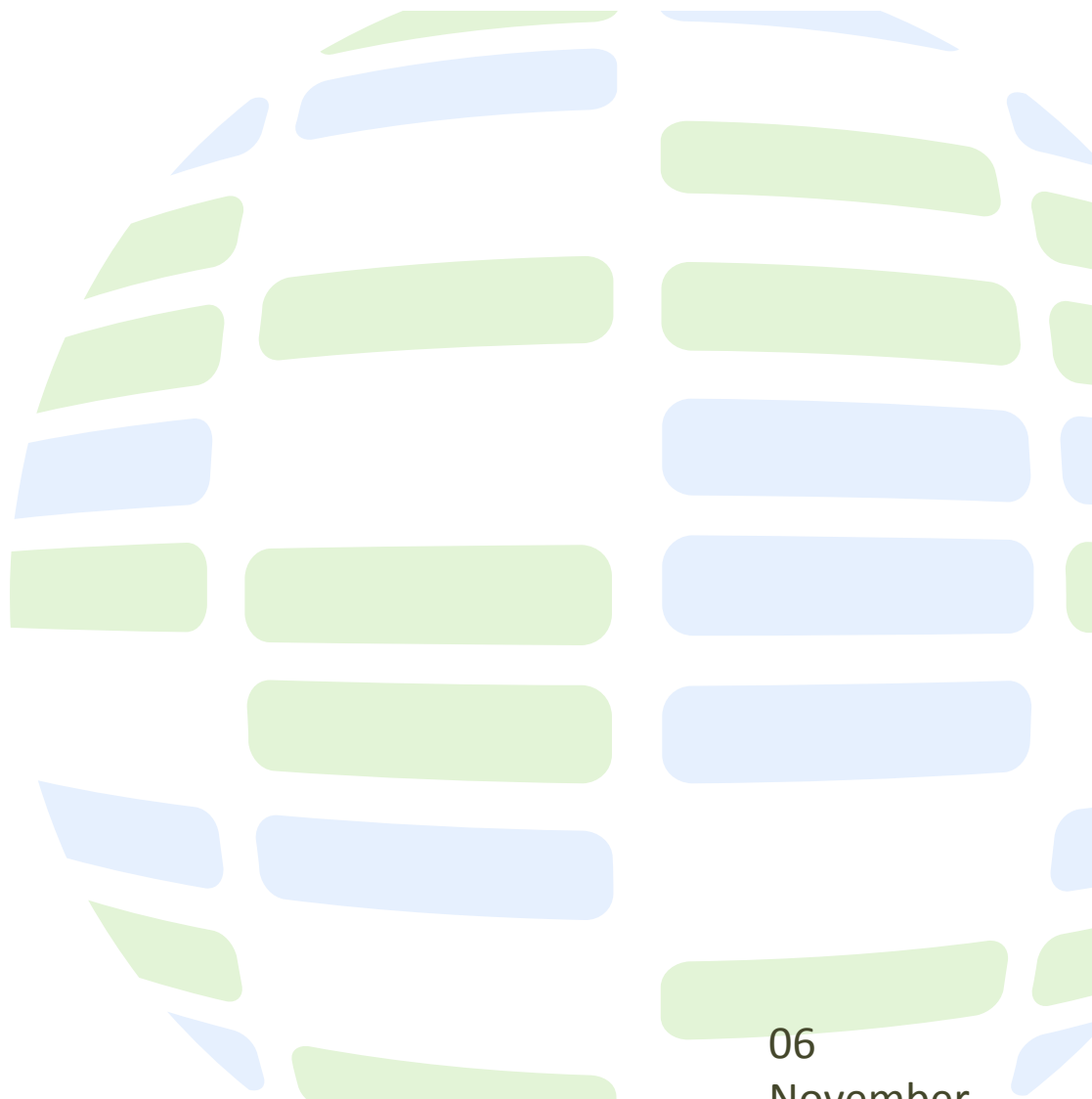
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