



Population sequencing and analysis projects at NYGC

Michael C. Zody

Scientific Director, Computational Biology, New York Genome Center

November 11, 2021



OVERVIEW

- Summary of large scale genome projects
 - TOPMed
 - CCDG
 - 1000 Genome high coverage
- Methods development for large scale projects
 - Absinthe insertion detector
 - Structural variant phasing and imputation

TOPMED

- Trans-Omics for Precision Medicine
- NHLBI project to create resources for deeply phenotyped cohorts
 - Whole genome sequencing for >130,000 samples of diverse ancestry
 - RNA-Seq, metabolomics, proteomics
- Flagship paper in Nature this year (Taliun et al., 2021)
 - Analyzed >53,000 genomes
 - >400M variants discovered (~50% singletons)
 - Imputation panel with >97,000 genomes
 - Discovery of >1000 non-reference sequences from AC = 1 to 100% AF

CCDG

- Centers for Common Disease Genomics
- NHGRI project to develop paradigms for understanding genetic architecture of common disease
 - Whole genome sequencing for >130,000 samples of diverse ancestry
 - Exome sequencing for an additional 198,000 samples
- Phenotypes include ASD, epilepsy, heart disease, stroke, IBD

CCDG ANALYSIS PLANS

- Whole genome sequencing now complete
- Final (“Freeze 3”) data set called
- Joint SNV/indel calling with GATK (Broad)
- Distributed SV call set (WashU, Baylor, NYGC)
 - Lumpy (deletions and inversions)
 - Absinthe (insertions)
 - Canvas + QuickKmer2 (depth of coverage/copy number)
 - Genotyping of long read derived variants with Paragraph
- SV calls will be genotyped on all samples and merged into a single set
- Imputation server based on the Michigan/TOPMed model
- Timeline for release in 2022

DATA AVAILABILITY FROM TOPMED AND CCDG

- Both projects intend to broadly share data
- Both projects consist of collections of older cohorts with a wide variety of patient consents ranging from general research to disease specific
- TOPMed data are currently available on a per cohort basis from dbGaP and BioData Catalyst
- CCDG data will be publicly available on AnVIL (access controlled through dbGaP)
- Imputation servers will be available as a service only (no downloadable panels) due to access restrictions

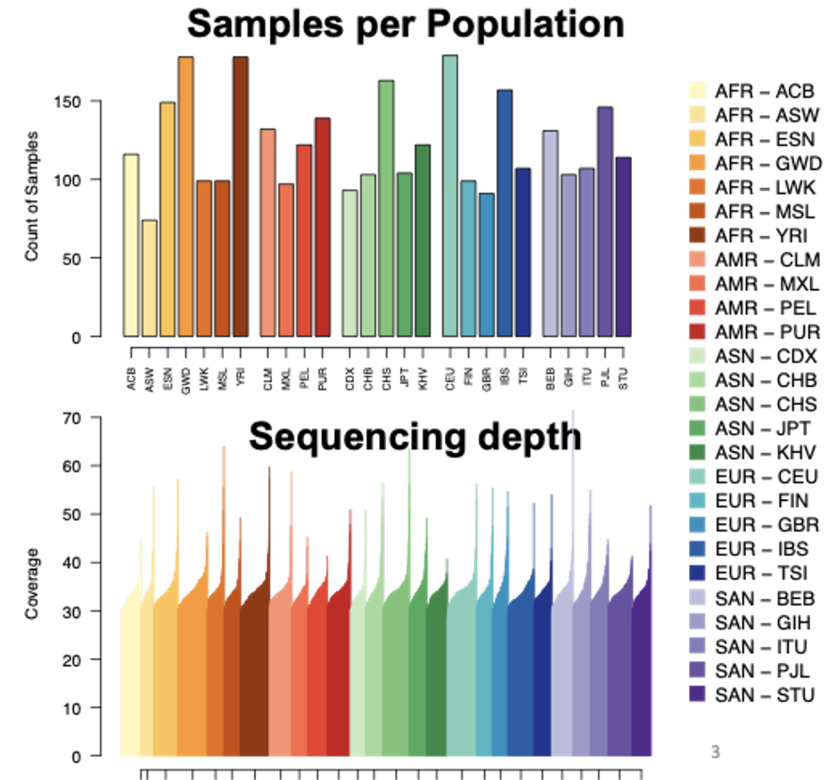
1000 GENOMES PROJECT SEQUENCING

- Supplement to CCDG
- 30x Illumina of all 2,504 phase 3 samples
- Additional 698 sample sequenced to complete 602 trios
- GATK joint calling for SNVs/indels
- Comprehensive combined SV calling from the HGSVC
- All data released through EBI/ISGR and NCBI:
<https://www.internationalgenome.org/data-portal/data-collection/30x-grch38>
- Data are also available on AnVIL (Google cloud) and AWS
- Preprint up on biorxiv (Byrska-Bishop, Evani, Zhao, *et al.*, 2021)

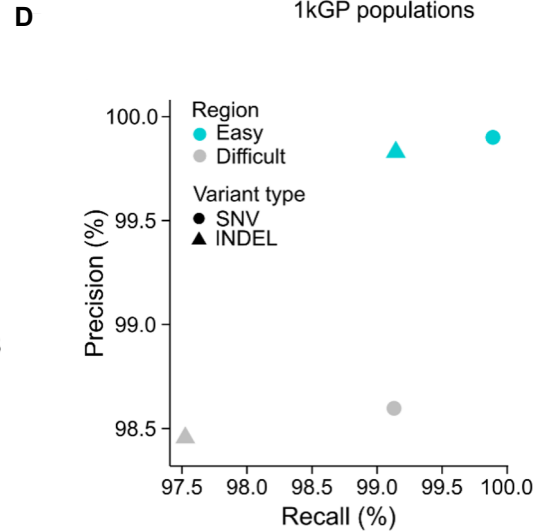
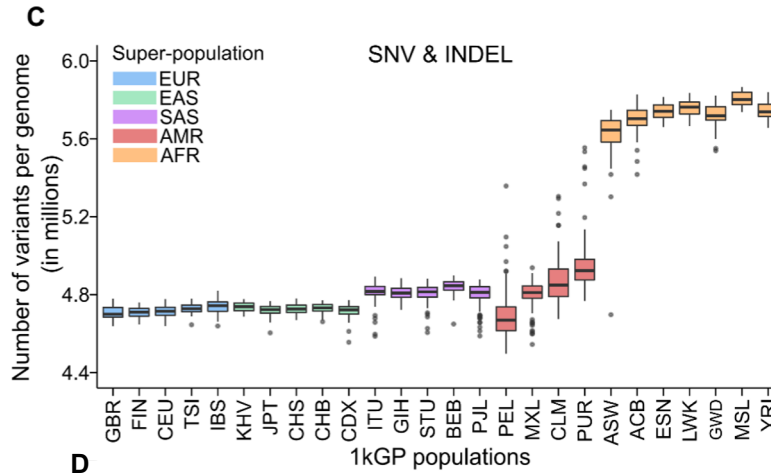
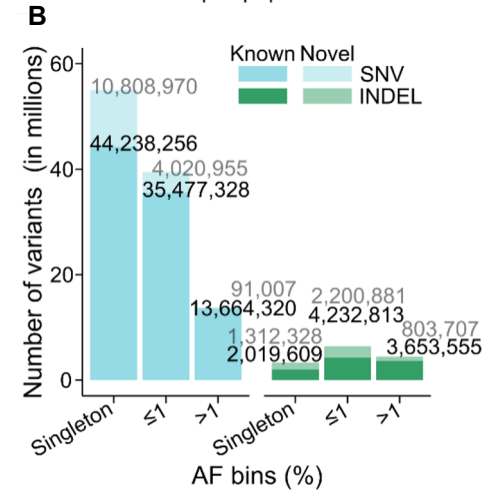
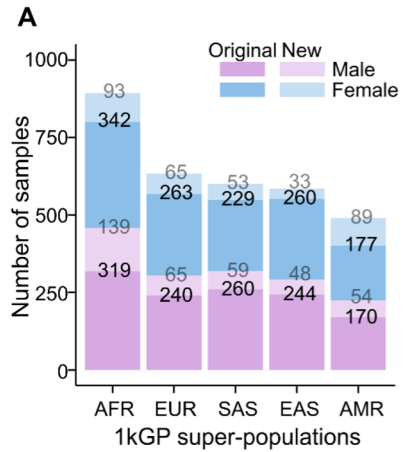
1000 GENOMES OVERVIEW

- 3,202 genomes (2,504 original + 698 new) collected from 26 populations, including:
 - 602 complete trios
 - 6 parent-child duos
- All samples were sequenced to a targeted depth of 30X by the NYGC.
- SNVs and INDELS were discovered using GATK's HaplotypeCaller; SVs were discovered with the GATK-SV pipeline^[1], the svtools pipeline^[2] and Absinthe^[3].
- 2,504 unrelated samples were previously sequenced to ~7.4X (phase 3 callset)^[4,5].

[1] Collins *et al.* 2020. Nature
 [2] Abel *et al.* 2020. Nature
 [3] Corvelo A. *in prep.*
 [4] The 1000 Genomes Project Consortium. 2015. Nature
 [5] Sudmant *et al.* 2015. Nature



SNV/INDEL DISCOVERY



Summary stats:

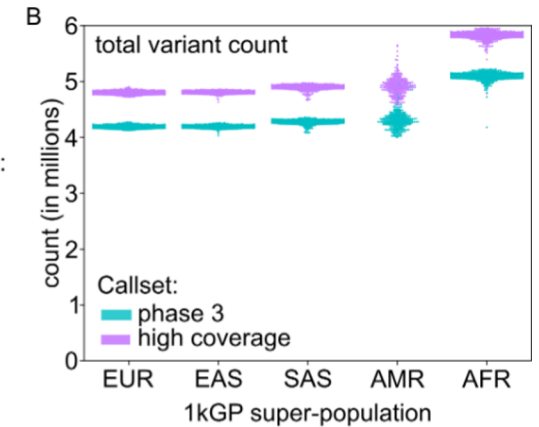
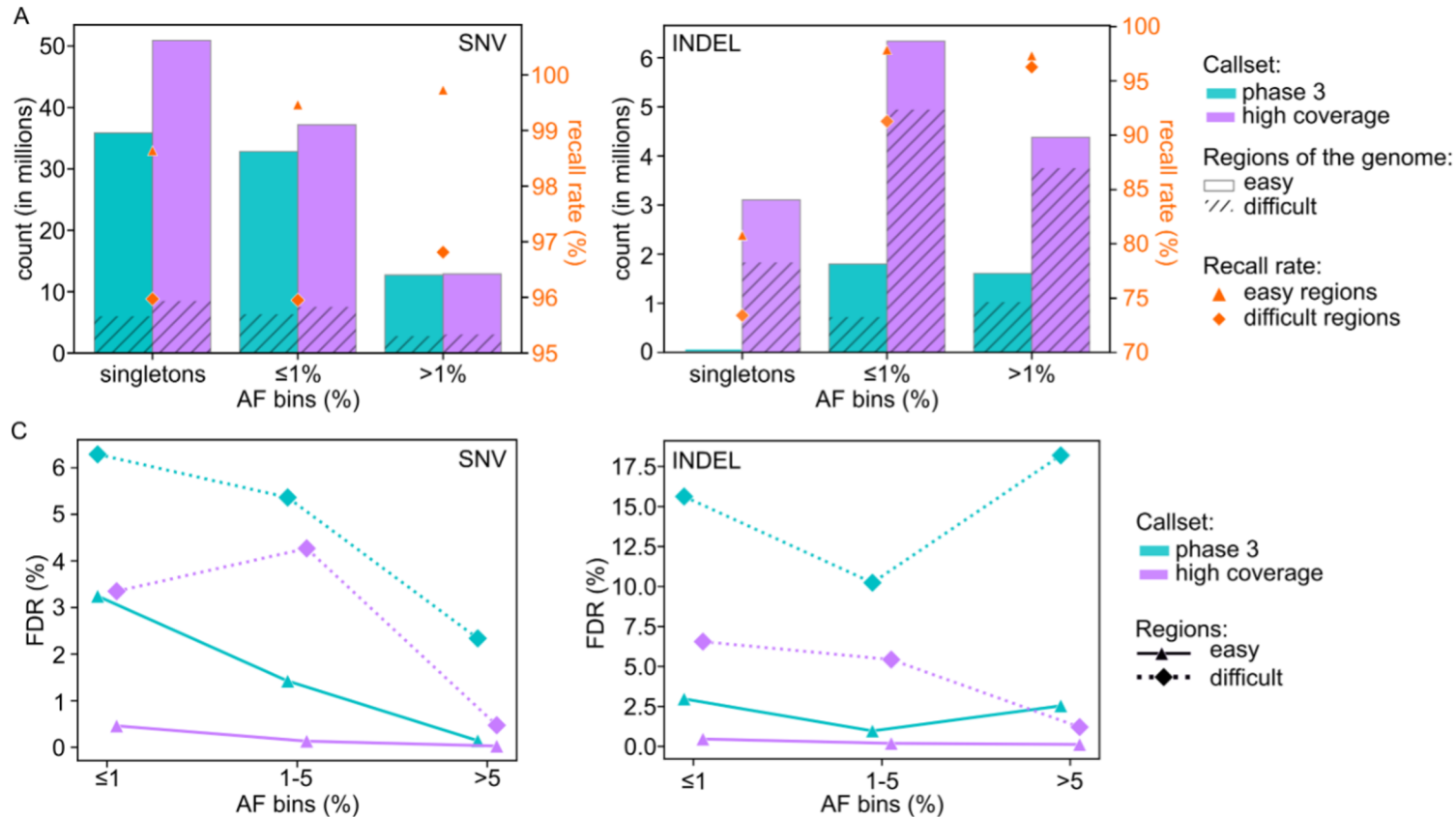
	Cohort level		Per sample (mean)	
	SNV	INDEL	SNV	INDEL
Total	111,048,944	14,435,076	4,080,992	871,923
Singletons	55,047,226	3,331,937	23,197	
Novel	14,920,932	4,316,916		

Comparison against the GIAB truth set:

Variant type	FDR (%)
SNV	0.3
INDEL	1.15

COMPARISON TO 1KG PHASE 3

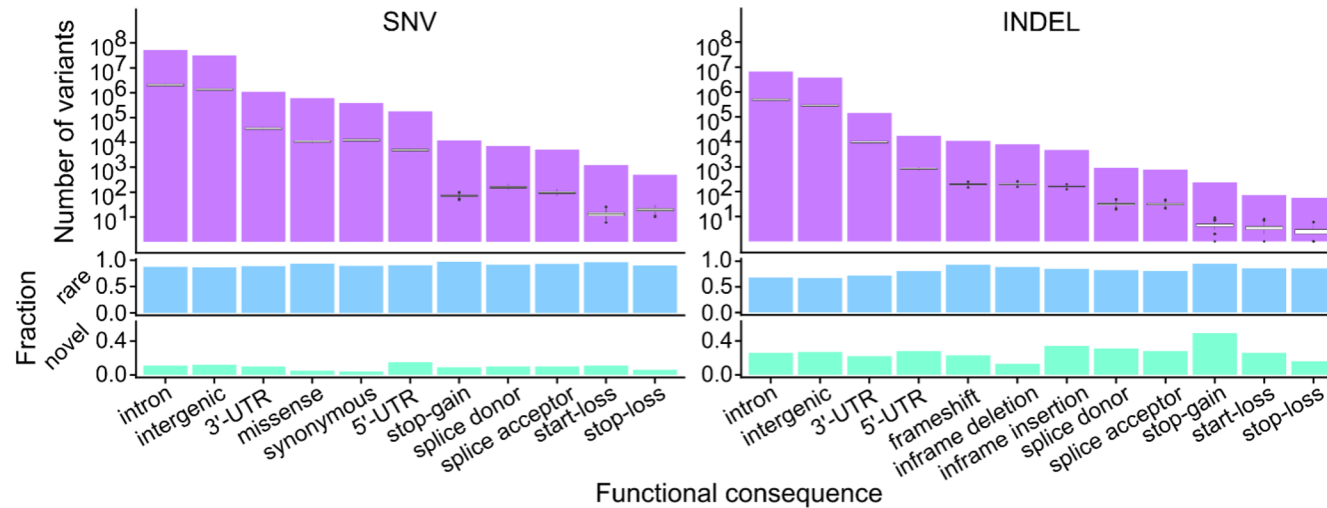
- Comparison restricted to the 2,504 samples shared between the two callsets.
- Used the GRCh38 lifted-over version of the phase 3 callset.



FDR (%):

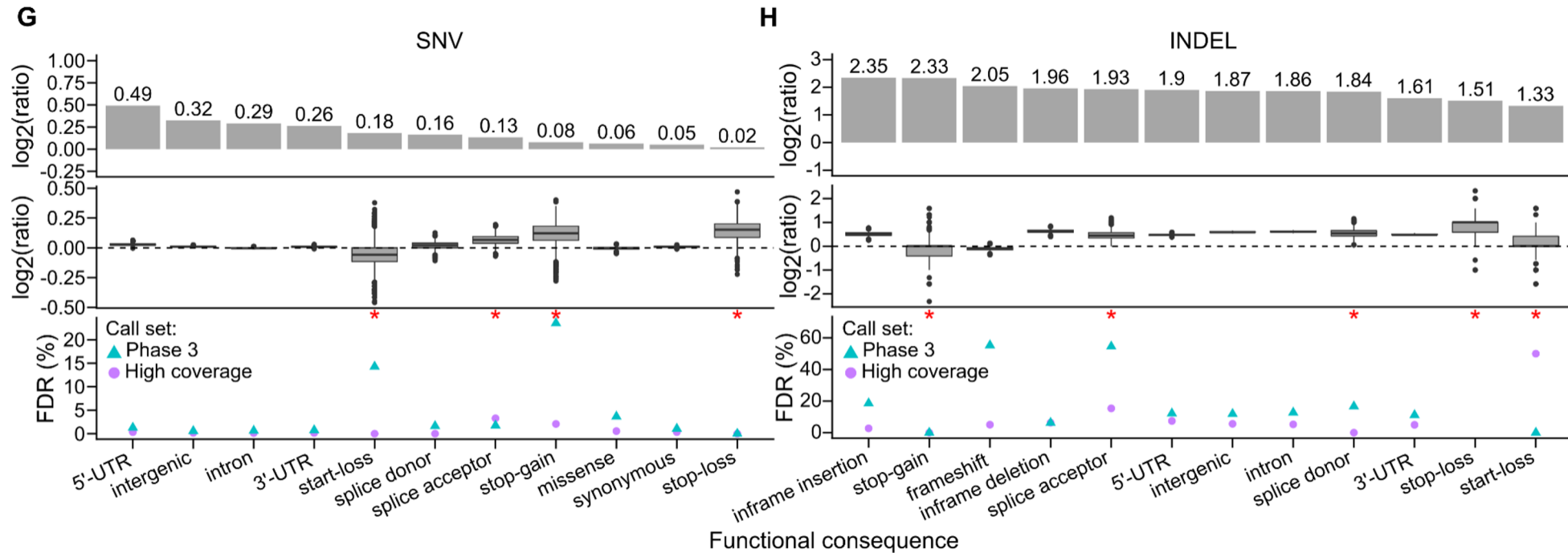
Variant type	Phase 3	High coverage
SNV	0.60	0.10
INDEL	12.40	1.10

VARIANT FUNCTION PREDICTION



- **Cohort-level total:**
 - 605,896 missense mutations,
 - 384,451 synonymous mutations,
 - 36,520 predicted loss of function variants (pLOF), defined as stop gained (n=12,181), frameshift (n=10,850), and splice mutations (n=13,489).
- **Genome-level average (MAF < 1%):**
 - 754 missense,
 - 569 synonymous,
 - 43 pLOFs (11 stop-gained, 14 frameshift, and 18 splice mutations).

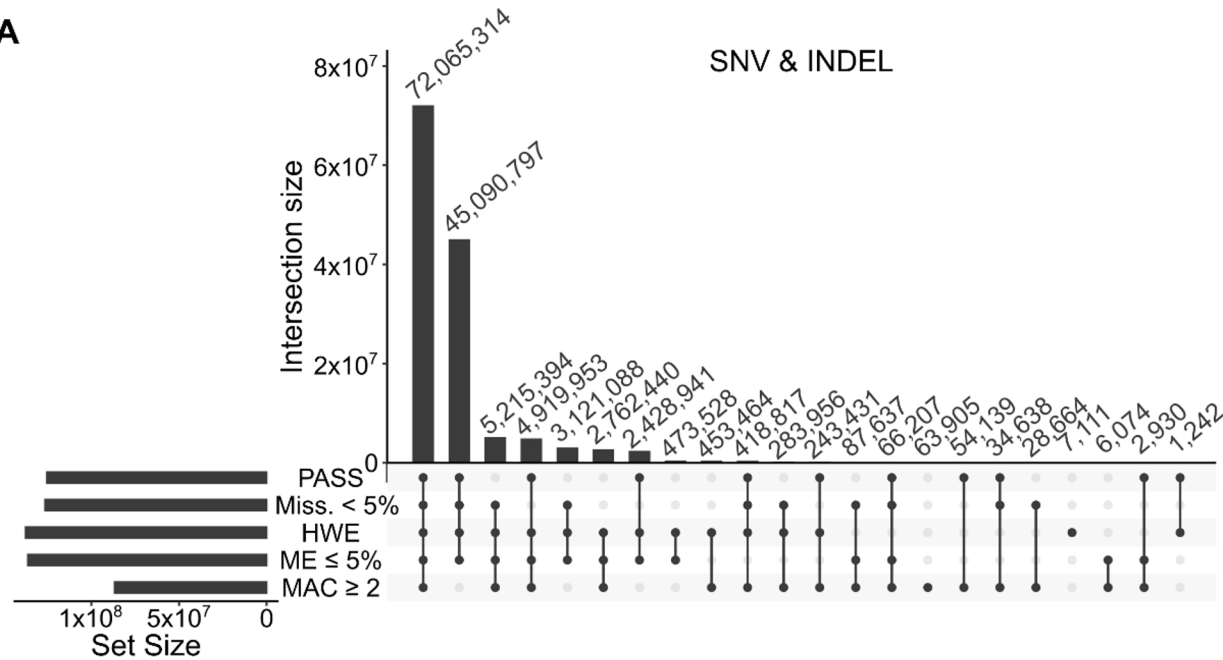
COMPARISON TO 1KG PHASE 3



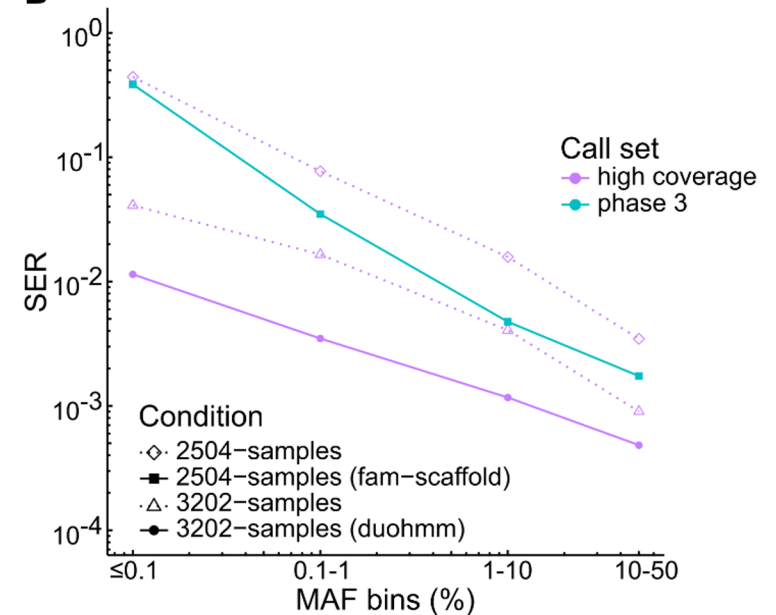
- **Cohort-level:**
 - SNVs: 1.01-1.41-fold increase in high coverage vs. phase 3.
 - INDELS: 2.52- and 13.48-fold increase in high coverage vs. phase 3.
- **Genome-level:**
 - SNVs: most categories show no significant difference, except for stop-gained (9% increase), stop-lost (11% increase), and start-lost (3% decrease).
 - INDELS: most categories show ~9-55% increase on average in the high coverage vs. phase 3, except for stop-gained and frameshift (3 and 7% decrease on average, respectively).

HAPLOTYPE PHASING OF SNV/INDEL

A



B



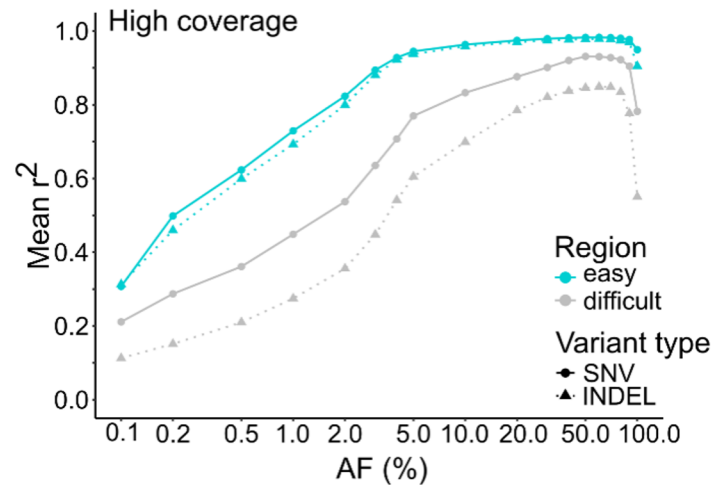
- Filtering criteria: VQSR PASS, missingness <5%, HWE PASS, ME ≤5%, MAC ≥2.
- Phasing performed using statistical phasing with pedigree-based correction (SHAPEIT2-duohmm) across autosomes (chrX was phased using Eagle2).

Delaneau, O. et al. Nat. Methods 9, 179–181 (2011);
 O’Connell, J. et al. PLoS Genet. 10, e1004234 (2014);
 Loh, P.-R., et al. Nat. Genet. 48, 1443–1448 (2016).

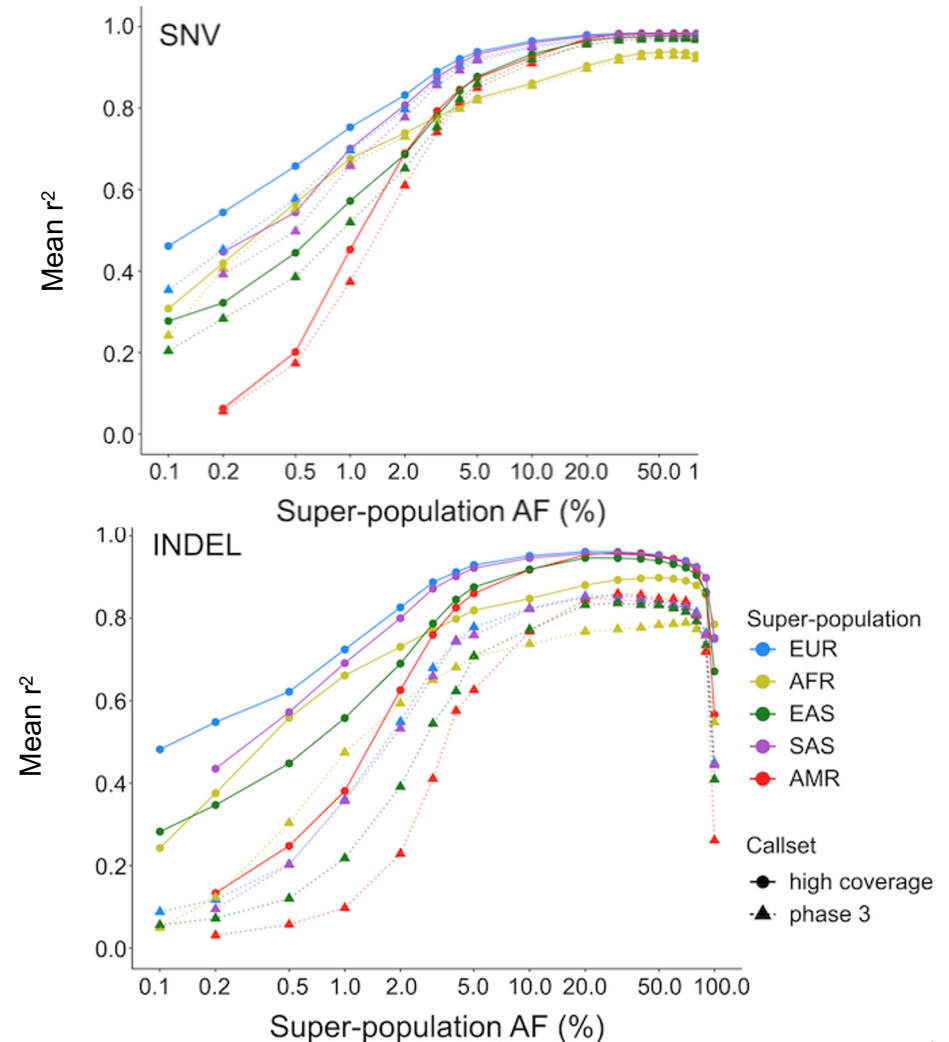
IMPUTATION PERFORMANCE

- Imputed a set of 279 diverse samples from the Simons Genome Diversity Project (SGDP) using IMPUTE2 software.
- Evaluated the accuracy of imputed genotypes by computing the squared correlation (r^2) between imputed allele dosages and dosages from WGS data across 110 samples, 22 from each of the five super-populations.

Performance of the high coverage panel stratified by variant type and genomic region:



Comparison against phase 3 (shared sites):

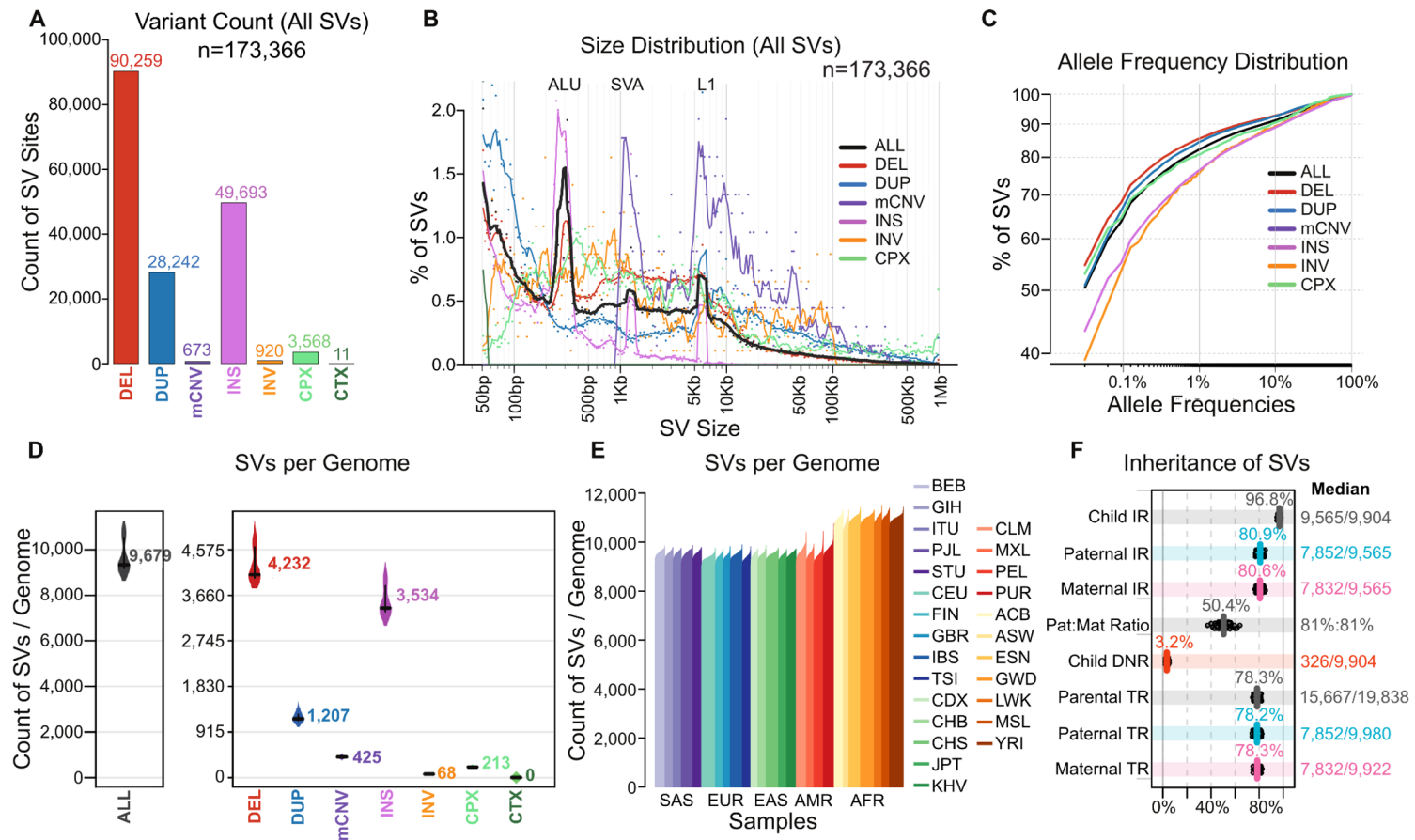


Howie, B.N. *et al.* PLoS Genet. 5, e1000529 (2009).

INTEGRATED STRUCTURAL VARIANT CALLS

SV callset integrated from GATK-SV, SVTools and Absinthe:

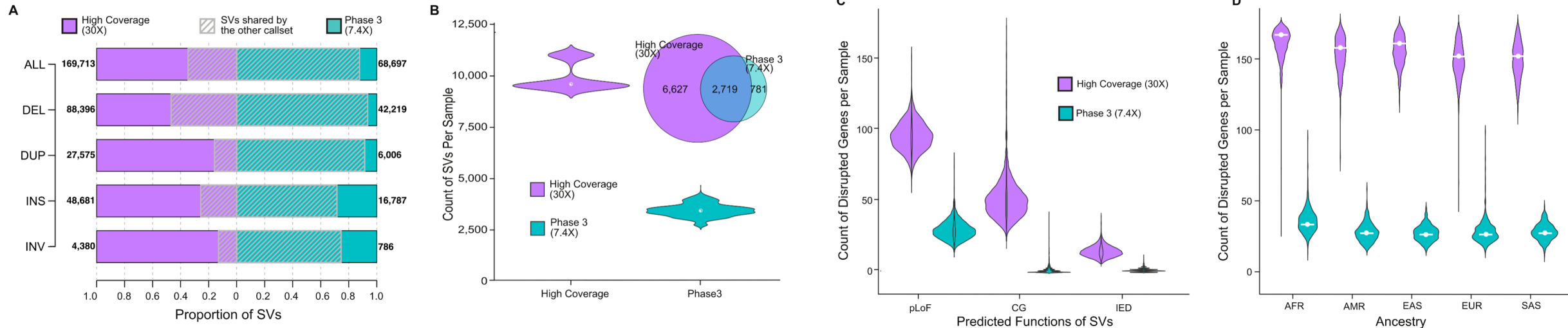
- A total of 173,366 SV sites across 3,202 samples in the high coverage callset.
- An average of 9,679 SVs per genome.
- More SVs are observed in African population.



INCREASED SV YIELD COMPARED TO PHASE 3

Increased sensitivity is observed in the SV callset from high-coverage (~35X) sequences than the 1KGP phase 3 callset (~7.4X):

- Over two times more SV sites are detected from the high-coverage sequences than 1kGP phase 3 (169,713 vs. 68,697).
- Increased sensitivity is also reflected in the SV count per sample.
- Most significant increase in sensitivity is reflected in small SVs < 250bp.
- More genes are altered by SVs in the new callset than 1kGP phase 3.
- More genes are altered in AFR population than others.



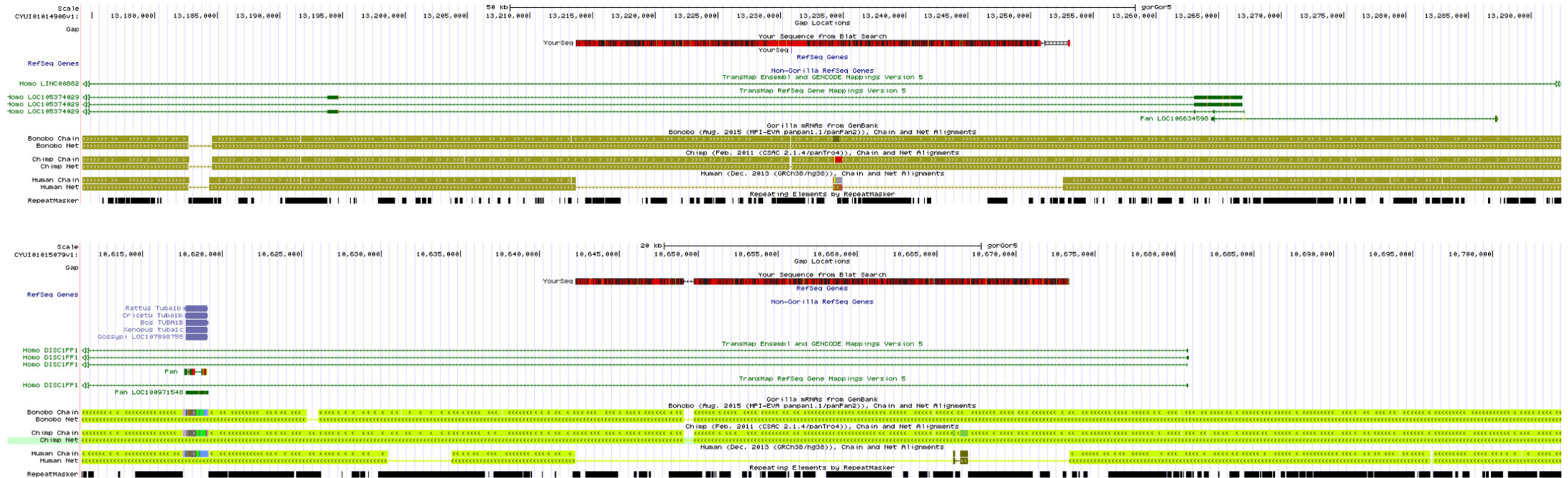
CONCLUSIONS

- We called **>111 million SNVs & >14 million INDELS** across the 3,202 samples with FDR of 0.3% and 1%, respectively.
- Relative to the phase 3 callset, we called **6% more SNVs and 48% more INDELS per genome**.
- The vast majority of the **new SNVs are in the rare MAF spectrum** ($AC \leq 2$).
- We observed **gains in INDEL counts across the entire MAF spectrum**, with gains in the rare end of the spectrum being the most pronounced.
- The phased high coverage SNV/INDEL panel exhibits **an order of magnitude higher phasing accuracy** as compared to the phase 3 dataset across the entire MAF spectrum.
- Improvements in small variant calling, coupled with higher phasing accuracy of the high coverage panel, translated into **significantly better imputation accuracy**, especially for INDELS, across all of the 1kGP super-populations.
- We called **173,366 SV** sites across 3,202 samples with $FDR \leq 3.2\%$
- More genes are altered by SVs in the high coverage call set as compared to phase 3

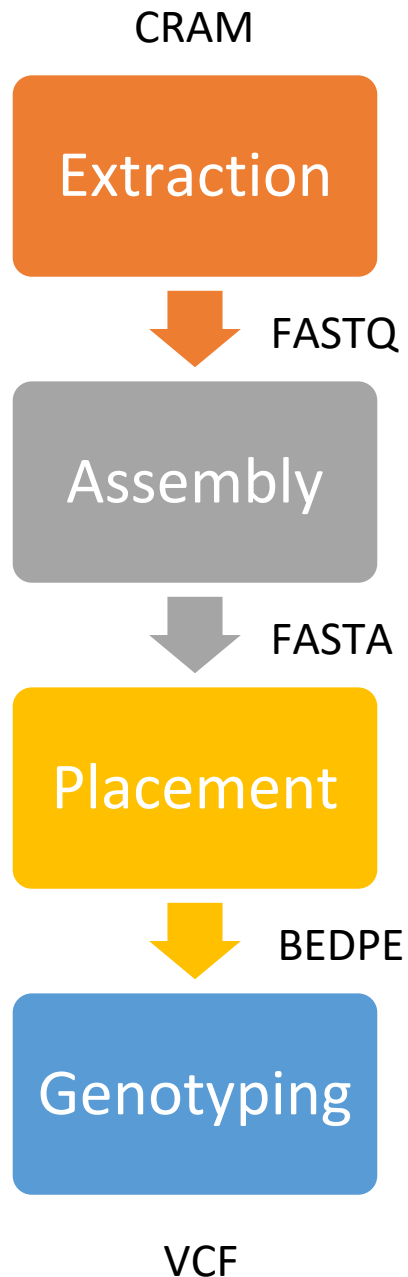
ABSINTHE INSERTION CALLING

- Calling “insertions” from short reads has traditionally been difficult
- Absinthe identifies reads that don’t map or mismap and assembles them
- The resulting contigs can then be placed back on the reference

EXAMPLES OF ASSEMBLED INSERTIONS



ABSINTHE PIPELINE



- Not properly mapped read-pairs
- phiX removal, adapter clipping, low quality base trimming

- *de novo*
- ABySS v2.0.2
- $k = 77$

- *ab initio*:
 - Flank maximal best hit pairs to GRCh38
 - Alignment with gap excision
- LiftOver:
 - Hominid alignment and reference-based scaffolding
 - Coordinate transposition to GRCh38
 - Alignment with gap excision

- Merging
- Paragraph v2.4b

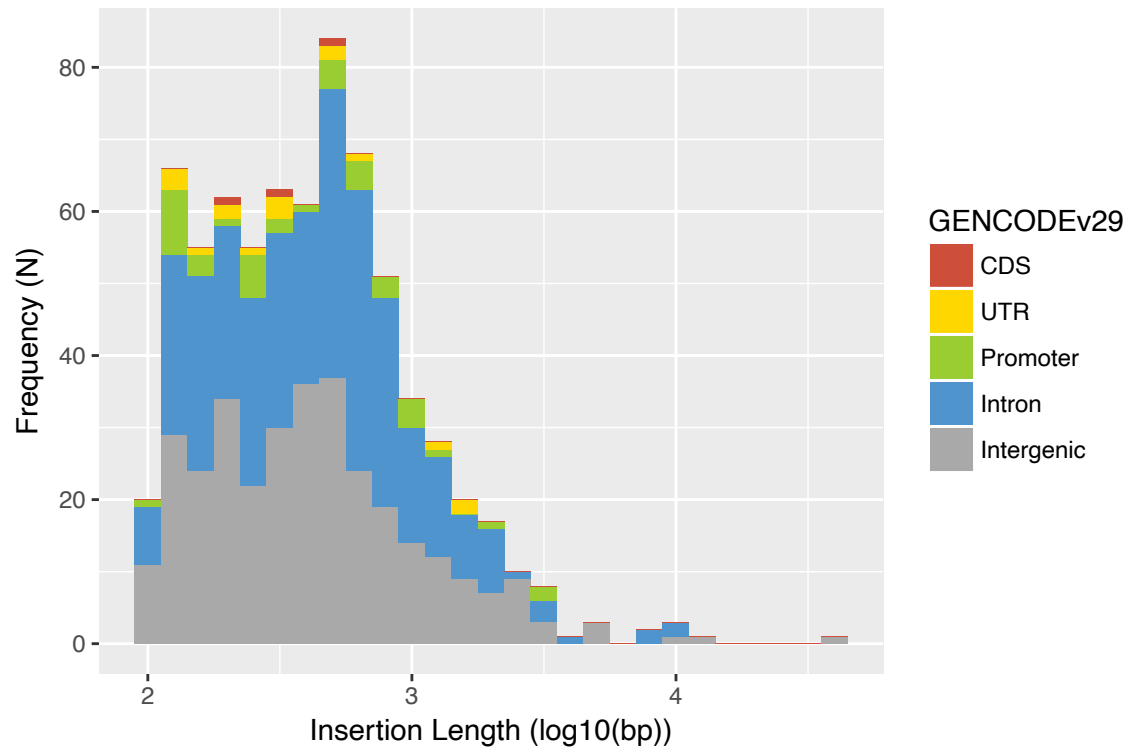
RESULTS FROM TOPMED

- 53,831 genomes (reads aligned to GRCh38)
- Genotype using Paragraph, rather than simply determining presence/absence (insertions only)

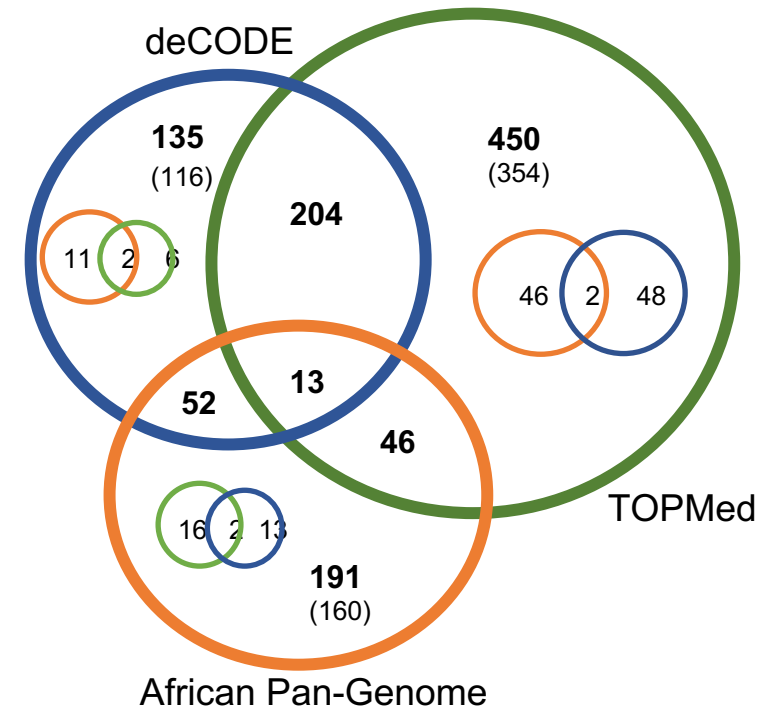
		53k GRCh38
Insertions	N	713
	(bp)	514,642
Breakends	(N)	304
	(bp)	186,343

RESULTS FROM TOPMED

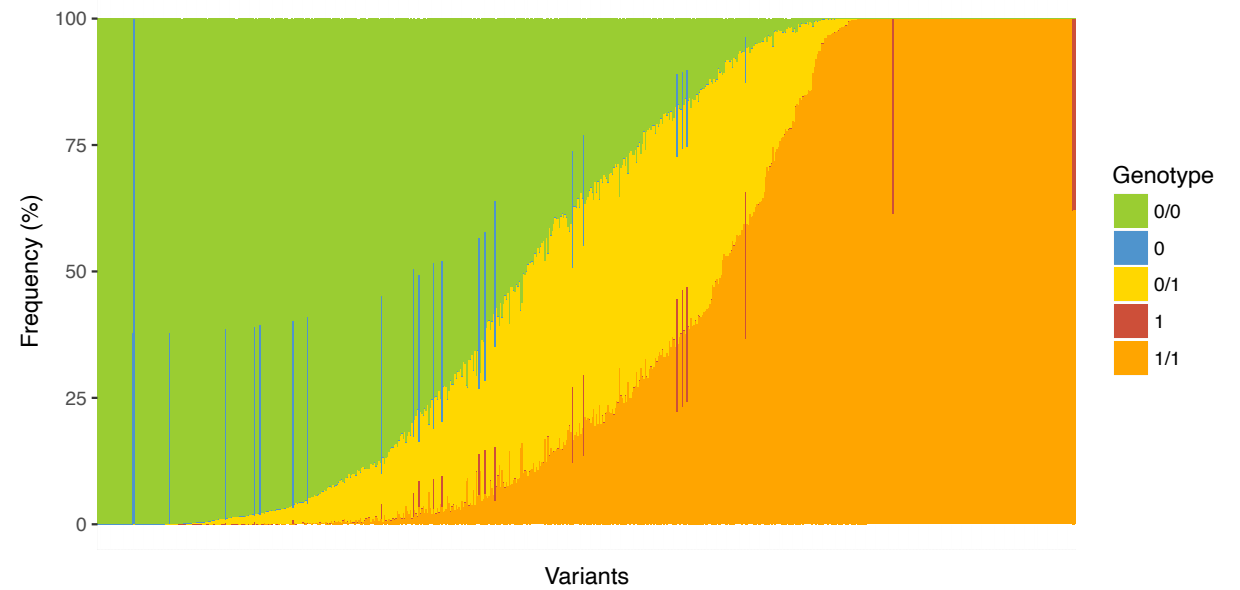
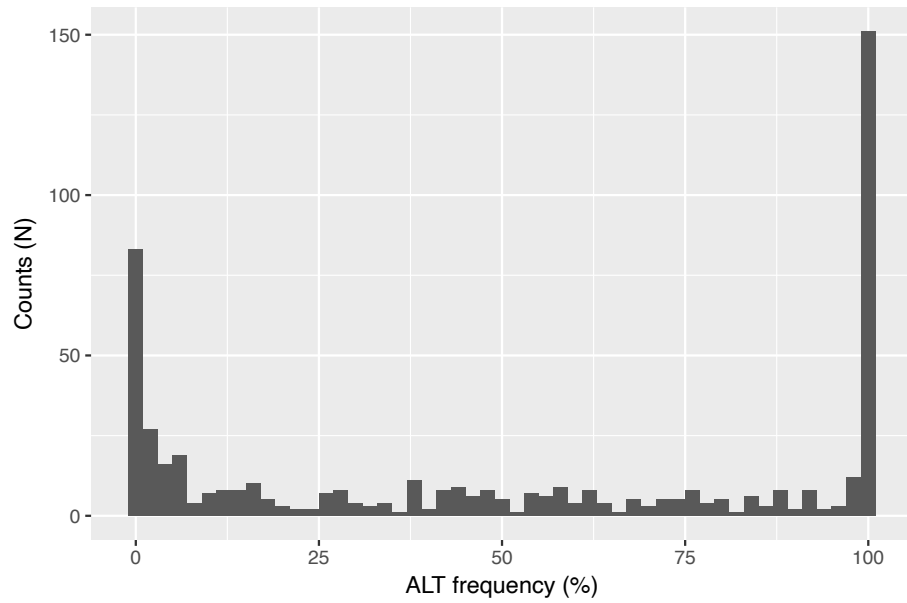
Length distribution



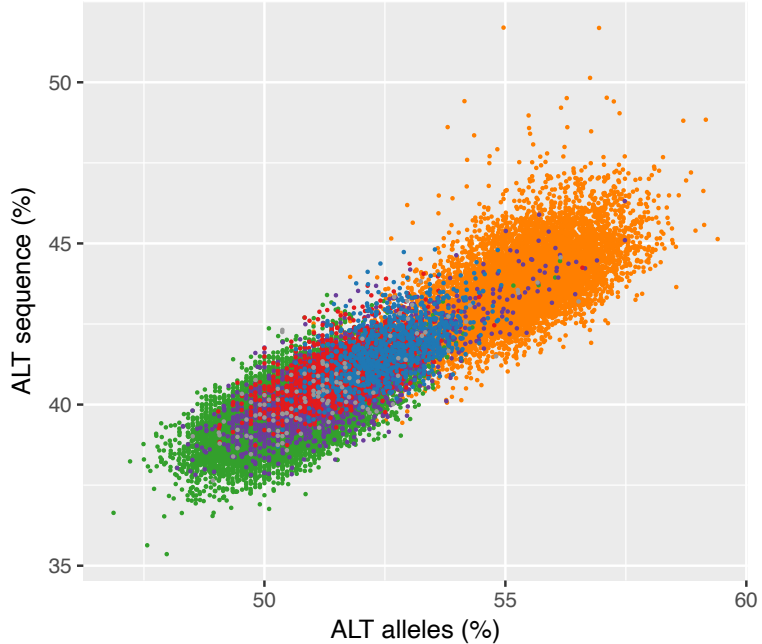
Positional concordance with insertions identified using short-read data from two previous studies



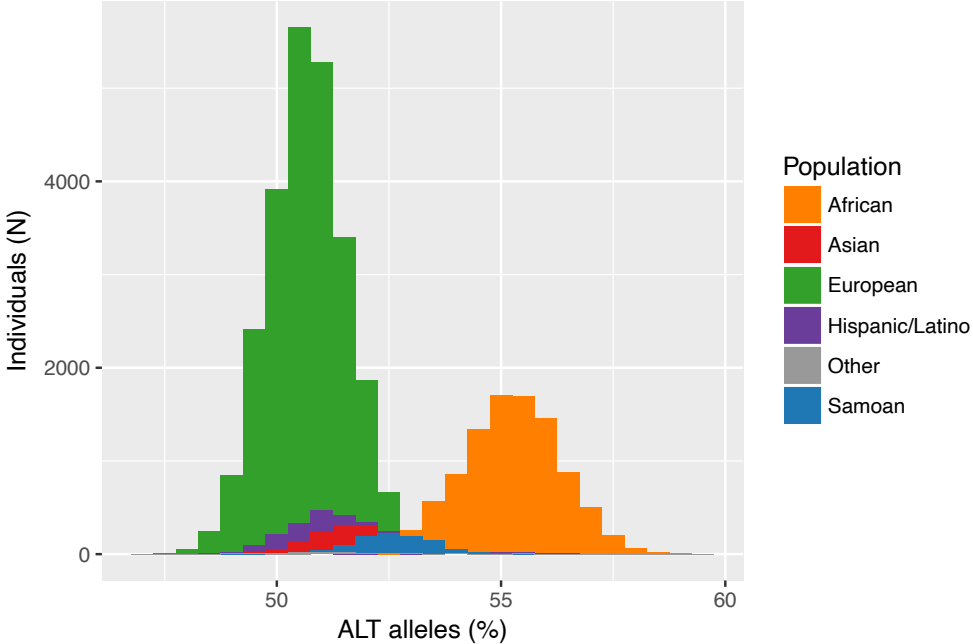
ALLELE AND GENOTYPE FREQUENCY



ALT ALLELE DISTRIBUTION BY ANCESTRY

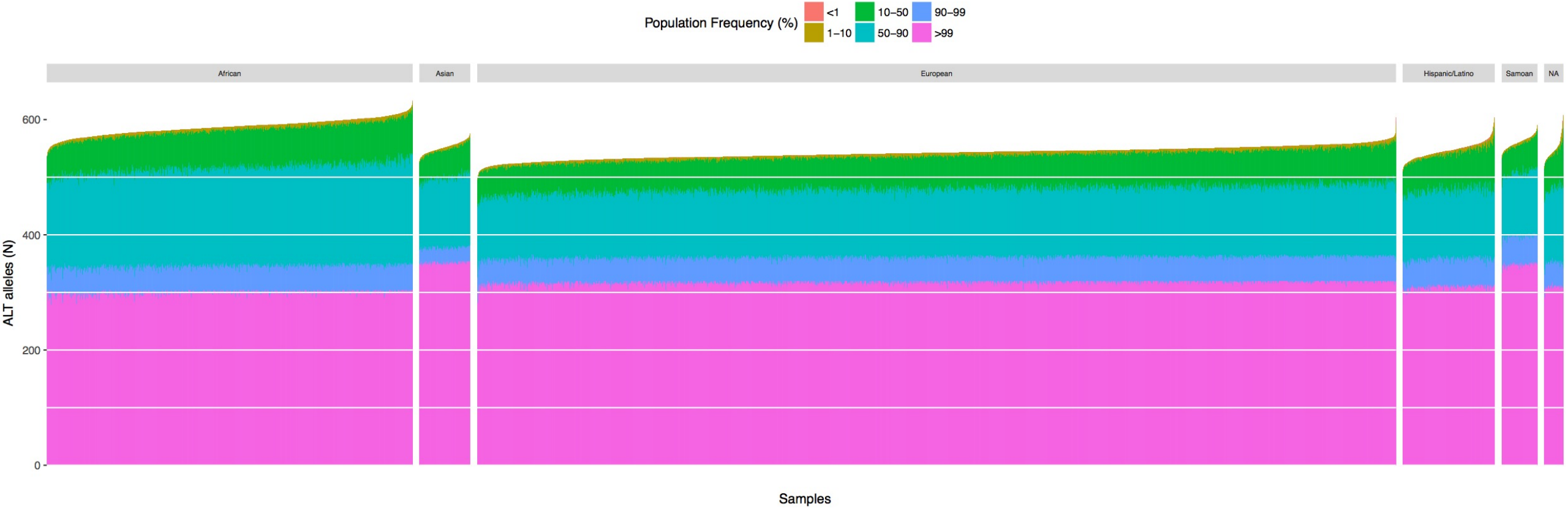


- Population
- African
 - Asian
 - European
 - Hispanic/Latino
 - Other
 - Samoan



- Population
- African
 - Asian
 - European
 - Hispanic/Latino
 - Other
 - Samoan

ALT FREQUENCY WITH POPULATIONS

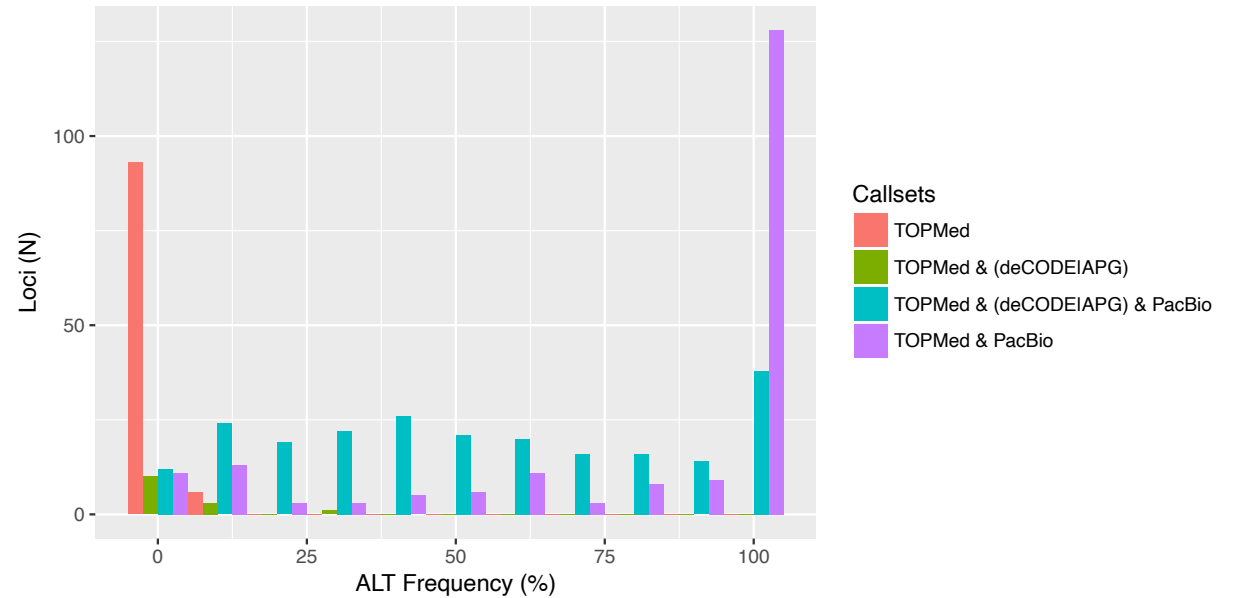


Higher fraction of >99% alleles in Asians and Samoans

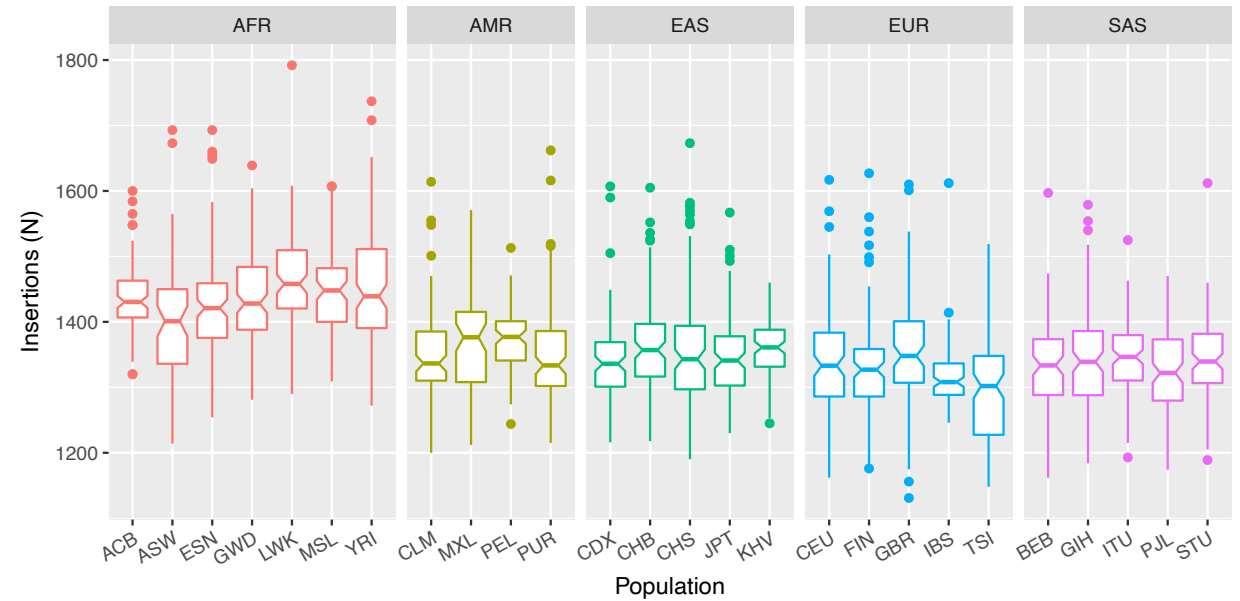
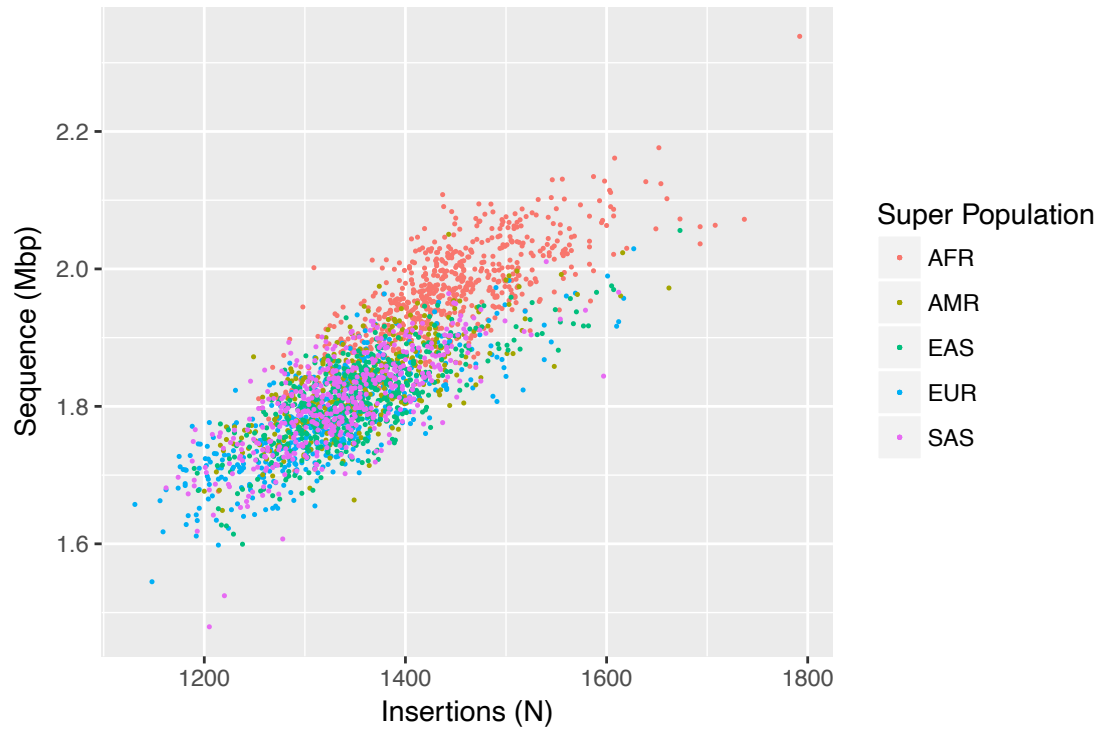
Excess ALT alleles observed in individuals of African ancestry fall in the frequency range of 10-90%

VALIDATION WITH LONG READS

ALT allele frequency by overlap with deCODE, APG and PacBio*
79% overlap PacBio insertions

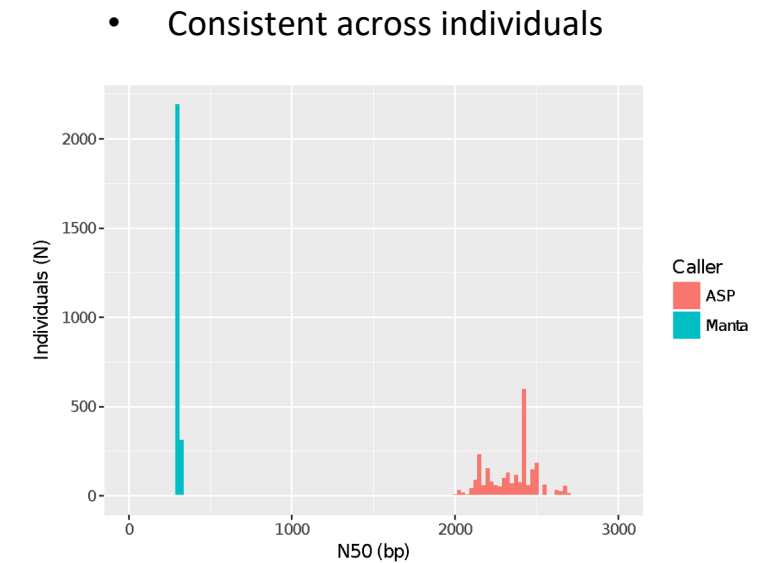
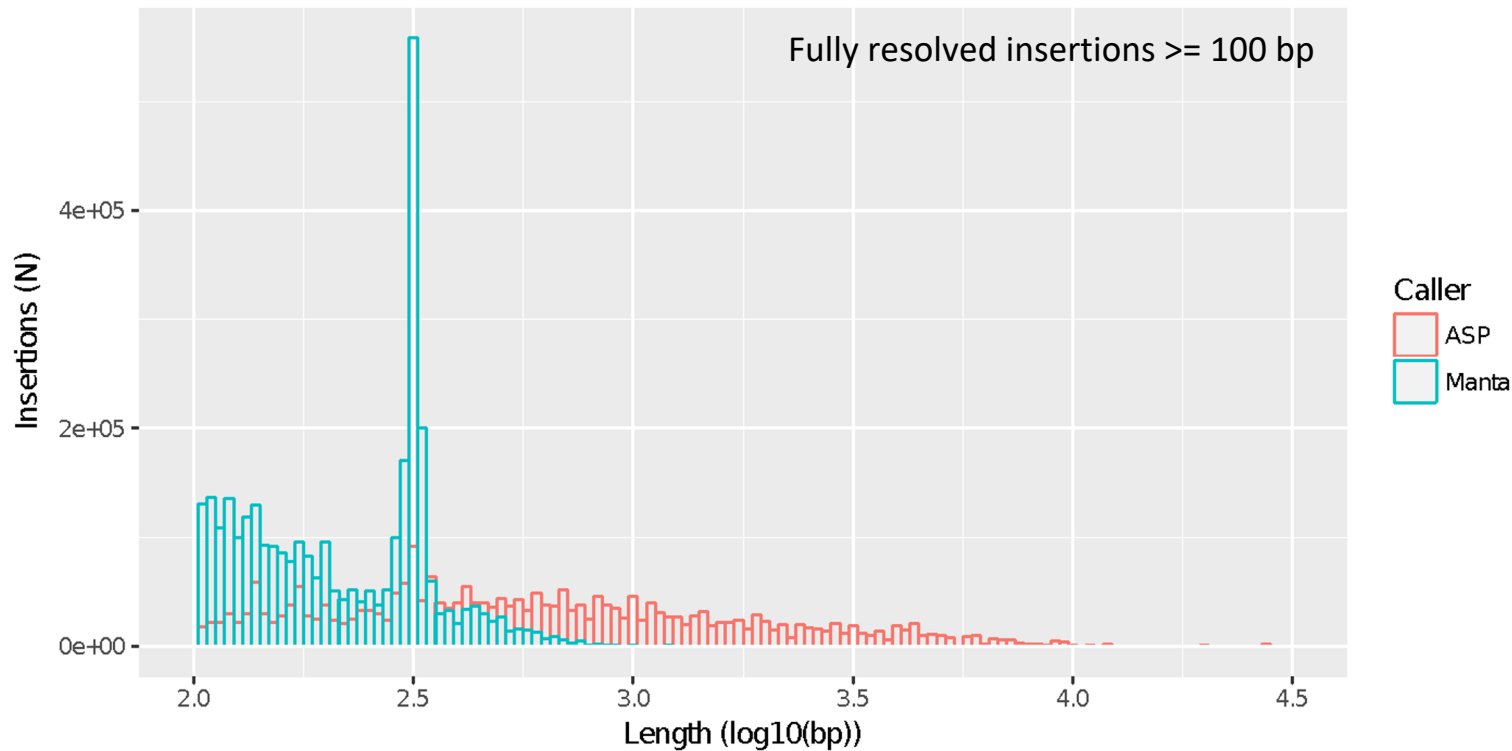


CALLING IN 1000 GENOMES



- 1,300-1,500 insertions per individual (1.6 – 2.2 Mbp)
- Larger number of insertions in individuals from African populations

INSERTION LENGTH DISTRIBUTION

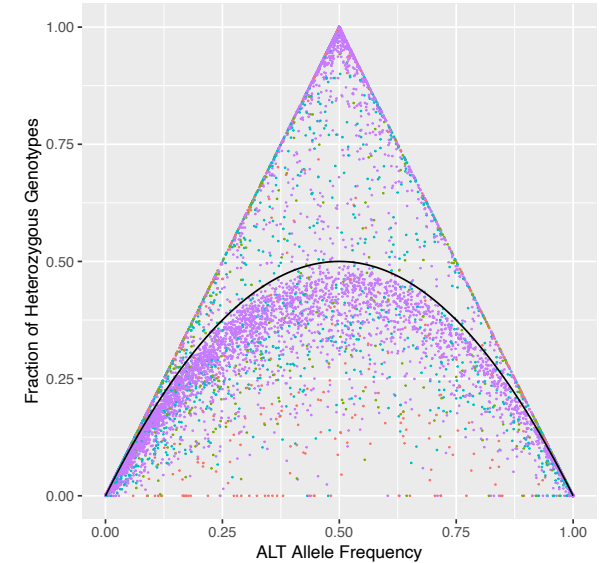
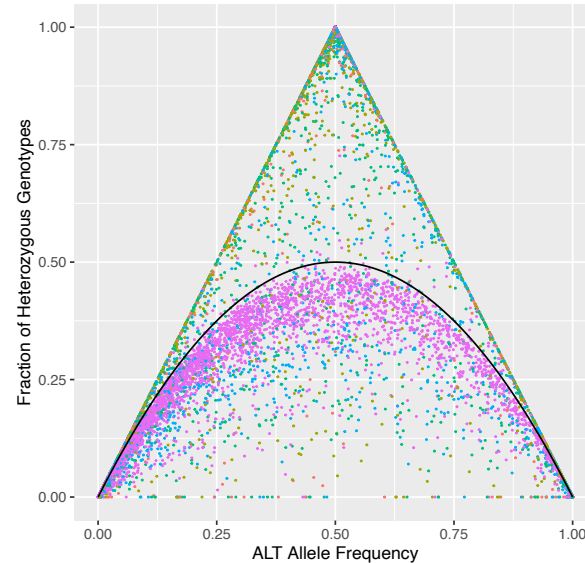


- Absinthe calls are a good complement to Manta's as they extend well into the range of 1Kb – 10Kbp
- Several fully resolved insertions are longer than 10Kbp

1000 GENOMES MERGED CALLSET

Merging:

- MSA-based
- Input:
 - 3,583,674 per-sample calls
 - Self-genotyped (1, 0/1, 1/1)
 - 657,757 distinct
 - 12,222 loci
- Output:
 - **12,704** insertions



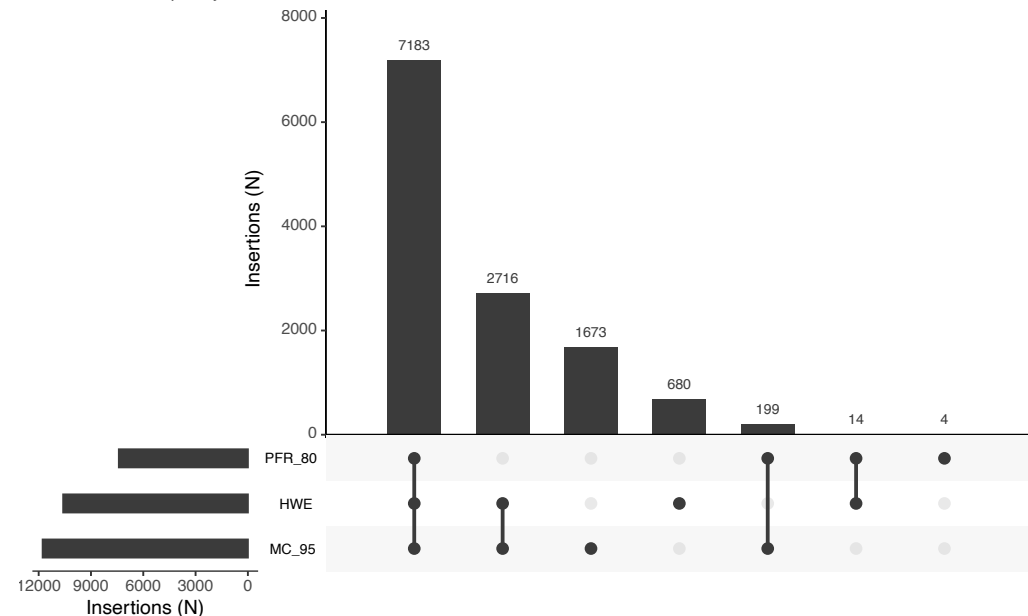
Genotyping:

- Paragraph (Chen et al, 2019)

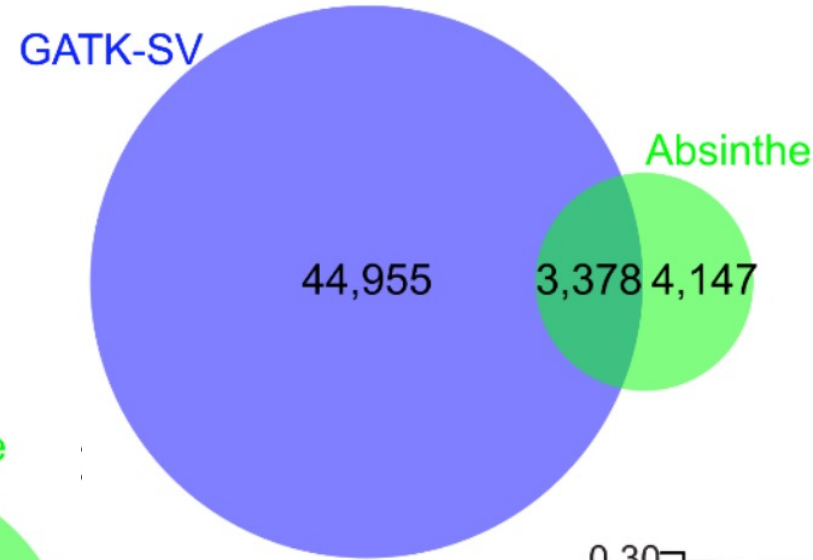
Filters:

- Super population PASS-filter rate [all ≥ 0.8]
- Super population HWE [any $> 10^{-6}$]
- Mendelian Consistency based on 602 trios [≥ 0.95]
- Output:

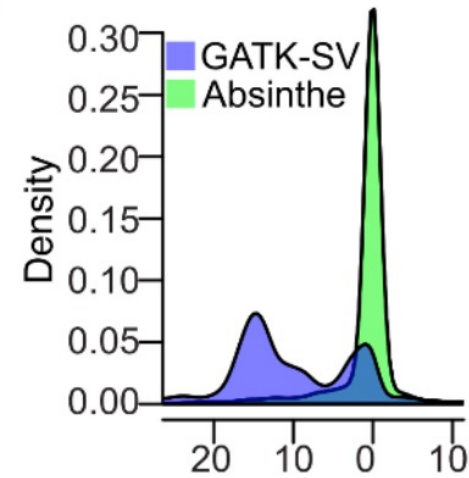
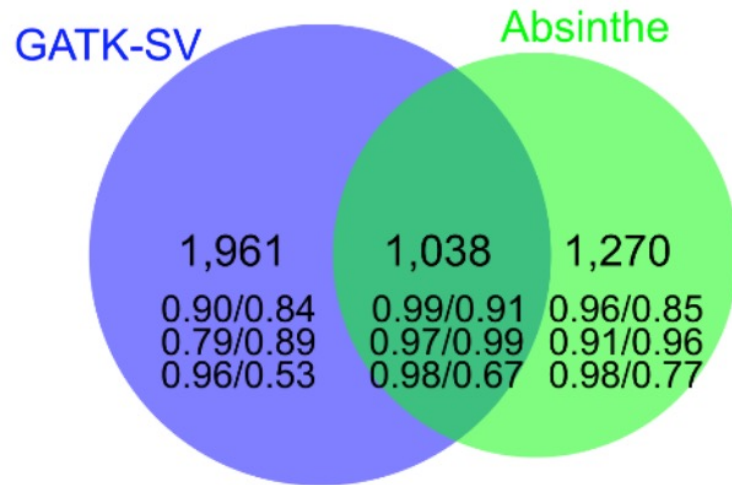
7,183 HQ genotyped insertions



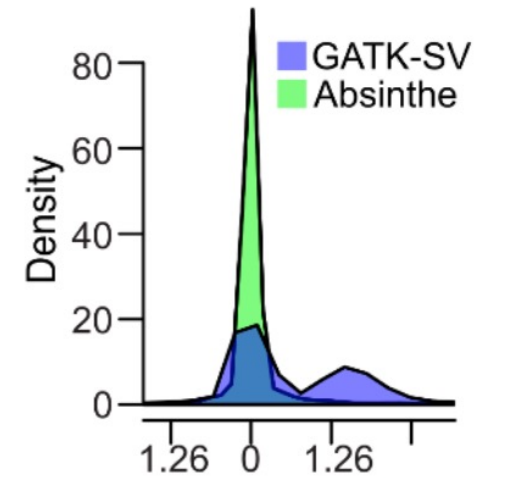
COMPARISON TO GATK-SV CALLS



Per sample averages



Breakpoint Distance(bp)
(PacBio - srWGS coordinates)



Ratio of Insert Length
(PacBio/srWGS length)

N insertions

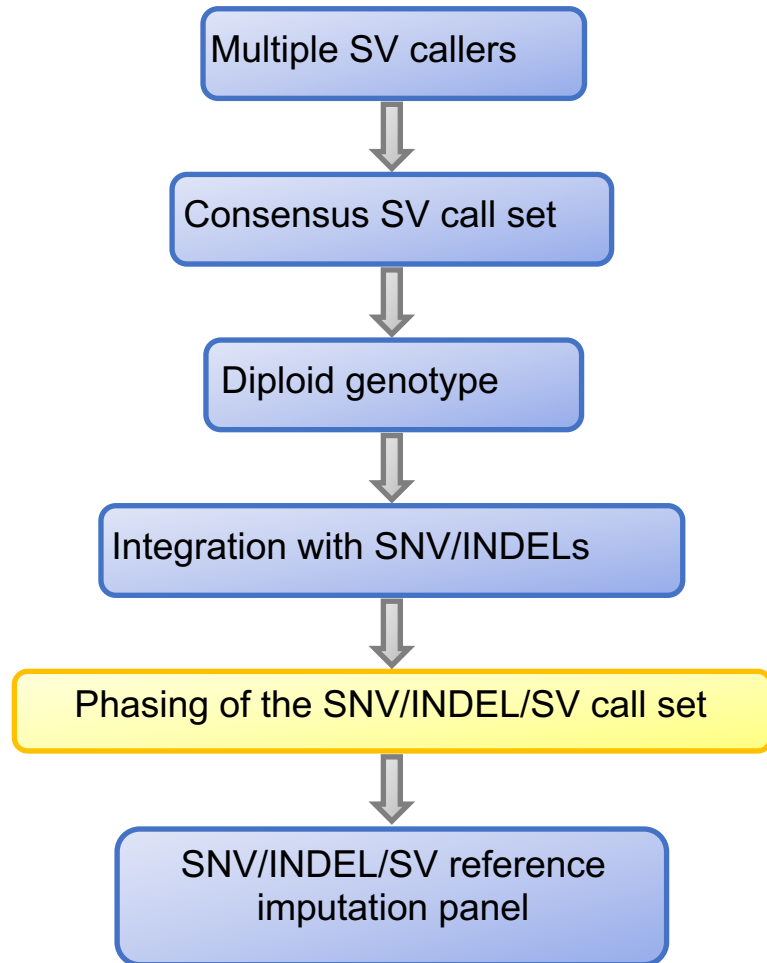
validated by VaPoR / assessable by VaPoR
in PacBio SVs (Ebert et al. 2021) / in PacBio SVs (Chaisson et al. 2019)

30 transmission rate / rate of bi-parentally inherited

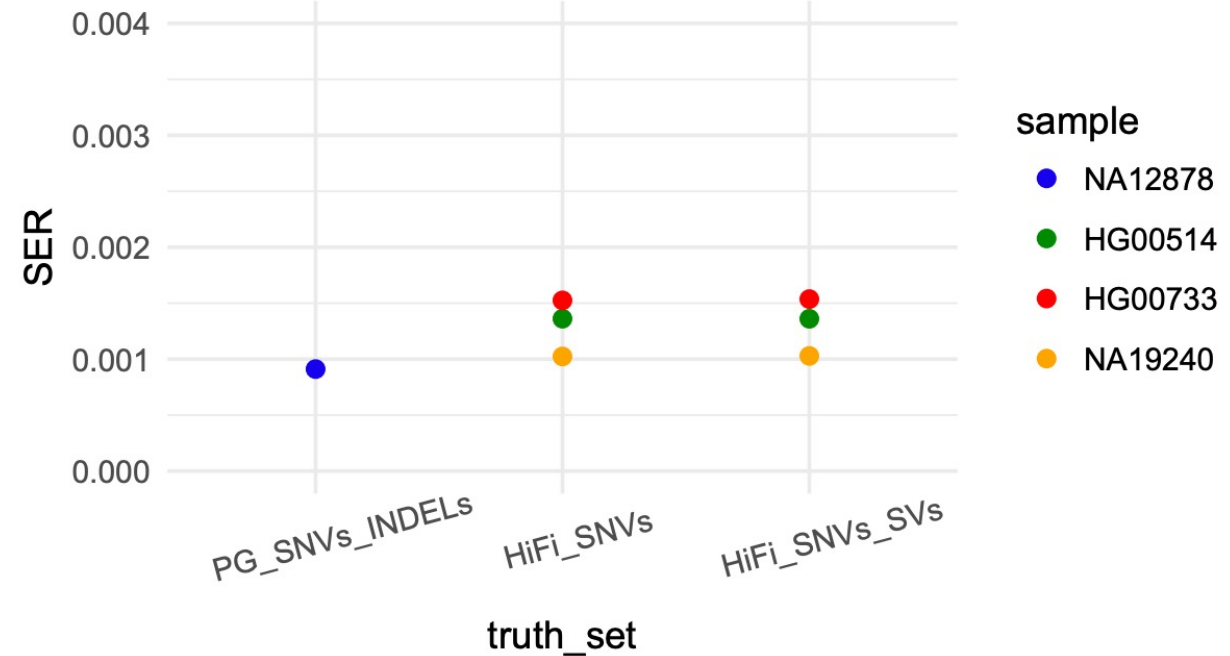
STRUCTURAL VARIANT IMPUTATION

- Imputation panels for SNVs and small indels have greatly improved our power to run associations for traits
- SVs are harder to call from sparse data than SNVs
- SVs have typically not been included on imputation panels
- Association of SVs to phenotype has typically been done case-by-case leveraging associations discovered from linked SNVs
- We would like to be able to directly associated SVs with phenotype

PHASING ACCURACY OF SVS



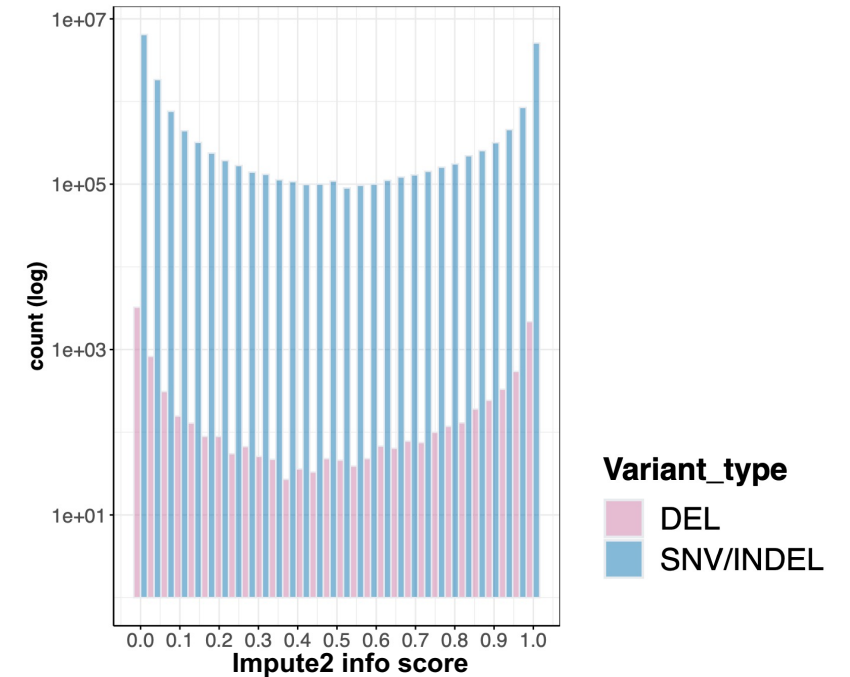
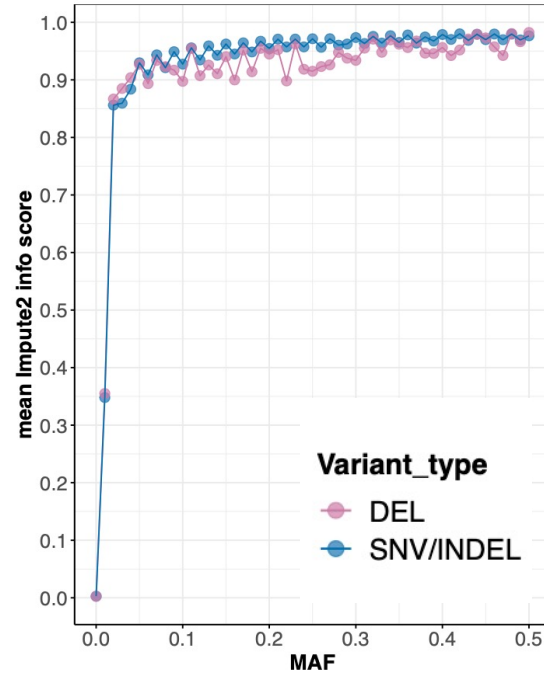
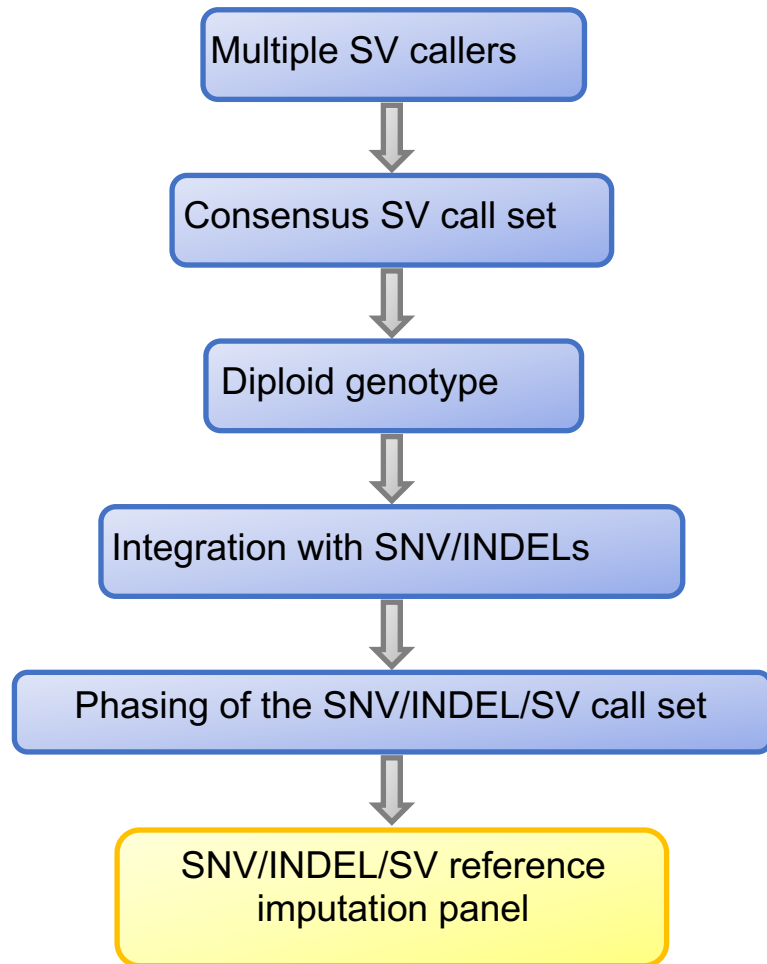
Using PacBio-HiFi haplotype-resolved SNV/SV call sets* to assess accuracy of SV phasing.



Δ SER for DELs = 0.012

* HGsvc, pre-publication.

EVALUATION OF IMPUTATION PERFORMANCE



SV GT concordance evaluation against the GIAB truth set*:

Imputed sample	Info score threshold	Sensitivity	Precision
HG002	≥ 0.5	98.12%	95.55%

* Zook JM *et al.* *Sci data*, 3:160025 (2016)

STRUCTURAL VARIATION IN ALZHEIMER'S

- Create a harmonized, publicly available SV call set from a 972 familial and 39,000 unrelated LOAD case-control ADSP dataset of multi-ethnic ancestry.
- Augment ADSP SV call-set in by using SVs derived from long-read sequencing data from 200 AD patients.
- Increase sample size by imputing SVs in individuals without WGS data from the AD Genetics Consortium (ADGC).
- Identify common and rare SVs associated with LOAD and related endophenotypes.

ACKNOWLEDGEMENTS

NYGC:

Marta Byrska-Bishop

André Corvelo

Uday Evani

Anna Basile

Wayne Clarke

Rajeeva Musunuri

Giuseppe Narzisi

Kshithija Nagulapalli

Alexi Runnels

Lara Winterkorn

Soren Germer

HGSVC:

Michael Talkowski

Xuefang Zhao

Harrison Brand

Ira Hall

Haley Abel

Allison Regier

Evan Eichler

Peter Audano

Susan Fairley

Ernesto Lowy-Gallego

Paul Flicek

AD SV Grant:

Badri Vardarajan

TOPMed Consortium:

Daniel Taliun

Gonçalo Abecasis

Funding: NHGRI, NIA