Introduction to Genome STRiP for discovery and genotyping of deletions

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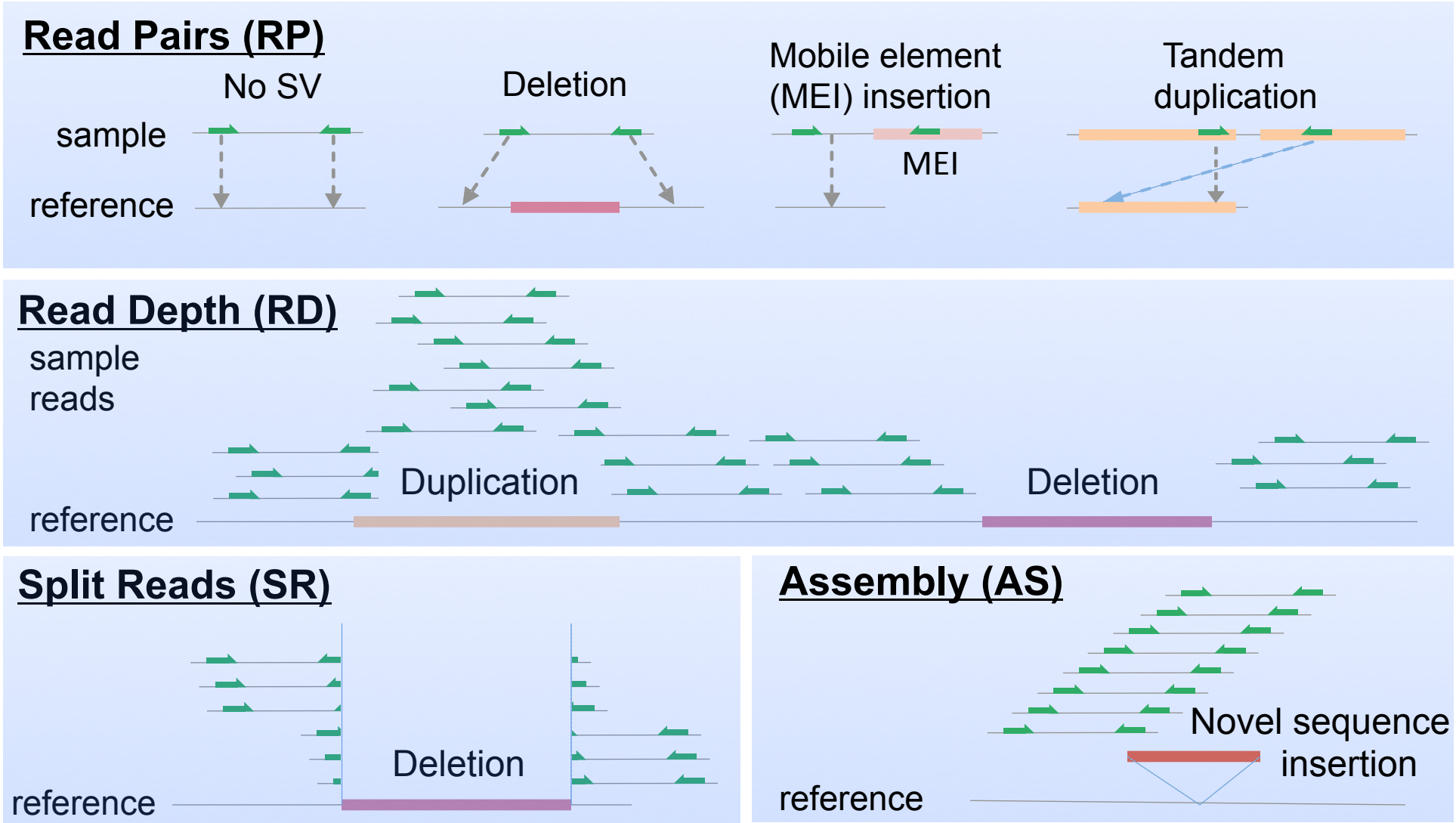
Outline

- Overview of structural variation calling
- Genome STRiP processing pipelines
- Techniques for quality control
- Genotyping novel sites in 1000 Genomes data
- Software and support

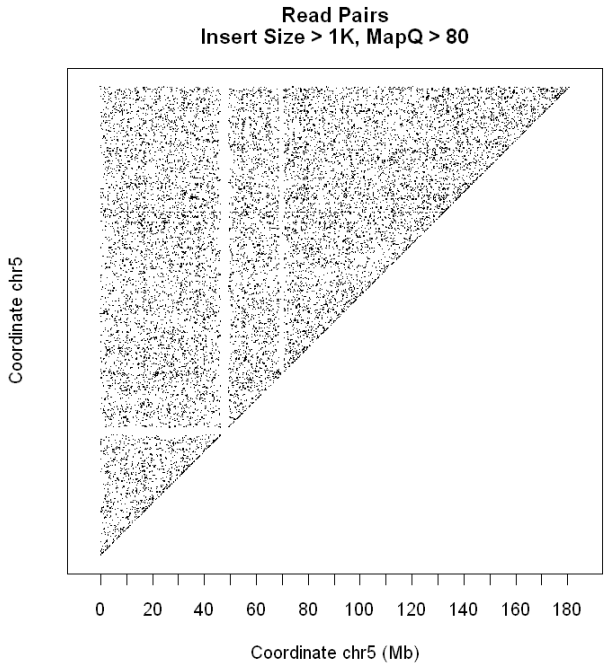
Ascertaining large variants from small reads



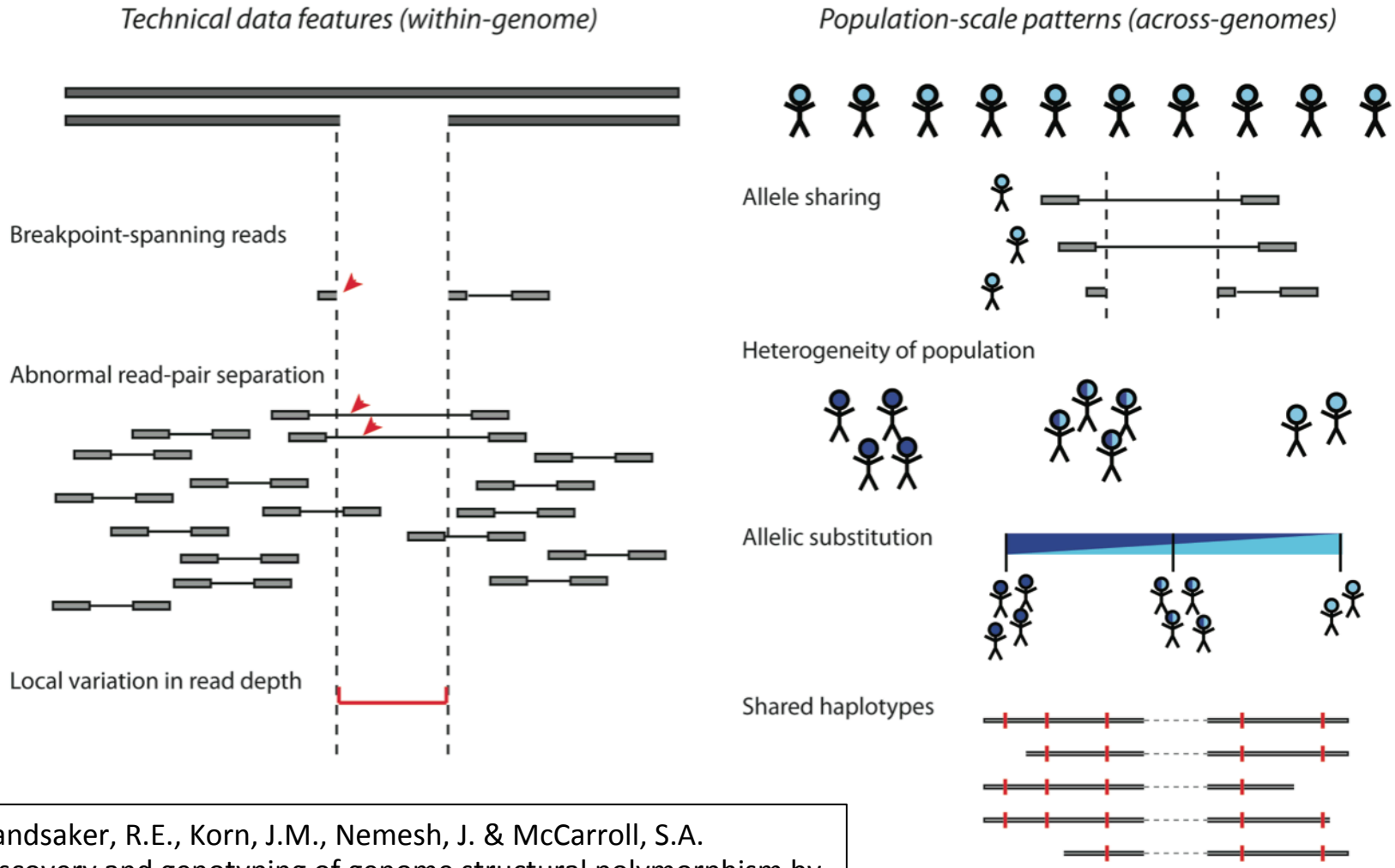
Graphics courtesy of Jan Korbel, Ryan Mills

Why is structural variation calling challenging?

- Artifacts abound
  - Millions of chimeric molecules generated during library construction
  - Read depth varies across the genome and across libraries
  - Alignment algorithms are misled by the genome's repeats
- Low-coverage sequencing
  - Data is not definitive in each genome
  - False discoveries can accumulate across genomes
- Deeply sequenced genomes
  - Increased depth can help, but methodology is more important



Discovery and genotyping are enhanced by combining technical and population-level features of a data set



Handsaker, R.E., Korn, J.M., Nemesh, J. & McCarroll, S.A. Discovery and genotyping of genome structural polymorphism by sequencing on a population scale. *Nat Genet* **43**, 269-76 (2011)

Genome STRucture in Populations

What is it?

Methods for discovering and genotyping large deletions from sequencing data

Our Focus

- Whole genome sequencing (shallow or deep)
- Using populations to inform calls in individuals
- Germline/somatic DNA (not tumor/normal)

Genome STRiP in 1000 Genomes Project

Discovery specificity

Consistently low false discovery rate (1.5% - 4.2%)

Lowest FDR in 1000G pilot and in phase 1

Discovery sensitivity

Best overall sensitivity on low coverage sequencing data (Mills, 2011)

Contributed over 80% of phase 1 deletion call set

Genotyping accuracy

Genotyping algorithm of choice for pilot and phase 1

Genotyping accuracy, 1000 Genomes Phase 1

Genotyped Sites	Evaluation Data	# Sites Evaluated	HOMREF (Conrad)	HET (Conrad)	HOMALT (Conrad)	OVERALL
14,422	Conrad 2010 80% RO 248 samples	1,092	99.92%	99.01%	99.47%	99.82%

Other large projects using Genome STRiP

Genome of the Netherlands (GoNL) *University of Groningen*
250 whole genomes in trios at 12x coverage (9,000x)

GoT2D (Type 2 Diabetes) *Oxford University*
2800 cases/controls at 4x coverage (11,000x)

UK10K Cohorts Project *Sanger Institute*
2453 individuals (so far, 4000 planned) at 6x coverage (14,000x)

1000 Genomes Project Phase 3 *Broad Institute*
2535 individuals at 4x coverage (10,000x)

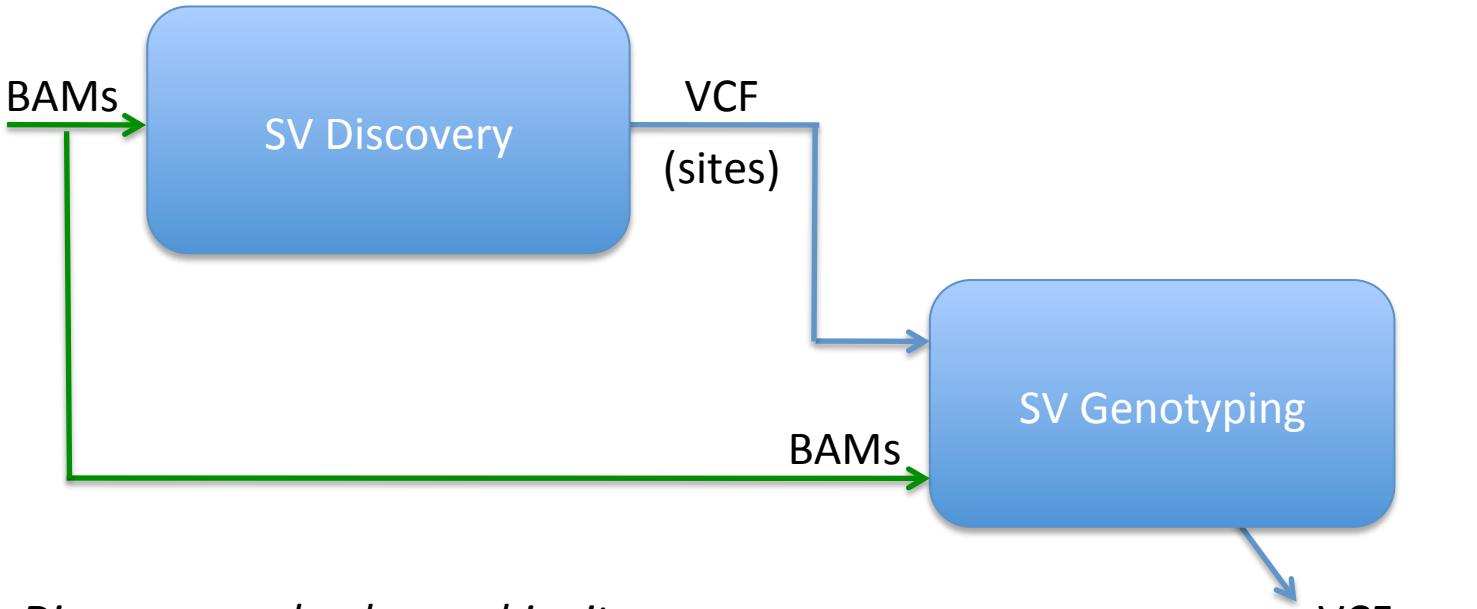
PROCESSING PIPELINE

Required inputs

Processing phases

- Preprocessing, Discovery, Genotyping

Discovery and genotyping are two distinct modules in Genome STRiP

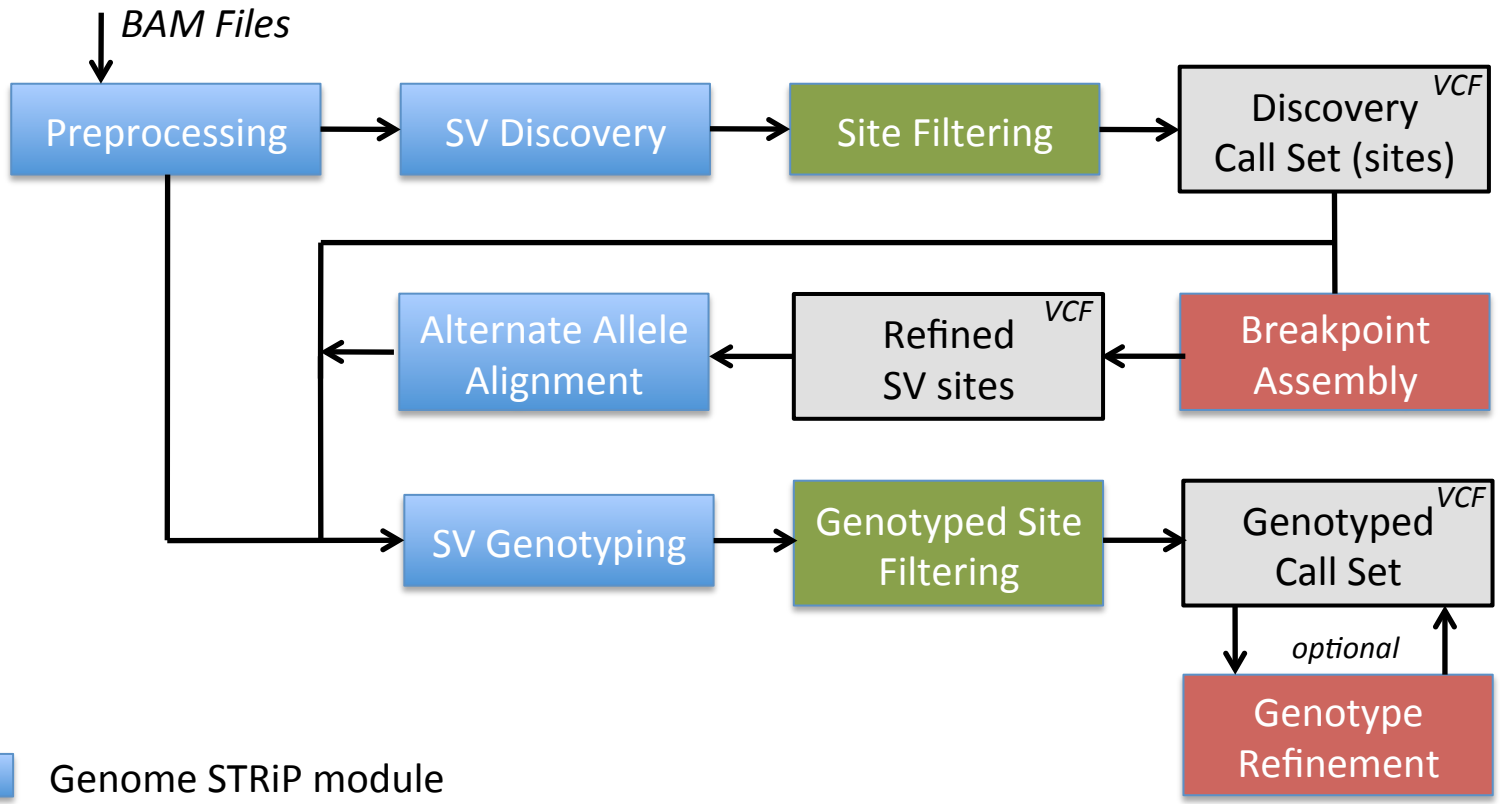


Discover novel polymorphic sites

Genotype known events in new data sets

Genotype call sets from multiple discovery methods

Detailed processing pipeline



Genome STRiP module

Requires user intervention, defaults or best-practices are available

Third party tool (e.g. TIGRA or velvet for assembly, BEAGLE for genotype refinement)

VCF data file

What inputs are needed to run Genome STRiP?

- "Analysis-ready" BAM files
  - Whole genome sequencing
  - Aligned, sorted, indexed, duplicates marked or removed
- Reference sequence
  - Indexed fasta file, must match *exactly* the reference used for alignment
- Alignability mask
  - Indicates which reference positions are uniquely alignable
  - Must be based on the same reference you are using
  - Commonly used masks are available for download
- CN2 mask
  - Flags regions unlikely to be copy-number polymorphic, used for estimating GC-bias
  - CN2 masks for common reference sequences are available for download
- Ploidy map
  - Required to process sex chromosomes
  - Indicates expected ploidy of positions on the reference, stratified by gender
- Gender map
  - Gender of each sample, required to call on sex chromosomes

Analysis-ready BAM files

- Reads aligned to reference sequence
- Sorted by coordinate and indexed
- MarkDuplicates is essential
- Indel realignment does not matter (with/without is ok)
- Key headers and tags must be present and consistent
  - Read group (RG tag), e.g. Illumina lane
  - Library (LB tag)
  - Sample (SM tag)
  - Platform (PL tag)
- GATK ReduceReads compression is not supported

Alignability Mask

- What is it?
  - An alignability mask indicates all sites on the reference that are uniquely alignable by a single, error-free read of length *k*
  - Generated by aligning *k*-mers centered on each base position back to the reference using bwa, test if *k*-mer aligns uniquely
  - Function of reference sequence and *k*
  - If you have multiple read lengths in your data, use the smallest as *k*
- Where do I get it?
  - Mask files are available for download (hg19, 1000G b36/b37)
    ftp://ftp.broadinstitute.org/pub/svtoolkit/svmasks/
- Building your own
  - See documentation for ComputeGenomeMask
  - Can be parallelized for scalability
  - Mask format is currently an indexed fasta file, but subject to change

CN2 Mask

- What is it?
  - Indicates sites on the reference unlikely to be copy-number polymorphic in most individuals
  - We use this when measuring GC-bias in read depth, following an approach similar to that described in Sudmant *et al.*, Science 2010
  - Excludes chrX, chrY, chrM, all unplaced contigs
  - Excludes regions with 200bp of UCSC-annotated repeats, segmental duplications or copy number variants from DGV
- Where do I get it?
  - Versions available for 1000G b36, b37

    ftp://ftp.broadinstitute.org/pub/svtoolkit/cn2masks/
  - Format is indexed fasta file, with 0 or 1 for each position
  - Bed files are also provided for convenience (or viewing), but not used during processing
- Building your own
  - No tools provided to build your own
  - Liftover might be an option for human sequence

Ploidy Map File

- What is it?
  - Simple text file of expected ploidy on reference sequence by gender
  - Used in newer versions of Genome STRiP
  - Not strictly necessary when processing autosome only (code generally assumes ploidy 2 if missing), but should be supplied
  - Beware: incorrect results on sex chromosomes if ploidy file is missing, may lead to incorrect QC statistics
- Where do I get it?
  - Example available for 1000G b37
    ftp://ftp.broadinstitute.org/pub/svtoolkit/ploidymaps/
- Building your own
  - Simple text file, whitespace delimited
  - Columns: chrom, start, end, gender, ploidy
  - Lines are matched in order, asterisks are wildcards

Example (1000G b37):

```
X   2699521   154931043   F   2
X   2699521   154931043   M   1
Y         1    59373566   F   0
Y         1    59373566   M   1
*         *           *   *   2
```

Gender Map File

- What is it?
  - Text file listing the gender of each sample in your dataset
  - Genome STRiP does not attempt to infer the gender of samples
- Where to I get it?
  - You have to generate it
- File format
  - Tab delimited text file, no header
  - Columns: sample ID, gender
  - Gender can be M/F, Male/Female or 1 (male) and 2 (female)
  - Sample ID in the file must match the sample ID in your BAM files

Example:

```
SAMPLE1    M
SAMPLE2    F
SAMPLE3    F
```

Configuration File

- What is it?
  - Specifies default values for many algorithm parameters used in Genome STRiP
- Why do I care?
  - Usually you don't (advanced users may want to override some settings)
- Where do I get it?
  - Standard configuration file comes with the Genome STRiP distribution
  - Best practice is to override select configuration file parameters using *-P key:value*

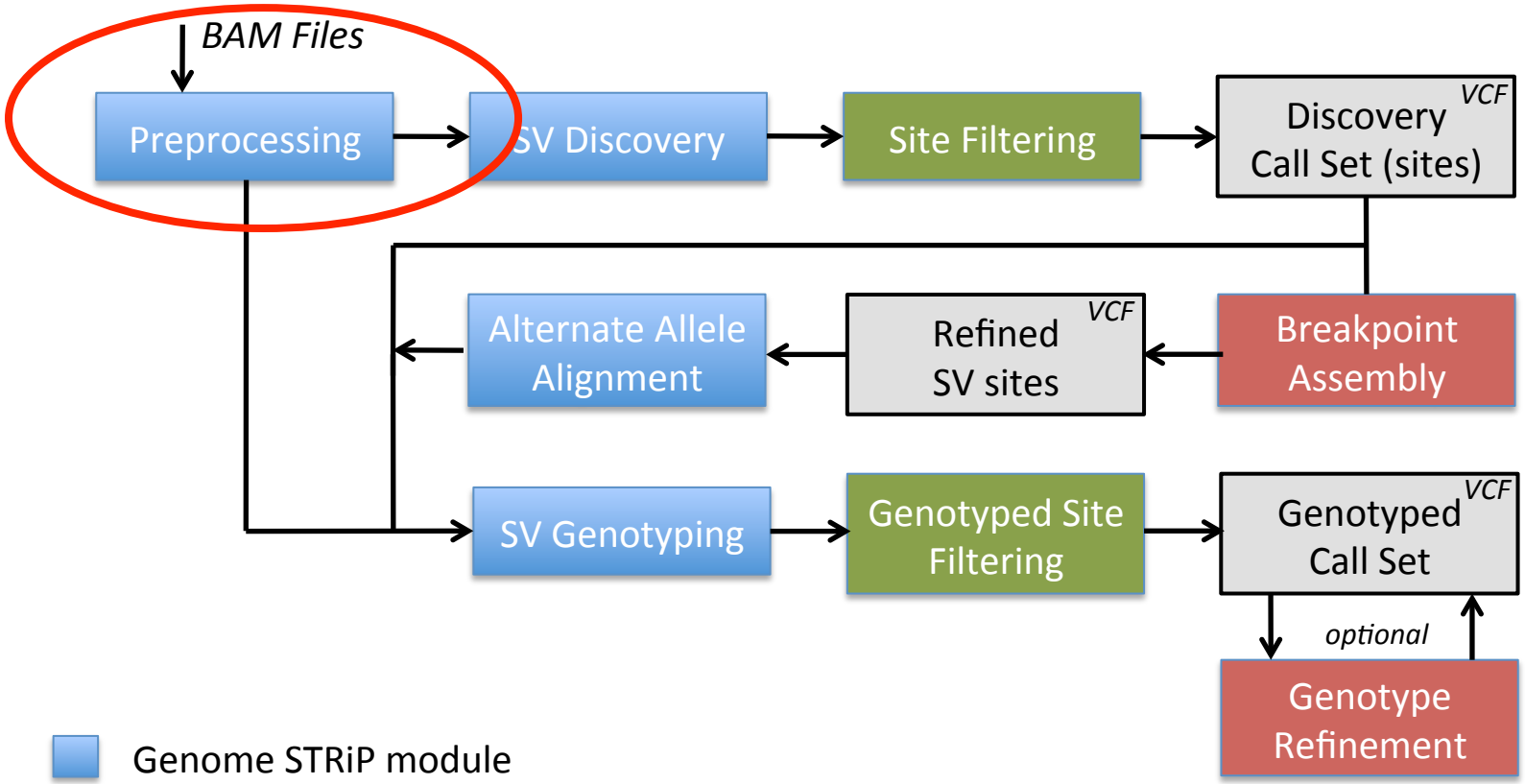
Example

Override genotyping.modules to disable split read genotyping

```
java –Xmx4g –cp ${classpath}
    org.broadinstitute.sting.queue.QCommandLine
    -S ${SV_DIR}/qscript/SVGenotyper.q
    -S ${SV_DIR}/qscript/SVQScript.q
    -P genotyping.modules:depth,pairs
    ...
```

Note: As of r1162, "genome sizes" are no longer read from the config file

Preprocessing



Genome STRiP module

Requires user intervention, defaults or best-practices are available

Third party tool (e.g. TIGRA or velvet for assembly, BEAGLE for genotype refinement)

VCF data file

Preprocessing

- What is pre-computed?
  - Effective genome sizes (from alignability mask)
  - Insert size distributions (per library)
  - Sequencing depth (per readgroup/library/sample)
    Expectation of fragments per base
  - Span coverage (per readgroup/library/sample)
    Expectation of paired reads crossing a breakpoint
  - GC bias (per library)
- Inputs: BAM files
- Outputs: Multiple files in metadata directory
- Workflow: Parallel per-BAM, then merged

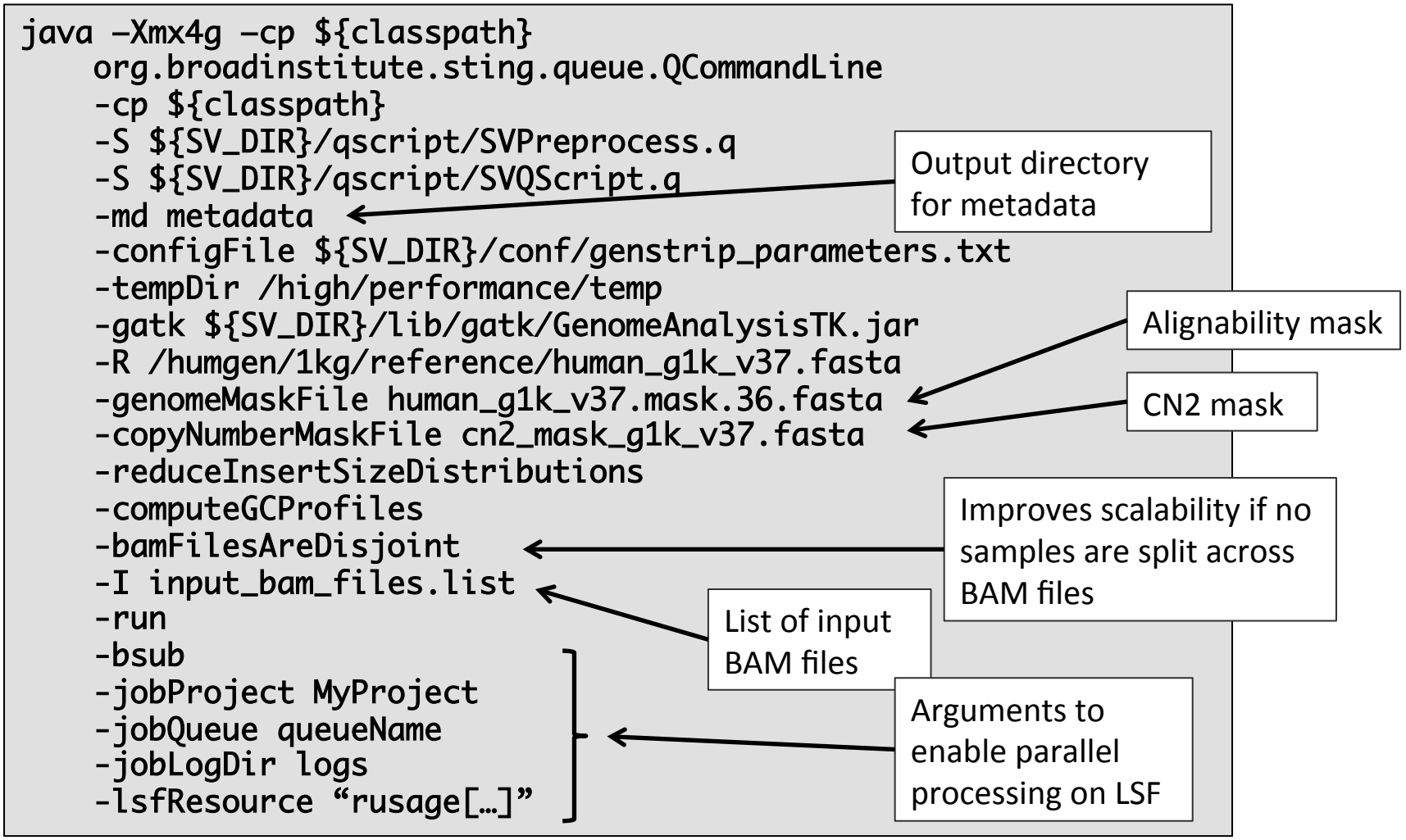
Metadata directory contents

- Genome sizes
  - genome_sizes.txt (simple text file)
- Insert size (fragment length) distributions
  - isd.spans.dat (text file of per-library statistics, useful for QC)
  - isd.hist.bin (full histograms, bulky but lossless)
  - isd.dist.bin (uses lossy accuracy-dependent compression)
  - Use *-reduceInsertSizeDistributions* to generate isd.dist.bin

    Important for scalability on large data sets
- Sequencing depth
  - Sub-directory metadata/depth and depth.dat summary file
- Span coverage
  - Measures total distance "spanned" between the two ends of paired-end reads
  - Sub-directory metadata/spans and spans.dat summary file
- GC-bias profiles
  - Enable with *-computeGCProfiles* command line argument (recommended)
  - Sub-directory metadata/gcprofile and summary gcprofiles.zip file

Running Queue script for preprocessing

```
java -Xmx4g -cp ${classpath}
    org.broadinstitute.sting.queue.QCommandLine
    -cp ${classpath}
    -S ${SV_DIR}/qscript/SVPreprocess.q
    -S ${SV_DIR}/qscript/SVQScript.q
    -md metadata
    -configFile ${SV_DIR}/conf/genstrip_parameters.txt
    -tempDir /high/performance/temp
    -gatk ${SV_DIR}/lib/gatk/GenomeAnalysisTK.jar
    -R /humgen/1kg/reference/human_g1k_v37.fasta
    -genomeMaskFile human_g1k_v37.mask.36.fasta
    -copyNumberMaskFile cn2_mask_g1k_v37.fasta
    -reduceInsertSizeDistributions
    -computeGCProfiles
    -bamFilesAreDisjoint
    -I input_bam_files.list
    -run
    -bsub
    -jobProject MyProject
    -jobQueue queueName
    -jobLogDir logs
    -lsfResource "rusage[…]"
```



Preprocessing options

Two-pass processing

- Starting in version r1068, there are two different strategies for parallelizing preprocessing
- Multi-pass: 5 passes over each input bam file
- Two-pass: First pass is for insert sizes, second pass computes everything else in parallel
- Two-pass is more I/O efficient and is generally faster, but not necessarily on the cloud
- Multi-pass can be faster if you change settings or add some new samples to an existing data set

Preprocessing options

Two-pass processing

Starting in version r1068, there are two strategies for parallelizing preprocessing.

- Multi-pass: Makes 4 passes over each input bam file
- Two-pass: First pass measures insert sizes, second pass computes everything else

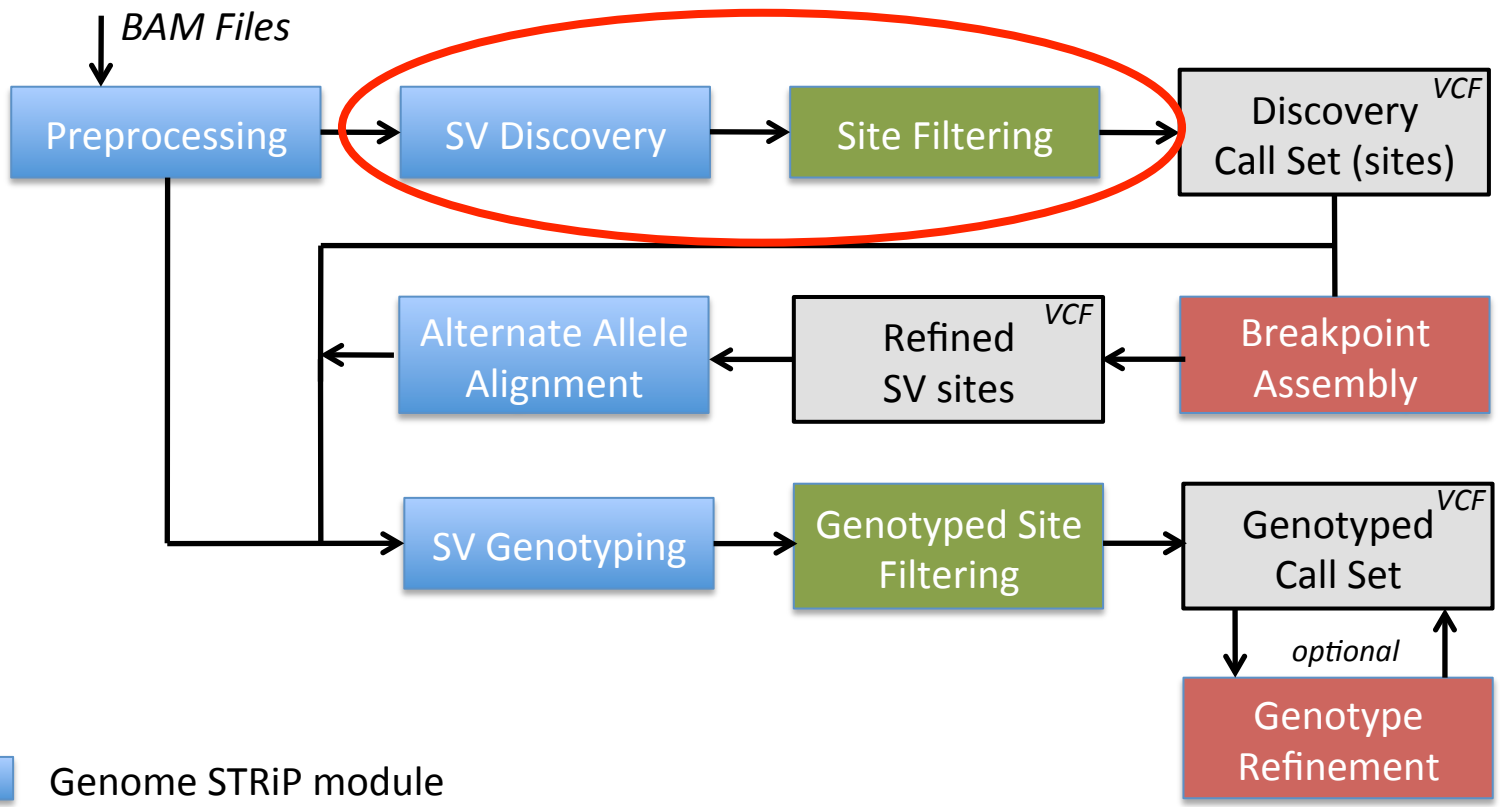
Two-pass is the default, pass *–useMultiStep* to SVPreprocess.q to force multi-pass.
Two-pass is more I/O efficient and uses less total CPU, but is less parallel.
Multi-pass can be better if you change settings or need to add some new samples to an existing data set.

Multiple metadata directories

Starting in version r1162, Genome STRiP discovery/genotyping allows you to reference multiple metadata directories.

- The samples in each metadata directory must be disjoint
- Can be used to preprocess samples in batches over time and then call together
- Each preprocessing run generates a single metadata directory
- Metadata directories used together must be consistent (same reference, same masks, same parameters, etc.)

Discovery



Genome STRiP module

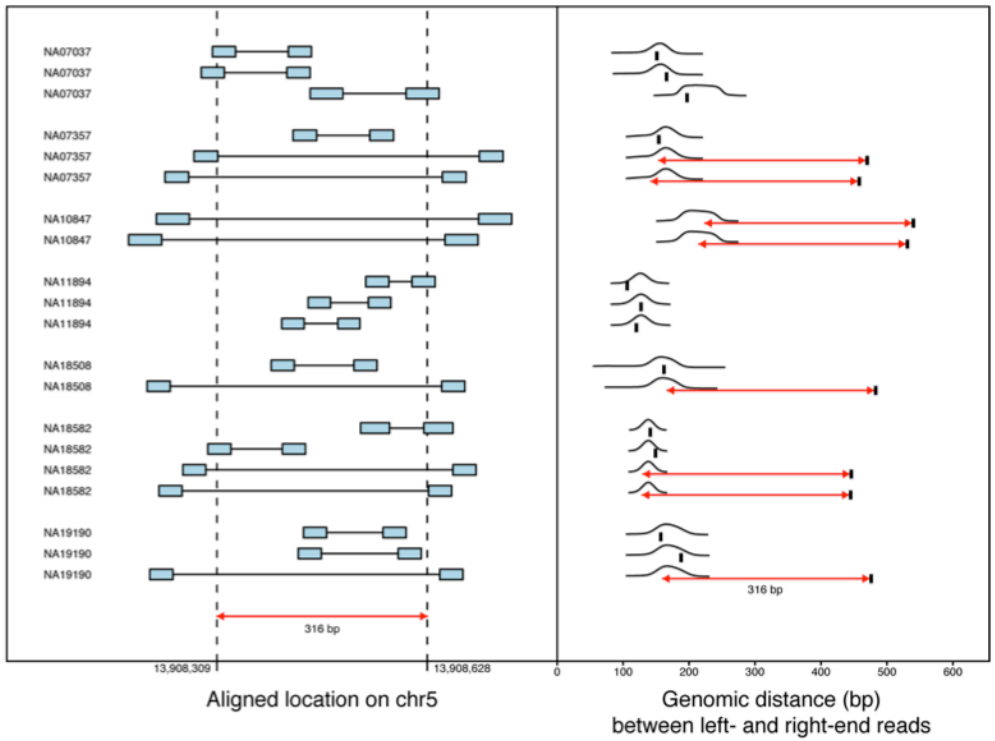
Requires user intervention, defaults or best-practices are available

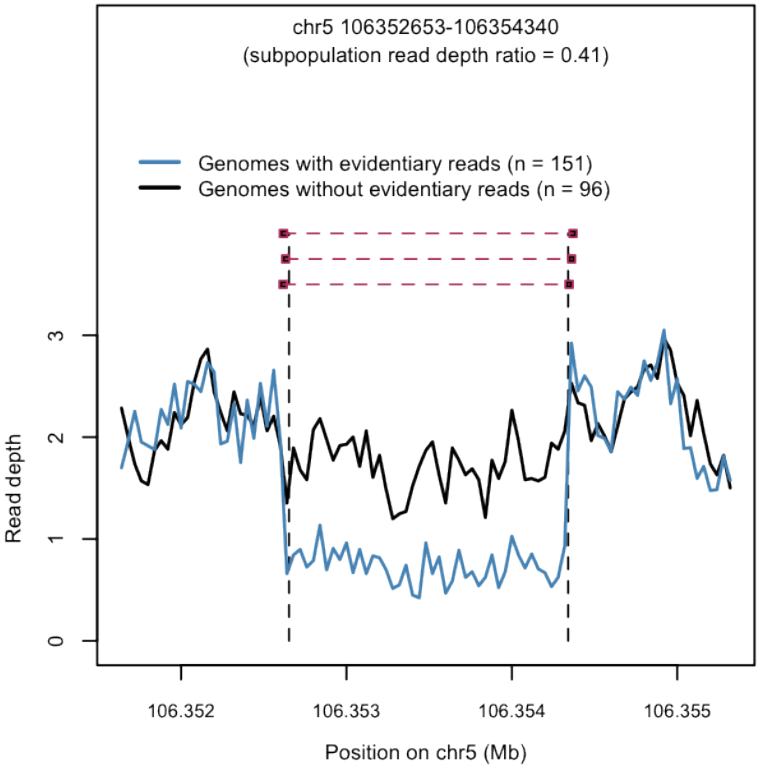
Third party tool (e.g. TIGRA or velvet for assembly, BEAGLE for genotype refinement)

VCF data file

SV Discovery

Deletion discovery integrates diverse features of the sequencing data, including aberrantly spaced read pairs, differential read depth, and distribution of evidence across multiple samples.





SV Discovery

- Inputs: BAM Files, metadata directory
- Outputs: Site VCF file
  - Contains records for all *evaluated* sites
  - The FILTER field tells whether a site is called as a true variant
  - Most evaluated sites are typically *not* called
  - The INFO field contains other quality metrics
- Auxilliary outputs: Multiple files in run directory
  - Used for QC and filtering
- Workflow: Parallel per-genome-locus and per-length-range, then merged
  - Tip: In the current implementation, discovery runs that include sites larger than 100Kb are significantly slower. If you have a large data set, it is recommended to do separate runs for events shorter than 100Kb and larger than 100Kb (your throughput will be more uniform)

Running Queue script for deletion discovery

```
java -Xmx4g -cp ${classpath}
    org.broadinstitute.sting.queue.QCommandLine
    -cp ${classpath}
    -S ${SV_DIR}/qscript/SVDiscovery.q
    -S ${SV_DIR}/qscript/SVQScript.q
    -md metadata
    -configFile ${SV_DIR}/conf/genstrip_parameters.txt
    -tempDir /high/performance/temp
    -gatk ${SV_DIR}/lib/gatk/GenomeAnalysisTK.jar
    -R /humgen/1kg/reference/human_g1k_v37.fasta
    -genomeMaskFile human_g1k_v37.mask.36.fasta
    -ploidyMapFile human_g1k_v37_ploidy.map
    -genderMapFile sample_gender.map
    -runDirectory run1
    -minimumSize 100
    -maximumSize 100000
    -I input_bam_files.list
    -O run1/deletions.discovery.vcf
    -jobProject MyProject
    -jobQueue queueName
    -jobLogDir run1/logs
    -windowSize 1000000
    -windowPadding 10000
```

Run directory for intermediate files

Parallelize based on event size

Output VCF file

Arguments for parallelization on a compute cluster

Discovery site filtering

Default filters

Filter name	Description
COVERAGE	Site has excessive read pileup
COHERENCE	Read pairs spacing is not consistent with a single segregating event
DEPTH	Read depth is not consistent with the read pair evidence across samples
DEPTHPVAL	Read depth differences are not significant
PAIRSPERSAMPLE	Read pair evidence is thinly distributed across samples

Not in default filter list, but recommended best-practices

ALPHASAT	Call is in regions of mostly alpha satellite repeat

Additional discovery site filters

For alpha satellite filtering, first run SVAnnotator to add annotations to the VCF about the repeat content of each evaluated site.

```
java -Xmx4g -cp ${classpath} org.broadinstitute.sv.main.SVAnnotator
  -R /humgen/1kg/reference/human_g1k_v37.fasta
  -A MobileElements
  -repeatTrackFile ucsc_repeats_g1k_v37.dat
  -vcf deletions.discovery.unfiltered.vcf
  -O deletions.discovery.annotated.vcf
```

Track file from UCSC browser, also available from our ftp site

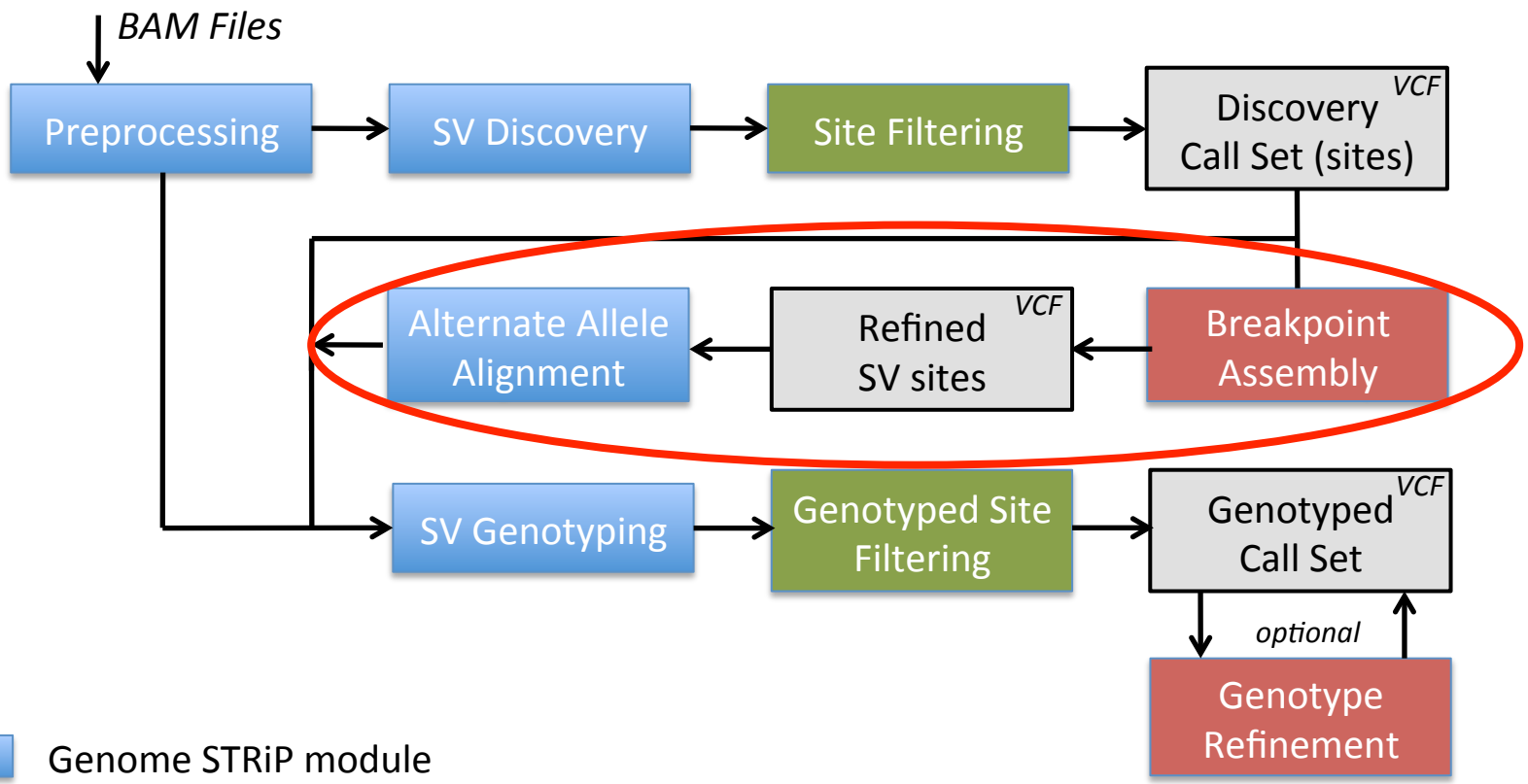
Then filter variants using GATK VariantFiltration.
This can also be combined with the default filters run from SVDiscovery.q.

```
java -Xmx4g -jar GenomeAnalysisTK.jar
  -T VariantFiltration
  -B:variant,VCF deletions.discovery.annotated.vcf
  -o deletions.discovery.vcf
  -R /humgen/1kg/reference/human_g1k_v37.fasta
  -filterName ALPHASAT -filter "GSALPHASATFRACTION > 0.90"
```

Best practice filters

In addition, if you have deep sequencing, you might consider a higher threshold than the default of 1.1 (mean aberrant read-pairs per sample) in the default PAIRSPERSAMPLE filter, but this threshold has worked well for both 4x sequencing (1000 Genomes) and 12x sequencing (GoNL).

Breakpoint assembly



Genome STRiP module

Requires user intervention, defaults or best-practices are available

Third party tool (e.g. TIGRA or velvet for assembly, BEAGLE for genotype refinement)

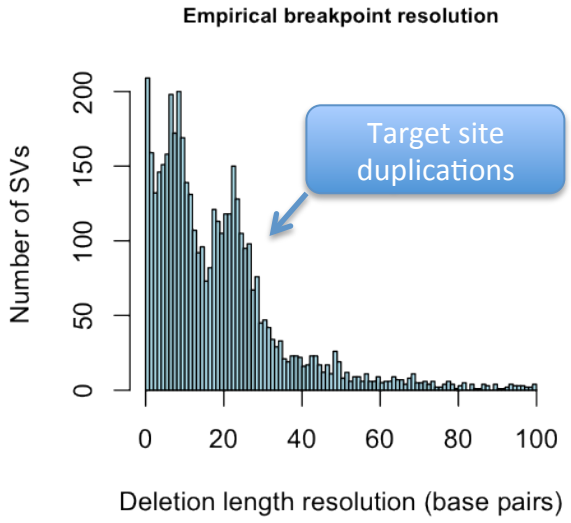
VCF data file

Breakpoint assembly

To determine precise breakpoints, use a third party tool (e.g. TIGRA-SV, velvet) or a catalog of known breakpoints (e.g. 1000 Genomes)

Genome STRiP generates calls with approximate coordinates (typically 10-20 bp resolution)

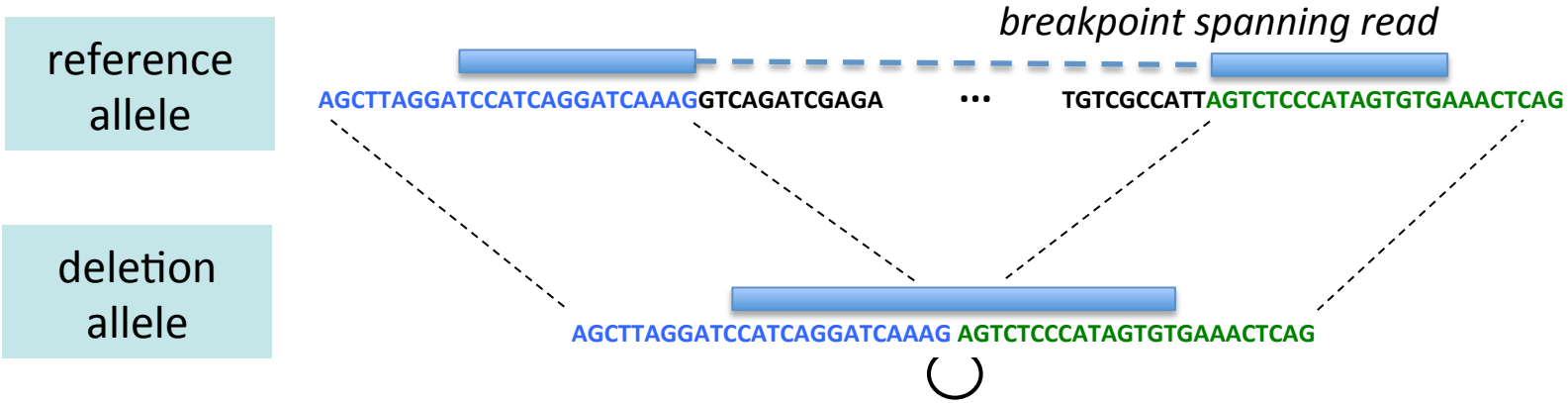
To utilize breakpoint-spanning reads in genotyping, you need exact breakpoint coordinates.



In the 1000 Genomes project, we used TIGRA-SV (WashU) and AGE (Yale) to assemble breakpoints for over half of the discovered deletion sites.

Alternate allele alignment

When precise alleles are available, we use breakpoint-spanning reads in genotyping.



There are three sources for breakpoint-spanning reads:

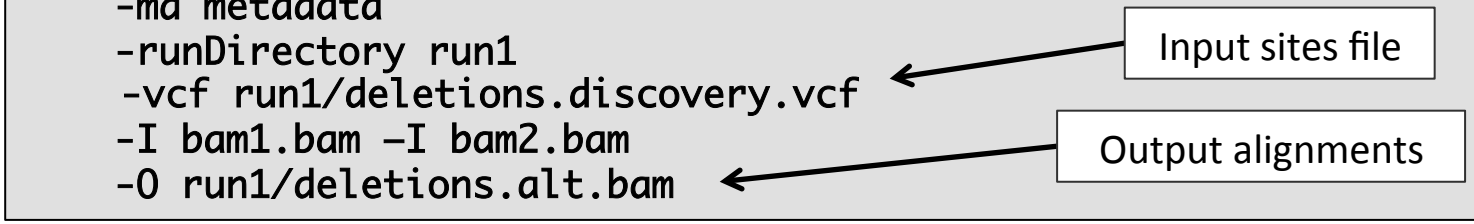
Source	How handled
"in-place" reads aligned at the breakpoint	Automatically realigned on-the-fly to alt allele during genotyping
unmapped mates in same BAM file where mate is aligned nearby	Automatically realigned on-the-fly during genotyping
completely unmapped reads	Requires alternate allele aligner Rarely useful with modern 100+ bp reads

Queue script for alt allele alignment

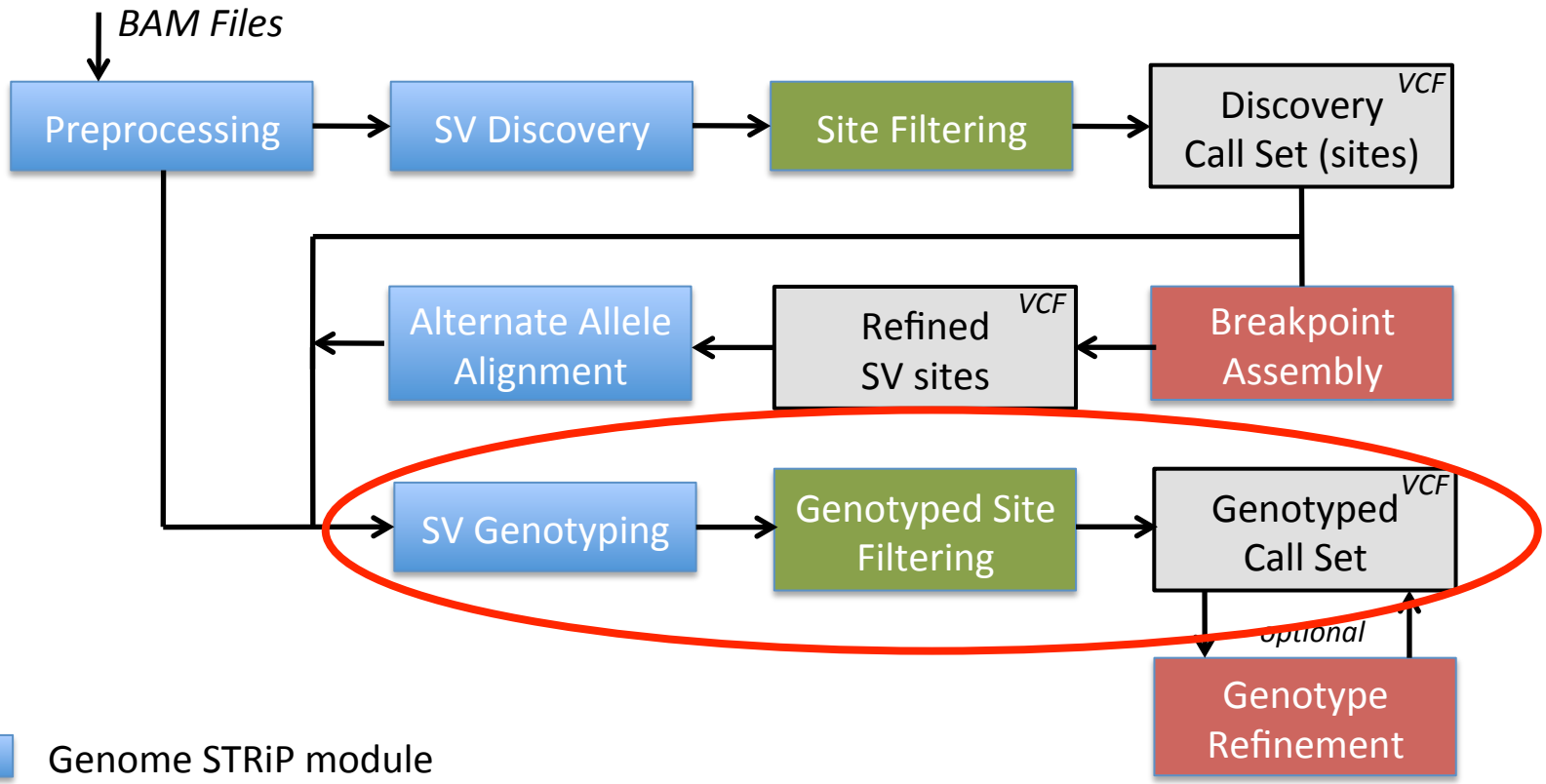
Inputs: VCF file containing SVs with exact alleles
BAM files containing unmapped reads
Outputs: BAM file containing alignments to alternate alleles

With longer reads (e.g. 100bp) there is only marginal benefit in running the alternate allele aligner step. Most reads will be in the main BAM files (often soft-clipped) and will be used automatically for breakpoint genotyping.

```
java –Xmx4g org.broadinstitute.sting.queue.QCommandLine
    ...
    -S ${SV_DIR}/qscript/SVAltAlign.q
    -R /humgen/1kg/reference/human_g1k_v37.fasta
    -md metadata
    -runDirectory run1
    -vcf run1/deletions.discovery.vcf
    -I bam1.bam –I bam2.bam
    -O run1/deletions.alt.bam
```



Genotyping



Genome STRiP module

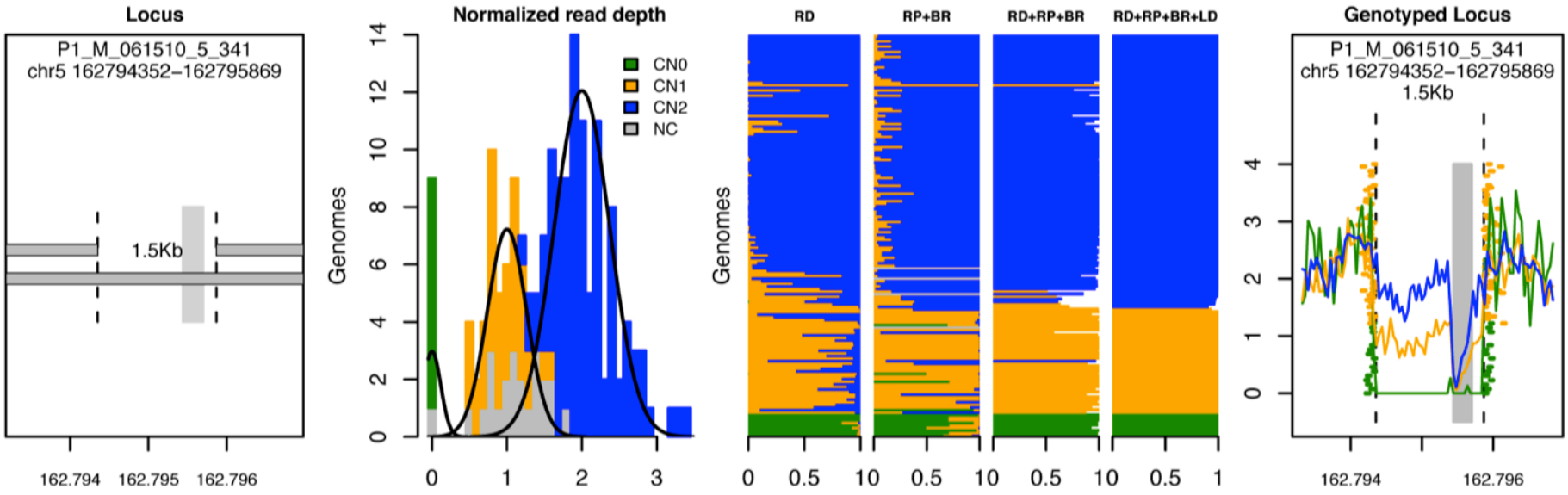
Requires user intervention, defaults or best-practices are available

Third party tool (e.g. TIGRA or velvet for assembly, BEAGLE for genotype refinement)

VCF data file

SV Genotyping

Genome STRiP integrated information from read depth, discordant read pairs and breakpoint spanning reads to genotype deletions.



Support for genotyping other types of variants (e.g. duplications) is under development.

SV Genotyping

- Inputs: BAM Files, VCF site file, metadata directory
  - Typically you want to extract only the PASS sites for genotyping if you are using sites generated from Genome STRiP discovery
  - Reformat VCF records to include specific alleles (rather than <DEL>) to enable breakpoint-based genotyping
- Outputs: Genotype VCF file
  - Contains records for all input sites
  - The FILTER field tells whether a site passes genotyping site filters
  - Non-passing sites will still contain genotype calls
- Auxilliary outputs: Multiple files in run directory
  - Used for QC and filtering
- Workflow: Parallel per-input-site from VCF (in batches), then merged

Running Queue script for deletion genotyping

```
java -Xmx4g -cp ${classpath}
    org.broadinstitute.sting.queue.QCommandLine
    -cp ${classpath}
    -S ${SV_DIR}/qscript/SVGenotyper.q
    -S ${SV_DIR}/qscript/SVQScript.q
    -md metadata
    -configFile ${SV_DIR}/conf/genstrip_parameters.txt
    -tempDir /high/performance/temp
    -gatk ${SV_DIR}/lib/gatk/GenomeAnalysisTK.jar
    -R /humgen/1kg/reference/human_g1k_v37.fasta
    -genomeMaskFile human_g1k_v37.mask.36.fasta
    -ploidyMapFile human_g1k_v37_ploidy.map
    -genderMapFile sample_gender.map
    -runDirectory run1
    -vcf deletions.sites.vcf
    -I input_bam_files.list
    -O run1/deletions.genotypes.vcf
    -jobProject MyProject
    -jobQueue queueName
    -jobLogDir run1/logs
    -parallelRecords 100
```

Run directory for intermediate files

Input sites file

Output VCF file

Arguments for parallelization

Genotype VCF for SVs

Sample VCF file (INFO field not shown)

```
CHROM  POS ID      REF   ALT QUAL FILTER INFO  FORMAT          NA10000

1  2918690 DEL_833 G   <DEL>  .   PASS    …    GT:FT:GL:GL0:GQ  0/1:PASS:-9.3,-0.0,-135.2:-11.1,-0.0,-13.0:93
```

FORMAT tag	Examples	Description
GT	0/1	Genotype (0 = reference, 1 = alt) But if FT is not PASS, should consider as no-call
FT	PASS, LowQual	Filter field for genotypes LowQual for sites with GQ < 13 (95% confidence)
GL/PL	-9.3,-0.0,-135.2	Genotype likelihoods (uses population frequency information extracted from read depth mixture model) PL is same information, but encoded as phred-scaled integer
GL0	-11.1,-0.0,-13.0	Genotype likelihoods with no frequency prior Use these values for LD-based genotype refinement
GQ	93	Genotype quality (phred scaled)

Genotyped Site Filtering

Current best practices are to apply a set of site-level filters after genotyping in the following categories*

- Data sufficiency
  - Filter sites with insufficient data to genotype accurately
- Genotype accuracy
  - Filter sites with low apparent genotype accuracy
- Non-variant sites
  - Flag sites that appear non-variant post-genotyping
- Duplicate sites
  - Flag apparent duplicate calls based on genotype likelihoods
  - Filter, retaining the site with the highest genotype quality

* Running these post-genotyping site filters is not done by default in the latest released version of Genome STRiP, but is a planned option for future releases.

Genotyped site filtering

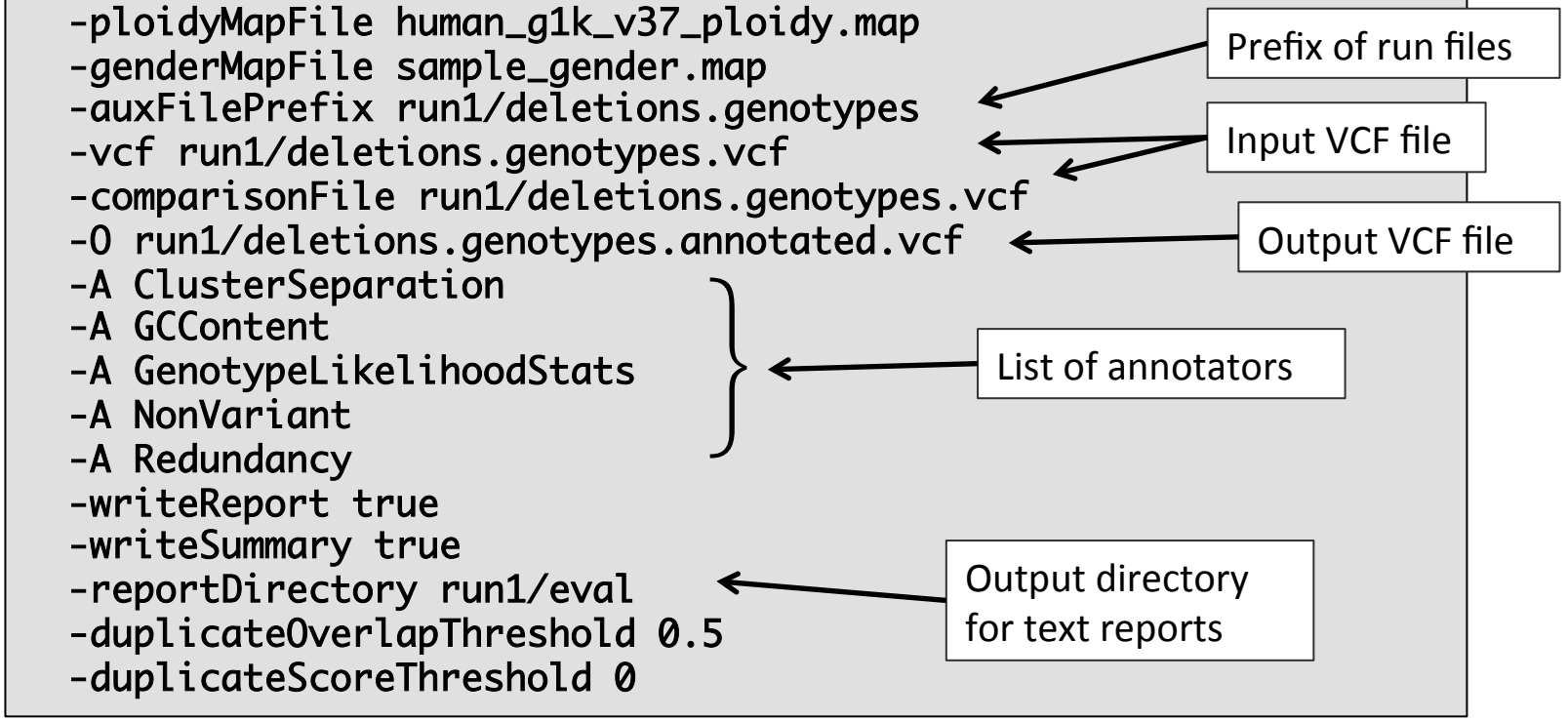
Current best-practice filters (from 1000G and other large projects)

Filter name	Description
ALIGNLENGTH	Site has insufficient alignable bases (default 200)
CLUSTERSEP	Read depth model shows insufficient cluster separation
GTDEPTH	Read depth at the site is too low or too high
INBREEDINGCOEFF	Filter site due to excess number of het calls
DUPLICATE	Site is apparent duplicate of another site
NONVARIANT	Site is likely non-variant based on genotype likelihoods

These filters are currently implemented by running annotators post-genotyping and then using GATK VariantFiltration. This strategy is applicable only when breakpoints are not being used for genotyping, and therefore is not enabled by default.

Annotations used for filtering

```
java -Xmx4g -cp ${classpath}
  org.broadinstitute.sv.main.SVAnnotator
  -md metadata
  -R /humgen/1kg/reference/human_g1k_v37.fasta
  -ploidyMapFile human_g1k_v37_ploidy.map
  -genderMapFile sample_gender.map
  -auxFilePrefix run1/deletions.genotypes
  -vcf run1/deletions.genotypes.vcf
  -comparisonFile run1/deletions.genotypes.vcf
  -O run1/deletions.genotypes.annotated.vcf
  -A ClusterSeparation
  -A GCContent
  -A GenotypeLikelihoodStats
  -A NonVariant
  -A Redundancy
  -writeReport true
  -writeSummary true
  -reportDirectory run1/eval
  -duplicateOverlapThreshold 0.5
  -duplicateScoreThreshold 0
```



Genotyped site filters

These post-genotyping filters are applicable only when breakpoints are not being used for genotyping. If you are genotyping with breakpoints, the depth-based filters should be applied to the read depth signal only, which currently requires custom post-processing.

Example of using GATK VariantFiltration to run best-practices filters, based on experiences from 1000 Genomes, GoNL and other projects.

```
java –Xmx4g –jar GenomeAnalysisTK.jar
  -T VariantFiltration
  -B:variant,VCF deletions.discovery.annotated.vcf
  -o deletions.discovery.vcf
  -R /humgen/1kg/reference/human_g1k_v37.fasta
  -filterName ALIGNLENGTH –filter “GSELENGTH < 200”
  -filterName CLUSTERSEP –filter “GSCLUSTERSEP == NA || GSCLUSTERSEP <= 2.0”
  -filterName GTDEPTH –filter “GSM1 == NA || GSM1 <= 0.5 || GSM1 >= 2.0”
  -filterName INBREEDINGCOEFF –filter “GLINBREEDINGCOEFF != NA &&
                                      GLINBREEDINGCOEFF < -0.15”
  -filterName NONVARIANT –filter “GSNONVARSCORE != NA && GSNONVARSCORE >= 13.0”
  -filterName DUPLICATE –filter “GSDUPLICATESCORE != NA && GSDUPLICATESCORE >= 0”
```

These default filters and thresholds are available in the auxilliary Q script ${SV_DIR}/qscript/SVGenotyperWithoutSplitReads.q

Genotype refinement

1000 Genomes used BEAGLE & MaCH for genotype refinement

Exploits LD between deletions and SNPs / small indels
Generated complete phased haplotypes at all sites
Can be computationally demanding

There is currently not an automated module in Genome STRiP to perform genotype refinement using LD.

QUALITY CONTROL

Brief overview of several QC methods

Illustrative examples from GoNL and 1000 Genomes Projects

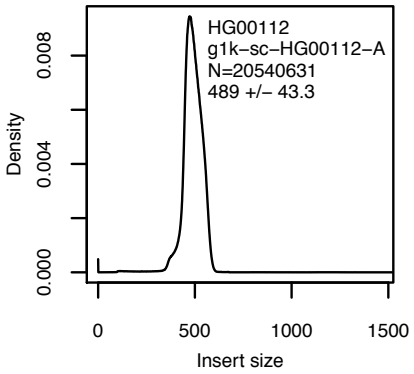
Sample QC: Insert size distributions

Review insert size distribution summary statistics: metadata/isd.stats.dat

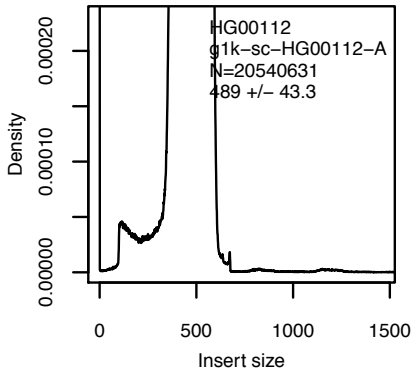
SAMPLE	LIBRARY	READGROUP	NPAIRS	NBINS	MEDIAN	RSD
HG00112	g1k-sc-HG00112-A	NA	20540631	49437	489	43.28
HG00112	g1k-sc-HG00112-D	NA	32894100	48657	417	34.03

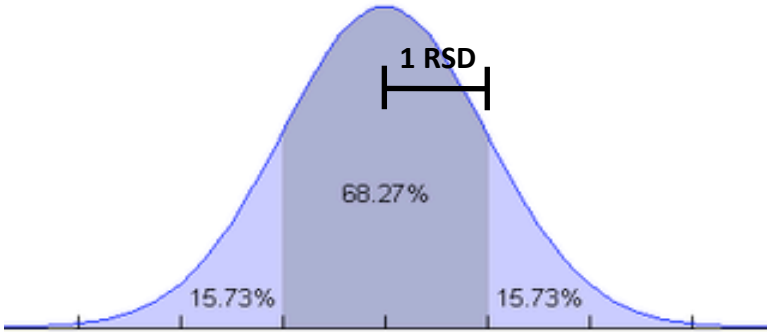
Distributions for individual libraries can be plotted using PlotInsertSizeDistributions. Two plots are generated for each library, one full scale and one expanded. Distributions with excess mass in the *right* tail are particularly problematic for discovery. The example below is relatively clean in the right tail.

Full scale



Expanded





RSD is a variance measure more robust to outliers.
RSD = 0.5 x width of middle 68.27% of the distribution.

Discovery QC
Filter summary from GoNL Project

The Genome of the Netherlands Project (GoNL) performed 12x whole-genome sequencing on 250 trios/quartets of Dutch ancestry

Discovery sites evaluated: 32,759
Passing discovery sites: 14,103 (43%)

FILTER	COUNTSINGLE	COUNTTOTAL
ALPHASAT	331	2060
COHERENCE	823	2461
COVERAGE	257	6846
DEPTH	3341	16829
DEPTHPVAL	2092	10834
PAIRSPERSAMPLE	251	7424
PASS	14103	14103

The rate of passing sites (43%) is higher than 4x sequencing projects like 1000 Genomes. The PAIRSPERSAMPLE filter rejected relatively few sites on its own, as would be expected for 12x sequencing.

Genotyping QC in GoNL
Filters, genotyped sites rate, genotype call rate

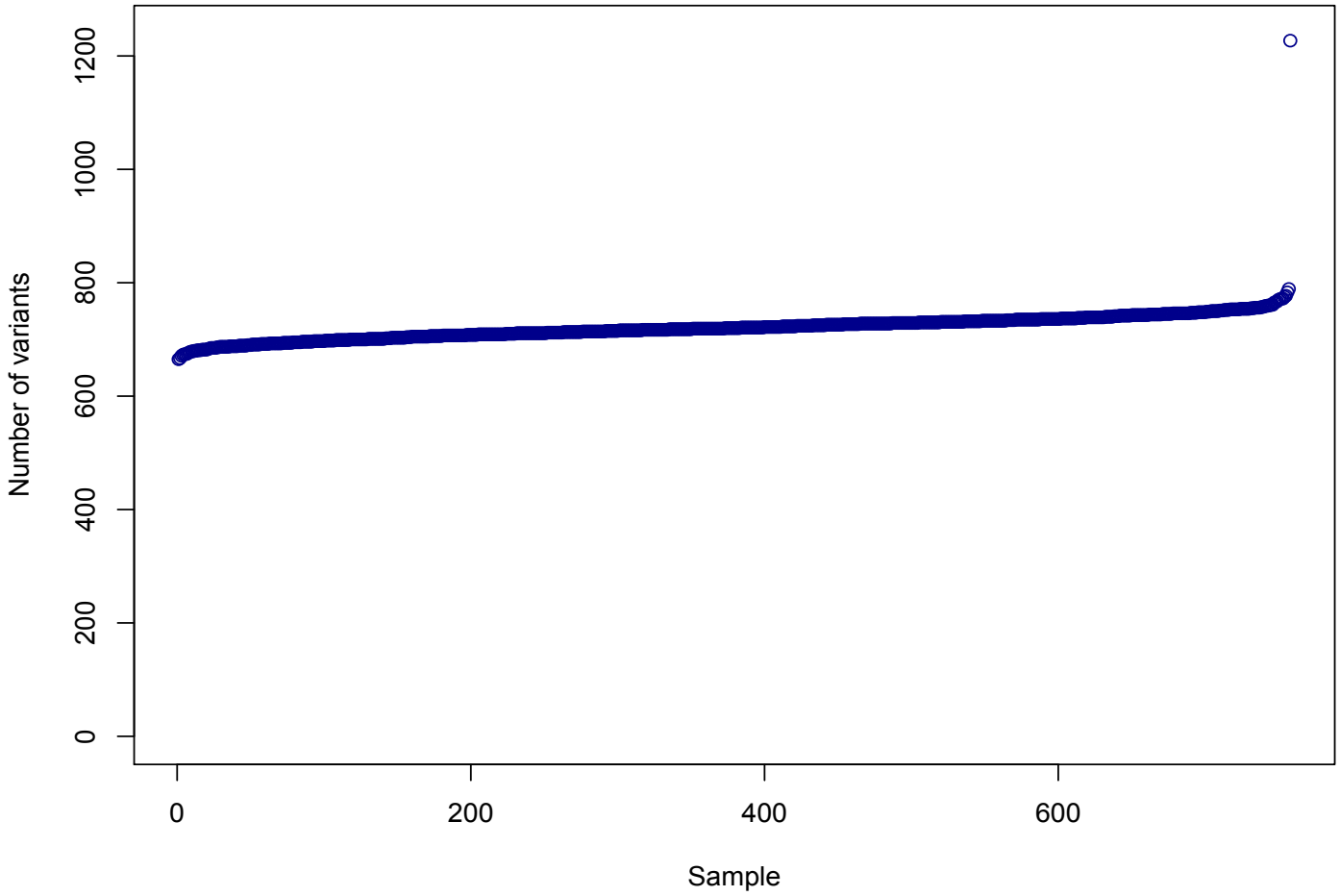
Passing discovery sites: 14,103
Genotyped sites: 9,352 (66%)

FILTER	COUNTSINGLE	COUNTTOTAL
ALIGNLENGTH	608	3351
CLUSTERSEP	252	3025
GTDEPTH	94	863
INBREEDINGCOEFF	477	2467
DUPLICATE	30	30
NONVARIANT	55	62
PASS	9352	9352

The genotyping rate (66%) is higher than 1000G Phase 1 (61%) – in both projects breakpoints were not used in genotyping, which means many shorter sites are called but not genotyped. At passing sites the genotype call rate (at 95% confidence) is 98.9% before genotype refinement, substantially higher than 1000G which used 4x sequencing.

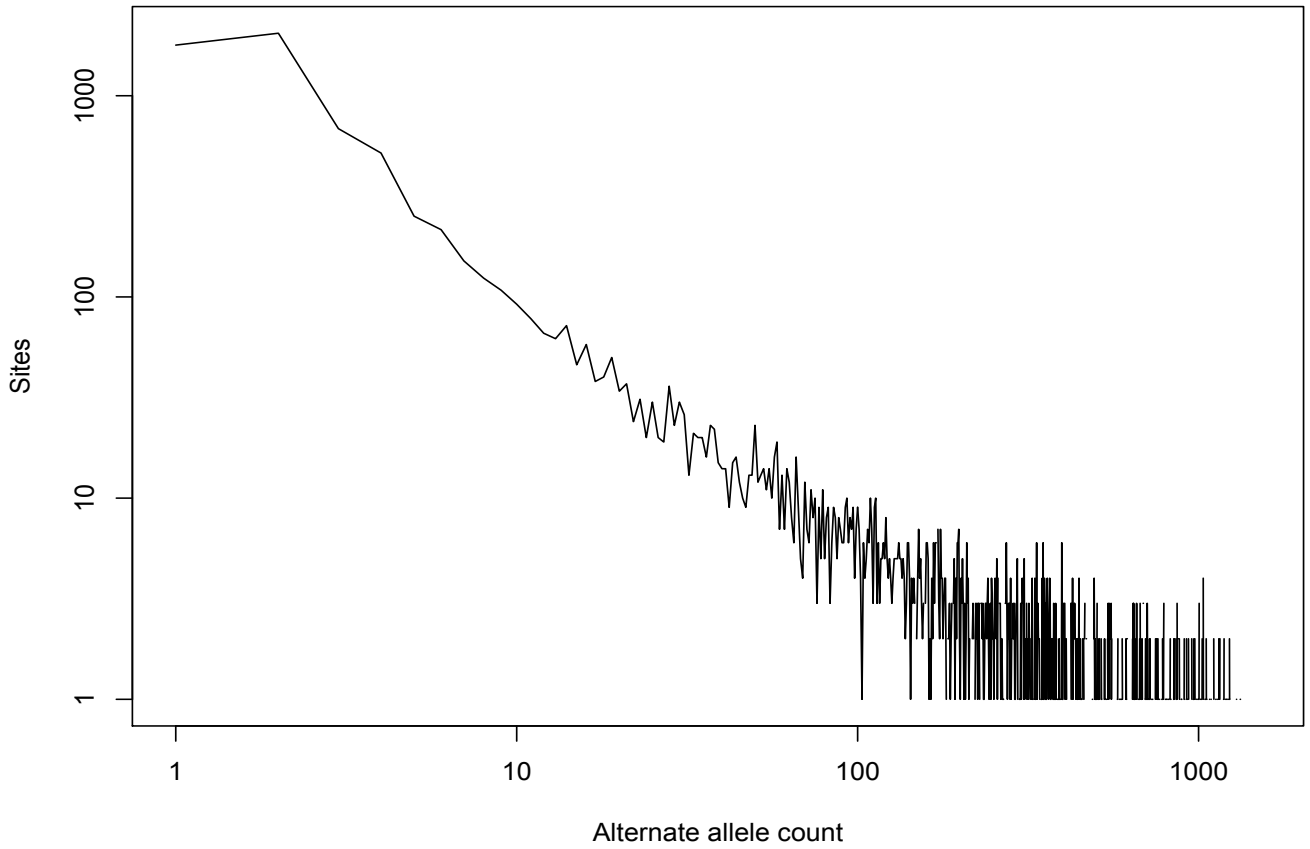
Genotype QC: Variants per sample

Data from GoNL: The distribution of variants per sample looks very uniform in this relatively homogeneous European cohort, with the exception of one clear outlier, which was flagged for investigation and subsequently removed.



Genotyping QC: Allele frequency spectrum

The overall spectrum looks roughly linear on this log/log plot. The apparently reduced power to call singletons is likely due to the trio design (the offspring should never have singletons).



Other Genotype QC Metrics

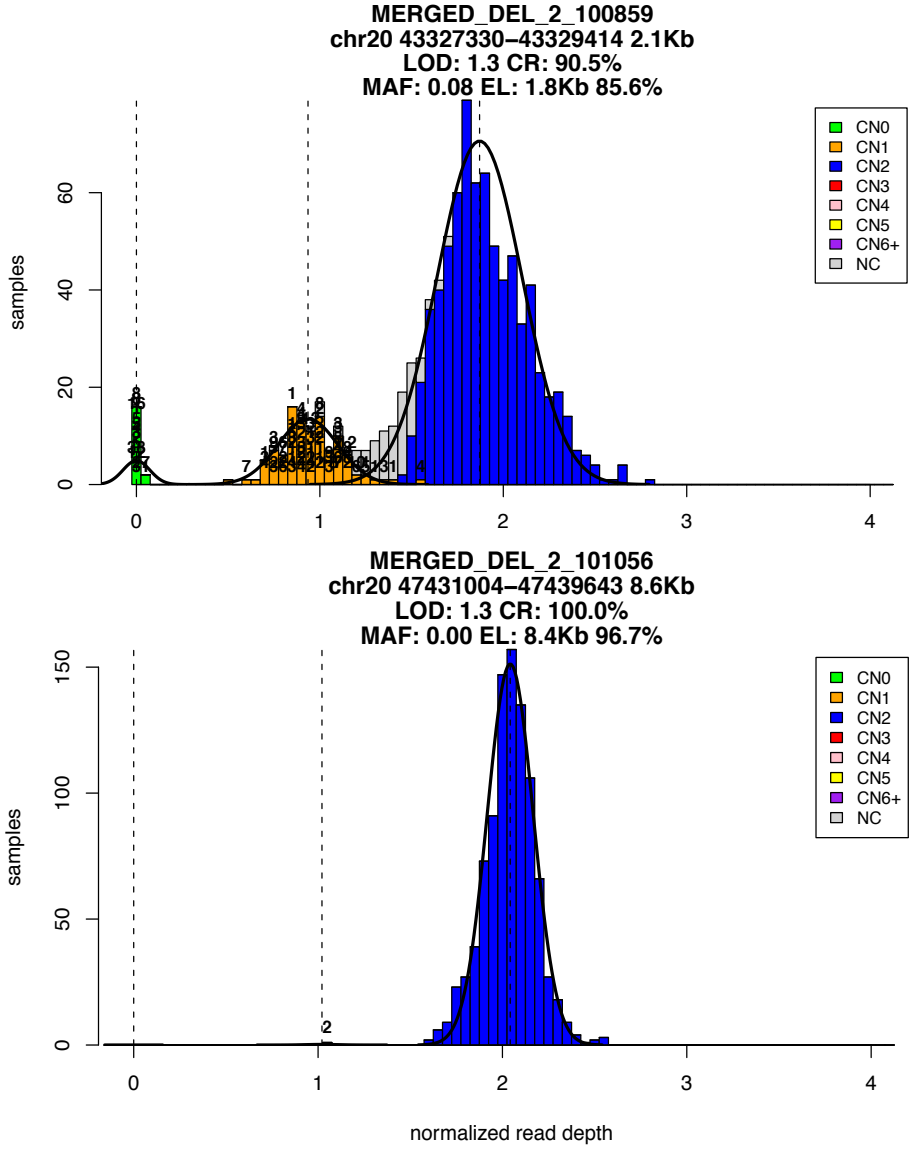
- Comparison to previously ascertained call sets
  - 1000 Genomes
  - ReciprocalOverlap annotator is a useful tool (compares two SV vcf files)
- Successful breakpoint assembly
- Lack of heterozygous SNPs in hemizygous regions
  - Individuals heterozygous for a deletion should not have heterzygous SNPs at the locus

Individual site QC: PlotGenotypingResults

PlotGenotypingResults is a utility program that generates PDF plots for individual sites, like the ones shown on the right (4x sequencing from 1000 Genomes). Inspection of individual sites is a valuable QC tool.

The top site is relatively short (2.1Kb) and high frequency. The small black numbers are supporting read pairs, seen in samples called het or homozygous deleted. Some samples (gray) are not confidently called in 4x sequencing at this length scale.

The bottom site is a larger (8.6Kb) singleton, with a call rate of 100%.



GENOTYPING NOVEL SITES IN 1000 GENOMES PHASE 1

Local genotyping with remote data access

Using Amazon Web Services (AWS)

Scenario

- You discover a large deletion in an individual or family
  - Potential rare mendelian syndrome
- Questions
  - How rare is this variant?
  - Was there any evidence for this polymorphism in the 1000 Genomes project?

- Comparing to the published 1000 Genomes call sets might miss something (false negatives in the 1000G analysis).

- Is there an inexpensive way to genotype this site in the 1000 Genomes cohort?

Inexpensive genotyping

- 1000 Genomes data is available publicly online

  Three convenient locations to serve you
  - Amazon S3 (in our experience, the most reliable)
  - NCBI over http or ftp
  - EBI over http or ftp
- Not much data is needed to genotype a single site, so remote access is practical
  - Genome STRiP now supports accessing remote bam files by URL (http, ftp or s3)
- We have pre-computed and packaged the necessary Genome STRiP metadata
  - Available for download from our FTP site (about 20Gb)

    ftp://ftp.broadinstitute.org/pub/svtoolkit/public_metadata/1000G_phase1_mdv1.tar.gz
- You can run the computation on your own hardware or on the Amazon cloud (AWS)
  - Genotyping a single site takes only a few minutes
  - Cookbook recipes and instructions are available on our web site

    http://www.broadinstitute.org/software/genomestrip

Genotyping inputs/outputs

Input

A minimal sites-only VCF file (from another tool or you can create by hand)

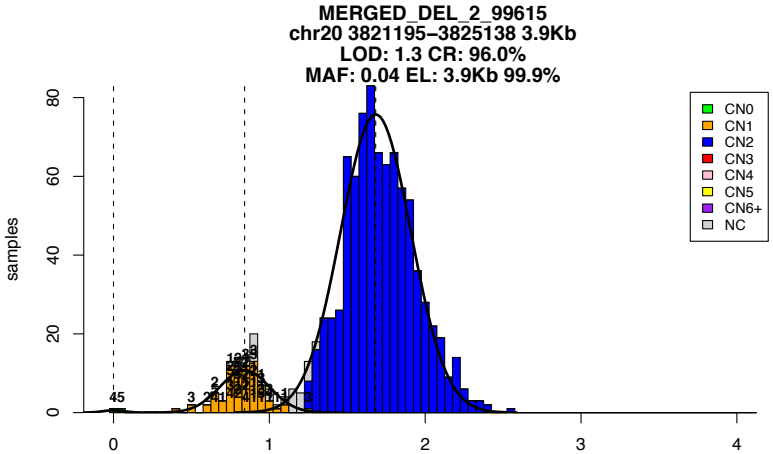
```
##fileformat=VCFv4.1
#CHROM POS  ID     REF ALT   QUAL FILTER INFO
20     1000 MYSITE N   <DEL> .    .      SVTYPE=DEL;END=2000
```

Important fields: ID, CHROM, POS, END, SVTYPE

Outputs

A genotype VCF file (includes genotypes for each sample)

```
##fileformat=VCFv4.1
#CHROM POS  ID     REF ALT   QUAL FILTER INFO                 FORMAT   HG00096      ...
20     1000 MYSITE N   <DEL> .    PASS   SVTYPE=DEL;END=2000 GT:FT:GQ 0/0:PASS:73  ...
```



Genotyping plot of each site (pdf file)

Refer to the VCF file format specification for additional details.

SOFTWARE AND SUPPORT

Software availability

Usage scenarios

Resource requirements

Software availability and support

Web site

http://www.broadinstitute.org/software/genomestrip

Documentation, FAQ, Cookbook

Software downloads

- You need to register in order to download (name, email, organization)
- Production release: Corresponds to 1000 Genomes Pilot
- Interim releases: More recent updates, supported, limited documentation

Most of the functionality discussed here is in the interim releases.

Support mailing list

http://sourceforge.net/projects/svtoolkit/support

GATK Support Forum

http://gatkforums.broadinstitute.org

Installation test

Used to validate correct installation

- Ten minute example run on toy data set
- People also use this as a recipe for production analyses (better to refer to the examples in this presentation)
- Most common pitfalls
  - Installtest runs single threaded, not parallel
  - Does not use –reduceInsertSizeDistributions; add this for scalability
  - Uses -L 1 (restrict analysis to chromosome 1) to speed up the run time; remove this for whole genome analysis (or change if your reference uses “chr1”)

Usage Scenarios

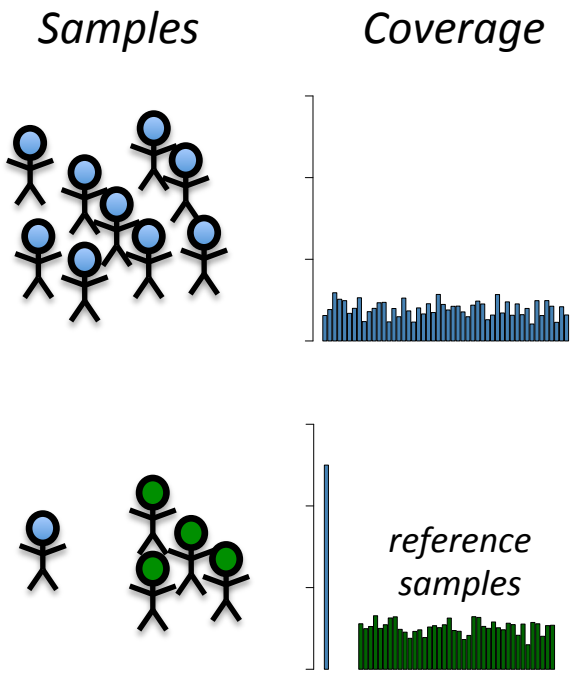
De novo deletion discovery and genotyping
Genotyping known events in new samples

<u>Whole Genome Population Sequencing</u>

Need 20-30+ samples for good results
Low or high coverage, can be variable

<u>Future Goals</u>

Deep coverage single individual
using 1000G reference samples
as background population



Resource requirements

Performance on some sample analyses (1000 Genomes pilot / phase 1)
All steps are highly parallel, designed for compute farms

Algorithm Step	Data Set Size	Run time (CPU days)
Preprocessing	672x (168 x 4x) 2.3Tb	11
Discovery	672x (168 x 4x) 2.3Tb	5
Alt allele alignment	672x (168 x 4x) 2.3Tb	4
Genotyping	22,000 sites x 168 samples	4
Preprocessing	3800x (946 x 4x) 17Tb	86
Discovery	3800x (946 x 4x) 17Tb	150
Genotyping	113,000 sites x 946 samples	360

A number of scalability improvements are under development

Summary

- Genome STRiP has performed well in the 1000 Genomes Project on deletion discovery and genotyping
- Genome STRiP has been used successfully in other large projects
- Common usage scenarios
  - De novo deletion discovery and genotyping in sequencing-based GWAS
  - Genotyping known deletions (e.g. from 1000 Genomes) in new samples
  - Efficient genotyping against public data sets (1000 Genomes Phase 1)
- Improvements are ongoing
  - Usability and scalability
  - Best-practices and tools for calling and QC
  - Pipelines for new variant types and usage scenarios

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