GRIDSS2: comprehensive characterisation of somatic structural variation using single breakend variants and structural variant phasing

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Abstract

GRIDSS2 is the first structural variant caller to explicitly report single breakends - breakpoints in which only one side can be unambiguously determined. By treating single breakends as a fundamental genomic rearrangement signal on par with breakpoints, GRIDSS2 can explain 47% of somatic centromeric copy number changes using single breakends to non-centromeric sequence, with chromosome 1 exhibiting a unique centromeric rearrangement signature. On a cohort of 3,782 deeply sequenced metastatic cancers, GRIDSS2 achieved an unprecedented 3.1% false negative rate and identified a novel 32-100bp duplication signature. Somatic structural variants are highly clustered with GRIDSS2 phasing 16% using just paired-end sequencing.

Keywords

Single breakends, somatic, structural variation

Background

The reliable detection of structural variants (SVs) is critical to understanding the role genome architecture plays in health and disease. This is especially important in cancer and precision medicine where structural variation can be a key driver mutation [1,2]. Over the past decade, many tools have been developed for the detection of genomic rearrangements, which have been the subject of recent extensive benchmarks [3,4]. These tools fall broadly into two camps: those that detect changes in DNA abundance, known as copy number variant or aberration (CNV/CNA) callers, and those that detect non-reference DNA adjacencies, known as structural variant (SV) or breakpoint callers. While CNAs and SVs are merely two different viewpoints of the underlying genomic rearrangements, the methods of detection are fundamentally different.

Here, we address the problem of SV detection and show that breakpoint detection alone is

insufficient for the comprehensive characterisation of somatic genomic rearrangements that

occur in cancer. A third genomic rearrangement primitive is essential: single breakends.

The Variant Call Format (VCF)[5] defines a single breakend as a breakpoint in which only one

side can be unambiguously placed. This can occur due to one of two reasons. Firstly, the

sequence on one side of the breakpoint could be absent from the reference. Either non-

reference sequence could be present due to the integration of foreign DNA (e.g. provirus) or the

reference could lack sequence present in the sample. Secondly, breakpoints into highly

repetitive regions cannot be unambiguously placed. Single breakends allow the representation

of such breakpoints. Such rearrangements are common in cancer and by reporting single

breakends the rearrangement landscape of regions previously considered inaccessible to short

read sequence can be explored.

Short read-based SV detection algorithms identify breakpoints by finding clusters of reads that

do not support the reference allele. Typically these use discordant read pairs [6], or split reads[7],

with some callers also considering reads with unmapped mates [8] and soft-clipped reads [9]. More

sophisticated callers incorporate assembly either through de novo assembly [10], targeted

breakpoint assembly [11], or breakend assembly [12]. These callers report breakpoints, that is,

novel adjacencies. When reads cannot be unambiguously mapped on either side, a breakpoint

call cannot be made and information is lost. Some callers have attempted to address this by

considering multiple alignment locations for each read [13] but this only works for regions with a

small number of potential alignment locations and has proven impractical for general use. Single

breakend calling has the potential to improve short read caller sensitivity above the 50%

reported in recent benchmarking[3–5].

As we move closer to a world in which the CNA and SV primitives can be reliably detected,

accurate interpretation of the causative biological events becomes increasingly possible by

integrated analysis of this knowledge. While progress has been made on derivative

chromosome reconstruction using long reads [14], reconstruction of complex events such as

chromothripsis has been problematic for short reads [15,16]. To date, SV phasing has been used to

reduce the complexity of reconstruction for long read based approaches [17] but has not been

done by short read callers. The ability of phase somatic structural variants is limited by the read

length and, for short read data, by the library fragment size - typically less than 500bp.

Here, we demonstrate the power of single breakend variant calling using GRIDSS2 - a somatic

structural variant caller that reports single breakends and phases nearby structural variants.

Running GRIDSS2 on 3,782 metastatic solid tumours with matched normal samples from the

Hartwig cohort we show that, due to the high prevalence of somatic breakpoints involving low-

mappability sequences, GRIDSS2 achieves a false negative rate lower than possible with a

traditional breakpoint-only caller. The precision and sensitivity of GRIDSS2 in conjunction with

single breakend variant calling and SV phasing lay a strong foundation for downstream tools

that enable a deeper understanding of the nature of somatic genomic rearrangements.

Results

GRIDSS2 utilises the same high-level approach as the first version of GRIDSS, assembling all

reads that potentially support a structural variant using a positional de Bruijn graph breakend

assembly algorithm[12]. Breakend contigs are then realigned back to the reference to identify

breakpoints and probabilistic structural variant calling is performed based on both the aligned

reads and assembled contigs. Single breakend variant calling uses the same probabilistic

variant calling approach as breakpoint calling, but instead of split reads, discordant read pairs,

and assembly contigs with chimeric alignments support, single breakends are called based on

soft-clipped reads, reads with unmapped or ambiguously mapping mates, and assemblies with

unmapped or ambiguously mapping breakend sequence (Figure 1a). SV phasing is performed

based on assembly contigs and the presence of transitive calls (Figure 1b). SVs are phased cis

if an assembly spans both breaks or a transitive call is found, and phased trans if an assembly

involves one SV but supports the reference at the other. Since assembly contig length is limited

by the library fragment size only nearby SVs can be phased. GRIDSS2 includes a 16-step

somatic filter specifically tuned for deeply sequenced tumour/normal samples.

a)

- Discordant read pairs
- Soft-clipped reads
- Split reads
- Reads with unmapped mates
- Indel-containing reads

Assemble

Breakend assembly of fragments
sharing anchoring alignment

Align to reference

Single alignment

Zero alignments
Multiple alignments

Split alignment

Breakpoint call

Single breakend call

Phased breakpoints

Annotate unplaced sequence with:
- candidate reference genome locations
- RepeatMasker repeat type and class
- viral sequence matches

b)

Spanning assemblies
phase adjacent calls cis

Imprecise transitive calls replaced by cis phasing of precise calls

Figure 1: GRIDSS2 overview. a) contigs are assembled from a single locus of reads

mutually supporting the same putative break junction. If the other side cannot be

uniquely determined, the contig supports a single breakend call at the break junction

position. If different portions of the contig sequence uniquely align to different genomic

loci, the assembly supports multiple cis phased breakpoints. b) Nearby structural

variants will have discordant read pairs spanning across multiple breakpoints. These

generate spurious transitive calls that are collapsed into the underlying breakpoints,

phasing them cis.

Benchmarking performance

To estimate precision and sensitivity of GRIDSS2, we used a recently generated "gold standard"

somatic SV truth set for the COLO829 melanoma cell line and the COLO829BL cell line, which

was derived from a normal cell from the same individual, using a combination of Illumina,

PacBio, Oxford Nanopore, 10X Genomics linked reads, and optical mapping followed by

targeted capture and PCR-based validations and manual curation [18]. To test sensitivity and

reproducibility, we ran GRIDSS2, Manta[11], svaba[19], and novobreak[20] on 3 independent

sequencing replicates of the COLO829T/COLO829BL matched tumour-normal cell lines

sequenced to a depth of 100x tumour and 40x normal coverage. GRIDSS2 achieved an

average sensitivity/precision of 94%/83% compared to 88%/52% for Manta, 75%/11% for svaba

and 70%/7% for novobreak (Figure 2a).

To evaluate performance at lower sequencing depths and sample purity, we use in-silico

downsampling and mixing to simulate a matched normal at 40x and a 60x tumour sample at

8%-100% purity corresponding to 5x, 10x, 15x, 20x, 25x, 30x, 45x, 50x, and 60x effective

tumour coverage. Above 10x effective tumour coverage GRIDSS2 achieved higher sensitivity

and specificity than the benchmarked callers. At 10x and below, GRIDSS2 retained higher

precision, but at lower sensitivity than Manta or svaba (Figure 2a).

Figure 2: Somatic benchmarks. a) COLO829T/BL tumour and blood cell lines were

sequenced in triplicate to 100x/40x. In-silico purity downsampling was performed at 40x

normal, and 60x tumour coverage. Results are compared against a PCR validated

somatic truth set generated from multiple sequencing technologies. b) GRIDSS2/Manta

validation results on 13 patient samples for 50bp+ events. c) GRIDSS2/Strelka

validation results for 32-50bp events. d) False negative rate (FNR) inferred from the

presence of SVs copy number transitions broken down by magnitude of copy number

change for 60 PCAWG samples. Comparison is between GRIDSS2/PURPLE and the

PCAWG consensus call set. e) Inferred FNR for 3,782 100x tumour samples from the

Hartwig cohort. Single breakend variant calling is crucial to the low FNR in this cohort. f)

Per sample counts of 32-100bp somatic tandem duplications in the Hartwig cohort.

These mutations are enriched in colorectal cancer and associated with ATM driver

mutations. g) Size distribution of small (32-100bp) tandem duplications across the

Hartwig cohort. This is a distinct signature not associated with microsatellite expansion.

h) Comparison of expected vs actual copy number changes for the Hartwig cohort. SV

inferred and actual copy number changes are closely correlated.

Validation on patient samples

To further validate somatic performance, we performed independent validation of GRIDSS2 and

manta breakpoint calls from 13 patient tumor samples from the Hartwig cohort [2,11,19,20] with a

high burden of structural variants. Since the default minimum reported event sizes of GRIDSS2

and Manta are 32 and 50bp respectively, we compared 32-50bp events to the short indel caller,

Strelka [21]. We used a hybrid capture approach with target probes flanking and overlapping

break-junctions to independently validate over 5,000 calls identified by any tool. 3,403 of 3,666

(93%) GRIDSS2 calls were validated compared to 2,685 of 4,299 (65%) for Manta (Figure 2b).

Of the private Manta calls not found by GRIDSS2, just 230 of 1777 (13%) were validated

compared to 836 of 1031 (81%) GRIDSS2 private calls. Imprecise (that is, not base-pair

accurate) Manta calls validated at a rate (40/288, 14%) similar to Manta private calls, whereas

GRIDSS2 reports only precise somatic SV. No imprecise GRIDSS2 calls passed somatic

filtering, whereas All validated imprecise Manta calls were called by GRIDSS2 precisely. In the

32-50bp range, 329 of 343 (96%) of GRIDSS2 calls validated against 142 of 182 (78%) for

Strelka (Figure 2c). 95% (219 of 232) of 32-50bp calls private to GRIDSS2 were validated,

compared to 47% (35 of 74) for Strelka. Notably, GRIDSS2 finds many short duplications of 32-

100 bases which are largely missed by both Strelka and Manta.

Novel somatic short duplication signature

In addition to reidentifying known kilobase and megabase length duplication signatures, we find

a signature consisting of short 32-100bp non-microsatellite tandem duplications (Figure 2f).

There is a median of 4 of these short (32-100bp) duplications per sample (Supplementary

Figure 1). They are not correlated with larger duplications (R=0.08), or total breakpoints

(R=0.10). Enrichment of samples with 15 or more short duplications is positively associated with

colorectal cancer (Figure 2g) (q=1.2 x 10^{-9}) and driver mutations in PARK2 (q=0.0003) and ATM

(q=0.008). Across the Hartwig cohort, 23 samples have driver mutations involving the disruption

of a tumour suppressor caused by small duplications.

These short tandem duplications are too large to be reliably called by most somatic indel callers,

but too short to be reliably called by many SV callers. In part this is due to the weak read pair

signal due to the short variant length, but also since most callers do not report variants shorter

170 than 50bp threshold used for variant databases such as dbVar. Popular callers such as lumpy [22]

171 and delly [23] do not call duplications shorter than 100 and 300bp respectively [4], and no

172 duplications shorter than 300bp were included in the PCAWG consensus call set[1].

173 Cohort-level FNR/FDR estimation using copy number consistency

174 Structural variant and copy number calls are intrinsically related. Any breakpoint must have

175 either a compensating breakpoint (for example, as with inversions), or a copy number change at

176 that SV position. Using this principle, we can estimate a false negative rate (FNR) from the

177 number of unexplained copy number transitions. To generate matching SV and copy number

178 calls, we ran GRIDSS2 and PURPLE [2] on 1,528 samples from the PCAWG WGS cohort and

179 compared results with the state-of-the-art PCAWG consensus call set[24]. Copy number

180 transitions in or within 100kb of centromeres or a gap in the reference genome were excluded.

181 Across the 1,528 samples, GRIDSS2 identified breakpoints for 84% of copy number transitions

182 and single breakends for a further 4.7%, with an estimated 11.2% FNR. The PCAWG

183 consensus call set identified breakpoints for 72% of copy number transitions (28% FNR). When

184 restricted to clonal copy number transitions, the estimated FNR for the PCAWG consensus

185 dropped to 14.2% and GRIDSS2 to 9.36% (Figure 2d), indicating robust subclonal GRIDSS2

186 performance.

187 To evaluate GRIDSS2 on high quality, deeply sequenced samples, GRIDSS2 and PURPLE

188 were run on 3,782 40x normal/100x tumour samples from the Hartwig cohort. Excluding those

189 occurring within 1kb of a gap in the reference genome, 153,231 of 1,954,548 (7.0%) copy

190 number transitions in the Hartwig cohort were explained only by single breakend variants and

191 68,171 (3.1%) lacked a corresponding GRIDSS SV (Figure 2e). The higher rate of single

192 breakend calling can be attributed to GRIDSS2 conservatively calling single breakends and the

greater sequencing depth in the Hartwig cohort. The 7.0% of copy number transitions in the

Hartwig cohort explained by single breakend variant calls represents a lower bound for the FNR

of an exclusively breakpoint-based caller. A FNR of 3.1% suggests that, on this cohort,

GRIDSS2 achieves a FNR lower than that possible for a breakpoint-based caller.

To demonstrate that this reduction in FNR does not come at the cost of a high false discovery

rate (FDR), we compared the change in copy number to the change expected based on the

variant allele fraction (VAF). For isolated breaks, the change in copy number should match the

variant copy number inferred from the variant allele fraction. Using a 3000bp threshold to ensure

at least one full 1kbp copy number bin between SVs, we find that the VAF-inferred SV copy

numbers reported by GRIDSS2 are consistent with the copy number changes with no

systematic bias in the VAF (Figure 2h). This trend remains true for subclonal variants although

the false discovery rate does go up. Assuming variants with a copy number change of less than

0.1 and a VAF inferred copy number of at least 0.25 are false positives, GRIDSS2 isolated SV

calls have an estimated FDR of 5.4%, with 74% of these subclonal, and single breakends

having twice the FDR of breakpoints. Extrapolating these to the rest of the cohort gives an

overall estimated FDR of 3.3%.

Resolving somatic centromeric rearrangements

Although only one side of single breakend variant calls can be uniquely placed, the assembled

sequencing flanking the break can be used to classify integrated provirus, mobile element

transposition, rearrangements involving centromeric and telomeric sequence, and other events.

RepeatMasker annotation reveals that the majority of somatic single breakend calls are caused

by SINE Alu, LINE L1HS insertions or rearrangements involving centromeric sequence, a

pattern shared between both the Hartwig and PCAWG cohorts (Figure 3a, Supplementary

Figure 2). Breakend assembly lengths for SINE single breakends are typically shorter than

150bp as assemblies longer than this can typically be resolved into breakpoint calls. Similarly,

the polyA repeat motif characteristic of LINE translocations[25] is also found in the shorter

breakend assemblies. Such assemblies are short as the de Bruijn graph assembler used

truncates assemblies at unresolved repeat loops and assemblies able to span the polyA tail are

able to be resolved as breakpoints.

91% of the Hartwig cohort samples contain at least one copy number transition occurring in

centromeric sequence. Being able to resolve the partners of the centromeric breaks explaining

these copy number changes is critical to the accurate reconstruction of the derivative

chromosomes. Single breakends into ALR/Alpha and HSATII centromeric repeats are able to

give significant insight into the nature of these centromeric breaks. As each human centromere

has a slightly different dominant repeat sequence, a mapping between each centromeric single

breakend and their most likely centromeric breakpoint partner is possible. To do this, we aligned

the single breakend sequences containing a centromeric or peri-centromeric repeat against the

hg38 reference genome using BLAT, annotating each with the most likely centromeric partner.

Using this approach, we were able to explain 5,614 of 11,996 (47%) centromeric copy number

changes, implying that approximately half of centromeric rearrangements are centromere to

centromere, and the remainder centromere to non-centromeric sequence. Of the 21,587

centromeric single breakends detected 3,148 (15%) had no copy centromeric copy number

change, 6,850 (32%) had no copy number change but had multiple single breakends linked to

the same chromosome, 3,358 (16%) had a single breakend associated with a centromere with

copy number change, and the remaining 8,231 (38%) associated with a centromere with copy

number change with multiple breakends mapping to that centromere in that sample.

Figure 3 Classification of single breakends. a) RepeatMasker annotations indicate the

majority of somatic single breakends are due to mobile element translocations, or

centromeric breaks. b) Most likely centromere for single breakends containing

centromeric or peri-centromeric repeats based on realignment of breakend sequence to

hg38. Shading indicates whether prediction is consistent with the copy number change

across the centromere. Chromosome 1 has an excess of inter-chromosomal breaks to

centromeric sequence. Chromosomes 13, 14, 15, 21, 22 have insufficient non-gap p-

arm sequence for a centomeric copy number change to be called. c) Location of single

breakends to centromeric sequence and corresponding centromere. Chromosome 1

has an excess of inter-chromosomal breaks to centromeric sequence, particularly to 5

and 19. d) Location of single breakends connected to centromeric sequence on the

same chromosome. Red events left of the chromosomes are intra-chromosomal, and

yellow events to the right are inter-chromosomal.

255 Novel centromeric break signature

The centromeric single breakend rate can be further broken down by chromosome (Figure 3b)

and based on the location of the single breakend (Figure 3c). Chromosome 1 is a clear outlier

with an overabundance of centromeric inter-chromosomal rearrangements, particularly to

chromosomes 5 and 19. Although the high level of sequence similarity between the

centromeres of 1, 5, and 19 [26] could be a cause of false positive predictions, this relationship

holds even when restricting the analysis to single breakends with an associated centromeric

copy number change (Supplementary Figure 3), implying that the centromeric similarity between

1, 5 and 19 results in an increased rate of centromeric rearrangements between these

chromosomes. In contrast, the lack of copy number supported single breakends to chromosome

13,14, 15, 21, and 22 centromeres is an artifact caused by missing p arm copy number due to

gaps in the reference genome. Similarly, the centromeric sequence homologies between 13, 14,

21, and 22 combined with the lack of confirmatory copy number support, make it difficult to

determine how much of the high inter-chromosomal centromeric rearrangement of chromosome

13 is due to misattribution of rearrangements to other chromosomes.

In general, intra-chromosomal single breakends to centromeric sequences occur close to the

centromere (Figure 3d), with this effect less pronounced for inter-chromosomal breaks.

Chromosome 1 is enriched for inter-chromosomal breaks, particularly to chromosomes 5 and

19, with inter-chromosomal breaks from these chromosomes to the centromere on 1

(Supplementary Figure 4) occurring in a pattern similar to the intra-chromosomal breaks of other

chromosomes.

Somatic phasing

The breakend assembly approach taken by GRIDSS2 also enables the assembly-based

phasing of nearby variants. When two structural variants occur in close proximity, they can be

phased as cis if the contig aligns across both, and trans if the contig aligning across one aligns

to the reference sequence at the other (Figure 4a). Segments shorter than 30bp are not typically

uniquely alignable by BWA and unaligned short DNA segments are treated as insert sequences

of an SV connecting the longer flanking segments. Since breakend assembly contig lengths are

limited by the fragment size distribution of the DNA library sequenced, only nearby variants can

be phased.

For the Hartwig cohort, variants could be phased up to around 500 base pairs. We found that

multiple nearby somatic structural variants are frequent, with 22% of all structural variants

having an adjacent variant within 1,000bp. This is far in excess of the 0.02% expected if the

breakpoints were uniformly randomly distributed (Figure 4c). Of these, GRIDSS2 was able to

phase 70% (Figure 4b) with 72% cis and 28% trans. This distribution is recapitulated in the

1,528 PCAWG samples and LINX classification of these structural variants indicate that that

phasable breakpoint clusters occur predominantly in LINE translocations (due to target site

duplication, and highly active donor elements) and highly complex rearrangement events

(Supplementary Figure 5). This phasing information greatly assists downstream derivative

chromosome reconstruction, as it exponentially reduces the number of possible paths through

the breakpoint graph.

Figure 4: Structural variant phasing. a) Phasing of structural variants can be determined

when breakend assembly contigs span multiple breakpoints. b) The majority of variants

within 600bp can be phased using breakend assembly. c) Somatic SVs are highly

clustered with 22% of all SVs in the Hartwig cohort potentially involving a DNA fragment

of 1kbp or less.

Impact on complex event resolution

To demonstrate the impact on downstream analysis of complex somatic genomic

rearrangements, we ran LINX [27], a rearrangement event interpretation and classification tool, on

the Hartwig and PCAWG cohorts. To fully resolve complex rearrangements structural variants

must be chained together to reconstruct the relevant portions of the derivative chromosomes

rearranged by the event. If there are errors in the SV call set, it is likely that many complex

events will not be able to be fully resolved. To evaluate the impact of FNR on this

reconstruction, we evaluated the portion of SVs resolved into long chains for the PCAWG and

Hartwig cohorts. In addition, we simulated the effect of increasing FNR by subsampling the

Hartwig call set (Figure 5a). 5.0% of SVs in the Hartwig cohort were reconstructed into chains of

20 SVs or more. Increasing the FNR reduces this to 3.6% of SVs at 5% FNR, 1.5% at 10%

FNR, 0.6% at 15% FNR, and 0.25% at 20% FNR. We previously estimated the PCAWG cohort

FNR at 11.2% and we find that the 1.27% of SVs in chains of 20 SVs or more closely match the

1.29% we expect from a simulated downsampling of the Hartwig cohort. This implies that the

PCAWG and Hartwig pancaner cohorts have a broadly similar composition of complex

rearrangements and the differences observed are primarily technical in nature. Small

improvements in FNR result in large increases in the ability for downstream tools to resolve

complex events. A sub-5% FNR is critical for large event reconstruction.

SV phasing can be critical to the correct interpretation of complex events. For LINX, SV phasing

is a critical first step in the chaining of SVs. Of the 486,632 links in the chains resolved by LINX

in the Hartwig cohort, 100,007 (21%) were due to GRIDSS2 SV phasing. For the PCAWG

cohort, 13,212 of 107,952 links (12%) were resolved by GRIDSS SV phasing, with the

difference primarily driven by shallower coverage and shorter library fragment sizes resulting in

shorter assembly contig lengths (Supplementary Figure 2), and the higher FNR. In some cases,

apparently complex events can be resolved to simple events containing additional short DNA

fragments purely through SV phasing (Figure 5b).

Finally, we use single breakend repeat annotations to identify instances of chromothripsis

overlapping centromeres. In the Hartwig cohort, LINX identifies 270 complex events with at least

10 breakends to centromeric sequence, 17 of which could be fully chained (Figure 5c). The

large number of events with many centromeric single breakends indicates a previously

unexplored level of centromeric involvement in complex rearrangements worthy of further

investigation.

Figure 5: Complex rearrangement interpretation. a) Impact of false negative rate (FNR)

on complex event resolution. The y-axis indicates the portion of structural variants that

form part of a resolved chain of SVs at least as long as the chain length indicated on the

x-axis. LINX results for GRIDSS2 calls on the Hartwig and PCAWG cohorts are shown

along with simulated results from downsampling the Hartwig cohort to the specified

FNRs. A low FNR is essential to accurate complex event resolution. b) Circos plot of

SMAD4 driver deletion event. The interpretation of this deletion is confounded by the

presence of 3 short fragments at the breakpoint. This event can be fully resolved by

GRIDSS2 SV phasing. Circos tracks from innermost to outermost are: single breakends

(open white circles) and breakpoints, LOH, copy number, connected segments, genes,

chromosome. b) Circos plot of chromothripsis overlapping centromeric sequence. This

event spans across the chromosome 7 centromere. A subset of the chromothriptic

fragments have been inserted into chromosome 4. Each SV chain is represented in a

different colour.

Discussion

Through cell line, patient validation, and cohort-level comparisons, we have shown GRIDSS2

has excellent somatic performance above 10x effective tumour coverage. The identification of a

small (32-100bp) duplication signature by GRIDSS2 highlights the importance of robust

software tested across a wide range of variant types and sizes. The presence of a signature

overlapping the widely accepted but arbitrary 50bp threshold separating indels from structural

variants suggests it is time to reconsider this threshold as the minimum reported event size for

structural variants.

Explicitly reporting and handling of single breakend variants represents a significant conceptual

advancement in the treatment of genomic rearrangements. Even though only the high-

mappability side can be unambiguously placed, sequence classification of the low-mappability

side produces useful results and meaningful insights. The identification of frequent somatic

centromeric rearrangements demonstrates the utility single breakend variant calling has in

regions of the genome traditionally considered inaccessible to short read sequencing. Single

breakend variant calling provides a framework for the reliable detection of LINE integration

without a specialised caller [28], for the detection of centromeric and telomeric viral integrations [29],

and for an entire ecosystem of tools that explicitly model the ambiguity they represent. As single

breakends comprise 7.0% of GRIDSS2 calls in the Hartwig cohort, any purely breakpoint-based

caller must have a false negative rate of at least 7.0%. GRIDSS2's 3.1% FNR may thus be

impossible to achieve for any breakpoint-based caller, at least for this cohort.

One biologically significant finding coming from GRIDSS2's ability to phase structural variants is

the degree to which somatic structural variants are clustered. In the Hartwig cohort of metastatic

solid tumours, 22% of somatic structural variants potentially involve DNA fragments of less than

1000bp with GRIDSS2 able to phase 70% of these. Long read sequencing is considered the

gold standard for structural variant phasing and phasability is indeed better: 10kb long reads

increase this theoretical phasability to 31%. The high indel error rate of PacBio and ONT

sequencing presents a current drawback of long read sequencing: simple long read based SV

detection approaches are likely to misidentify complex rearrangements involving nearby cis-

phased SVs. Without HiFi sequencing or error correction prior to alignment, the short DNA

segments between the SVs will be unmappable and the long read caller will report a transitive

call between the flanking segments as outlined in Figure 1b. On COLO829, we found 5

instances (of 67 true positives) in which GRIDSS2 based on short-read sequencing was able to

correctly place a short DNA segment that the three long-read callers were not. Care must be

taken when comparing or combining short and long read variant calls to ensure the different

representations of the same event are reconciled and cis phased. GRIDSS2's ability to phase

breakpoints involving short DNA fragments is of great utility to downstream rearrangement

event classification and karyotype reconstruction as it exponentially reduces the number of

possible paths through the breakpoint graph. The highly clustered nature of somatic SVs means

that short read sequencing is surprisingly competitive when it comes to phasing somatic

variants. Sophisticated analysis and interpretation of somatic genomic rearrangements does not

necessarily require long read sequencing.

Single breakend variant calling enables a sensitivity and specificity unprecedented amongst

short read-based somatic structural variant callers, facilitating the resolution of highly complex

rearrangements. While breakpoints and copy number segments are widely adopted

fundamental genomic rearrangement signals, single breakends have been hitherto unutilised.

Their introduction enables the ambiguities present in low mappability regions to be explicitly

modelled without compromising FNR or FDR and their potential extends far beyond the

examples presented here. GRIDSS2 demonstrates that single breakend variant calling is

essential to the comprehensive characterisation of somatic structural variation from short read

sequencing data. Combining single breakend variant calling and structural variant phasing with

low FNR and FDR, GRIDSS2 represents a foundation upon which sophisticated somatic

analysis can be performed.

Methods

411 GRIDSS2 extends the GRIDSS[12] software suite with additional features, tools and capabilities.

412 GRIDSS2 is composed of the following 5 phases: (i) preprocessing, (ii) assembly, (iii) variant

413 calling, (iv) annotation, and (v) somatic filtering.

Preprocessing

GRIDSS2 takes one or more aligned samples in the SAM/BAM[30] file format. These files are pre-

processed on a per-file basis and all reads supporting a structural variant are extracted, and all

fields or tags referring to another record are corrected. Reads with chimeric alignments (i.e. split

reads), reads with a soft or hard clipped alignment CIGAR of at least 5bp, read pairs in which

only one read is mapped, and discordant read pairs are extracted. The library insert size

distribution is estimated from the first 10,000,000 reads using picard tools

(http://broadinstitute.github.io/picard) and read pairs considered discordant if they fall outside

the 99.5% distribution of fragment size lengths. The clipped bases of soft clipped non-chimeric

reads are realigned to the reference genome using bwa mem [31] and converted to a chimeric

split read alignments if an alignment is found. Inconsistencies in the mate chromosome and

position are corrected (since tools such as GATK indel realignment adjust read alignment

positions without updating the mate record), hard clips converted to soft clips, the NM, SA, MC,

MQ tags recalculated, and the R2 tag is populated. Improving performance over GRIDSS,

GRIDSS2 performs this in a two-pass manner with samtools[30] used for name/coordinate sorting

the output of the first/second pass respectively.

As with GRIDSS, reads with low alignment sequence entropy and reads with a mapping quality

(mapq) less than 20 (c.f. mapq<10 GRIDSS) are treated as unmapped, soft-clipped reads with

clipped sequence having high homology with standard adapter sequences are ignored, reads

marked as duplicates, and regions above 50,000x (c.f. 10,000x) coverage are ignored. Read

alignments containing an insertion or deletion under 5bp are considered consistent with the

reference.

Assembly

GRIDSS2 uses the same genome-wide positional de Bruijn graph break-end assembler used by

GRIDSS. Reads are split into kmers and associated positions based on the anchoring

alignment: kmers from split reads must be assembled only with kmers at the positions inferred

by the anchoring alignment, and kmers of unmapped mate reads are assembled at any position

consistent with the library fragment size distribution and the anchoring read alignment position.

For assembly purpose, split reads and indel alignments are considered multiple soft clipped

alignments, and discordant read pairs are treated as multiple read pairs with one read aligned.

The output of the assembly is a set of 'soft-clipped' contigs with anchoring bases supporting the

reference, and non-reference bases supporting a putative breakpoint at a given position. This

contig is iteratively realigned to the reference using bwa mem and converted to a split read

alignment. Assemblies longer than the 1.5x maximum fragment size distribution, as well as

assemblies supporting the reference sequence are ignored. Assembly alignments with a mapq

of less than 20 are treated as unmapped. GRIDSS2 introduces a number of refinements to the

assembly calling process.

Assembly support is tracked per base pair. Fragments are considered to support a breakpoint

only if the fragment support spans at least one base pair beyond any breakpoint homology on

both sides. This ensures that when a single contig spans both a germline indel and a somatic

SV, the fragments originating from the matched normal sample will not be considered as

supporting the somatic breakpoint. This also improves variant allele fraction calculations in

regions of complex rearrangement.

GRIDSS2 performs compound realignment of the entire assembly contig. BWA is used to align

the entire assembly contig. Assembly contig bases which are soft clipped in the primary

alignment reported by BWA are fed back to BWA for realignment. This process is repeated until

either all bases are aligned, or no alignment can be found for the remaining bases. Assembly

contigs that do not overlap with the locus of origin of the assembly are filtered out. To ensure

that valid assemblies are not unnecessarily filtered, GRIDSS 2 includes both reads of each

fragment in the assembly, and up to 300bp of anchoring reference-supporting sequence is

included in the assembled contig. The remaining contigs are treated as split read alignments. To

rectify over-alignment of the primary alignment location in the presence of imperfect breakpoint

microhomology, the bounds of each split are adjusted to minimise the edit distance to the

reference. The originating alignment is tracked using OA SAM tag and contigs that do not

partially align to the originating assembly location and strand are filtered.

gridss.SoftClippedToSplitReads invokes bwa, -L 0,0 is added to the command line to remove

the soft-clipping alignment penalty. This prevents 1bp non-template inserted sequences being

over-aligned and reported as clean breakpoints with a flanking SNV.

Worse-case assembly performance has been improved by adding an assembly graph path

count threshold. Generating 3 assembly contigs with more than 50,000 alternative paths

through the assembly graph without advancing the assembly window will flush the assembly

window. The maximum assembly window size has been reduced by 2.5x and more aggressive

assembly read downsampling in high coverage regions is performed.

The presence of a contig with at least three non-overlapping alignments results in the

breakpoints supported by that assembly being phased cis. If the initially soft-clipped portion of

an assembly realigns across one breakpoint but not another, these breakpoints are phased

trans.

Variant Calling

Breakpoints are called using a probabilistic model based on the empirical distribution of CIGAR

operators, the library fragment size distribution, and mapping rate. Each read/read pair is given

a phred-scaled quality score based on the mapping quality and the probability of encountering

that read/read pair given the library empirical distribution. Split reads and soft clipped reads use

the distribution of soft clipping CIGAR operators. Discordant read pairs use the discordant

mapping mate if distal or the library fragment size distribution if falling within the range reported

by Picard tools CollectInsertSizeMetrics. Reads with unmapped mates use the unmapped mate

fragment mapping rate, and indels based on rate of alignments with insertion/deletion CIGAR

elements of matching lengths. As with GRIDSS, split reads and breakpoint-supporting

assemblies incorporate the mapping quality scores on both sides of the supported break.

The key novel feature of the GRIDSS2 variant calling processing is the reporting of single

breakend variants. Single breakends variants are called based on supporting soft clipped reads,

assembly contigs, and reads with unmapped mates. Single breakend calling uses the same

two-pass approach as breakpoint calling with all maximum cliques first calculated, then

evidence uniquely assigned to the highest scoring clique.

In addition to single breakend variant calling, the variant caller has been improved by: reducing

the default minimum called event to 10bp; preferentially allocating reads to variants supported

by an assembly containing the read; requiring two supporting fragments to call a variant; and

excluding inversion-like breakpoints from the minimum variant size filter to prevent filtering of

foldback inversions.

Annotation

GRIDSS2 provides a full per-sample breakdown of all supporting evidence for each variant through the following VCF INFO and FORMAT fields:

- AS, RAS, CAS: counts of assembly contigs supporting a breakpoint originating locally, from the other side of the breakpoint, and from another location respectively. CAS assemblies support multiple variants and provide linking information about those variants.

- ASSR, ASRP: total number of split/soft clipped/indel-containing reads, and discordant read pairs/reads with unmapped mate contributing to any breakpoint-supporting assembly contig at the breakpoint location. Note that read/read pairs that are assembled into a contig but whose interval of support does not span the breakpoint are not counted. The interval of support for a read/read pair is defined as the interval between the first and the last contig base for which that read/read pair contributed to the assembly.

- SR, RP, IC: counts of split reads and discordantly mapped read pairs, and indel-containing reads that directly support the breakpoint.

- BA: counts of assembly contigs support a single breakend at this position. Such contigs are aligned only to the local breakend with the breakend sequence either aligning ambiguously, or unable to be aligned to the reference genome by bwa.

- BASSR, BASRP: total number of split reads or soft clipped reads, and discordant read pairs or reads contributing to any breakend-supporting assembly contig at the variant location.

- BSC, BUM: counts of soft-clipped reads, and reads with unmapped mates at the variant location

- ASQ, RASQ, CASQ, SRQ, RPQ, IQ, BAQ, BSCQ, BUMQ: corresponding quality score contribution for the supporting evidence.

- QUAL, BQ: total contribution to the breakpoint/breakend quality score.

- BANRP, BANSR, BANRPQ, BANSRQ: counts of read pairs/split reads not supporting this breakpoint but assembled into a contig that supports this breakpoint and their corresponding assembly quality score contribution.

- REF/REFPAIR: count of reads/read pairs spanning the local variant position that support the reference allele. Only reads/read pairs that span across the breakpoint microhomology interval (if present) are counted.

- VF/BVF: count of unique fragments supporting the breakpoint/breakend. By tracking unique fragments supporting the variant, a more accurate variant allele fraction can be calculated. This approach prevents double-counting of discordantly mapped fragments for which one of the reads contains a split read alignment. A fragment can support a variant either directly through split read, soft clipped read or discordant alignment of a read pair, indirectly through incorporation of one or both of the constituent reads in an assembly supporting the variant, or both directly and indirectly.

- RF: count of unique fragments supporting the reference allele.

- CQ: variant quality score prior to evidence reallocation.

- BEALN: Potential alignment locations of breakend sequence as determined by *gridss.AnnotateInsertedSequence.*

- BEID, BEIDL, BEIDH: identifiers of assembly contigs and the corresponding local and remote alignment base offsets. Single breakend variants do not have a remote breakend, and

only breakpoint variants include breakpoint-supporting assemblies. Variants containing the

same BEID are phased cis.

● CIPOS: For IMPRECISE variants, CIPOS encodes the interval in which the breakpoint could

occur and for precise variants, CIPOS encodes the homology interval.

● CIRPOS: corresponding CIPOS of the remote breakend.

● IHOMPOS: interval of inexact homology. A Smith-Waterman alignment of the breakpoint

sequence against the reference sequence is performed at both breakends. The reference and

breakpoint sequence are extended 300bp from the break on either side with the reference

extended an additional 10bp to account for potential indels in the alignment. The homology

length is the length that the sequence alignment could be extended from the common sequence

into the breakpoint/reference sequence. Alignments containing a soft clip on the common

sequence side are classified as alignment errors and ignored. The SSW library[32] is used for

which we implemented a JNI wrapper. Alignment scored 1, -4, 6, 1 for match, mismatch, gap

open, and gap extend respectively which correspond to bwa mem alignment scores.

● SC: CIGAR encoding of the anchoring bases that at least one read/read pair/assembly is

aligned to and supports the variant. This is encoded as a CIGAR string with a match for each

anchoring base that provides support for the variant call, XNX for the interval over which the

breakpoint could occur (due to microhomology or an imprecise call), and a deletion CIGAR

element for any intervals over which there is no support (such as a small flanking deletion).

Variants with an anchoring SC 10bp further from the break than a nearby variant are considered

to be phased trans.

● SB: Strand bias of the reads supporting the variant. 1 indicates that reads would be aligned

to the positive strand if the reference was changed to the variant allele. 0 indicates that reads

bases would be aligned to the negative strand if the reference was changed to the variant allele.

Strand bias is calculated purely from supporting reads and exclude read pair support since

these are intrinsically 100% strand bias. Note that reads both directly supporting the variant and

supporting via assembly will be double-counted. Both breakpoint and breakend supporting

reads are included.

● IMPRECISE, HOMLEN, HOMSEQ, PARID, EVENT, CIEND, END, and SVTYPE fields

carry their usual meaning as per the VCF file format specifications.

● MQ, MQN, MQX, BMQ, BMQN, BMQX mean, min, and max MAPQ score of

reads/assembly contigs providing breakpoint/breakend support.

After initial annotation, *gridss.AnnotateInsertedSequence* aligns any single breakend sequences

or non-template inserted breakpoint sequences to an arbitrary reference genome and adds an

annotation reporting the alignment location. Integrated viral sequence is identified by aligning to

a reference of viral sequences. By default, the same reference as the input files were aligned to

is used. If a RepeatMasker bed file generated by BedOps[33] rmsk2bed is supplied, inserted

sequences will be annotated with the RepeatMasker class and type corresponding to the

BEALN alignments.

Somatic filtering

By default, GRIDSS2 is a sensitive caller and reports all putative variants supported by at least

two well-mapped reads. To generate a set of high and low confidence somatic call sets, a

somatic filter was developed. Variants with 3% of the supporting reads originating from the

normal, or deletion or duplication breakpoints under 1000bp that have any direct split read

support in the normal, are hard filtered. Variants are classified as low confidence if any of the

following conditions are met: breakend coverage of less than 8 fragments in the normal; allelic

fraction of less than 0.5% in the tumour; imprecise variant call; breakend variants without an

assembly containing at least one discordant read pair; single breakends with a poly-C or poly-G

run of at least 16bp in the breakend sequence; deletion or duplication breakpoints under 1000bp

with a split read strand bias of 0.95 or greater; breakpoints with a microhomology of over 50bp;

breakpoints with an inexact microhomology of over 50bp which are not deletion or duplications

under 1000bp; deletion or duplication breakpoints under 1000bp with no split read support either

directly, or through assembly; breakpoints with no discordant read pair support (either directly,

or via assembly) which are not deletions or duplications under 1000bp; deletion or duplication

breakpoints under 1000bp that have any direct split read support in the normal; 100-800bp

deletion breakpoints with an inexact microhomology length of 6bp or greater; inversion-like

breakpoints 40bp or less that have at least 6bp of microhomology; deletion-like breakpoints

under 1000bp whose length of sequence inserted at the breakpoint is within 5bp of the deletion

length, except those whose edit distance to the deleted bases is at least 0.5 per base, and less

than 0.2 per base to the reverse complement. Breakpoint variants are filtered if either breakend

is filtered.

Somatic variants are panel-of-normal (PON) filtered if a match within 2bp was found in a panel

of normals. The default hg19 was constructed from the 40x coverage WGS matched normals for

3,972 patients from the Hartwig cohort using the *gridss.GeneratePonBedpe* utility. If multiple

samples for a patient existed, only the normal for the first sample was included in the PON.

Variants were aggregated across samples using the default setting of ignoring the FILTER field,

and excluding imprecise calls and breakpoints/single breakends with a QUAL score of less than

75/428.

Viral insertions are annotated using *gridss.AnnotateInsertedSequence*. Single breakend

sequences and non-template inserted sequences that do not have an alignment to the

reference genome were aligned to a set of human viral reference sequences. Viral reference

sequences were obtained from the virus host database [34] and filtered to include only viruses

associated with the homo sapiens taxid of 9606. The viral sequences were then masked using

RepeatMasker with "-no_is -s -noint -norna -species human" parameters. Generation scripts can

be found at https://github.com/hartwigmedical/scripts/tree/master/virus.

Assembly linking: pairs of breakpoints mutually supported by a common assembly contig were

annotated as linked by assembly. For assembly contigs spanning more than 2 breakpoints,

each adjacent pair was linked with a unique identifier to enable unambiguous traversal of the

breakpoint graph.

Transitive linking: chains of precise breakpoint variants were phased trans if an imprecise

spanning transitive breakpoint call could be found. To identify transitive calls, a breadth-first

search over the breakpoint graph was performed. Variants were considered transitive if the start

and end breakends overlapped the start and end breakpoints in a path of precise breakpoint

calls. Paths were limited to 1,000bp and 4 segments, with each segment required to be at least

20bp in length. Paths could not self-intersect. To prevent exponential runtime in highly

rearranged genomes, at most 100,000 paths and at most 1,000 paths per starting breakpoint

were considered.

Simple inversion annotation: pairs of breakpoints with orientations consistent with a simple

inversion were annotated as simple inversions if the matching breakends were within 35bp on

both sides, no other simple event annotation could be applied, and fragments supporting the

constituent variants differed by at most threefold.

Templated insertion annotation: breakend/breakpoint and breakend/breakend pairs were

annotated as simple templated insertions if the breakends had opposite orientations, were

within 35bp, no other simple event annotation could be applied, and fragments supporting the

constituent variants differed by at most threefold.

Reciprocal translocation: breakpoint/breakpoint pairs were annotated as reciprocal

translocations if the breakends on both sides had opposite orientations, were within 35bp, no

other simple event annotation could be applied, and fragments supporting the constituent

variants differed by at most threefold.

Equivalent: variants were annotated as equivalent if variants had a breakend within 5bp of each

other and they shared a common breakend sequence. Breakend sequences were truncated to

the length of the shorter sequence and were considered matching when the per-base edit

distance between breakend sequences was 0.1 or less. For the purposes of this comparison,

the nominal breakend sequence was used for single breakends, and the reference sequence of

the partner breakend was used for breakpoint variants. For breakpoint variants, the length of the

breakend sequence was the maximum of 20 bases, and the width of the interval over which the

fragments supporting the partner breakend had anchoring alignments.

Finally, a quality filter was applied to breakpoint variants with a QUAL score of less than 350

and single breakend variants with a QUAL score of less than 1000. Variants linked to a variant

passing the qual filter other than through equivalence were rescued from the quality filtering and

were considered to have passed regardless of the actual variant quality score. For each input

file, two output files were generated: a high confidence call set containing calls passing all filters

and a low confidence call set containing all calls except those failing the normal support filter or

short events with split read support in the normal.

Independent validation of SV calls

13 samples from the Hartwig metastatic cancer cohort were selected for capture panel

validation of the structural variant calls. Each variant called in GRIDSS2 was compared with

variants called from Manta (for variants longer than 50 bases) and/or Strelka (for variants from

32-50 bases in length) to determine if the variant is shared or private. Variants were marked as

matching GRIDSS2 if the start and end chromosomes and orientation both matched and start

end positions (including confidence intervals) were within 20 bases of each other.Hybrid capture

probes were created for each of the shared and private variants. For each breakpoint variant 3

probes of 120 bases each were created: Two reference probes leading up to the breakends

from either side, as well as another SV probe going through the structural variant with the break

junction close to the middle of the probe. The reference probes were designed to end 20 bases

before the start of each structural variant breakend. The SV probe consists of any insert

sequence flanked by equal number of bases from each side of the structural variant. For each

single breakend variant 2 probes were created: One reference probe as described above and

one SV probe which includes no more than 60 bases of the insert sequence with the remainder

coming from the reference leading to the break point.

Together, this created a total of 17,125 capture probes of 120 nt in length, targeting 5,821

break-junctions (see supplementary table 1) which were ordered as custom target capture

probes from Twist Biosciences (catalog ID 100533). For each of the 13 samples, 50 ng input

DNA was used for indexed library construction with enzymatic fragmentation (Twist kit catalog

IDs 100253, 100255 and 100401) according to the manufacturer's protocol. A bead-based size

selection was performed after PCR to remove the remaining larger fragments (>700bp).

Multiplexed hybridization was performed using the Twist Hybridization (ID100254), Blockers

(PN100856) and Wash Kits (PN100214, 100215, 100216)) using Dynabeads™ MyOne™

Streptavidin T1 (Invitrogen PN65604D) following standard Twist protocol. Enriched library

molecules were amplified by PCR for 11 cycles and sequenced on the Illumina NextSeq500 2x

150bp High Output run according to manufacturer's standard protocol.

We created a set of predicted alternate contigs from the shared and private structural variant

calls using the same technique from above for generating the (non-reference) SV probes and

added these to the reference genome. We then mapped each of the reads from the capture

panel output with BWA to a hybrid genome including both the GRCH37 reference genome and

the novel alternate contig.

We assessed the viability of the probes by mapping each of the 120 base SV probes to the

2,000 base alternate contigs to determine its mapping quality. Of the 5,821 SV probes, 80 had 0

mapping quality and 5,377 had a perfect mapping quality of 60. Probes with a mapping quality

of less than 20 were ignored as well as 77 micro-satellite probes that were unable to be

unambiguously validated. Resultant BAM files were examined for evidence of the SV alternate

contigs in the SV source sample BAM as well as the BAM files of each of the other samples as

controls for systemic effects for each of the predicted variants. Specifically, the read depth on

the alternate contig at the variant location was used to assess the validation status of the

variant. SV calls were marked as validated if all the following criteria were met (and not

validated otherwise):

● At least 2 reads were mapped to the alternate contig in the predicted sample

● The support rate for the alternate contig was significantly higher (Poisson model, p=0.001) in
the predicted sample than the maximum of the other 13 samples

● <40 reads in total across all 13 control samples were mapped to the predicted alternate
contig

Comparison to PCAWG

Copy number data was obtained as for the Hartwig cohort running PURPLE [2] with default

settings. PCAWG consensus SV and CN calls were obtained from

https://dcc.icgc.org/releases/PCAWG/consensus_sv, and

https://dcc.icgc.org/releases/PCAWG/consensus_cnv. Copy number transitions were matched

with structural variants with a 100kb margin for PCAWG calls, and a 0bp margin for

GRIDSS2/PURPLE. Copy number transitions in or within 100kb of centromeres or a gap in the

reference genome were excluded from analysis. Copy number transitions matched by both a

single breakend and a breakpoint, were considered breakpoint matches.

Hartwig metastatic tumour cohort

GRIDSS2 was run on 3,782 paired tumour/normal samples from the Hartwig Medical

Foundation cohort of metastatic solid cancers with a 32bp minimum event size. Samples were

aligned with bwa against a GRCH37 reference genome containing only primary contigs. Single

breakend RepeatMasker annotations were obtained by running

gridss.AnnotateInsertedSequence against the UCSC GRCH37 (hg19) RepeatMasker track

downloaded from http://hgdownload.cse.ucsc.edu/goldenpath/hg19/bigZips/hg19.fa.out.gz after

converting to BED formart using bedops *rmsk2bed*.

Hartwig copy number was determined by PURPLE. Since PURPLE infers the copy number of

short segments by the VAF of the flanking SVs, the copy number of these segments do not

represent an independent validation of the SV. As such, FDR was segments from only the

breakpoints in which the copy number of all 4 flanking segments was determined of depth of

coverage and SNP BAF were considered.

Single breakends with a RepeatMasker annotation associated with centromeric or

pericentromeric repeats were considered centromeric single breakends. The matching

chromosome was considered to be the chromosome for which the BLAT[35] based score

score=(1000-((9-floor(Qsize/100))*mismatch+Qcount+Tcount))*min(match/Qsize,1)) is at least

900 and at least 25 higher than the best alignment on a different chromosome when aligning

against hg38. A script for annotating likely centromere can be found in

example/annotate_most_likely_centromere.R in the GRIDSS repository.

Phasability of the Hartwig cohort was calculated by determining, for each break junction, the

length of the DNA segment if it was phased with the first break junction encountered in the

appropriate orientation. Known phasing information was ignored for this analysis. Expected

phasability was calculated by simulating 3,782 randomly fragmented paired genomes with the

same number of break junctions as the corresponding Hartwig sample.

For both the PCAWG and Hartwig cohort, rearrangement event classifications were obtained by

running LINX[28] 1.12 on the GRIDSS/PURPLE outputs. Simulated FNR results were obtained by

random subsampling of the Hartwig GRIDS2 SV calls and breaking LINX chains whenever a SV

was excluded from the subsampling.

COLO829 somatic benchmark

The COLO829T/COLO829BL cell lines (ATCC® CRL-1974™ and 1980™ respectively) were

each sequenced three times to 100x/40x using the HMF workflow [2,4] and aligned against

GRCH37 without alt contigs using BWA 0.7.15. GRIDSS 2.9.3, Manta 1.5.0, svaba 1.1.0, and

smufin 0.9.3 [36] were run with default parameters. Programs were allocated 8,16,16,20 cores

and 32, 32, 50, 500Gb of memory respectively. No smufin results were obtained in any replicate

as smufin failed to complete in the 100,000 CPU hours/3 months wall time allocated.

Call matching was performed using the StructuralVariantAnnotation BioConductor package

(DOI 10.18129/B9.bioc.StructuralVariantAnnotation). A 100bp matching margin was allowed

around the break junction position. Tandem duplication calls matched with insertion calls if the

size difference between the duplication and insertion was within 25bp. False positive calls under

50bp were filtered after matching so as not to penalise a caller reporting an event slightly larger

than 50bp in the truth set, but slightly smaller than 50bp in the call set. If multiple calls in a call

set matched a single truth set call, all except the highest QUAL call were ignored.

762   COLO829 truth set generation

The COLO829 somatic SV truthset was generated using an orthogonal sequencing strategy.

We sequenced the COLO829BL and COLO829T cell lines using Illumina HiSeqX (ENA run

accessions ERR2752449 and ERR2752450 for COLO829BL and COLO829T, respectively),

Oxford Nanopore (ERR2752451 and ERR2752452), Pacific Biosciences (ERR2752447 and

ERR2752448) and 10X genomics (ERR2820166 and ERR2820167). All data is grouped under

ENA study accession PRJEB27698.

Raw data was analysed for structural variants using the following tools:

-   Illumina data was mapped using BWA 0.7.5, SV calling was performed with GRIDSS 2.0.1

and somatic SVs were filtered using gridss_somatic_filter.R.

-   Nanopore data was mapped using NGMLR 0.2.6 and SV calling was performed with both

NanoSV 1.2.0 and Sniffles 1.0.8 separately for COLO829T and COLO829BL. All SVs were

merged with an overlap window of 200 base pairs using SURVIVOR 1.0.6 and SVs not present

in COLO829BL were kept.

-    Pacbio data was mapped using minimap2 2.11-r797 and SVs were called using pbsv 2.1.0

in joint calling mode for COLO829T and COLO829BL. Only SVs with no evidence in

COLO829BL were kept.

-    10X genomics data was processed using Longranger 2.2.2 with default settings for

COLO829BL and somatic mode for COLO829T. SV calls for both cell lines were merged with an

overlap window of 200 base pairs using SURVIVOR 1.0.6 and SVs not present in COLO829BL

were kept.

Somatic SV calls for each technology were merged with an overlap window of 200 base pairs

using SURVIVOR 1.0.6. and all candidate breakpoints were subjected to independent validation

by targeted capture and/or PCR-based approaches. SVs detected with two or more techniques

that failed in these validation experiments were curated by manual inspection of the mapped

reads using IGV [37]. A total of 69 SVs were finally considered as true somatic SVs for

COLO829T.

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Author Contributions

880 DLC designed and implemented GRIDSS2. DLC, PP, JB, CS designed and performed dry lab

881 experiments and analysis. JEV provided COLO829 golden reference data. NB performed

882 independent break junction validation experiments. AH, RJ generated the

883 GRIDSS2/PURPLE/LINX TCGA call set. ATP, PP, EC designed and supervised experiments.

884 DLC, ATP, PP, EC contributed to writing of the manuscript.

Competing Interests

886 The authors declare no competing financial interests.

Availability of data and materials

GRIDSS2 source code is available as free and open source software from

https://github.com/PapenfussLab/gridss under a GPLv3 license. GRIDSS2 releases are

available as a github release, bioconda package, and docker image. Analysis scripts used to

generate results are available from

https://github.com/PapenfussLab/gridss/tree/master/scripts/gridss2_manuscript.

Hartwig cohort data was obtained from the Hartwig Medical Foundation (Data request DR-005).

Standardized procedures and request forms for access to this data can be found at

https://www.hartwigmedicalfoundation.nl/en.

Raw and analyzed data for the creation of the COLO829T/COLO829BL tumor/normal cell line

pair structural variant truth set are available grouped under ENA study accession PRJEB27698.

The COLO829 truth VCF is available from

https://github.com/UMCUGenetics/COLO829_somaticSV.

Capture panel validations of 13 patient tumor samples are available under the controlled access

dataset accession EGAD00001005525.

Supplementary Figures

907   Supplementary Figure 1: Distribution of 32-100bp duplication events per cancer type.

911   Supplementary Figure 2: Single breakend RepeatMasker annotation for 1,528 PCAWG

912   samples.

Supplementary Figure 3: Heatmap of single breakends to centromeric sequence by
chromosomal arm. The x axis indicates the arm of the single breakend with the y axis indicating
the most likely chromosomal arm the single breakend is connected to based on the breakend
sequence, and the copy number profile across the centromere. ? indicates the arm is unknown
due to the lack of copy number change across the centromere, and p and q indicate a
centromeric copy number gain to that arm.

925   Supplementary Figure 4: location of single breakends associated with the chromosome 1

926   centromere. Red indicates a single breakend associated with centromeric sequence on the

927   same chromosome, yellow indicates a single breakend associated with centromeric sequence

928   on chromosome 1. Single breakends to chromosome 1 occurring on chromosomes 5 and 19

929   follow a similar distribution to intra-chromosomal single breakends.

Supplementary Figure 5: a) Phasability and LINX classification of structural variants in the
Hartwig and PCAWG cohorts. Phasable variants are predominantly LINE translocation or form
part of complex events. b) Number of SVs in complex event clusters containing phasable SVs.
Most phasable SVs occur in highly complex events.