Discovery and characterization of Alu repeat sequences via precise local read assembly

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ABSTRACT

Alu insertions have contributed to >11% of the human genome and ~30–35 Alu subfamilies remain actively mobile, yet the characterization of polymorphic Alu insertions from short-read data remains a challenge. We build on existing computational methods to combine Alu detection and de novo assembly of WGS data as a means to reconstruct the full sequence of insertion events from Illumina paired end reads. Comparison with published calls obtained using PacBio long-reads indicates a false discovery rate below 5%, at the cost of reduced sensitivity due to the colocation of reference and non-reference repeats. We generate a highly accurate call set of 1614 completely assembled Alu variants from 53 samples from the Human Genome Diversity Project (HGDP) panel. We utilize the reconstructed alternative insertion haplotypes to genotype 1010 fully assembled insertions, obtaining >99% agreement with genotypes obtained by PCR. In our assembled sequences, we find evidence of premature insertion mechanisms and observe 5’ truncation in 16% of Alu Ya5 and Alu Yb8 insertions. The sites of truncation coincide with stem-loop structures and SRP9/14 binding sites in the Alu RNA, implicating L1 ORF2p pausing in the generation of 5’ truncations. Additionally, we identified variable Alu J and Alu S elements that likely arose due to non-retrotransposition mechanisms.

INTRODUCTION

Mobile elements (MEs) are discrete fragments of nuclear DNA that are capable of copied movement to other chromosomal locations within the genome (1). In humans, the ~300 bp Alu retroelements are the most successful and ubiquitous MEs, collectively amounting to >1 million genome copies and accounting for >11% of the nuclear genome (2,3). The vast majority of Alu insertions represent events that occurred in the germline or early during embryogenesis (4) millions of years ago and now exist as non-functional elements that are highly mutated and no longer capable of mobilization (3). However, subsets of MEs, including Alu and its autonomous partner L1Hs, remain active and continue to contribute to new ME insertions (MEIs), resulting in genomic variation between individuals (5) and between somatic tissues within an individual (6,7).

The human genome contains elements derived from the Alu Y, Alu S and Alu J lineages, which can be further stratified into more than ~35 subfamilies based on sequence diversity and diagnostic mutations (2,5,8). Most human Alu elements are from the youngest lineage, Alu Y, whose members have been most actively mobilized during primate evolution (5,9). Of these, the Alu Ya5 and Alu Yb8 subfamilies have contributed to the bulk of insertions in humans (10–14), although polymorphic insertions from >20 other Alu Y and >6 Alu S subfamilies have also been reported (5,15), implying polymorphic insertions of other lineages may still be segregating. In contemporary humans, the retrotransposition of active Alu copies results in de novo germline insertions at a frequency of ~1:20 live births (11,16). Over 60 novel Alu insertions have been shown to cause mutations leading to disease (17–19); older, existing Alu insertions in the genome have also been shown to facilitate the formation of subsequent rearrangements by providing a template of sequence utilized in non-allelic homologous recombination, replication-template switching and the repair of double-strand breaks (2,18–26). Thus, Alu insertions continue to shape the genomic landscape and are recognized as profound mediators of genomic structural variation.

Active copies of Alu are non-autonomous but contain an internal RNA Pol III promoter (27). Mediated by L1 encoded enzymes, Alu transcripts are mobilized by a ‘copy-and-paste’ mechanism referred to as target primed reverse transcription (TPRT) (7,28). TPRT involves the reverse transcription of a single stranded Alu RNA to a double stranded DNA copy, during which two staggered single-stranded breaks are introduced in the target DNA of ~5 to ~25 bp that are later filled by cellular machinery. The result-
ing structure consists of a new Alu flanked by characteristic target site duplications (TSDs) and a poly-A tail of variable length. Together these serve as hallmarks of retrotransposition. Integration of the new copy is permanent, although Alu can be removed by otherwise encompassing deletions, or very rare recombination-mediated excision events (29). Complete TPRT of a full-length element is responsible for the majority of Alu insertions. A minority of insertions have signatures of 5′ truncation but otherwise appear as standard retrotransposition events, indicating their movement by premature TPRT (21,30,31).

The primary difficulty in identifying novel Alu insertion loci stems from the abundant copy-number of these elements in primate genomes. Various approaches for large-scale analyses of Alu and other ME types have been developed that utilize next generation sequencing platforms. Scaled sequencing of targeted Alu junction libraries has permitted genome-wide detection, as implemented in techniques such as Transposon-Seq (32) and ME-Scan (33,34). Such targeted methods offer high specificity and sensitivity, but are restricted by the primers used for detection and are generally subfamily-specific. A broader detection of Alu variant locations is possible by using computational methods to search Illumina whole genome sequence (WGS) paired reads by ‘anchored’ mapping. This method seeks to identify discordant read pairs where one read maps uniquely to the reference (i.e. the ‘anchor’) and its mate maps to the element type in query (10,35,36). However, in considering read pair data alone, in their simplest form these methods are limited in the recovery of variant-genome junctions that might otherwise be captured from split read information. Specialized algorithms now offer improved breakpoint accuracy from WGS data by consideration of split-reads, soft-clipped reads and unmapped reads during variant detection (36–39). Beyond these basic requirements, existing programs differ in the implementation of additional filters and read-support criteria to identify the subset of calls likely to be true. For most callers this includes removing MEI candidate calls that are located near reference MEs of the same class due to their higher likelihood of being false predictions (36,38–41,37). Thus, as in all methods, a trade-off exists between sensitivity and specificity.

Having been mostly applied to high-coverage WGS, these methods require modification when applied to lower coverage data. Generally, these read-based detection methods have been developed in the context of ME variant loci discovery and reporting. Assembly approaches have received increasing application to next-generation sequencing data for breakpoint identification. Tools such as TIGRA (42), a modification of the SGA assembler used in HYDRA-MULTI (43,44), and the use of a de bruijn graph-based approach in SVMerge (45) have been developed to assemble structural variant breakpoints from population scale or heterogeneous tumour sequencing studies. Assembly-based approaches have also led to increased sensitivity and specificity for the detection of SNPs and small indels (46–50) and mobile elements (38). The sequence of polymorphic Alus has also been inferred by remapping supporting reads to an element reference followed by the construction of a consensus sequence (14). Since sequence changes within elements can determine their activity (9), the assembly of insertion sequences can better inform our understanding of element proliferation. Additionally, insertion haplotype reconstruction offers a direct way to determine genotypes across samples based on an analysis of sequence data supporting each allele.

Here, we utilize a classic overlap-layout-consensus assembly strategy applied to ME-insertion supporting reads to completely reconstruct and characterize Alu insertions. We apply this approach to pooled WGS data, from 53 individuals in 7 geographically diverse populations from the Human Genome Diversity Project (HGDP) panel (51,52). We assess the limitations of this approach by cross-comparison with Alu insertions identified from PacBio long sequencing reads in a complete hydatidiform mole (CHM1) (53). Finally we demonstrate the ability to obtain accurate genotypes based on explicit mapping to reconstructed reference and alternative alleles. We present the analysis of 1,614 fully reconstructed Alu insertions from these samples, including breakpoint refinement and genotyping of 1010 insertions. These results provide a basis for future study of such MEIs in human disease and population variation, and should facilitate similar analyses in relevant non-human models.

**MATERIALS AND METHODS**

**Samples**

We analysed whole genome, 2 × 101 bp Illumina read sequence data from a subset of 53 samples and 7 populations from the HGDP: Cambodia (HGDP00711, HGDP00712, HGDP00713, HGDP00715, HGDP00716, HGDP00719, HGDP00720, HGDP00721), Pathan (HGDP00213, HGDP00222, HGDP00232, HGDP00237, HGDP00239, HGDP00243, HGDP00247, HGDP00258), Yakut (HGDP00948, HGDP00950, HGDP00955, HGDP00959, HGDP00960, HGDP00963, HGDP00964, HGDP00967), Maya (HGDP00854, HGDP00855, HGDP00856, HGDP00857, HGDP00858, HGDP00860, HGDP00868, HGDP00877), Mbuti Pygmy (HGDP00449, HGDP00456, HGDP00462, HGDP00471, HGDP00474, HGDP00476, HGDP01081), Mozabite (HGDP01258, HGDP01259, HGDP01262, HGDP01264, HGDP01267, HGDP01274, HGDP01275, HGDP01277) and San (HGDP00987, HGDP00991, HGDP00992, HGDP01029, HGDP01032, HGDP01036). WGS data was processed using BWA, GATK (54) and Picard (http://picard.sourceforge.net) as described previously (52) and is available at the Sequence Read Archive under accession SRP036155. Final datasets are ∼7x coverage per sample. For analysis of CHM1 we utilized Illumina data obtained under accession SRX652547, and ERX009608 for analysis of NA18506. Reads were mapped, respectively, to the hg19/GrCh37 or hg18/build36 genomes as appropriate using the same procedures described above.

**Non-reference Alu discovery**

RetroSeq ‘discover’ was run on individual BAM files for each sample to identify discordant read pairs with one read mapping uniquely to the reference genome and its pair to an Alu consensus or to an annotated Alu present in the reference. A FASTA file of Alu consensus sequences was ob-
tained from RepBase (version 18.04) (55). Reference Alu elements were excluded using hg19 RepeatMasker (56) annotations. Candidate loci were assessed using RetroSeq ‘call’. For this analysis, we combined the supporting read information discovered in each individual and ran the ‘call’ phase on a combined BAM consisting of all samples. A minimum of two supporting read pairs was required per call (reads). A maximum read-depth of 1000 (depth; default is 200) was utilized for regions surrounding each call in order to accommodate the increased coverage of the merged BAM. Finally, any output call within 500 bp of an annotated Alu insertion was excluded using the bedtools window command (57) and RepeatMasker hg19 reference annotations (56). Unless otherwise noted, any other RetroSeq options were run at the default settings. Final Alu calls having met the further criteria of a filter tag FL = 6, 7 or 8 were selected for subsequent analysis.

Assembly of non-reference Alu elements

De novo assembly of insertion-supporting reads for each candidate insertion was performed using CAP3 (58). For each candidate insertion, a window of 200 bp was defined around the predicted breakpoint, from which we extracted read-pairs reported to support an insertion at that site based on RetroSeq outputs, and intersecting read-pairs with a soft-clipped segment from the BAM corresponding to each sample. For soft-clipped reads we required the clipped portion to be ≥20 bp with a mean quality ≥20. Using CAP3 we then assembled the extracted reads per site, with parameters chosen to account for shorter matches (–c 25 –j 31 –o 16 –s 251 –z 1 –c 10). CAP3 utilizes read-pair information to report scaffolds of contigs that are linked together but without assembled overlap. We merged such contigs together, separated by 300 ‘N’ characters to represent sequence gaps in the assembly. The resulting contigs and scaffolds were analysed using RepeatMasker (56), from which 2971 candidate assembled sites were identified that contain an Alu element (≥30 bp match) and at least 30 bp of flanking non-gap sequence. This pipeline is available on GitHub at https://github.com/KiddLab/insertion-assembly.

Breakpoint determination

Breakpoints for the assembled Alu variants were recovered utilizing a multiple alignment-based approach similar to an approach previously described (59,60). Orientation of candidate insertion sequences relative to the reference genome was determined using BLAT (61). Candidate breakpoints were then identified using mirorpeats (62) followed by a semi-automated parsing process. A global alignment was obtained for sequences from the two insertion breakpoints to the corresponding segment on the reference genome using stretcher (63) with default parameters, to generate pairwise alignments for two sequences aligned independently against the third (i.e. reference) sequence. A 3-way alignment was created from the two pair-wise alignments by inserting gaps as appropriate. Alignment columns were scored as a match among all three sequences (*), a match between the left (‘1’) or right (‘2’) insertion breakpoint and the reference, or mismatch among all sequences (‘N’). We then computed an alignment score across the left and right breakpoint sequence, with matches between the target sequence and the genome sequence (‘1’ or ‘*’ for the left breakpoint and ‘2’ or ‘*’ for the right breakpoint) resulting in a score of +1, a sequence mismatch among all three sequences a score of −1, and a match among the reference and the other breakpoint a score of −3. The same procedure was applied to the right breakpoint, except the score was tabulated from right to left across the 3-way alignment. Visualizations of the resulting aligned sequences with breakpoint annotations were constructed and subjected to manual review. When necessary, the sequences extracted for breakpoint alignment were adjusted and the alignment and scoring scheme described above then repeated until a final curated set of 1614 assembled insertions was obtained. The breakpoint was then interpreted as the position where the maximum cumulative score was reached respective to the reference, from which regions of overlapping sequence on the reference allele were determined. Of note, (i) overlapping regions are defined from the alignment itself, without regard to the element boundaries, and (ii) in scoring, a small degree of divergence among regions of overlap is permitted, in some cases, resulting in the identification of longer segments with less than 100% identity. To define candidate TSDs, segments of identical sequence were identified within the region of sequence overlap by extending towards the element 3’, and a mismatch was observed.

Sub-family assignment and analysis

A multiple sequence alignment was constructed from 45 Alu consensus sequences corresponding to active elements reported by (5) and all subfamilies identified in the assembled set by RepeatMasker. Consensus sequences were obtained from RepBase (version 18.04) (55) and aligned using MUSCLE v3.8.31 (64) run with default parameters. Each of the of 1614 assembled Alu sequences was extracted and separately aligned to the consensus profile alignment. The proportion of sequence differences between each element and each sub-family consensus was tabulated, excluding the poly-A tail. Elements equally distant from multiple subfamilies were deemed to be unclassified. Alignments for the 1010 sites suitable for genotyping were utilized to assess the extent of the recovered element length relative to the sub-family consensus.

Validation

A total of 66 assembled Alu insertions were validated by Sanger sequencing. This includes 20 sites selected randomly using a custom python script, 33 sites chosen based on characteristics of the assembled insertions, and 13 sites representing members of AluS and AluJ lineages. Chromosomal coordinates for each insertion were considered based on unique mapping of CAP3 assembled contigs and subsequent breakpoint analysis to the hg19 build. We extracted ∼500 bp in either direction of each insertion from the hg19 reference (http://genome.ucsc.edu/). The sequence was masked using RepeatMasker, and primers designed to include at least 150 bp in either direction of the predicted insertion, avoiding masked sequence when possible. Each
primer set was analysed by in silico PCR to the hg19 reference to ensure site-specific target amplification predictions overlapping each breakpoint, and to infer product size predictions for either allele. All primers were designed using Primer3v.0.4.0 (65) and purchased from IDT. Loci examined, primers, and samples analysed are summarized in Supplementary Table S2.

All PCRs were performed with ~50 ng of genomic DNA as template along with 1.5–2.5 μM Mg ++, 200μM dNTPs, 0.2 μM each primer and 2.5 U Platinum Taq Polymerase (Invitrogen). Reactions were run under conditions of 2 min denaturation at 95 °C; 35 cycles of [95 °C 30 s, 55–59 °C 30 s, 72°C 2 min]; and a final extension at 72°C for 10 min. For each PCR reaction, 10 μl were analysed by electrophoresis in 1% agarose in 1x TBE. Products from at least one positive reaction per locus were sequenced. When possible, PCR products from a homozygous individual were sequenced; otherwise the insertion-supporting fragment was gel-extracted (Qiagen), and the products eluted in water and subjected to sequencing. Sequencing was performed with primers situated both upstream and downstream of the insertion to account for uncertainty introduced from polymerase slippage at the poly-A tails and ensure amplification across both predicted breakpoints. Traces obtained for each insertion allele were aligned to the corresponding reference allele and CAP3 assembled contig in order to confirm the presence of the Alu insertion, breakpoints, and agreement in nucleotide sequence between the validated and assembled insertion.

Comparison with previous studies
For comparison with Chaisson et al. calls on CHM1 (53), we utilized insertion positions based on hg19 coordinates from http://eichlerlab.gs.washington.edu/publications/chm1-structural-variation/. These data consisted of 1254 total calls classified as ‘AluYSimple’, ‘AluSimple’, ‘AluSTR’ or ‘AluMosaic’. Overlapping calls were counted from intersection of any call located within 100 bp. The 1727 Alu insertion calls for NA18506 were based on alu-detect analysis (37) of ERX009608 as obtained from the Sequence Read Archive. Consistent with that study, calls were relative to the hg18 genome assembly, and any call located within 100 bp was counted as an intersecting site.

Comparison with primate data
The reference and insertion alleles for each AluS or AluI element were searched against primate reference genomes using BLAT (61) on the UCSC Genome Browser (http://genome.ucsc.edu/) (66). Insertion coordinates were converted to hg18 coordinates using liftOver and compared with polymorphic insertions reported in (67). Intersections within 500 bp were reported.

Construction of random insertion site distributions
We sampled 1614 insertion positions throughout the genome matching the known sequence preferences of the L1 endonuclease. To reflect the known L1 EN sequence preferences, we constructed a position probability matrix (PPM) using 99 5-bp L1 EN nic-site sequences reported in Gilbert et al. (68), with a pseudocount of 1 added to each column to permit the sampling of unobserved sequences. To create each randomly placed insertion we first randomly (uniformly) selected a position in the reference genome sequence and an insertion strand (+ or −). We then extracted the corresponding 5-bp nic-site, calculated its probability $P$ using the computed PPM, sampled a random number $p$ (from a uniform distribution) and accepted the site if $p < P$. This process was repeated until 1614 sites were accepted, to generate one set of randomly sampled insertion positions and then repeated to generate 200 sets of randomly sampled insertions. For some analysis we imposed an exclusion mask that mirrored our discovery criteria such that positions within 500 bp of an Alu element in the reference genome were not accepted. Scripts for randomly sampling positions based on a PPM are available at https://github.com/KiddLab/random-sample-by-ppm.

Genotyping
We performed in silico genotyping by mapping relevant reads to a representation of the complete insertion and reference alleles for each site. The reference allele consisted of 600 bp of sequence upstream and downstream of the start and end of any inferred TSD extracted from the hg19 reference. Based on the aligned breakpoints, insertion alleles were created by replacing the appropriate portion of this sequence with insertion sequence, accounting for inferred TSDs or target site deletions. For each site, these insertion and reference alleles constituted the target genome for mapping of reads. A BWA index was constructed from each (bwa version 0.5.9). Mapping and analysis was performed separately for each sample and each site. We extracted read-pairs with at least one read having an original mapping within the coordinates of the targeted reference allele with a MAPQ ≥ 20. The extracted read-pairs were then aligned to the site reference and alternative sequences using bwa align and bwa sampe (version 0.5.9). We then calculated genotype likelihoods based on the number of read pairs mapping to the insertion or reference alleles, considering the resulting MAPQ values as error probabilities as previously described (69). Read-pairs with equal mappings between reference and insertion sequences have a MAPQ of 0 and do not contribute.

Genotypes were obtained from the resulting raw genotype likelihoods using one of two approaches. For sites on the autosomes and the pseudoautosomal region of the X chromosome, genotype likelihoods for Alu insertions were processed, along with previously calculated SNP genotypes using LD-aware refinement using Beagle 3.3.2 (with options maxlr = 5000, iteration = 10, nsamples = 30, maxwindow = 2000) (70–72). For sites on the X chromosome, genotypes were obtained using a ploidy-aware expectation-maximization (EM) algorithm that utilized the genotype likelihoods and assumes Hardy–Weinberg Equilibrium across all 53 samples. Briefly, we followed (69) to estimate the allele frequency for each site via EM. Using the estimated allele frequency, we then determined genotype prior probabilities for X-linked alleles assuming Hardy–
Weinberg equilibrium. These genotype priors were then combined with the already computed genotype likelihoods to identify the sample genotype with the highest posterior probability. Principal component analysis was performed on the resulting autosomal genotypes using the smartpca program from the EIGENSOFT package (75).

Genotype validation

In order to validate in silico genotyping and permit estimation of genotyping accuracy, a subset of 11 insertion loci were screened from a panel of 10 individuals utilizing gel band assays, for a total of 110 predicted genotypes (Supplementary Table S6). Locus-specific primer sets flanking each insertion locus were designed as above (see Validation) and utilized for PCR amplification of each sample. All reactions were performed with 50 ng genomic DNA, in cycling conditions of 2 min at 95 °C; 35× [95°C 30 s, 55–59°C 30 s, 72°C 1 min], and a final 72°C extension of 3 min. 10 μl were analysed by electrophoresis in 1.2% agarose in 1x TBE, and results interpreted by banding patterns that supported either the unoccupied or insertion allele, as based on predicted band sizes from in silico PCR and size information for the assembled insertion at that site.

RESULTS

Precise assembly of full-length Alu variants using read data

To generate an accurate and highly specific collection of non-reference polymorphic Alu variants from population-scale WGS data we combined methods utilizing read-based discovery of all possible insertion sites with de novo local assembly of supporting reads (Figure 1). We utilized WGS data from a subset of the HGDP collection, specifically consisting of 2 × 101 bp paired-end libraries from 53 individuals across seven populations exhibiting a cline of diversity reflecting the major migration of humans out of Africa, with a median coverage of ~7x per genome. Given the coverage levels, we anticipated insertions that were private to a single individual were likely to be missed. However, we reasoned that borrowing read information across samples would increase our ability to detect rarer insertions that were nonetheless present in multiple samples, and pooled the data into a single merged BAM from all individuals for an effective coverage of ~429x. Candidate Alu insertions were then identified by applying RetroSeq (36) to the merged BAM. This particular program implements standard approaches to identity MEI-supporting read signatures, with performance characteristics that are comparable to other existing callers (e.g. (36,39,40)) and assigns a quality score to each call based on the number and mapping characteristics of supporting reads (Figure 1, Materials and Methods). Based on these outputs we considered calls with a score of 6 or higher for further filtering. To minimize false calls associated with reference elements, we removed any candidate call that mapped within 500 bp of an annotated Alu in the human reference. After exclusion, this resulted in 41 365 putative Alu insertions with an assigned quality score 6 or higher.

We then attempted to reconstruct as many individual insertion variants as possible, including the complete Alu sequence, its breakpoints and contiguous flanking sequence for each site. While recent efforts in short read assembly have focused on a de brujin graph approach (74–76), we reasoned a local assembly using an overlap-layout-consensus approach would take full advantage of our data. For these purposes we utilized the program CAP3 (58) that was originally developed for the assembly of large-insert clones sequenced using capillary sequencing, but has also been applied to de novo assembly of short read RNA-seq (77) and metagenomic sequence data (78). For each putative site, we retrieved any insertion-supporting read pairs as reported by RetroSeq and proximal soft-clipped reads, and then performed de novo assemblies with CAP3 (Figure 1) run with parameters adjusted for joining smaller overlaps that could be expected from 101 bp reads (also see Materials and Methods). The resulting scaffolds were filtered to require the presence of ≥30 bp of Alu sequence (with ≥90% nucleotide identity) and recovery of ≥30 bp of flanking non-gap sequence at one end, resulting in 2971 candidate assemblies. Assembled scaffolds having satisfied all the above
criteria were interpreted to represent the presence of a true non-reference insertion.

The resulting assemblies were aligned to the reference genome and putative breakpoints identified using a semi-automated procedure supplemented by manual curating. For these purposes, we adapted a procedure previously developed for the analysis of structural variant breakpoints represented in finished fosmid clone sequences (59) (also see Materials and Methods). A total of 1614 Alu-containing contigs were reconstructed, each having the complete associated insertion and at least one breakpoint with ≥30 bp of mapped flanking sequence (Supplementary Table S1A, Figure S1). Utilizing the locally assembled scaffolds, we determined regions of overlapping sequence on the reference allele (also see Materials and Methods; Supplementary Figure S2). We then identified a set of 1010 of the 1614 Alu insertions that had both breakpoints flanked by at least 100 bp of non-gap assembled sequences. Given the quality of these assembled calls, we refined the initial regions of overlapping sequence to identify candidate TSDs based on segments of exact nucleotide matches (Supplementary Figure S3). These 1010 insertions were deemed to be of the highest quality and suitable for in silico genotyping (see below), and had sizes ranging from 77–495 bp (median 315 bp) with refined candidate TSDs up to 50 bp (median 14 bp).

**Sensitivity and specificity of insertion discovery using short-read assembly**

A comprehensive assessment of the performance of MEI callers is hindered by the lack of an orthogonal ‘gold standard’ call set for formal comparison. To better assess the potential limitations of Alu discovery using Illumina (2 × 100 bp) paired-end short reads, we applied our approach to two additional samples, which have been extensively characterized (Supplementary Table S1B). The first comparison utilized Illumina WGS data generated from a complete hydatidiform mole (CHM1) (53). This sample offers particular advantage for these purposes, as it is essentially haploid throughout its genome and has been subjected to extensive characterization using long PacBio sequencing reads (mean 5.8 kbp). In their analysis, Chaisson et al. reported 1254 Alu insertions from this sample; of which 911 intersected within 100 bp of a candidate call based on the Illumina sequence data (Table 1). However, we note these raw RetroSeq calls include 18 501 predictions, implying an extremely high false discovery rate (FDR). Considering only the highest level of RetroSeq support still implies an FDR greater than 80%.

Mapping into highly repetitive regions using shorter sequencing reads results in an unacceptably high degree of false calls and necessitates filtering out regions near reference elements, a step common to most mobile element callers (36,38,39,41). The longer reads available from the CHM1 sample permitted interrogation of genome intervals harbouring a high number of repetitive elements. In fact, 54% of Alu insertion calls by Chaisson et al. are located within 500 bp of an Alu sequence present in the human reference genome. Limiting the analysis to only those calls ≥500 bp from any reference Alu results in an increase in both sensitivity and precision. Requiring successful element assembly further increases the precision: 446 of 468 of our assembled calls intersect (within 100 bp) with CHM1 reported calls. Assuming the remaining calls are all errors implies a FDR below 5%, however, we note additional analysis suggests this may be an over estimate (see Discussion).

To further investigate the precision of our assembly approach, we applied this approach to 2 × 100 bp Illumina WGS data from the Yoruba sample NA18506, an individual that has been analysed using multiple MEI callers (35,37,38). Again excluding calls within 500 bp of an annotated insertion, initial filtering of RetroSeq calls on the NA18506 sample resulted in 1375 putative Alu insertions having a quality score of 6 or higher. A total of 820 Alu insertions were fully assembled, of which 774 intersect with calls reported by alu-detect (out of 1727 total calls) (37), again implying a FDR around 5%.

**Validation of assembled HGDP insertions**

In validation, our specific goals were to demonstrate (i) the presence of the Alu at that chromosomal location, (ii) agreement between the assembled sequence with the cognate validated sequence for the insertion, and (iii) contiguous sequence of each insertion with its mapped flanking regions. To assess the accuracy of our approach, we experimentally validated a set of 20 insertions that were randomly selected from our total set of 1614, without bias to frequency or queried sample. We further validated 33 sites with unusual breakpoint characteristics and 13 sites derived from the AluS or AluJ lineages (Supplementary Tables S2 and S3). Detailed alignments corresponding to individual insertions, including visualized trace information for all sites, are provided in Supplementary Figures S4–S6 and properties for all validated sites are summarized in Supplementary Table S3. Examples of three representative insertions are highlighted in Figure 2, illustrating the recovery of mapped Alu-containing contigs, breakpoint estimations at those sites, and alignment of the deduced nucleotide sequence to the corresponding CAP3 assembly.

The presence of each of the 20 randomly selected sites was validated in at least two individuals predicted to have the insertion. Inspection of individual nucleotide alignments confirmed the sequence accuracy of each assembled element (base changes relative to subfamily consensus sequences are shaded in each alignment in Supplementary Figure S4); however, there were differences observed in breakpoints and flanking sequences for some sites. The randomly selected set of 20 includes 13 insertions deemed suitable for genotyping. The predicted sequence was confirmed for each of these 13 elements. For one site, located at chr6:131113270, the candidate TSD could be extended to include additional sequence that was originally excluded due to ambiguities in alignment gap placement. Seven of the 20 sites had predicted breakpoints within 100 bp of a gap in the CAP3 assembly. Experimental validation resolved the sequence corresponding to the gaps for 6 of these 7 insertions. For one site, at chr6:156541100, the CAP3 assembly gap was within the element poly-A tail and could not be resolved. The predicted breakpoint for 5 of these 7 insertions was confirmed, including 2 sites with additional predicted sequence embedded within the poly-A tail that could not be confirmed.
Figure 2. Sequence analysis of assembled non-reference \textit{Alu} insertions. Breakpoints were determined based on alignment the 5' and 3' edge of each insertion sequence with the corresponding sequence from the hg19 reference. \textit{Miropeats} annotation, aligned breakpoints and Sanger sequences are shown for three
We also validated a set of 33 calls biased to include insertions predicted to have unusual breakpoint characteristics (i.e. 0–3 bp TSDs, regions of sequence overlap larger than 20 bp, sites with corresponding target site deletions and insertions with predicted 5′ truncations). We successfully confirmed the presence and predicted nucleotide sequence of each assembled Alu element (nucleotide alignments are in Supplementary Figure S5). Overall, 24/33 insertions had breakpoints that precisely agreed with the corresponding assembly (e.g. see Figure 2A). Of the 26 of these 33 sites utilized for in silico genotyping, 3 were found to have incorrect breakpoints, in each case due to the absence of target site sequence adjacent to the Alu poly-A tail in the CAP3 assembly. The 24 fully validated sequences include a full-length element having a 20 bp candidate TSD within a 46 bp region of overlap which likely represents deletion of a pre-existing insertion (chr18:74638702), 3 assembled insertions that were correctly predicted to have short target site deletions relative to the pre-insertion allele ranging in size from 1 to 6 bp (at positions chr6:164161904, chr11:26601646 and chr12:73056650; see also Figure 2B and Supplementary Figure S5), and several elements with evidence of 5′ truncation (also detailed further below). We were also able to precisely reconstruct insertions that were within other repetitive sequence classes (e.g. see Figure 2C).

Finally, we validated each of the 13 assembled elements belonging to the AluS and AluJ lineages (Supplementary Table S3). We confirmed both the variable presence and derived subfamily of the element for each of these sites (detailed alignments are provided in Supplementary Figure S6). One of the 13 contained an assembly gap near the predicted breakpoint. As anticipated, several insertions had sequence characteristics indicating variable presence caused by non-retrotransposition mechanisms (i.e. involved in encompassing deletions or other endonuclease-independent processes) that were captured in our assembly and experimentally validated; these particular sites are detailed further below.

Characteristics of assembled insertions

Given the accuracy of our assemblies, we sought to more comprehensively characterize our set of reconstructed Alu insertions. Previous studies of full-length polymorphic elements have been mostly limited to insertions from an assembled reference genome (13,79), examination of trace archive data (12), insertions having been captured in relatively long read data (10), or consensus sequence generation based on re-mapping of reads (14). By making use of WGS data in de novo assembly, the insertion sequence itself is accurately reconstructed for analysis. Thus, utilizing our assembled contigs, we readily extracted the corresponding 1614 Alu nucleotide sequences and characterized each in terms of subfamily distributions and element properties.

Based on sequence divergence from Alu subfamily consensus sequences obtained from RepBase (55), we were able to assign 1465 (90.7%) of our insertions to one of 30 subfamilies (Table 2). We found 149 elements that were equally diverged from more than one subfamily consensus and could not be conclusively classified. Insertions from AluY subfamilies made up >99% of all assigned calls, with AluYa5 and Yb8 collectively representing more than half (62.7%) of the set. This observation was anticipated, given that AluY insertions have contributed to nearly all Alu genomic variation in humans, with AluYa5 and Yb8 being the most active subfamilies (5,9); and is broadly consistent with the sub-family assignments obtained by Mustafa et al. (14). Also as expected, assembled elements derived from the older AluS and AluJ lineages were a minority, together representing less than 1% of calls that could be assigned to a subfamily and similar to previous analysis of representative intact polymorphic Alu in humans (10–13).
Table 2. Classification of assembled \textit{Alu} variants

| Subfamily | Count | % Total | % Assigned |
|-----------|-------|---------|------------|
| \textit{Alu}Y |       |         |            |
| \textit{Alu}Y | 87    | 5.4%    | 6.0%       |
| \textit{Alu}Ya1 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Ya4 | 85    | 5.3%    | 5.9%       |
| \textit{Alu}Ya5 | 538   | 33.3%   | 37.1%      |
| \textit{Alu}Ya6 | 6     | 0.4%    | 0.4%       |
| \textit{Alu}Ya8 | 6     | 0.4%    | 0.4%       |
| \textit{Alu}Yb3a1 | 6     | 0.4%    | 0.4%       |
| \textit{Alu}Yb8 | 377   | 23.4%   | 26.0%      |
| \textit{Alu}Yb9 | 50    | 3.1%    | 3.4%       |
| \textit{Alu}Yc1 | 107   | 6.6%    | 7.4%       |
| \textit{Alu}Yc2 | 12    | 0.7%    | 0.8%       |
| \textit{Alu}Yd8 | 9     | 0.6%    | 0.6%       |
| \textit{Alu}Ye5 | 71    | 4.4%    | 4.9%       |
| \textit{Alu}Yf1 | 7     | 0.4%    | 0.5%       |
| \textit{Alu}Yg6 | 50    | 3.1%    | 3.4%       |
| \textit{Alu}Yh3 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Yh7 | 5     | 0.3%    | 0.3%       |
| \textit{Alu}Yi6 | 17    | 1.1%    | 1.2%       |
| \textit{Alu}Yi6.4d | 7    | 0.4%    | 0.5%       |
| \textit{Alu}Yj4 | 2     | 0.1%    | 0.1%       |
| \textit{Alu}Yk11 | 4    | 0.2%    | 0.3%       |
| \textit{Alu}Yk12 | 5    | 0.3%    | 0.3%       |
| \textit{Alu}Yk13 | 5    | 0.3%    | 0.3%       |
| \textit{Alu}S |       |         |            |
| \textit{Alu}Sc | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Sc8 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Sp | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Sq | 2     | 0.1%    | 0.1%       |
| \textit{Alu}Sq2 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Sx1 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}Sx3 | 3     | 0.2%    | 0.2%       |
| \textit{Alu}Sx6 | 1     | 0.1%    | 0.1%       |
| \textit{Alu}J |       |         |            |
| \textit{Alu}Jb | 2     | 0.1%    | 0.1%       |
| Unclassified | 149   | 9.2%    | 10.3%      |
| Total | 1614 |         |            |
| Total Classified | 1452 |          |            |

Since \textit{Alu}S and \textit{Alu}J are considered to be functionally inactive in humans (5,80), we examined each of the 13 validated sites from each of these families to infer possible mechanisms giving rise to these particular variants (Supplementary Figures S6 and S7). Several of these insertions had sequence characteristics indicative of a deletion, including extended but non-identical sequence flanking the element that exists as a single copy in the reference (e.g. \textit{Alu}Sq at chr12:26958660 and \textit{Alu}Sx3 at chr17:46617220) or the presence of remnant ‘scar’ sequence indicative of a deletion (\textit{Alu}Sq2 at chr10:68049106). Of the remaining elements, one (\textit{Alu}Sc8, chr6:31296783) is found within the HLA region on chromosome 6 and present on alternative HLA haplotypes, suggesting it has been maintained as polymorphic in the human population (81). Surprisingly, four of the elements have nearly identical matches with \textit{Alu}J or \textit{Alu}S sequences located elsewhere in the genome (Supplementary Figure S7), two of which also include matching non-\textit{Alu} sequence, suggestive of their presence by non-TPRT mechanisms.

We searched for evidence of each of the \textit{Alu}S and \textit{Alu}J insertions in non-human primates, utilizing reference genome sequences and intersection with polymorphic insertions reported in chimpanzees, gorillas and orangutans by Hormozdiari \textit{et al.} (67) (also see Materials and Methods). We found that 7 of the 13 elements were present in the genomes of Great Apes and other primates, with 3 also reported as polymorphic in Hormozdiari \textit{et al.} (also refer to Supplementary Table S5). The presence of these shared sites again implies that the variable presence of these elements in humans is not caused by new retrotransposition events, but rather is a consequence of other processes.

Insertions with 5’ truncation

To assess the length distribution of non-reference \textit{Alu} variants from our call set, we focused on insertions assembled from the \textit{Alu}Ya5 and \textit{Alu}Yb8 subfamilies. We reasoned that analysis of these particular subfamilies should provide the most informative resource for comparison given their representation as the majority of identified variants. We further limited analysis to those \textit{Alu} that were suitable for genotyping, as insertions that do not meet our criteria for genotyping may erroneously appear to be truncated due to an incomplete breakpoint assembly. This resulted in an analysis set of 351 \textit{Alu}Ya5 and 215 \textit{Alu}Yb8 insertions. Based on nucleotide alignments of the assembled insertions against their respective consensus, we examined the collective coverage of assembled elements, per subfamily, in comparison to the nucleotide positions relative to their respective consensus (also refer to the plots in Figure 3A and B).
Figure 3. Length distribution of assembled AluYa5 and AluYb8 insertions. (A) Scaled representation of the AluYa5 and AluYb8 consensus and element properties. The Alu is comprised of two arms (left, blue; right, grey) joined by an A-rich region and having a 3′ poly-A tail. The A and B boxes indicate promoter regions. A 31 bp insertion distinguishes the arms. AluYb8 has a 3′ 7 bp insertion relative to Ya5; the sequences are otherwise structurally conserved. The gold bar shows bases within SRP/14 binding sites. (B) The size distribution of 351 AluYa5 and 215 AluYb8 assembled insertions relative to the respective subfamily consensus. The red dashed lines indicate peaks in truncation near 45 bp and 170 bp. The number of assembled insertions containing an aligned nucleotide is shown against the corresponding position in the consensus.

We observed that 84.9% of AluYa5 (298/351) and 81.4% AluYb8 (175/215) variants were full-length, or within at least 5 bp of being full-length, consistent with previous reports of the genome-wide distribution of full-length Alu (31,82). Comparing the length distribution of all insertions revealed a detectable minority of 5′ truncations that were present in both subfamilies and exhibited a similar distribution of the apparent truncation point, as is shown in Figure 3B. More specifically, a subset of insertions from either subfamily was truncated ~8–45 bp from the consensus start (9.9% or 35/351 AluYa5, and 13.4% or 29/215 AluYb8 insertions), and a second subset was truncated ~55–170 bp from the consensus start (5.1% or 18/351 AluYa5, and 5% or 11/215 AluYb8) (Table 3). Besides having apparent 5′ truncation, all but two of these assembled insertions displayed characteristics of standard retrotransposition, including short flanking duplicated sequence and a poly-A tail (insertions at chr13:86166445 and chr11:26601646; also see Supplementary Table S1). We note the observed distribution is similar to that from the previous analysis of 10 062 reference human Alu (as extracted from NCBI build33) (82), and of 1402 intact polymorphic Alu from the then-current dbRIP (31); aspects of both are addressed further in the Discussion.

L1 and Alu insertions that are truncated but otherwise standard (i.e. have poly-A tails and flanking TSDs) are thought to arise from incomplete or premature TPRT (30,31,82,83). One mechanism thought to contribute to 5′ truncations is a microhomology-mediated pairing of nucleotides at the genomic target 5′ end with the nascent Alu mRNA, thus promoting the premature completion of TPRT (31), and in turn leaving a detectable signature. We manually examined each three way alignment of the 53 AluYa5 and 40 AluYb8 assembled 5′ truncation events for such evidence, specifically searching for nucleotides at the 5′ break that were shared with the respective Alu consensus at that position (31). We observed a subset of insertions with detected microhomology, with 40.9% of truncation having 1 bp of matching sequence, and 15.1% of all truncations with ≥2 bp shared at the 5′ break (details are summarized in Supplementary Table S4), though we note limitations of interpreting a single shared nucleotide as a ‘true’ instance of microhomology. Given this observation, the data indicate premature TPRT may account for a proportion of the truncated insertions.
Table 3. Truncation analysis of Alu variants

| Subfamily | Start     | Count | % Total  |
|-----------|-----------|-------|----------|
| AluYa5    | 1–5 bp    | 298   | 84.90%   |
|           | 7–45 bp   | 35    | 9.97%    |
|           | 55–171 bp | 18    | 5.12%    |
| AluYb8    | 1–5 bp    | 175   | 81.39%   |
|           | 7–45 bp   | 29    | 13.48%   |
|           | 55–171 bp | 11    | 5.11%    |

Insertion breakpoint distribution

We analysed the distribution of assembled insertions based on Genecode v19 annotations (84) relative to simulated placement of insertions at randomly sampled genomic locations, utilizing a PPM based on L1 EN preferred nic sites as reported by Gilbert et al. 2005 (68) and excluding positions within 500 bp of annotated Alus in the reference. Of the 1614 assembled insertions, 865 (~53.5%) were found within genes, of which 643 (~39.8%) were located within protein coding genes. Although slightly higher than reported in previous analysis (10,11), these values are consistent with our simulated placements (87.5% of simulated insertion sets had ≥865 placements within genes, and 82% had ≥643 placements within protein coding genes). Just 10 insertions (~0.61% of all calls) were found within exons, all of which were located in untranslated regions and therefore would not be predicted to disrupt coding sequence. This value is lower than that obtained in our simulation (mean of 32 sites with 0 of 200 random simulations having ≤10 insertions), indicating potential selection against retrotransposition into exons and other coding sequence, and consistent with previous studies indicating exonic depletion of Alu (10,11,34,35).

A total of 708 (~43.8%) of our assembled insertions were located within repetitive sequence. The majority of these insertions were found within other retrotransposon-derived elements (459, or ~28.4% were in LINEs and 124, or ~7.6% in LTRs), and in DNA transposons (69, or ~4.2%); 22 insertions were found in minor or unknown repetitive classes. This distribution is also consistent with that observed in previous survey of non-reference Alu insertions (10). Since we excluded any candidate call that was near an annotated Alu prior to assembly, no insertion from our callset was recovered within or near any existing Alu, though a handful of insertions (33, or ~0.02%) were observed within non-Alu SINE classes (e.g. the Mir, FLAM or FRAM groups). These results are broadly consistent with simulated insertions, except that we observe a deficit of insertions within other SINE classes and within satellite sequences (2.5% of simulations resulted in ≤33 insertions in SINEs and 0% of simulations resulted in ≤2 insertions in satellite sequence). Separate simulations permitting placement near annotated Alus in the human reference show that 48% of insertions sampled based on the calculated PPM would be within 500 bp of a reference Alu and subsequently excluded from our analysis. However, we note that none of our simulated data sets reached the level of 54% of insertions near other Alu elements that is observed in the Chaisson et al. data.

Genotyping

We identified a subset of 1010 insertions that had both breakpoints at least 100 bp away from an assembly gap that were suitable for in silico genotyping using Illumina sequencing reads (Supplementary Table S6, Figure S8). We compared the inferred genotypes for 11 autosomal sites with PCR-based genotyping across 10 samples, and found a total concordance rate of 99% (109/110) (Figure 4A and B; predicted genotypes are in Supplementary Table S7 for direct comparison). The only error among the tested calls occurred when the inferred genotype was homozygous for the insertion allele, while PCR genotyping indicates that the site is heterozygous (chr1:19550721; HGDP00476). Finally, we performed a Principle Component Analysis (PCA) of the autosomal genotypes across all 53 samples (73). As expected, individual samples largely cluster together by population with the first PC separating African from non-African samples (Figure 4C). This result further confirms the high accuracy of the inferred genotypes.

DISCUSSION

We utilized Illumina WGS paired reads to fully reconstitute a high-specificity set of 1614 non-reference Alu insertions from a subset of 53 genetically diverse individuals in 7 global populations from the HGDP (51,52). Experimental interrogation of a total of 66 sites confirmed the presence and deduced nucleotide sequence of a non-reference Alu, at each predicted site. The majority of events appear to result from retrotransposition characterized by apparent TSDs within an expected size range and the presence of poly-A tails, including several insertions with variable 5′ truncations (also see Supplementary Figures S4–S6). Mustafa et al. also reconstructed polymorphic Alu insertions from population samples (14). Our method, which requires assembly across breakpoints of each insertion, recovers a smaller number of variants per sample than Mustafa et al., which created a consensus based on remapping reads for each element. In both studies, AluYa5, Yb8 and Yb9 account for the majority of calls, but, we recover proportionally more calls classified as Yb8 and fewer Yb9, consistent with the observation that AluY insertions have contributed to nearly all Alu genomic variation in humans, with AluYa5 and Yb8 being the most active subfamilies (5,9).

We confirmed assembled insertions with aberrant breakpoint characteristics, including breakpoints with deleted sequence relative to the hg19 reference (chr6:164161904, chr11:26601646 and chr12:73056650), as well as insertions for which the TSD was absent (e.g. at chr17:46617220).
Figure 4. Genotyping of a subset of non-reference \textit{Alu} insertions. Genotype validation was performed for 11 sites across 10 individuals. (A) Strategy for primer design and allele detection. A single primer set was used for genotyping each locus, designed to target within 250 bp of the assembled insertion coordinates relative to hg19. (B) Genotyping from PCR screens and band scoring. Banding patterns supporting the unoccupied or \textit{Alu}-containing allele were assessed following locus-specific PCR; predicted band sizes were estimated by \textit{in silico} PCR analysis and mapped \textit{Alu} coordinates per site. The chromosomal location of each \textit{Alu} is indicated at left. ‘+’ or ‘−’ shows the relative position of each allele. Sample information is provided for population (above) and for each individual (below). (C) Principle Component Analysis (PCA) was performed on genotype matrix for 1010 autosomal sites genotypes across 53 populations. A projection of the samples onto the first two Principal Components is shown.

For 1010 insertions that had at least 100 bp of assembled sequence flanking both sides, we obtained a high level of breakpoint accuracy, having perfect agreement at 46 of a total of 51 sites tested (90.2%), including those with predicted aberrant breakpoints and/or 5’ truncation. Analysis of SNPs has demonstrated that improvements in accuracy can be obtained by separating the ‘discovery’ and ‘genotyping’ phases of analysis (85). We therefore performed genotyping by determining genotype-likelihoods based on remapping Illumina read-pairs to the reconstructed reference and alternative haplotypes, achieving an estimated 99% genotype concordance (109 of 110 genotypes analyzed).

For each of the 66 validated insertions, comparisons with Sanger sequences of those sites revealed that the correct nucleotide sequence of the \textit{Alu} insertion itself was obtained in assembly. However, a closer comparison of breakpoints at individual sites indicated that elements located near edges of assembled contigs (e.g. excluding the complete predicted TSD length) were more likely to have incompletely assembled breakpoints. Further examination of the individual reads supporting the assembled contig indicated that this was due to aberrant joining or incomplete TSD capture of reads that covered the poly-A tract (e.g. refer to trace data from insertions at chr1:102849294, chr12:99227704, chr22:26997608 and chrX:5781742 in the Supplement). An illustrative example of this comes from our assembly of the previously reported \textit{Y} \textit{Alu} Polymorphic element (YAP) (86) located at chrY:21611993, which contained an incomplete 3’ TSD. Capillary sequencing in sample HGDP00213 revealed the correct 11 bp TSD (5’ AAAGAAATATA), and confirmed the presence of YAP-specific nucleotide markers (at bases 64, 207, 243 and 268 relative to the \textit{Alu}Y8b consensus), as recovered by our CAP3 assembly and consistent with previous reports (Supplementary Figure S5.33) (86). Even when alleles are fully (and correctly) reconstructed by read assembly, we note that interpretation of the variant may not be clear. The assembled \textit{Alu}Sx3 at chr1:35425392 exemplifies this complexity (Supplementary Figure S6.8). At this site, our breakpoint predictions were inaccurate due to the presence of concomitant variation at this site relative to the hg19 reference, as revealed by sequencing in other individuals without the insertion to better reconstruct the structure of the pre-insertion allele (Supplementary Figure S9). Given the structure associated with this insertion, we suggest its variable presence is the result of an encompassing deletion. Notably, the CAP3 assembled insertion and additional non-reference proximal genomic sequence was found to be in complete agreement with corresponding Sanger reads, despite the presence of this surrounding structural variation relative to the genome reference.

Our validations of 11 \textit{Alu}S and 2 \textit{Alu}J insertions (Supplementary Table S3) correctly confirmed their bimorphic presence and high level of divergence from their respective subfamily consensus (Supplementary Figure S6). Surprisingly, we identified four of these elements (including both assembled \textit{Alu}Jb copies) that had nearly identical matches with \textit{Alu}S or \textit{Alu}J sequences located elsewhere in the genome (Supplementary Table S3 and Figure S7) implying a common source. As these elements have shared accumulated mismatches from their subfamily consensus
in divergences ranging from \(\sim 13.2\%\) to 14.2\%, we sug-
uggest these particular elements are likely the consequence of
non-TPRT processes. Two of these elements lacked TSDs
and/or poly-A tail (AluJb at chr12:73056650 and AluSx3
at chr17:460617220). The remaining two elements had adja-
cent non-Alu sequence that can be mapped contiguously to
other locations within the human reference genome (AluSp
at chr3:110413394 and AluJb at chr5:172054822). Multi-
ple non-TPRT mechanisms may have resulted in the du-
plication of these sequences. For example, they may have
served as donor sequences utilized in replication template
switching (87,88) or in templated double-strand break re-
pair (89) as has been reported for CRISPR/Cas induced le-
sions (90). Several insertions had sequence characteristics
indicative of subsequent deletion in the genome reference
sequence, such as the presence of regions of overlap of un-
usual length (i.e. 21 bp overlap for the AluSx1 located at
chr2:161952317, 98 bp for the AluSx3 at chr11:35425492),
additional flanking non-Alu sequence that does not map to
the human reference at that location (refer to alignments
for sites chr1:232869263, chr6:5348761 and chr12:26958660
in Supplementary Figure S6), and/or coincidence presence
with bimorphic insertions reported in the genomes of Great
Apes and other primates (67) (Supplementary Table S5). In
addition to the 7 AluS variants classified as likely deletions,
we identified 8 sites having stretches of identical candidate
TSDs that were at least 15 bp shorter than the correspond-
ing regions of sequence overlap, suggesting that \(\sim 1\%\) of the
assembled events may actually represent deletions.

Having an assembled, high-specificity call set of non-
reference Alu variants permitted analysis of element prop-
erties. Performing this step was meant to take particular ad-
vantage of these data, as existing MEI callers are generally
designed to catalogue events detected from read-based sig-
natures within the data. Examination of our assembled in-
sertions suggested the majority of elements exhibited prop-
erties consistent with classical retrotransposition, specifi-
cally being full length and the presence of a TSD and poly-
A tail of variable length. However, our analysis of the length
distribution of the reconstructed AluYa5 and Yb8 inser-
tions also revealed that 93 elements (\(\sim 16.4\%\)) of this subset
had evidence of having been 5’ truncated, despite appearing
otherwise standard, indicating insertion by mechanisms in-
volveing premature TPRT. We also observed evidence of
at least two groups of this subset, respectively, truncated \(< 45
bp and \(< 170 bp from the canonical 5’ edge (Table 3
and Figure 3).

These data are consistent with a previous manual cu-
rator of 1402 intact polymorphic Alu from dbRIP that
characterized full-length elements available at the time
(31). In that study the authors identified 115 elements
\(\sim 8.2\%\) with apparent 5’ truncations \(\sim 8-45 \) bp from the
Alu start \(\sim 8.2\%\) and 89 elements had \(\sim 55-171 \) bp trun-
cations \(6.3\%\) (31). The authors proposed a model of
microhomology-mediated nucleotide pairing of the 5’ end
of the genomic strand with the Alu RNA, having observed
41.2\% events with nucleotides at the 5’ break shared with
the Alu consensus at that position. However, a single shared
base supported the majority of the truncations; consider-
ing \(\geq 2 \) bp accounted for 16.7% of their observed events. We
searched our own data corresponding to all 5’ truncation
events, and observed similar levels of putative microhomol-
ogy: 15.1\% had at least 2 shared bases at the 5’ edge, and
40.9\% of insertions shared 1 base; although tentatively con-
sidered to represent true cases of microhomology. Another
study reported similar instances of Alu truncation events
(1005/10 062 or \(\sim 10.5\%\)), but found little to no statistical
support for base overlap at the 5’ breaks \((\sim 29\% \) 1 bp; \(\sim 13\%\)
\(\geq 2 \) bp) (82). Given that the 5’ Alu end is particularly GC
rich, this suggests such a ‘mis’-pairing during TPRT would
account for a minority of observed truncations. In support,
we examined the nick site for truncations with and without
putative signatures of microhomology and found no differ-
eence in preference, further confirming that both classes con-
tained the canonical L1 ORF2 protein (ORF2p) nick site,
5’ T\(_x\)/A\(_x\) (the ’/’ indicating the site of cleavage) (91,92).

We note that secondary structure of the Alu RNA it-
self may drive the non-random distribution of 5’ truncation
points. The bases associated with the points of truncation,
near \(\sim 45 \) bp and \(\sim 170 \) bp from the Alu start are also co-
incident with the predicted hairpin structure in the folded
RNA (9). The Alu RNA is reverse transcribed by the L1
encoded ORF2p, which pauses at sites of RNA secondary
structure such as poly-purine tracts and stem-loops (93).
Additionally, both truncation regions are located directly
3’ to predicted SRP9/14 binding locations (7,94). Although
SRP9/14 binding is necessary for efficient retrotransposi-
tion, the younger Alu\(_S\) and Alu\(_Y\) subfamilies contain nu-
cleotide substitutions that reduce SRP9/14 binding affinity,
suggesting that efficient displacement of bound SRP9/14
is important for the successful propagation of these ele-
ments (5,9). This suggests that the characteristic location
of 5’ truncations may be a consequence of ORF2p pausing
and premature disengaging from the Alu RNA during re-
verse transcription. Regardless, the data indicate premature
TPRT may account for a subset of the truncated insertions
(21,30,83).

Although our assembled calls are of high quality, our dis-
cover process suffers from the same limitations that are
common to other studies utilizing NGS. Because of the vari-
ability in coverage across samples, we are likely missing sites
present in only one or a small number of the analysed sam-
plies therefore biasing our call-set towards common inser-
tions. For example, of the 994 genotyped sites on the au-
tosomes or in the X-par region, we predict no singleton genotypes, 5 sites with an allele count of 2, and 147 sites
with an allele count \(\leq 10 \) across all 53 individuals (Sup-
lementary Figure S8). In addition to discovery sensitivity,
the genotypes inferred in samples having lower coverage are ex-
pected to be less accurate. By requiring element assembly,
we focus on a highly reliable call set that will have reduced
sensitivity. To further explore these issues, we compared our
approach with Alu insertions identified in the CHM1 hyda-
tidiform mole using PacBio reads (33). This analysis high-
lights trade-offs in sensitivity and specificity that are inher-
et in any discovery approach and clearly demonstrates the
challenges for discovering Alu insertions that are coincident
with existing Alu elements in the reference (Table 1).

The observed distribution of insertion sites distant from
reference elements is broadly consistent with expectations
based on the known L1 EN nic-site sequence preferences.
We note that our simulation approach, based on a 5-bp mo-
tif determined from cell-culture insertion assays, assumes statistical independence among the positions at the cleavage site. David et al. (37) have also analysed insertion positions of a large set of polymorphic Alus with respect to sequence motifs, reaching broadly similar conclusions. Their approach directly considers explicit 6-bp sequences by normalizing the number of discovered insertions with a given nic-site by the frequency of the observed 6-mer in the reference genome. The David et al. analysis, which included a method for discovering polymorphic insertions near existing Alu elements, did not detect an enrichment of insertions near other Alus beyond that predicted by the observed 6-mer sequence preferences. In contrast, the level of coincident insertions observed in the Chaisson et al. long read PacBio data exceed that predicted by our 5-bp model of nic site preference. This may result from preferences that are not modelled by the 5-bp motif as well as difficulty in detecting Alu insertions within clusters of reference Alus using shorter-read length data (37).

Considering insertions near other Alus results in thousands of false calls, necessitating subsequent filtering steps (36,38,39,37). When considering only insertions that are distant from existing reference Alus, our assembly approach has a moderate sensitivity (77%), and an FDR less than 5%. The true false-call rate of the insertions assembled from Illumina data is likely to be lower. Of the 16 assembled insertions that do not intersect with Chaisson et al. calls, three correspond to Alu insertions reported by Chaisson et al. that are near existing Alu elements, but remained in our call set because the initial RetroSeq prediction was at least 500 bp away from a reference element. An additional three calls correspond to more complex variants reported by Chaisson et al. involving Alu and other repetitive sequence. Counting these six sites reduces the apparent FDR of the assembly-based approach to 3.4%. Since the Chaisson et al. call set is itself likely to have missed some calls due to variable coverage and mapping ambiguities, we consider these rates to be merely approximations.

Despite the ability of local-assembly approaches to recover Alu insertions with high precision, it is clear that analysis of insertions, and other types of structural variation, within highly repetitive sequence using comparatively short reads remains a major challenge. Analysis of insertion site preferences, population diversity and insertion rates across individuals and somatic tissues should be cognizant of the severe challenges posed for accurate variant detection in repetitive regions.

**DATA ACCESS**

The HGDP Alu sequence data from this study is available in the NCBI Database of genomic structural variation (dbVar: [http://www.ncbi.nlm.nih.gov/dbvar/](http://www.ncbi.nlm.nih.gov/dbvar/)) under accession nstd109 and is also available in Supplemental Table S1A. The pipeline for Alu assembly and breakpoint analysis is available at [https://github.com/KiddLab/insertion-assembly](https://github.com/KiddLab/insertion-assembly). Scripts for randomly sampling positions based on a PPM are available at [https://github.com/KiddLab/random-sample-by-ppm](https://github.com/KiddLab/random-sample-by-ppm).

**SUPPLEMENTARY DATA**

Supplementary Data are available at NAR Online.

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