Structural variation (SV), including insertions and deletions (indels), is a primary mechanism of genome evolution. However, the mechanism by which SV contributes to epigenome evolution is poorly understood. In this study, we characterized the association between lineage-specific indels and epigenome differences between human and chimpanzee to investigate how SVs might have shaped the epigenetic landscape. By intersecting medium-to-large human–chimpanzee indels (20 bp–50 kb) with putative promoters and enhancers in cranial neural crest cells (CNCCs) and repressed regions in induced pluripotent cells (iPSCs), we found that 12% of indels overlap putative regulatory and repressed regions (RRRs), and 15% of these indels are associated with lineage-biased RRRs. Indel-associated putative enhancer and repressive regions are approximately 1.3 times and approximately three times as likely to be lineage-biased, respectively, as those not associated with indels. We found a twofold enrichment of medium-sized indels (20–50 bp) in CpG island (CGI)-containing promoters than expected by chance. Lastly, from human-specific transposable element insertions, we identified putative regulatory elements, including NR2F1-bound putative CNCC enhancers derived from SVAs and putative iPSC promoters derived from LTRs. Our results show that different types of indels are associated with specific epigenomic diversity between human and chimpanzee.

[Supplemental material is available for this article.]

The question of what makes us uniquely human has long been of interest (Darwin 1871). Comparative genomics has sought the genetic basis of human-specific traits (King and Wilson 1975; The Chimpanzee Sequencing and Analysis Consortium 2005; Wall 2013; Rogers and Gibbs 2014; Kronenberg et al. 2018), including human-specific gene gain/loss or regions under accelerated evolution (Enard et al. 2002; Pollard et al. 2006; Zhu et al. 2007; Franchini and Pollard 2017; Atkinson et al. 2018; Fiddes et al. 2018; Florio et al. 2018; Suzuki et al. 2018). In addition, epigenetic and transcriptomic differences also contribute to human-specific phenotypes (Pai et al. 2011; Hernando-Herraez et al. 2013; Gallego Prescott et al. 2015; Romero et al. 2015; Trizzino et al. 2017; Danko et al. 2018; Ward et al. 2018; Eres et al. 2019). However, how structural variations (SVs) affect human-specific functions is just beginning to be explored (Gordon et al. 2016; Fudenberg and Pollard 2019).

SVs, which include deletions, duplications, inversions, insertions, and translocations, are responsible for the majority of genetic differences within populations and between species. The 1000 Genomes Project estimated that an individual carries a median of 8.9 Mb of SVs versus 3.6 Mb of single-nucleotide variants (SNVs) (Sudmant et al. 2015). Long-read sequencing of two haploid human genomes revealed that the majority of SVs were novel, suggesting that the impact of SVs is underestimated (Huddleston et al. 2017). SVs also contribute to inter-species divergence. In 2005, The Chimpanzee Sequencing and Analysis Consortium reported ~90 Mb of insertions or deletions (indels) between human and chimpanzee; in contrast, SNVs constituted only ~35 Mb (The Chimpanzee Sequencing and Analysis Consortium 2005).

Noncoding cis-regulatory elements (CREs) play a critical role in gene regulation (The ENCODE Project Consortium 2004). One powerful method to identify putative functional elements is epigenomic profiling (Ernst and Kellis 2012; Roadmap Epigenomics Consortium et al. 2015). For example, H3K4me3 is usually associated with promoters, and H3K27ac is associated with both active promoters and active enhancers. In contrast, H3K9me3 is associated with heterochromatin, a repressed state characterized by densely packed DNA and low gene expression (Becker et al. 2016). By applying epigenomic profiling to related organisms (“comparative epigenomics”), we can compare epigenetic signature across species between syntenic regions and investigate the birth and death of regulatory elements during evolution (Xiao et al. 2012; Lowdon et al. 2016). Although studies have investigated enhancer evolution between human and chimpanzee (Gallego Romero et al. 2015; Prescott et al. 2015; Trizzino et al. 2017; Ward et al. 2018), the impact of SV on these elements has not been studied.

Here, we developed a novel computational strategy to define syntenic regions that contain indels. By using publicly available epigenomic data sets from human and chimpanzee, we defined the association between indels and epigenetic differences between the two species. We explored both epigenomic conservation and innovation in association with medium-to-large indels (20 bp–50 kb), as well as how lineage-specific transposable element (TE) insertions contribute to new putative functional elements. Our findings indicate that SVs and epigenomic changes between human and chimpanzee are significantly interrelated.

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Results
Development of a novel method to find orthologous regions overlapping large indels

Conventional comparative genomic/epigenomic methods rely on tools such as UCSC liftOver to retrieve syntenic regions between species based on alignments between genome assemblies (Kuhn et al. 2013). However, these tools usually fail to return syntenic regions when the synteny is disrupted by medium-to-large SVs. To overcome this obstacle, we developed a new pipeline that combines CrossMap (Zhao et al. 2014), an alternative to liftOver, with our newly developed tool called OrthoINDEL (Methods). Instead of filtering syntenic regions using the minimum percentage of bases that can be converted to the new assembly, CrossMap outputs the syntenic region as multiple blocks split by alignment gaps. OrthoINDEL then concatenates the fragmented orthologs output by CrossMap if they are continuous or separated only by indels (Fig. 1). This way, our pipeline stringently converts regions from the source genome to their syntenic coordinates in the target genome even if they overlap large indels. In contrast, without sacrificing specificity, the UCSC liftOver dismisses regions with a large fraction absent in the target genome. To illustrate this improvement, we performed the same genomic coordinates conversion from human to chimpanzee using either OrthoINDEL or liftOver. We found OrthoINDEL successfully converted approximately 1000 more regions that were dismissed as “partially deleted” or “split in new” by liftOver (Supplemental Fig. S1; Supplemental Table S1). Thus, OrthoINDEL is better suited to retrieve syntenic regions overlapping indels.

Indel-associated enhancers and H3Kme3 regions are more likely to be lineage-biased

By analyzing two human–chimpanzee comparison ChIP-seq data sets (Prescott et al. 2015; Ward et al. 2018), we identified about 15,000 putative promoters (H3K4me3 ChIP-seq peaks), about 27,000 putative enhancers (H3K27ac peaks outside of H3K4me3 peaks), and about 31,000 H3K9me3 regions (H3K9me3 broad peaks) in each species that constitute ~40 Mb, 67 Mb, and 300 Mb, respectively (Fig. 2A). Together, we define them as putative regulatory and repressed regions (RRRs). We found support for ~88% of putative CNCC promoters by overlapping them with annotated FANTOM5 or GENCODE promoters (~84% and ~85% in GENCODE and FANTOM5, respectively) (The FANTOM Consortium and the RIKEN PMI and CLST [DG1] 2014; Frankish et al. 2019).

Next, we applied the OrthoINDEL pipeline to establish the syntenic relationship between human and chimpanzee RRRs. We further classified regions as human-biased, chimpanzee-biased, or invariant based on both peak-calling and ChIP-seq read difference defined by DESeq2 (Methods) (Supplemental Fig. S2; Supplemental Table S2; Landt et al. 2012; Love et al. 2014). A data-hub displaying the processed data over human and chimpanzee syntenic regions is available at the WashU Epigenome Browser (Li et al. 2019; https://epigenomemegateway.wustl.edu/browser/?sessionFile=https://wangftp.wustl.edu/~xzhuo/CNCC/publication Session.json).

We annotated all indels >20 bp between the human (hg38) and chimpanzee (panTro5) reference genomes using the DASVC pipeline (Methods) (Gordon et al. 2016). In total, we defined 193,180 medium-to-large indels (20 bp–50 kbp) encompassing 95.8 Mb (~3% of the human haploid genome). We selected approximately 127,000 of them (42.2 Mb) with a defined ancestral state (using gorilla genome as an outgroup) and located within nonrepetitive regions for epigenomic analysis (Methods) (Supplemental Fig. S3). We also annotated TE-derived insertions within these indels (Methods) (Fig. 2B; Supplemental Table S3). The overall number and length of our indels were in excellent agreement with previously published results (Supplemental Fig. S4; The Chimpanzee Sequencing and Analysis Consortium 2005; Kronenberg et al. 2018).

Next, we characterized the association between indels and putative RRRs. We identified approximately 15,000 indel-overlapping putative RRRs by intersecting their coordinates using BEDTools (Quinlan and Hall 2010) (Supplemental Table S4). Indels are slightly depleted in putative RRRs instead of being uniformly distributed in the genome (Fisher’s exact test enrichment ratio 0.94, P-value 2.4 × 10−9). We found that ~88% (112,433/127,350) of indels do not overlap any putative RRRs in the study (Supplemental Table S5), 10% of indels overlap with invariant elements, and the remaining 1.8% (2267 indels) are associated with lineage-biased elements (Fig. 2C). An association between a human-specific indel and a human-biased putative RRR (135 with enhancer and 801 with H3K9me3 heterochromatin) could suggest that the indel may have created the RRR in the human lineage. On the other hand, association between a human-specific indel and a chimpanzee-biased element (149 with enhancer and
In accordance with previously reported findings, we found that all except five putative promoters are invariant between the two species (Fig. 2C). In contrast, ~85% of putative enhancer and repressed regions are invariant (Fig. 2C; Prescott et al. 2015; Ward et al. 2018). Compared with those not associated with indels, putative enhancers associated with indels are ~30% more likely to be lineage-biased (Fisher’s exact test P-value 4.7 × 10^{-6}), and H3K9me3 regions associated with indels are about three times as likely to be lineage-biased (Fisher’s exact test P-value 10^{-760}) (Fig. 2C). This result suggests that indels have a moderate association with putative enhancers and a strong association with H3K9me3 regions.

The enrichment of different indels with putative RRRs

We sought to understand if different size categories of indels had different association with putative RRRs. We divided non-TE-derived indels to four groups (20–50 bp, 50–500 bp, 500–5 kb, 5–50 kb) and separated TE-derived insertions by TE class. We defined indels from 20–50 bp as medium-sized indels and defined indels ≥50 bp as large indels following conventions (The 1000 Genomes Project Consortium 2015; Kronenberg et al. 2018). We calculated the enrichment of the intersection between these indel categories with putative RRRs over the genomic background using Fisher’s exact test (Fig. 4). We plotted the enrichment ratio and P-value of all pairs with at least one intersection (Fig. 4).

We noticed three main trends. First, medium-sized indels (20–50 bp) are enriched in invariant putative CNCC promoters. In contrast, indels ≥500 bp are depleted in invariant putative promoters. Second, we found indels ≥5 kb are depleted in invariant putative CNCC enhancers. Third, in H3K9me3-repressed regions, lineage-specific sequences ≥500 bp (insertions ≥500 bp in that lineage and deletions ≥500 bp in the other lineage) are enriched for H3K9me3 regions from the same lineage (Fig. 4). As an example, we found that human lineage insertions and chimpanzee lineage deletions ≥500 bp are enriched in human-biased H3K9me3 regions.

None of the lineage-biased putative CNCC promoters overlap an indel (Fig. 4), and the enrichment/depletion of indels with putative invariant promoters is almost identical to their enrichment/depletion with all putative CNCC promoters.

We separated TE-derived insertions by TE class and performed the same enrichment analysis (Fig. 4). Lineage-specific SVA insertions are significantly enriched in both putative lineage-biased CNCC enhancers and iPSC H3K9me3-repressed regions for both species, which implies that newly inserted SVA elements may have provided CNCC-specific enhancers and been targeted by the repressive marks in both species (Fig. 4). ERV insertions are only enriched as lineage-biased H3K9me3-repressed regions in the chimpanzee lineage, contributed mainly by the H3K9me3-modified chimpanzee-specific PTERV regions (Supplemental Table S4; Yohn et al. 2005). Previous studies in fly and yeast suggest that repressive marks can spread beyond the heterochromatin boundary and affect nearby genes (Elgin and Reuter 2013; Obersiebig et al. 2016; Greenstein et al. 2018). However, we found here that H3K9me3 marks rarely expand beyond 2 kb outside newly inserted TE boundaries (Supplemental Fig. S5). In general, out of the SVs associated with biased chromatin marks, insertions tend to be associated with creation, rather than destruction, of putative RRRs.

We further explore the enrichment of indels from 20–50 bp with putative CNCC promoters and the enrichment of SVA insertions with putative CNCC enhancers in the following sections.
Enrichment of medium-sized indels within CpG islands

Promoters are considered conserved elements owing to their low nucleotide substitution rate (The FANTOM Consortium and the RIKEN PMI and CLST [DGT] 2014). We also found both putative CNCC promoters and GENCODE-annotated promoters are conserved using a phastCons score (Supplemental Fig. S6A,B). In contrast, using Fisher’s exact test, we found that medium-sized indels (20–50 bp) are highly enriched within putative CNCC promoters, whereas indels >500 bp are depleted within putative CNCC promoters (Fig. 4). To characterize the relationship between indel size and their enrichment in promoters at a finer resolution, we separated indels <500 bp by size at 50-bp intervals and tested their enrichment with putative CNCC promoters using Fisher’s exact test. We found that 50- to 100-bp indels are barely enriched, and all indel bins >100 bp are not statistically enriched (Supplemental Fig. S7).

To validate our observation, we further calculated indel frequency around annotated genes for 20- to 50-bp and 50- to 100-bp indels. Again, we found that 20- to 50-bp indels, but not 50- to 100-bp indels, have elevated frequency immediately upstream of transcription start sites (Fig. 5A). We also observed slightly lower indel frequency in more conserved promoters (phastCons > 0.2) than in less conserved promoters (phastCons < 0.2), indicating
that conserved promoters have fewer indels (Supplemental Fig. S6C,D). To test whether the enrichment of 20- to 50-bp indels in promoters also exists in the human population, we extracted variants from The 1000 Genomes Project Phase 3 release (Sudmant et al. 2015; The 1000 Genomes Project Consortium 2015) and repeated the analysis. Consistent with our inter-species observation, only medium-sized indels (20–50 bp) within the human population are enriched in the promoters (Fig. 5B).

Seventy percent of human gene promoters contain CpG islands (CGIs) (Saxonov et al. 2006). Many indels in promoters are located within UCSC annotated CGIs, suggesting that indels within CGIs could be driving the observed high indel rate in promoters (Kuhn et al. 2013). We separated promoters into CGI and non-CGI promoters and found that 20- to 50-bp indels are enriched only in CGI promoters (enrichment ratio 2.60, \( P \)-value \( 10^{-762} \)) but not in non-CGI promoters (enrichment ratio 1.06, \( P \)-value 0.02).

Because indels >150 bp are not enriched within putative CNCC promoters (Supplemental Fig. S7), we analyzed the enrichment of 20- to 50-bp, 50- to 100-bp, and 100- to 150-bp indels with promoters and CGIs. To verify the Fisher’s exact test results (Fig. 4; Supplemental Fig. S7), we directly compared observed number of intersections with the expected distribution based on random sampling (Methods) (Fig. 5C). CGIs outside promoters are enriched for all three sizes of indels (\( P \)-value < 0.001, permutation test), and the enrichment of indels in CGIs within promoters decreases as a function of indel size from 20 bp to 150 bp. In contrast, promoters without CGIs are barely enriched for any indel category (Fig. 5C). Thus, consistent with previous findings, our data suggest

Figure 5. Enrichment of 20- to 50-bp indels in promoters/CGI. (A) Frequency of human–chimpanzee indels of sizes 20–50 bp and 50–100 bp in and around annotated human genes. Metaplots display genes with introns removed and 5-kb flanking regions surrounding the transcription start site and end site. (B) Frequency of human population indels found in The 1000 Genomes Project around the same promoter regions described in A. (C) Comparing the observed number of indels intersecting with CpG island within promoters, CpG island outside of promoters and promoters without CpG island with the same intersections between randomly shuffled indels and the three CpG island or promoter regions. All indels were shuffled 1000 times. The density distributions of the numbers of shuffled indel intersections are illustrated by black lines. The observed numbers are indicated with a red vertical line with the observation/expectation ratio (O/E) and \( P \)-value at the top of each graph.
that CGIs are hot spots for indels in evolution (Tian et al. 2011; Kiktev et al. 2018).

Indels from 100–150 bp are less enriched in CGIs within promoters than the smaller 20–50 bp and 50–100 bp indels, potentially because they may drastically influence regulatory activity and have thus been selected against during evolution. Medium-sized indels (20–50 bp), on the other hand, seem to have been tolerated, as promoters with CGIs are highly epigenetically conserved between humans and chimpanzees (Fig. 4).

Lastly, we characterized the enrichment of indels in CGIs within the human population using the variants from The 1000 Genomes Project Phase 3 release and found the same pattern (Supplemental Fig. S8; The 1000 Genomes Project Consortium 2015). We also characterized the CGI enrichment of small indels (<20 bp). Small indels (1–20 bp), especially indels with length ≤6 bp, are depleted in CGIs. Because 98% of The 1000 Genome Project indels are <20 bp, our conclusion is consistent with the previous observation that indels are depleted within CGIs (Neininger et al. 2019), but highlights the novel finding of enrichment of medium-sized indels in CGIs.

**Lineage-specific TE insertion events.**

Lineage-specific SVA insertions are highly enriched in both putative enhancers and repressed regions (Fig. 4). For the 37 SVA elements that overlap with human-biased putative enhancer regions, the aggregated H3K27ac ChIP-seq profile shows areas of elevated signal similar to that in putative CNCC enhancers (Supplemental Fig. S9A,B). To better illustrate their enrichment pattern, we plotted aggregated ChIP-seq profiles around all human-specific SVAs and compared them to the profiles of their orthologous preinsertion sites in chimpanzees (Fig. 6A). In addition to the previously described ChIP-seq data, we also included iPSC H3K27ac from both species and human iPSC H3K4me3 (The ENCODE Project Consortium 2004; Gallego Romero et al. 2015). We also analyzed human-specific LTR5 insertions in a similar fashion (Supplemental Fig. S9C,D).

We used the mappability score as a measurement of repetitiveness and the propensity of a genomic region to produce uniquely mappable reads (Derrien et al. 2012). Lineage-specific SVA and LTR5 have low mappability scores (Fig. 6A). Therefore, their corresponding ChIP-seq signals were likely underestimated. Indeed, with the exception of H3K9me3 ChIP-seq, which was sequenced using 100 bp paired-end reads, all other ChIP-seq data sets generated using single-end reads have close to zero signal over the low mappability regions. Nevertheless, we found a strong H3K27ac signature, suggesting putative enhancer activity on the 3′ flanking region of SVA in CNCC and a strong promoter signature (H3K4me3 and H3K27ac) on both flanking regions of LTR5 in iPSCs in human but not in chimpanzee. Conversely, we found H3K4me3 ChIP-seq signal 3′ of LTR5 insertions in CNCC, but these were found in both human and chimpanzee, suggesting that in this case the epigenomic mark may be independent of the TE insertion (Fig. 6A).

To understand how epigenetic profiles might evolve over time, we extended our analysis to related TE insertions that are shared by both species. SVA proliferated and diverged in the human genome in a similar fashion as the amplification of L1 (Khan et al. 2006; Hancks and Kazazian 2016). SVAs in the human genome were classified into six subfamilies (SVA-A to SVA-F). The expansion of subfamilies from SVA-A to SVA-D predates the human–chimpanzee split, and SVA-E and SVA-F expanded after the human–chimpanzee divergence (Wang et al. 2005). We defined all human-specific SVAs as SVA-human and classified human–chimpanzee shared SVAs based on their subfamilies (from SVA-A to SVA-D). We plotted H3K27ac, H3K9me3, and mappability profiles of different SVA subfamilies in Figure 6B. We found that the 3′ boundary H3K27ac signal in CNCC decreases as SVA subfamily ages, with the greatest signal in human-specific subfamilies. In contrast, H3K9me3 signal intensifies with increasing SVA age in iPSCs (Fig. 6B).

Next, we similarly classified LTR5s based on lineage specificity and subfamilies (from oldest to youngest: LTR5A-shared, LTR5B-shared, and LTR5Hs-shared to LTR5-human) and performed the same analysis using H3K4me3 data in iPSCs. Similar to SVAs, the strength of the LTR5 promoter signature in iPSCs is negatively correlated with age. However, LTR5s are not marked by H3K9me3 in iPSCs (Fig. 6C).

NR2F1 binding is correlated with enhancer signature on the 3′ end of SVA in CNCC

Our data thus far predicted a putative enhancer within SVAs close to their 3′ end. The H3K27ac signal is likely “hidden” owing to low mappability, but part of it extends into mappable regions, as we observed in Figure 6. However, we cannot observe the boundary ChIP-seq signal at the 5′ end of a full-length (1.6-kb) SVA, which might be too long for the enhancer signal to extend beyond. To determine the precise location of the putative enhancer, we extracted all SVAs with complete 3′ ends in the human genome, sorted them based on size, anchored them at the 3′ end, and annotated them with CNCC H3K27ac signal (Fig. 7A). Again, we observed the elevated H3K27ac signal on the 3′ end of SVAs (Fig. 7A). However, once the SVA length was reduced to 300–500 bp, a similar boundary H3K27ac signal on the 5′ end emerged. This result is consistent with the hypothesis that the enhancer signature originating within SVAs can extend beyond the low mappability region. The boundary H3K27ac signal disappeared on both ends of shorter SVA copies, suggesting that further truncation resulted in a loss of the internal enhancer (Fig. 7A,B). Zooming in onto the 5′ end of these shorter SVA fragments revealed a strong NR2F1 binding motif (Fig. 7B,C). The disappearance of the boundary H3K27ac signal correlated with the truncation of the NR2F1 motif (Fish’s exact test P-value 7 × 10−35) (Fig. 7C). NR2F1 is a critical regulator in CNCC (Rada-Iglesias et al. 2012). Indeed, the NR2F1 and H3K27ac ChIP-seq signals co-occur in SVAs in CNCC (Fig. 7B; Prescott et al. 2015). However, not every putative human-biased CNCC SVA enhancer has a related NR2F1 peak (Fig. 7D,E), suggesting the possible involvement of other transcription factors in these SVA-derived putative enhancers.

**Discussion**

In this study, we systematically characterized how medium-to-large indels are correlated with differences in the epigenome in the human and chimpanzee lineages. We found that indels are enriched in putative lineage-biased enhancers and H3K9me3-repressed regions. We should note that all genomic enrichments were estimated using a whole-genome random distribution as the background. However, genomic features are not randomly distributed and the assumption may not always be appropriate. Yokoyama et al. (2014) described a substitution-based framework to model birth-death of lineage-specific functional elements. We
showed here that in addition to substitutions, indels can also contribute to the birth-death of regulatory and repressed elements. Our strategy is readily applicable to other comparative epigenomic data sets, and we have made our processed data available in our comparative browser (Li et al. 2019).

Mutation rate varies depending on genomic region. It has been reported that the substitution rate in closed chromatin regions is higher than the rate in open chromatin regions (Fortin and Hansen 2015; Makova and Hardison 2015). Lunter et al. (2006) found the highest indel rates in regions with extremely high and extremely low GC content. By using macaque as an outgroup, Kvikstad et al. (2007) reported a curvilinear relationship between human lineage indel rate and GC content, and they also found weak anticorrelation between insertion rate and number...

Figure 6. Chip-seq read count, normalized to reads per genomic content, surrounding repetitive TE insertions reveals putative hidden CREs. (A) Chip-seq signal profiles and 50-bp mappability over all human-specific SVA and LTR5 insertions in the human genome (top) and their orthologous preinsertion sites in the chimpanzee genome (bottom), with 1-kb flanking regions. (B) Profile of CNCC H3K27ac Chip-seq, iPSC H3K9me3 Chip-seq, and 50-bp mappability around SVA insertions in the human genome, with 1-kb flanking regions. SVA subfamilies are distinguished by color with copy numbers inside parenthesis. (C) Profile of iPSC H3K4me3 Chip-seq, iPSC H3K9me3 Chip-seq, and 50-bp mappability around LTR5 insertions in the human genome, with 1-kb flanking regions. LTR5 subfamilies are distinguished by color with copy numbers labeled.
of CGIs at 1-Mb genomic windows. Specific to CGIs, the substitution rate at CpG sites is higher than the mutation rate at other sites because of the high deamination rate of 5-methyl cytosine (Coulondre et al. 1978). Cohen et al. (2011) found that the lack of methylation of CGIs can explain their maintenance without implying purifying selection in primates. However, CGIs are often associated with genome instability (Deaton and Bird 2011; Du et al. 2014).

We found that CGIs are hotspots for medium/large indels in hominids. Kiktev et al. (2018) showed that the high-GC region in yeast has a high deletion/duplication rate resulting from DNA polymerase slippage. Thus, analogous to well-characterized SVs in human coding exons (Montgomery et al. 2013; Challis et al. 2015), high-GC regions are prone to forming a single-stranded DNA loop owing to polymerase slippage during DNA replication (Tian et al. 2011), which likely results in the increased indel rate. However, GC-rich regions are prone to sequencing error, and we cannot completely rule out the possibility that some indels we called were caused by the difficulty of sequencing and calling variants in these GC-rich regions.

Figure 7. NR2F1 binding profile in CNCCs coincides with the putative SVA enhancer profile. (A) Heatmaps of H3K27ac (50 bp single-end) and NR2F1 (202 bp paired-end) ChIP-seq signal in CNCC over all SVA elements with complete 3′ ends (outlined by dotted line) and 3-kb flanking regions in the human genome sorted by length (top to bottom, longest to shortest). (B) Zoomed-in view of the heatmaps displaying 53 5′-truncated SVA elements from 587 bp to 300 bp boxed in A. High H3K27ac and NR2F1 signal are visible on both ends, and flanking H3K27ac ChIP-seq signal correlates with NR2F1 ChIP-seq signal. (C) Nucleotide sequence alignment of the 10 truncated SVA elements boxed in B. Elements with strong CNCC H3K27ac signals are marked by checkmarks on the right; elements without CNCC H3K27ac signal are marked by a cross. The NR2F1 binding motif is provided at the bottom of the alignment. (D) A human SVA insertion associated with a human-biased enhancer and NR2F1 binding. (E) A human SVA insertion associated with a human-biased enhancer but no NR2F1 binding. Note the low mappability, indicating high repetitiveness, over the SVA insertions and the difference in NR2F1 ChIP-seq peaks between E and F.
By comparing sequence conservation with epi-conservation, Xiao et al. (2012) reported elevated epi-conservation of H3K27ac, H3K27me3, and methylated CpGs (but not H3K4me3) in rapidly evolving sequences and proposed that epi-conservation could buffer some deleterious mutations. Here, we report the conservation of H3K4me3 marks between human and chimpanzee despite an elevated rate of medium-sized indels, further supporting the concept of epi-conservation and its potential role in buffering the impact of mutations.

Britten and Davidson (1971) proposed the gene battery model in the 1970s to explain the evolution of regulatory networks. They proposed that a single “activator gene” can control a “battery of genes” by interacting with diffused repetitive sequences throughout the genome. Since then, TEs have been repeatedly shown to contribute novel regulatory elements (Wang et al. 2007; Feschotte 2008; Lynch et al. 2011; Chuong et al. 2017). However, in primates, especially in the human lineage, there have been conflicting reports about the regulatory role of TEs. Trizzino et al. (2017) found specific TE subfamilies enriched in liver enhancers. On the other hand, Ward et al. (2018) could not find significant contribution of TEs to gene regulation in pluripotent stem cells. We report clear signals of TE-derived, tissue-specific putative enhancers and promoters unique to human or chimpanzee. We identified LTR5 as putative promoters in iPSCs, whereas Fuentes et al. (2018) found them to have enhancer activity in human embryonal carcinoma NCCIT cells. We could not find a large impact on nearby gene expression associated with these TE-derived enhancers with our limited data set. It is possible that these new enhancers do not regulate the closest genes. Alternatively, they may provide functional redundancy instead of inventing new regulation (Osterwalder et al. 2018; Choudhary et al. 2020). We also report that TE-associated heterochromatin displays limited spreading (Supplemental Fig. S5). We found a rapid conversion of TE epigenetic modification from active to repressive states as a function of age in young TEs. This discovery echoes a previous report of the transition of repressive marks from cytosine methylation to H3K9me3 as ERVs age in the human genome (Ohtani et al. 2018). These data suggest that although many TEs carry regulatory elements, the host rapidly and continuously silences such activity during evolution.

Most genomic analyses rely on second-generation sequencing, which produces short reads, restricting our ability to detect signals from repetitive, low mappability regions. To overcome this limitation, we investigated the epigenetic signal not only from within TEs but also from flanking regions. By comparing TE insertions with orthologous preinsertion sites in another species, we can infer that the epigenetic signal originates from these highly repetitive regions. Our approach expands the application of second-generation sequencing and reveals that there are more potentially functional elements hidden in unmapped territories. However, the sensitivity of our method is limited by the distance of the element to the boundary, read length, DNA fragment size, and other factors.

Only four different TEs have been actively transposing in the human lineage. Of the four, we found that two are associated with putative enhancers and promoters using data from only two cell types. Although most TEs are neutrally evolving in the genome, our discovery suggested that many CRES carried by TEs were active upon insertion. Our finding begs more thorough investigation of CRES derived from recently inserted TEs in more cell types and between different species.

Methods

Indel identification

We applied the DASVC tool to annotate indels between humans and chimpanzees (Gordon et al. 2016). The DASVC tool was downloaded from (https://github.com/zeeev/DASVC) and was used with default parameters to identify 20-bp-to-50-kb indels between the hg38 and panTro5 genomes. We further processed DASVC output using a Python script (refine_calledSV.py) to remove segmental replacements and extract the exact coordinates in both species. To identify indels that occurred in mappable regions, we calculated the average 75-bp mappability score of the flanking 200 bp of all indels in both species and selected those indels with a mappability score > 0.7 in both flanking regions in both species (Derrien et al. 2012).

To differentiate deletion in one lineage from insertion in the other, we identified orthologous coordinates for all indels in the gorilla reference genome (gorGor5). We considered an indel to be a deletion if the indel region is present in the gorilla genome and to be an insertion if the region is absent in the gorilla.

A new chimpanzee reference genome panTro6 was published shortly after we started this project (Kronenberg et al. 2018). PanTro6 closed 52% of remaining gaps, but it does not refute the high-quality panTro5 reference genome in assembled regions. Therefore, the conclusions we reach here using panTro5 should remain valid.

TE-derived insertion annotation

To annotate TE-derived insertions within the identified indels, we intersected the indel list with RepeatMasker annotations for both the hg38 and panTro5 genomes using BEDTools (Smit et al. 2013–2015; Quinlan and Hall 2010). To avoid calling fragmented TEs as separate TE insertion events, we defined an indel as a TE-derived insertion if it was derived from a single TE insertion event. Indels containing only an Alu element, a full-length endogenous retrovirus (ERV), or a solo long terminal repeat (LTR) were counted as Alu/ERV insertions. Solitary LTRs were included because they are derived from an ERV insertion followed by nonhomologous recombination (Mager and Stoye 2015). Because of the prevalence of 5′ end truncation during target-primed reverse transcription (TPRT) (Luan et al. 1993), we also tolerated incomplete 5′ ends in defining L1 and SVA insertions. One limitation to our rigorous approach is that we did not annotate lineage-specific solo-LTRs, in which part of the solo-LTR aligned with the 5′ end LTR of the full-length ERV and part aligned with the 3′ end, as TE-derived insertions. In addition to the known actively transposing TE subfamilies described above, we also identified a few lineage-specific LTR12C elements (Supplemental Table S3).

Because TE insertions are homoplasy-free and unidirectional (Bashir et al. 2005; Ray et al. 2006), the orthologous regions corresponding to most human-/chimpanzee-specific TE insertions should be found as preinsertion sites in the gorilla genome. As expected, the orthologous locations of 98% (15,803 of 16,068) of lineage-specific TE insertions are preinsertion sites in the gorilla genome (Supplemental Fig. S4B), whereas the remaining 2% were present as TE insertions. These cases could be explained by incomplete lineage sorting, as shown previously (Ray et al. 2006; Kronenberg et al. 2018).

Peak calling and cross-species comparison

We downloaded both CNCC and iPSC raw read FASTQ files from NCBI Gene Expression Omnibus (GEO; https://www.ncbi.nlm.nih.gov/geo/) accessions GSE70751, GSE61343, and GSE96712
We applied CrossMap to identify orthologous segments for each peak in the other species (Zhao et al. 2014). For each peak region, CrossMap outputs all fragmented orthologous loci separated by any indel >1 bp. We developed a new tool, OrthoINDEL, that processes fragmented syntenic regions from the CrossMap output file. OrthoINDEL uses two parameters to filter fragmented regions. A maximum distance of 50 kb was used to filter out fragments with too large a separation. We used a minimum distance of 50 bp to define continuous fragments. To be defined as an indel, the split fragments were required to be continuous in one species. Because we focus on indels, our pipeline removes other SVs including inversions. Lastly, we filtered out regions with average 50-bp mappability <0.7 in either species (compared with indel identification, we used more stringent mappability criteria here to eliminate false-positive lineage-biased RRRs) (Derrien et al. 2012). The last filtering step is critical to filter out false-positive lineage-biased regions (Supplemental Fig. S10). With this pipeline, we defined stringent 1:1 syntenic regions between human and chimpanzee tolerating indels up to 50 kb that can be converted reciprocally using OrthoINDEL.

Peak calling is sensitive to sequencing coverage and background signal. To better differentiate invariant peaks from lineage-biased peaks, we counted the number of reads mapped to each peak in both species and applied DESeq2 to quantify peak intensity difference (Love et al. 2014). We classified regions as “human-biased” if a peak was called only in the human genome by MACS2 and the number of ChIP-seq reads in the human peak is significantly higher than the number in its orthologous region in the chimpanzee genome (DESeq2, q < 0.0001); “chimpanzee-biased” regions were defined in a similar fashion. Lastly, we classified regions as “invariant” if ChIP-seq peaks were called in both species or if a peak was called in only one species, but the difference of ChIP-seq reads number between syntenic regions is not significant.

Enrichment analysis
We calculated the number of indel-RRR overlaps using the BEDTools intersect function (Quinlan and Hall 2010). To perform Fisher’s exact test with the hg38 genome as background, we used the BEDTools fisher function. The permutation test used BEDTools to shuffle indel coordinates and intersect with CGIs and promoters (Supplemental Methods). Briefly, we counted the number of intersections of human–chimpanzee indels ranging from 20–50 bp, 50–100 bp, and 100–150 bp with CGIs within promoters, CGIs outside of promoters, and promoters without CGIs, respectively. We then randomly shuffled indel coordinates 1000 times and repeated the intersection.

Identification of transcription factor binding motifs
We used FIMO from the MEME suite to find potential transcription factor binding sites within SVA elements (Bailey et al. 2009; Grant et al. 2011).

Data visualization
We generated bigWig files using methylQA and displayed them on the WashU Epigenome Browser (Li et al. 2015). All data are visualized on the WashU Epigenome Browser (Li et al. 2019). All ChIP-seq data were normalized to reads per genomic content (RPGC) using deepTools bamcoverage (Ramírez et al. 2016). Binding profiles and heatmaps were generated using deepTools2 (Supplemental Methods; Ramírez et al. 2016).
